DEVELOPMENT AND APPLICATION OF NEW ANALYTICAL EXPRESSIONS FOR DRAINABLE AND FILLABLE POROSITY IN SHALLOW UNCONFINED AQUIFERS

By

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To my parents, wife and sisters
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DEVELOPMENT AND APPLICATION OF NEW ANALYTICAL EXPRESSIONS FOR DRAINABLE AND FILLABLE POROSITY IN SHALLOW UNCONFINED AQUIFERS

By

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In shallow unconfined aquifers, the response of the phreatic surface to input and output water fluxes is controlled by two distinct storage parameters, drainable ($\lambda_d$) and fillable porosity ($\lambda_f$), which are applicable for water table (WT) drawdown and rise respectively. Although it has long been recognized that these parameters are affected by the soil-moisture flux in the unsaturated zone, very few studies have attempted incorporating these effects during $\lambda_d$ and $\lambda_f$ estimations. Moreover, only the $\lambda_d$ estimated from hydrostatic soil moisture profile is commonly used in most studies assuming implicitly that the two parameters are equal.

In this research, separate analytical expressions for $\lambda_d$ and $\lambda_f$ were developed by accounting for the dynamic soil moisture conditions in the unsaturated zone. The new expressions were then tested by simulating field WT dynamics, estimating the groundwater evapotranspiration (ET) from diurnal WT fluctuations, and developing a subsurface irrigation and drainage model under a WT management system in Florida. It was found that the occurrence of unsaturated zone fluxes due to evapotranspiration (ET) and recharge ($R_e$) can result in significantly different values of $\lambda_d$ and $\lambda_f$ for a given WT elevation, even when hysteresis is neglected. Implementation of these expressions in the hydraulic groundwater model to simulate the hourly WT dynamics produced significantly improved results as compared to a hydrostatic-$\lambda_d$. The analytical expressions also
enabled the development of a more complete and accurate method to estimate groundwater ET \((ET_g)\) from diurnal WT fluctuations, corroborating the results of simulation. The subsurface irrigation and drainage model, implemented with \(\lambda_d\) and \(\lambda_f\), showed that the model was capable of simulating WT drawdown and rise during respective drainage and irrigation events in a crop field.

Soil moisture in the field is always under a continuous redistribution phase, rarely reaching the hydrostatic condition, which therefore stresses the critical importance of incorporating the effects of moisture fluxes in \(\lambda_d\) and \(\lambda_f\) to study the hydrology of shallow phreatic aquifers. This research shows that incorporation of soil moisture dynamics in \(\lambda_d\) and \(\lambda_f\) estimations not only improves our ability to better understand and explain the hydrology but also aid in developing more robust and complete methods to study it.
Shallow water table (WT) areas commonly exist in many parts of the world where the water table remains at a small depth from the surface during most parts of the year. These areas include wetlands, riparian zones, forests as well as intensively cultivated agricultural lands. Regardless of the vegetation, management type, or climate, the hydrology of shallow WT areas is mostly dictated by the dynamics of phreatic surface [Nachabe, 2002]. The growth and development of plants in these areas are directly influenced by the dynamics of WT [Bierkens, 1998; Nachabe, 2002] because of its contribution in the soil moisture fluxes due to evaporation and root water uptake [Nichols, 1993, 1994; Loheide et al., 2005; Naumburg et al., 2005; Shah et al., 2007; Butler et al., 2007; Laio et al, 2009 etc.]. During precipitation the water table can quickly rise to the surface inducing rapid runoff [Gillham, 1984; Abdul and Gilham, 1989; Troch, 1992; Hilberts et al., 2007 etc.]. In natural environments, such rapid changes in WT elevations in response to the different hydrologic fluxes affect the growth and development characteristics of plants thus acting as a force that can control the species composition of the area. In agricultural land settings, such shallow WT poses a considerable challenge to crop production practices. Since most of the crops cannot sustain high levels of soil moisture, artificial drainage of the land is required for optimum aeration of the root-zone (Ritzema, 1994; Skaggs, 1991). On the other hand, the nature of the vegetation, characteristics of the aquifers, and human management practices largely affect the behavior of WT itself. It is, therefore, critically important to understand the interactions among the phreatic surface, associated hydrologic components, soil, and the plant characteristics that affect WT dynamics in natural as well as managed environments.
**Unconfined Aquifer Storage Parameters: Drainable and Fillable Porosity**

Shallow WT fluctuations in unconfined aquifers are conventionally estimated by means of a storage parameter called the drainable porosity, also termed as specific yield [Tang and Skaggs, 1975; Nachabe, 2002]. Drainable porosity ($\lambda_d$) represents the amount of water released by the aquifer when the WT drops by a unit distance [Bouwer, 1978; Freeze and Cherry, 1979; Neuman 1987; Nachabe, 2002; Hilberts et al., 2005; Laio et al., 2009]. Drainable porosity is the primary storage parameter of hydraulic groundwater models which have been used extensively to study the drainage of agricultural land [e.g. Glover, 1964; Moody, 1966; van Schilfgaarde, 1974, 1991; Bouarfa and Zimmer, 2000, etc.] as well as hillslope drainage [Childs, 1971; Brutsaert, 1994; Verhoest et al., 2002; Singh and Jaiswal, 2006; Hilberts et al., 2005, 2007, etc.].

Analogous to the concept of $\lambda_d$, another storage parameter of unconfined aquifer can be defined; the fillable porosity ($\lambda_f$), which is the amount of water that is stored or absorbed by an aquifer per unit rise in water table [Wilson et al., 1980; Bouwer, 1978; Sophocleous, 1991; Healy and Cook, 2002; Park and Parker, 2008]. Therefore, for every given WT position, both drainable and fillable porosity should be defined because the WT has the potential to move in both directions (upwards and downwards) depending on the conditions at the aquifer boundaries. These two parameters, therefore, respectively dictate recession and rise of the phreatic surface due to evapotranspiration (ET), rainfall and groundwater flow and are integral parameters of hydraulic groundwater flow models [e.g., Chapman and Dressler, 1984; Brutsaert, 1994; Hilberts et al., 2005]. It is well-accepted in the literature that fillable porosity $\lambda_f$ is normally smaller than $\lambda_d$ due to hysteresis in soil moisture retention [Bouwer and Jackson, 1974; Bouwer, 1978; Sophocleous, 1991]. However, hysteretic effects have rarely been considered in hydraulic groundwater models, treating the two parameters essentially as one [e.g., Troch et al., 2003; Hilberts et al., 2005, Pumo et al., 2009, etc.]. Stauffer et al. [1992] was the first and perhaps the
only study to apply a “hysteretic storativity concept” to an unconfined groundwater flow model. He applied the Maulem’s, dependent domain theory of hysteresis [Mualem, 1984] to a soil moisture retention curve model and numerically estimated $\lambda_d$ and $\lambda_f$ during model simulation. After Stauffer et al [1992], hysteretic storage parameters have seldom been developed or applied in hydraulic groundwater flow models and the $\lambda_d$ and $\lambda_f$ have been treated as the common, single parameter. It is possibly due to the complex nature of soil moisture hysteresis which is difficult to incorporate in such hydraulic groundwater models. However, under shallow WT, hysteresis effect may not be as significant as compared to other factors. It has also been shown by previous studies that hysteretic effects on soil moisture status and flow may not be as significant in such fields with shallow WT and without ponding conditions [van Dam et al., 1996].

Despite the extensive use, the nature of the application of $\lambda_d$ in WT dynamics studies, however, has not been consistent. Many studies have treated $\lambda_d$ as a constant parameter [Glover, 1966; Skaggs, 1980, 1991; Singh and Jaiswal, 2006] although it has long been recognized that $\lambda_d$ is a dynamic parameter that depends on the moisture status of the soil profile [Childs, 1960]. According to Nachabe [2002], $\lambda_d$ is constant only if the aquifer response is linear, i.e., the volume of drainage changes linearly with WT fluctuation, which does not occur in shallow WT environments [Duke, 1972; Nachabe, 2002]. Some studies estimated $\lambda_d$ by recording the rise in WT during a brief rainfall period and dividing the amount of rainfall by the rise in WT [e.g., Gerla, 1992]. On the other hand, several studies have also used a variable $\lambda_d$ estimated using the soil moisture retention curve to study the subsurface lateral flow and WT dynamics [e.g., Duke, 1972; Bierkens, 1998; Hilberts et al, 2005, 2007; Laio et al., 2009; Pumo et al., 2010, etc.], which is also the most reliable method.
It is well known that the WT dynamics in unconfined aquifers is significantly influenced by the capillary fringe [Cartwright et al., 2006; Hilberts and Troch, 2006; Nielsen and Perrochet, 2000]. Hence the magnitudes of $\lambda_d$ and $\lambda_f$ parameters also depend on both the position of the WT and the soil moisture status in the unsaturated zone [Hillel, 1998]. In an attempt to incorporate these effects, different expressions for $\lambda_d$ that account for soil capillary properties have been developed in the past. The earliest of such expressions was given by Duke [1972] who developed an equation for $\lambda_d$ as a function of WT depth using Brooks-Corey (B-C) soil water retention model [Brooks and Corey, 1964]. Pandey et al. [1992] developed an empirical equation to estimate $\lambda_d$ to apply in subsurface drainage problems. Bierkens [1998] developed another analytical expression for $\lambda_d$, as a function of WT depth, by integrating the hydrostatic soil moisture profile above the WT. His expression was based on a variant of the van Genuchten [1980] soil water retention curve model. Nachabe [2002] developed an expression for $\lambda_d$ also using B-C retention model. His study suggested that $\lambda_d$ was significantly affected by soil capillarity and also depended on the duration of drainage. Hilberts et al. [2005] developed an expression analogous to Bierkens’ equation using the same concepts and applied in a hillslope subsurface storage and WT dynamics model [Troch et al., 2003; Hilberts et al., 2007].

Despite the difference in their approaches, all of these methods were based on one common assumption: hydrostatic conditions (zero-flux) in the unsaturated zone above the water table. This assumption facilitated the analytical integration of the unsaturated soil moisture profile thus the resulting expressions that partially accounted for the effects of unsaturated zone moisture retention in estimated $\lambda_d$ values. Although such hydrostatic moisture profile based expressions have been successfully used in several studies, it is well known that the storage parameters of unconfined aquifers are also significantly affected by moisture flux in the
unsaturated zone [Childs, 1960; Chapman and Dressler, 1984; Chapman, 1995; Tritscher et al., 2000; Brutsaert, 2005]. Tritscher et al. [2000] used a two-dimensional flow model to show that the error associated with estimation of λ_d from static moisture profile may be as high as 35% when the infiltration rates are high. Chapman and Dressler [1984] showed that fillable porosity was directly affected by the vertical recharge rate (R_e) and approached zero as R_e increased. Thus the application of λ_d expressions based on static moisture profile should ideally be limited to hydrostatic conditions. However, this condition rarely holds in most field conditions, unless the soil is very coarse [Bear and Cheng, 2008]. Assumption of hydrostatic moisture profile especially doesn’t hold in shallow WT environments such as wetlands, riparian areas and crop fields with controlled WT where a significant portion of the ET demand is directly fulfilled by WT [Loheide et al., 2005; Nachabe, 2005; Butler et al., 2007].

Another important limitation of λ_d expressions based on the hydrostatic assumption is that they result in the same value for both λ_d and λ_f for a given WT depth. These parameters have been treated as common in all of the hydraulic groundwater models and their derivatives [e.g., Troch, 2003; Hilberts et al., 2005; Pumo et al., 2009, etc.] which may potentially lead to erroneous results especially during water table rise scenarios.

Recently there have been attempts to incorporate the effect of unsaturated zone moisture fluxes in λ_d of unconfined aquifers. Laio et al. [2009] used B-C soil moisture retention model to develop an approximate analytical expression for λ_d assuming steady-steady state soil moisture profile above WT. Their solution was based on steady upward flow from WT in presence of roots described by an exponential distribution function. The final expression was expressed in terms of a critical WT depth term, representing the point of transition from shallow to deep WT. However, this critical WT depth lacked a closed-form expression and was approximated by a
hypergeometric function. Although their method was a step forward towards incorporating the effect of unsaturated zone hydrodynamics in $\lambda_d$ expressions these authors also failed to investigate the effect of dynamic soil moisture conditions on $\lambda_f$.

Theoretically, $\lambda_f$ is a parameter related to the storage-deficit of the aquifer, which can potentially behave differently than $\lambda_d$ in presence of such fluxes because of the inequality in the change in volume of soil moisture during a unit rise and a unit drawdown in WT respectively. Drainable porosity describes the volume of water lost when the WT moves down by a unit distance from a given position. On the other hand, fillable porosity is the volume of water that is taken by the aquifer when the WT moves upward from the same given position. Therefore, in presence of unsaturated zone fluxes (ET or $R_e$) in the soil, these changes will be unequal which can leads to different magnitudes of $\lambda_d$ and $\lambda_f$ parameters for the same WT elevation. However, theoretical investigation of the behavior of aquifer fillable porosity during such dynamic conditions has never been investigated. Since the shallow WT can fluctuate rapidly even in a short period of time, it is important to consider separate $\lambda_d$ and $\lambda_f$ parameters to describe corresponding WT behavior especially in presence of unsaturated zone fluxes due to ET and $R_e$.

**Applications of $\lambda_d$ and $\lambda_f$**

**Modeling Shallow Water Table Dynamics**

The fundamental processes that determine dynamics of WT in aquifers are described by the principles of water movement through the soil. The most rigorous and exact approach to describe these process is by solving the Richards’ equation which governs the flow through variably saturated porous media [Pikul et al., 1974; Tang and Skaggs, 1977]. The Richards’ equation has been used extensively to model the saturated-unsaturated flow at various spatial scales in natural as well as managed environments [e.g., Freeze, 1971; Feddes et al., 1974; Taylor and Luthin, 1969; Khan and Rushton, 1995, 1996; van Dam and Feddes, 2000 etc.].
However, due to high nonlinearity and associated difficulties of solving this equation, simpler approaches are usually sought [Pikul et al., 1974; Hilberts et al., 2005]. One such assumption most commonly employed in shallow unconfined aquifers is based on the *Dupuit-Forchheimer* (D-F) theory. This theory assumes that in shallow phreatic aquifers, the vertical hydraulic gradient is small enough to be neglected and the flow occurs essentially below the WT in horizontal direction only [Childs, 1971; van Schilfgaarde et al., 1974; Skaggs, 1991]. The D-F theory holds true when the lateral dimension of the aquifer is very large as compared to its vertical extent which normally results in small inclinations in the water table profile [van Schilfgaarde, 1974]. When D-F assumptions are valid, the lateral soil moisture flux through the soil is governed by the Darcy equation as

\[ q = -K_s h \frac{dh}{dx} \]  

(1-1)

where, \( q \) = subsurface lateral flux; \( h \) = the height of WT from the reference datum (usually the impermeable barrier at the bottom of the aquifer), \( K_s \) is the lateral saturated hydraulic conductivity and \( x \) is the space coordinate. Conservation of mass requires that

\[ \lambda \frac{dh}{dt} = -\frac{dq}{dx} + \mu \]

Or,

\[ \lambda \frac{dh}{dt} = K_s \frac{d}{dx} \left( h \frac{dh}{dx} \right) + \mu \]  

(1-2)

where, \( \lambda \) = storage parameter of the aquifer and \( \mu \) is the sink (ET) or source (\( R_s \)) system.

Equation 1-2 is known as the Boussinesq equation in which term \( \lambda \) can be either \( \lambda_d \) or \( \lambda_f \) depending on whether the aquifer is in the state of WT drawdown or WT rise respectively.

During drainage, the direction of flow is towards the ditches, which drives the WT downwards and therefore, \( \lambda = \lambda_d \). If the flow is towards the aquifer, \( \lambda = \lambda_f \) because the rate of WT rise
depends on the storage deficit of the aquifer. Note that the last term in equation 1-2 can either be ET or $R_e$. These components have opposite effects on the WT of the aquifer and therefore, they are dictated respectively by $\lambda_d$ and $\lambda_f$. However, this particular approach has never been adopted in hydraulic groundwater modeling studies [e.g., Hornberger et al., 1970; Brutsaert, 1994; Skaggs, 1991; Troch et al., 2003 etc.] which may potentially lead to erroneous estimations of WT fluctuations.

**Estimation of Groundwater Evapotranspiration ($ET_g$)**

In shallow groundwater environments such as riparian zone and wetlands, evapotranspiration is a significant component of the water balance [Dolan et al., 1984; Loheide et al., 2005; Luo and Sophocleous, 2010] where the shallow WT normally supports most of the plant water requirement [Shah et al., 2007; Gribovszki et al., 2007; Lowry and Loheide, 2010]. Consequently a distinct diurnal fluctuation pattern in the water table (WT) hydrograph is usually observed [Schilling, 2007; Butler et al., 2007; Nachabe et al., 2005; Loheide, 2008]. This diurnal water table fluctuation (DWTF) corresponds to the 24-hour variation cycles in the climatic variables, solar radiation, temperature and humidity [Gribovszki et al., 2007; Gribovszki et al., 2010].

Since there is normally a direct link between the vegetation water uptake and the water table (WT), ET loss from the shallow WT environments may be estimated from the soil water balance. A general expression for $ET_g$ from soil water balance can be expressed as

$$ET_g = q + R_e - \lambda \frac{dh}{dt}$$

(1-3)

where,

$h = \text{water table elevation}$

$\lambda = \text{the storage parameter of the aquifer}$

$q = \text{groundwater flux}$,
\( R_e = \) rainfall recharge, and
\( ETg = \) groundwater evapotranspiration loss

The last term in the right hand side of equation 1-3 gives the total change in soil moisture storage per unit area expressed in term of WT elevation. White [1932] was the first to analyze and report such direct link between ET and WT fluctuations in shallow water table areas. His method, however, was based on different simplifying assumptions. The most critical assumption of White’s original method was that the \( \lambda \) parameter was assumed constant. He also assumed that the rate of groundwater flow to the area was constant over the 24 hour period and could be estimated by the rate of WT rise between midnight and 4:00 am when \( ETg \) is negligible or zero. The original White’s equation for estimating \( ETg \) can be expressed as

\[
ET_g = \lambda_d (24r + \Delta S)
\]  

(1-4)

where, \( \lambda_d \) is the drainable porosity (or specific yield) of the aquifer, \( r \) is the rate of WT rise between midnight and 4:00 AM when the ET is considered zero, and \( \Delta S \) is the rise or fall of WT during the 24-hour period. Equation 1-4 suggests that, even if the underlying assumptions hold true, White’s original method is applicable only at daily time scales since it considers constant groundwater flow over the 24-hour period. However, subdaily scale ET estimation may be necessary in many situations which cannot be obtained from equation 1-4.

Following White, many have investigated the DWTF phenomenon to estimate the shallow groundwater consumption by vegetation [e.g., Dolan et al., 1984; Gerla, 1992; Loheide et al., 2005, Nachabe et al., 2005, Lautz, 2008, Butler et al., 2007, Gribovszki et al., 2007; Loheide, 2008; Mould et al., 2010, etc.]. Some of these studies also evaluated the underlying
assumptions in equation 1-3 and found that the assumptions of constant groundwater flow as well as constant $\lambda_d$ are normally not met in field conditions [Gerla, 1992; Loheide et al., 2005; Gribovszki et al., 2007]. Gerla [1992] reported that assuming a constant $\lambda_d$ in equation resulted in excessively large $ET_g$ estimation in a wetland environment. According to Loheide et al. [2005], the assumption of constant $\lambda_d$ creates problems in the White method and showed, with the help of soil moisture flow simulations, that $\lambda_d$ was critically important in accurately estimating $ET_g$ from DWTF. Loheide [2008] proposed a modified White method to estimate $ET_g$ at subdaily time scales considering variable groundwater flow rates. Gribovszki et al. [2007] presented a similar modification to Loheide [2008] by using successive steady-state groundwater flow and a WT dependent $\lambda_d$ expression developed by Nachabe [2005].

While these studies have shown that implementation of variable groundwater flow and variable $\lambda_d$ can generally resolve the inherent limitations of the White method and allow for subdaily scale $ET_g$ estimations [Gribovszki et al., 2007; Loheide, 2008], their study neglected the effect of unsaturated flux in $\lambda_d$ estimations. As a result, these methods cannot accurately reflect rapid WT fluctuations shown by shallow phreatic aquifers during ET and precipitation. Especially during and after rainfall events, the assumption of hydrostatic moisture profile is severely disrupted because of rapid soil moisture redistribution. Therefore, the actual $\lambda_d$ value may differ significantly from the estimated hydrostatic-$\lambda_d$ [Tritscher et al., 2000] consequently resulting in potentially erroneous $ET_g$ estimations. In order to avoid these complications during rainfall, most of the White based $ET_g$ estimation methods presented in the literature typically omitted the days with precipitation in their calculations [e.g. Dolan et al., 1984; Gribovszki et al., 2007; Schilling, 2007]. As a result, these estimations are usually discontinuous at smaller time
scales (i.e. hourly or daily) although fairly good estimates may be obtained on larger time scales (e.g. monthly average) if the rainfall frequency is low.

The potential impact of rainfall on the underlying assumptions of White based $ET_g$ estimation methods results due to the disturbance in the DWTF signal of the WT. However, the literature provides very little work on the extent to which the $ET_g$ estimates are affected when days with rainfall are included in the calculation. In one study, Gerla [1992] included the days with precipitation to estimate ET using the White method. He modified the original White method by accounting for the effect of infiltration on water table rise thus extending the use of the method even during precipitation events. However, his study not only used a $\lambda_d$ value estimated from the WT rise during rainfall but also failed to implement separate $\lambda_d$ and $\lambda_f$ during $ET_g$ calculations. Theoretically, use of these distinct $\lambda_d$ and $\lambda_f$ parameters should enable the estimation of $ET_g$ from soil moisture balance continuously regardless of the rainfall events because it is possible to estimate the contribution of rainfall to change in WT if reliable estimates of recharge ($R_e$) and corresponding $\lambda_f$ are available.

Hypotheses and Objectives

The main hypotheses of this research are

1. Drainable and fillable porosity of shallow unconfined aquifers can behave in significantly different manner in presence of unsaturated-zone vertical fluxes due to recharge and evapotranspiration respectively.

2. Dynamics of shallow WT in unconfined aquifers can be understood and modeled significantly better when both drainable and fillable porosity parameters are implemented in the hydraulic groundwater model (Boussinesq equation).

3. Estimates of direct groundwater evapotranspiration based on the White [1932] method can be reliably extended to rainfall periods as well, if reliable estimate of fillable porosity is available.

The major objectives of the research are
1. To develop separate analytical expressions for the drainable and fillable porosity of unconfined aquifer in presence of hydrodynamic conditions and compare with a previous expression estimated under hydrostatic conditions.

2. To implement the flux-dependent $\lambda_d$ and $\lambda_f$ parameters in the hydraulic groundwater model to simulate shallow WT dynamics and compare with the model implemented only with hydrostatic based $\lambda_d$ parameter.

3. To develop a modified White [1932] based method to estimate groundwater ET from diurnal WT fluctuations using the flux dependent $\lambda_d$ and $\lambda_f$ parameters.

4. To develop and test a shallow WT dynamics model, implementing the flux dependent $\lambda_d$ and $\lambda_f$ in order to study the effectiveness of WT control method under a subsurface drainage and irrigation system.

**Outline of Dissertation**

This dissertation presents the development of analytical expressions for the storage parameters $\lambda_d$ and $\lambda_f$ and their application in estimation of $ET_g$, and development of a WT dynamics model for agricultural fields with WT management systems. In Chapter 2, analytical expressions for $\lambda_d$ and $\lambda_f$ are developed following their proper definitions from the literature. Firstly the generalized analytical expressions of the two parameters under hydrodynamic conditions are presented. More specific expressions that account for the steady-state unsaturated flow to and from WT are then developed. These expressions are then implemented in the Boussinesq equation to simulate the WT dynamics in a field site in northeast Florida and compared with the field data. Specific comparisons are also presented with the Boussinesq equation implemented with the single $\lambda_d$ parameter estimated from the hydrostatic moisture profile.

In Chapter 3, the $\lambda_d$ and $\lambda_f$ expressions are applied to estimate $ET_g$ using the observed diurnal fluctuations in WT data. The common White based method of estimating $ET_g$ is modified by implementing the two parameters in order to apply the method even during and after rainfall periods which is a significant advantage over the previously available methods.
Application of the modified method during rainfall periods is enabled by the use of recharge dependent $\lambda_f$ which allowed for removing the effect of recharge from observed WT fluctuation time series. The modified method is then used to estimate $ET_g$ from a potato field in northeast Florida during 100 days period over the two year period from 2010 to 2011. The results are then compared with the standard Penman-Monteith ET estimates and also the $ET_g$ estimates from a previously available single, hydrostatic-$\lambda_d$ parameter method.

Chapter 4 describes the development, application and evaluation of a WT dynamics model for agricultural fields under a WT management system typically practiced in Florida. Conceptual models of the WT movement during subsurface irrigation and drainage via ditches and furrows are presented. The 1D Boussinesq equation that governs the groundwater movement in shallow unconfined aquifer also governs the flow in these ditch drained lands. The governing equation implemented with separate $\lambda_d$ and $\lambda_f$ is solved numerically by implicit finite difference method to simulate WT drawdown during drainage and WT rise during irrigation. The model results are compared to the WT data from two fields during two consecutive potato seasons in 2010 and 2011. The sensitivity of the model to some important field and soil parameters is also assessed. Simulated behavior of WT under an alternative potentially more water saving management scenario is also presented in this chapter. Finally in Chapter, 5 the summary of the major findings and the main conclusions of the research are presented.
Background

In many low-lying, humid regions where shallow water tables (WTs) are common, vegetation growth is largely driven by the water table dynamics [Bierkens, 1998; Nachabe, 2002]. It is therefore important to understand WT dynamics for efficient management of crops as well as natural vegetation. Shallow WT fluctuations in unconfined aquifers are generally studied by means of a storage parameter called the drainable porosity (or specific yield). Drainable porosity ($\lambda_d$) represents the amount of water released by the aquifer when the water table drops by a unit distance [Bouwer, 1978; Freeze and Cherry, 1979; Neuman 1987]. It is estimated as the change in total water storage of the aquifer, $S$, per unit drop in WT elevation [Bear 1972]:

$$\lambda_d(h) = \frac{\Delta S}{\Delta h}$$  \hspace{1cm} (2-1)

Where, $h$ is height of the phreatic surface above a reference datum. Analogous to the concept of $\lambda_d$ is the fillable porosity ($\lambda_f$), which can be defined as the amount of water that is stored or absorbed by an aquifer per unit rise in water table [Wilson et al., 1980; Bouwer, 1978; Sophocleous, 1991; Park and Parker, 2008]. It can be estimated, in a similar manner to $\lambda_d$, as the change in total storage deficit of the aquifer, $D_s$, per unit rise in WT elevation:

$$\lambda_f(h) = -\frac{\Delta D_s}{\Delta h}$$  \hspace{1cm} (2-2)

Drainable and fillable porosity play key roles in WT fluctuations due to evapotranspiration (ET), rainfall, and are integral parameters of Boussinesq-type groundwater flow models [e.g., Chapman and Dressler, 1984; Brutsaert, 1994; Chapman, 2005; Hilberts et al., 2005]. These
parameters are also required when ET is estimated from groundwater fluctuations [Loheide et al., 2005; Gribovszki et al., 2007; Loheide, 2008].

In shallow phreatic aquifers, WT fluctuations are significantly influenced by the soil capillary retention [Childs, 1960; Gillham, 1984; Nielsen and Perrochet, 2000; Healy and Cook, 2002; Hilberts and Troch, 2006; Cartwright et al., 2006]. Hence the magnitudes of $\lambda_d$ and $\lambda_f$ depend on the position of WT and soil moisture status in the unsaturated zone [Hillel, 1998], rather than being unique for a given soil type [Hilberts et al., 2005]. While existing analytical expressions for $\lambda_d$ do account for the effect of unsaturated-zone soil moisture [e.g., Duke, 1972; Bierkens, 1998; Nachabe, 2002; Hilberts et al., 2005], these methods assume hydrostatic conditions in the soil profile.

Although $\lambda_d$ expressions derived using static moisture profiles have been successfully used in several studies, it is well known that the storage parameters of unconfined aquifers are also significantly affected by moisture flux in the unsaturated zone [Childs, 1960; Chapman and Dressler, 1984; Chapman, 1995; Tritscher et al., 2000; Brutsaert, 2005]. Tritscher et al. [2000] used a two-dimensional flow model to show that the error associated with estimation of $\lambda_d$ from static moisture profile may be as high as 35% when the infiltration rates are high. Chapman and Dressler [1984] showed that fillable porosity was directly affected by the vertical recharge rate ($R_e$) and approached zero as the recharge rate increased. Thus the application of $\lambda_d$ expressions based on static moisture profile should ideally be limited to hydrostatic conditions. However, this condition rarely holds in most field situations, unless the soil is very coarse [Bear and Cheng, 2008]. The assumption of hydrostatic pressure distribution is especially inappropriate in shallow WT environments such as wetlands, riparian areas and crop fields with controlled WT where a
significant portion of the ET demand is directly fulfilled by the water table [Loheide et al., 2005; Nachabe, 2005; Butler et al., 2007].

Another important limitation of $\lambda_d$ expressions based on the hydrostatic assumption is that they result in the same value for both $\lambda_d$ and $\lambda_f$ for a given WT depth. Normally due to hysteresis in moisture retention and release by soils, $\lambda_f$ is regarded as smaller than $\lambda_d$ [Bouwer and Jackson, 1974; Bouwer, 1978; Sophocleous, 1991]. As we show below, as the moisture profile deviates from a hydrostatic pressure distribution, different $\lambda_d$ and $\lambda_f$ values result for a given WT depth, even when hysteresis is neglected. Yet, separate $\lambda_d$ and $\lambda_f$ parameters for water table decline and rise have not previously been implemented in unconfined groundwater flow studies. These parameters have also been treated as common in all of the hydraulic groundwater models and their derivatives [e.g., Troch, 2003; Hilberts et al., 2005; Pumo et al., 2009, etc.] which may potentially lead to erroneous results especially during water table rise scenarios.

Laio et al. [2009] recently developed an approach for incorporating dynamic moisture profiles explicitly into $\lambda_d$ estimations. These authors used the soil moisture retention model of Brooks and Corey [1964] to develop an approximate analytical expression for $\lambda_d$, accounting for both recharge and root uptake. Their relation was expressed in terms of a critical WT depth term, representing the point of transition from shallow to deep WT. However, this critical WT depth lacked a closed-form expression and was approximated by a hypergeometric function. Also, these authors did not separately consider both $\lambda_d$ and $\lambda_f$.

The main objective of this study was to develop closed-form expressions for $\lambda_d$ and $\lambda_f$ as separate aquifer parameters that account for vertical soil moisture flow in the unsaturated zone. Assuming no hysteresis in soil moisture retention, steady-state vertical soil moisture fluxes (ET and recharge) were considered in successive times to estimate the moisture profile above the
water table. The moisture profile was then integrated to obtain $S$ and $D_s$. Here, we use a variant of the *van Genuchten* [1980] retention model [*Troch*, 1992; *Troch et al.*, 1993] which enables the analytical solutions for both $S$ and $D_s$. Differentiating these according to Equations 1 and 2 results in generalized expressions for $\lambda_d$ and $\lambda_f$ in terms of the matric suction at the soil surface ($\psi_T$) and soil hydraulic parameters. This study therefore provides an expression for a separate $\lambda_f$ parameter for a given WT depth, in addition to $\lambda_d$, which has not been previously investigated. The effects of unsaturated soil moisture flux to and from the water table are accounted for by estimating $\psi_T$ under successive-steady flux ET or recharge ($R_e$) using the hydraulic conductivity-model of *Gardner* [1958]. The final $\lambda_d$ and $\lambda_f$ expressions are flexible enough to incorporate any predefined ET and $R_e$ functions. Finally, the analytical expressions are implemented in a one dimensional hydraulic groundwater model to simulate shallow WT dynamics in a 15-ha potato field in northeast Florida, managed under a WT control system. Simulated hourly WT dynamics are compared for models implemented with only a single, hydrostatic profile-based expression for $\lambda_d$, and with distinct $\lambda_d$ and $\lambda_f$ expressions developed in this study based on dynamic moisture profile.

**Theory**

**Saturated and Unsaturated Soil Water Storage**

We consider an unconfined aquifer with a shallow WT and plants growing at the soil surface. The aquifer is assumed to be incompressible with no hysteresis in the soil moisture retention behavior. The water balance components of the cross-section of such an aquifer with a shallow WT bounded by a ditch or stream are illustrated in Figure 2-1. A conceptual representation of saturated ($S_s$) and unsaturated storage ($U_s$), storage deficit ($D_s$), and the changes in these quantities due to WT decline or rise is presented in Figure 2-2. The unsaturated soil water storage is normally calculated by integrating the moisture profile from the WT to the soil
surface with respect to depth \( z \). However, analytical integration is possible only when the pressure distribution in the unsaturated zone is hydrostatic so that the matric suction can be expressed as height above the WT \([e.g., \text{Bierkens} 1998; \text{Hilberts et al.}, 2005]\). Under dynamic conditions, \( \psi \) is a non-linear function of \( z \) and depends on the magnitude of the flux, precluding analytical integration of the moisture profile with respect to depth. Since soil water content in the unsaturated zone is directly related to the soil matric suction \( (\psi) \) at each respective depth, we integrated with respect to \( \psi_z = \psi(z) \). Integrating the moisture profile with respect to \( \psi_z \) from the limit of WT depth \( (\psi_z = 0) \) to the soil surface \( (\psi_z = \psi_r) \) yields the total unsaturated storage. Therefore, from Figure 2-2 the total soil water storage \( (S) \) of the profile can be estimated by taking the sum of the area \( (\theta_s - \theta_r)h \) and area under \( ABDE \) as

\[
S = (\theta_s - \theta_r)h + \left[ -\int_0^{\psi_r} (\theta(\psi_z) - \theta_r) \, d\psi_z \right]
\]  

where \( \psi_z \) is the matric suction at respective depth \( z \), \( \theta_s \) and \( \theta_r \) are the saturated and residual soil water contents respectively, \( h \) is height of the phreatic surface above the reference datum, and \( \psi_r \) corresponds to the matric suction at the soil surface. The two terms on the right hand side of equation 2-3 respectively represent the saturated and unsaturated water storage fractions of aquifer vertical cross-section. Integration of the second term in equation 2-3 requires a constitutive \( \theta(\psi) \) relationship. In this study we used a modified \textit{van Genuchten} [1980] retention model introduced by \textit{Troch} [1992], which has been used in different studies \([e.g., \text{Bierkens}, 1998; \text{Hilberts et al.}, 2005, 2007] \):

\[
\theta(\psi) = \theta_r + (\theta_s - \theta_r) \left[ \frac{1}{(1 + (\alpha \psi)^n)^{1+\frac{1}{n}}} \right]^{1+\frac{1}{n}}
\]  

where \( \alpha \) and \( n \) are soil-specific parameters of the modified model. This modified relationship enables integration of equation 3 which gives
\[
S = \left( \theta_s - \theta_r \right) \left\{ h - \psi_T \left[ 1 + \left( a \psi_T \right)^n \right]^{\frac{1}{n}} \right\}
\]  
(2-5)

where \( a' \) and \( n' \) are the parameters of the modified van Genuchten model.

**Total Storage Deficit**

The total water storage deficit of the aquifer vertical cross-section can be estimated by taking the area under \( BCD \) in Figure 2-2 as

\[
D_s = -\int_0^{\psi_T} (\theta_s - \theta(\psi_x)) d\psi_x
\]

(2-6)

Substituting \( \theta(\psi_x) \) in equation 6 with the modified van Genuchten retention model and integrating yields

\[
D_s = (\theta_s - \theta_r)(\psi_T) \left[ -1 + \left( 1 + \left( a(\psi_T) \right)^n \right) \frac{1}{n} \right]
\]

(2-7)

It also follows from equations 2-5 and 2-7 that \( S + D_s = (\theta_s - \theta_r)[(-\psi_T) + h] \), which gives the total area of the profile above the reference datum.

**General Expressions for Drainable and Fillable Porosity**

From Equations 2-5 and 2-7, \( \lambda_d \) and \( \lambda_f \) can be obtained by taking the derivative of \( S \) and \( D_s \) with respect to \( h \):

\[
\lambda_d = (\theta_s - \theta_r) \left\{ 1 - \left( \frac{d(\psi_T)}{dh} \right) \left[ 1 + \left( a \psi_T \right)^n \right] \left( \frac{n+1}{n} \right) \right\}
\]

(2-8)

\[
\lambda_f = (\theta_s - \theta_r) \left( \frac{d(\psi_T)}{dh} \right) \left[ 1 - \left( 1 + \left( a \psi_T \right)^n \right) \frac{n+1}{n} \right]
\]

(2-9)

It also follows from equations 2-8 and 2-9 that, for a given water table depth, \( \lambda_d \) and \( \lambda_f \) can be related as

\[
\lambda_f = \lambda_d - (\theta_s - \theta_r) \left( 1 - \frac{d(\psi_T)}{dh} \right)
\]

(2-10)

Equations 2-8 and 2-9 are generalized expressions for \( \lambda_d \) and \( \lambda_f \) as functions of the WT elevation. Note that both equations 2-8 and 2-9 can be defined for every WT elevation because at every
point below the surface, the phreatic surface has a potential to move either upwards or downwards based on the combined effect of the water balance components (see Figure 2-1). In both equations 2-8 and 2-9, the critical variable is $\psi_T$, matric suction at the soil surface. Under hydrostatic condition, the magnitude of $\psi_T$ is simply the height of the soil surface above the WT, i.e., $\psi_T = h - H$. When this is substituted in equations 2-8 and 2-9, we get $\frac{d(\psi_T)}{dh} = 1$, yielding the drainable porosity expression derived by Bierkens [1998] and Hilberts et al. [2005]:

$$\lambda_d(h) = \lambda_f(h) = (\theta_s - \theta_r) \left\{ 1 - \left( 1 + \left( \alpha(h - H) \right)^n \right)^{-\frac{n+1}{n}} \right\}$$  

(2-11)

Under dynamic soil moisture conditions, on the other hand, the relationship between $\psi_T$ and $h$ becomes much more complex. Due to highly nonlinear nature of soil water movement, normally it is not possible to obtain an explicit relationship between $h$ and $\psi_T$ under transient flow conditions. The simplified approach used here is to assume steady state flux above the water table, which enables direct calculation of $\psi_T$.

**Steady Flow from/to the Water Table (Evaporation, Root Water Uptake and Infiltration)**

In this section, expressions for $\psi_T$ are derived considering steady state evaporation or infiltration under shallow water table conditions.

**Evaporation and root water uptake**

From Equations 2-5 and 2-7, $\lambda_d$ and $\lambda_f$ can be obtained by taking the derivative of $S$ and $D_s$ with respect to $h$:

$$\lambda_d = (\theta_s - \theta_r) \left\{ 1 - \left( \frac{d(\psi_T)}{dh} \right) \left( 1 + \left( \alpha \psi_T \right)^n \right)^{-\frac{n+1}{n}} \right\}$$  

(2-8)

$$\lambda_f = (\theta_s - \theta_r) \left( \frac{d(\psi_T)}{dh} \right) \left[ 1 - (1 + \left( \alpha \psi_T \right)^n)^{-\frac{n+1}{n}} \right]$$  

(2-9)
It also follows from equations 2-8 and 2-9 that, for a given water table depth, \( \lambda_d \) and \( \lambda_f \) can be related as

\[
\lambda_f = \lambda_d - (\theta_s - \theta_r) \left( 1 - \frac{d(\psi_T)}{dh} \right)
\]

(2-10)

Equations 2-8 and 2-9 are generalized expressions for \( \lambda_d \) and \( \lambda_f \) as functions of the WT elevation. Note that both equations 2-8 and 2-9 can be defined for every WT elevation because at every point below the surface, the phreatic surface has a potential to move either upwards or downwards based on the combined effect of the water balance components (see Figure 2-1). In both equations 2-8 and 2-9, the critical variable is \( \psi_T \), matric suction at the soil surface. Under hydrostatic condition, the magnitude of \( \psi_T \) is simply the height of the soil surface above the WT, i.e., \( \psi_T = h - H \). When this is substituted in equations 2-8 and 2-9, we get \( \frac{d(\psi_T)}{dh} = 1 \), yielding the drainable porosity expression derived by Bierkens [1998] and Hilberts et al. [2005]:

\[
\lambda_d(h) = \lambda_f(h) = (\theta_s - \theta_r) \left\{ 1 - \left( 1 + \left( \alpha(h - H) \right)^{n} \right)^{-\frac{(n+1)}{n}} \right\}
\]

(2-11)

Under dynamic soil moisture conditions, on the other hand, the relationship between \( \psi_T \) and \( h \) becomes much more complex. Due to highly nonlinear nature of soil water movement, normally it is not possible to obtain an explicit relationship between \( h \) and \( \psi_T \) under transient flow conditions. The simplified approach used here is to assume steady state flux above the water table, which enables direct calculation of \( \psi_T \).

**Evapotranspiration approximation based on water table elevation**

In case of natural shallow WT environments such as wetlands and riparian zones, most of the vegetation is well adapted to perpetual saturated or near-saturated conditions, thus showing little or no reduction in root water uptake due to high soil moisture [Laio et al., 2009; Pumo et
al., 2009; Tamea et al., 2009, 2010]. In such environments, it is reasonable to assume that

\[ ET \approx q_0 \] which yields

\[ \psi_T = \frac{1}{a_G} \ln \left( \frac{(K_s + ET) \exp(a_G (h - H)) - ET}{K_s} \right) \] (2-16)

Both linear [Harbaugh, 2005] and exponential [Shah et al., 2007] relations have been developed to estimate ET from water table elevation. In this study we used the exponential function given by Shah et al. [2007] since it considers both soil and vegetation type:

\[ ET = PET \times \exp(-\beta (d' - h)) \quad h \leq d' \] (2-17)

\[ ET = PET \quad h > d' \] (2-18)

where \( d' \) and \( \beta \) are parameters that respectively describe ET ‘transition depth’ and the rate of reduction in ET as the WT falls below \( d' \) (Figure 2-1). It was also shown by Shah et al. [2007] that the magnitude of \( \beta \) and \( d' \) depend both on the type of soil and vegetation. Smaller \( \beta \) and larger \( d' \) values correspond to finer soils and deeper rooted vegetation and vice versa. Figure 2-3 illustrates the relationship in 2-17 and 2-18 for clay, loam, and sand soils under typical grass type vegetation based on the parameter values reported by Shah et al. [2007]. Note that for clay soil the reduction in the rate of ET with decreasing WT is much smaller followed by loam and sand.

Estimation of the ET parameters for our field site is discussed in the parameter estimation section. The parameter values for the ET vs WT relationship for major soil textural classes under different vegetation types can be found in Shah et al. [2007]. Once the ET is estimated from equations 2-17 and 2-18, \( \psi_T \) can be easily calculated by using equation 2-16. Comparison between the \( \psi_T \) value calculated by equation 2-16 and the values obtained from numerical solution of Richards equation, using HYDRUS 1D model [Simunek et al., 1998], for estimated ET values for our study area is presented in Figure 2-4. Note that the numerical solution and the
steady state analytical solution do not differ significantly, suggesting that our assumption holds nearly true during ET from the water table.

**Water table recharge during rainfall**

Water table recharge during rainfall begins once the wetting front reaches the phreatic surface [Woods *et al*., 1997] resulting in its rise. Depending on the depth to WT and the soil hydraulic properties, it can be assumed that a fraction of total rainfall during a time-step (e.g., daily, hourly) contributes to recharge [Park and Parker, 2008] and subsequent water table rise.

Under shallow WT conditions, almost all of the infiltrated water contributes to recharge instantaneously [Novakowski and Gillham, 1988; Hilberts *et al*., 2007] because the capillary fringe is usually close to the surface [Laio *et al*., 2009]. As the WT depth increases, more infiltrated water is held in the unsaturated zone, thus reducing the fraction of recharge during the same interval of time. A simplified approximation method would be to estimate recharge as a function of precipitation and WT depth without having to explicitly model the unsaturated flow processes. For example, Woods *et al*.[1997] estimated spatially variable recharge as a function of effective precipitation and WT depth in a hillslope. However, their recharge function was solely dependent on the WT elevation regardless of the soil type. Here, we assume that the recharge is spatially uniform throughout the field site and depends on the soil type, in addition to the position of the phreatic surface. In this study, we propose that the recharge fraction of rainfall at a given time step can be estimated by an exponential function expressed as

\[
R_e = \alpha P = \begin{cases} 
P \cdot \exp(-\beta_r (d_r - h)) & h \leq d_r \\
0 & h > d_r 
\end{cases}
\]

(2-19)

(2-20)

where \(R_e\) is the recharge rate, \(P\) is the rainfall amount received during the given time step, and \(\beta_r\) and \(d_r\) are the parameters that determine what fraction of rainfall is converted to recharge
depending on the soil type. Note that the unsaturated flow processes are not explicitly considered when equation 2-19 is used to estimate WT recharge. Therefore, it cannot be effectively applied at very short time steps [Woods, et al., 1997]. However, if the WT is close to the surface, hourly time step may be adequate since instantaneous recharge can be assumed [e.g., Laio et al., 2009].

Now, if we treat \( R_e \) (-ve) as the steady flux in successive times steps, \( \psi_T \) is given as

\[
\psi_T = \frac{1}{a_G} \ln \left( \frac{(K_s+R_e) \exp(a_G(h-H)) - R_e}{K_s} \right) \tag{2-21}
\]

Note that because we make the simplifying assumption of constant flux, \( R_e \), throughout the unsaturated zone, \( \psi_T \) given by equation 2-21 may not correspond to values measured during rainfall infiltration. Under field conditions, \( \psi_T \) is likely to reach zero (or positive) before the water table experiences any recharge. Substituting a positive value of \( \psi_T \) in equations 2-8 and 2-9 would then produce physically impossible negative \( \lambda_d \) or \( \lambda_f \). The \( \psi_T \) values obtained from the HYDRUS 1D model and equations 2-21 are compared in Figure 2-4 for a constant rainfall rate of 0.03 cm/hr at two initial WT depths. The two solutions produced similar \( \psi_T \) values for the shallow water tables when the rainfall rate was small.

\( \lambda_d \) and \( \lambda_f \) during ET or Recharge from the Water Table

Differentiating equation (2-16) and (2-21) with respect \( h \) yields

\[
\frac{d\psi_T}{dh} = \frac{1}{a_G} \frac{d}{dh} \left[ \ln \left( \frac{(K_s+\mu)\exp(a_G(h-H)) - \mu}{K_s} \right) \right] = \frac{(K_s+\mu)\exp(a_G(h-H))}{(K_s+\mu)\exp(a_G(h-H)) - \mu} \tag{2-22}
\]

where \( \mu \) represents either ET (+ ve) or recharge, \( R_e \) (-ve). Combining equation 2-22 with equations 2-8 and 2-9 gives the final expressions for \( \lambda_d \) and \( \lambda_f \) under successive steady-state fluxes from the water table

\[
\lambda_d(h) = (\theta_s - \theta_r) \left[ 1 - \left( \frac{(K_s+\mu)\exp(a_G(h-H))}{(K_s+\mu)\exp(a_G(h-H)) - \mu} \right) \left( 1 + \left( \frac{a}{a_G} \ln \left( \frac{(K_s+\mu)\exp(a_G(h-H)) - \mu}{K_s} \right) \right)^n \right)^{-\frac{n+1}{n}} \right] \tag{2-23}
\]
\(\lambda_f(h) = (\theta_s - \theta_r) \left( \left( \frac{(K_s + \mu \exp(a_G(h-H))}{K_s + \mu \exp(a_G(h-H)) - \mu} \right) \left[ 1 + \left( \frac{d}{a_G} \right) \ln \left( \frac{(K_s + \mu \exp(a_G(h-H)) - \mu}{K_s} \right) \right]^{n} \right) \) \quad (2-24)

It can be seen from equations 2-23 and 2-24 that \(\lambda_d \neq \lambda_f\). The disparity between the two parameters is determined by the soil hydraulic parameters, and the unsaturated zone flux. The direction of the flux determines which storage parameter attains the larger value at a given water table depth while the hydraulic parameters determine how different the two parameters will be.

Note that, under hydraulic equilibrium where \(\mu = 0; \frac{d(\psi_T)}{dh} = 1\); and \(\psi_T = h - H\) which reduces both equations 2-23 and 2-24 to 2-11, which is the expression developed by Hilberts et al. [2005] and Bierkens [1998]. In general, fillable porosity will be greater than drainable porosity when ET is occurring from the WT while the reverse is true under conditions of WT recharge. The behavior of \(\lambda_d\) and \(\lambda_f\) for different soil textures under ET and \(R_e\) is discussed further in the results section.

**Implementation of \(\lambda_d\) and \(\lambda_f\) in a 1D Groundwater Model**

Movement of water through soils and other porous media are most accurately represented by the Richards equation. However, due to high nonlinearity and associated difficulties of solving this equation, simpler approaches are usually sought [Pikul et al., 1974]. The overall water balance per unit area of an aquifer can be expressed as

\[
\frac{ds}{dt} = \lambda \frac{dh}{dt} = q + ET - R_e
\]  
(2-25)

Note that in equation 2-25 only a single porosity parameter \(\lambda\) is implemented, which is the conventional approach. In equation 2-25, substitution for the groundwater flux, assuming horizontal flow [Bear and Cheng, 2008] and neglecting vertical flow, yields the one dimensional Boussinesq equation.
\[
\frac{dh}{dt} = \frac{1}{\lambda} \left[ K_s \left( \frac{d}{dx} \left( h \frac{dh}{dx} \right) \right) + ET - R_e \right] \tag{2-26}
\]

where the first term on the right gives lateral the groundwater flux. However, using only a single \( \lambda \) in equation 2-26 to express both ET and \( R_e \) in terms of WT elevation is theoretically inconsistent. Therefore, implementation of separate \( \lambda_d \) and \( \lambda_f \) developed in this study requires reformulation of equation 2-26, depending on the direction of groundwater flow:

\[
\frac{dh}{dt} = \frac{1}{\lambda_f} \left[ K_s \left( \frac{d}{dx} \left( h \frac{dh}{dx} \right) \right) \right] - \frac{ET}{\lambda_d} + \frac{R_e}{\lambda_f} \tag{2-27}
\]
during groundwater inflow towards the aquifer from the lateral boundaries, and

\[
\frac{dh}{dt} = \frac{1}{\lambda_d} \left[ K_s \left( \frac{d}{dx} \left( h \frac{dh}{dx} \right) \right) \right] - R_e \left( \frac{1}{\lambda_d} - \frac{1}{\lambda_f} \right) - \frac{ET}{\lambda_d} \tag{2-28}
\]
during groundwater outflow from the aquifer towards the boundaries. Note that only one or both of the storage parameters may appear in the governing equation depending on which unsaturated fluxes are active. For example, when groundwater is flowing into the system and ET = 0, then the governing relation (equation 2-27) will include only fillable porosity. But for ET > 0, both \( \lambda_d \) and \( \lambda_f \) appear in the governing equation because groundwater inflow raises the WT (controlled by \( \lambda_f \)) while ET tends to lower the WT (controlled by \( \lambda_d \)). Hence, the response of the phreatic surface will depend on the net difference between the magnitude of groundwater inflow and the ET loss. A common example of this condition is the diurnal fluctuations normally observed in shallow WT environments [e.g., Loheide et al., 2005; Gribovszki et al., 2007; Loheide, 2008] where the WT normally drops during the day time (ET > groundwater inflow) and recovers during the night time (ET \( \approx \) 0).

In equation 2-28, note that the recharge component is linked with both \( \lambda_d \) and \( \lambda_f \). An example scenario governed by equation 28 occurs in artificially drained fields where the water table is usually lowered before anticipated rainfall events in order to avoid the inundation of the
field. While the groundwater flow direction at the lateral boundaries is outward, the response of the water table (rise or drop) will depend on the magnitude of the recharge at the phreatic surface. When $R_e > 0$, the magnitude of $\lambda_d$ increases because at each time step there will be some accretion of moisture [Chapman and Dressler, 1984] which needs to be removed from the aquifer to bring the same decline in the water table as compared to when $R_e = 0$. Therefore, at each time step with $R_e > 0$, the total amount of moisture added by recharge to the system will be partitioned into two fractions: (a) the fraction which contributes to WT rise and is controlled by $\lambda_f$, and (b) the fraction which is drained from the profile as the WT drops due to groundwater outflow, and is controlled by $\lambda_d$. Therefore, the net effect of this $R_e$ is determined by the difference between the reciprocals of $\lambda_d$ and $\lambda_f$.

Methods

Study Site

The study field chosen to test the expressions for $\lambda_d$ and $\lambda_f$ was located in northeast Florida (29.694’N, 81.446’W, Figure 2-5). This area is a low (<10 masl), flat landscape with a naturally shallow water table (< 2m from surface). A locally important crop is potato grown under a WT management system known as seepage irrigation [Smajstrla, 2000; Munoz et al., 2008; Acharya and Mylavarapu, 2011] wherein the WT is maintained close to the crop root-zone by applying water from shallow furrows spaced uniformly at 18-20m between main lateral ditches (1.5-2.0m deep) at the edges of fields on the order of 3-15 ha. The flow system in these agricultural fields therefore can be appropriately represented by the Boussinesq equation (i.e., equations 2-27 and 2-28). The soil in the field site is highly sandy classified as Ellzey fine sand series (sandy, siliceous, hyperthermic, Arenic Endoaqualf) [NRCS, 1999] with some clay content at depths 75-120cm [Acharya and Mylavarapu, 2011].
Water Table and Weather Data

Three shallow (1.2m deep) wells were installed in a 15-ha field (Figure 2-5) prior to the Spring 2010 potato season (February to June). Hourly WT data were collected from each well using data-logging pressure transducers. In the 2011 potato season, hourly WT data were collected from 3 new wells installed in the same field at approximately the same locations. In both years, hourly PET estimates were calculated using the Penman-Monteith [Monteith, 1965] equation based on the weather data collected from a Florida Automated Weather Network [FAWN, 2011] weather station located adjacent to the field site. Hourly PET estimates were converted to the actual potato ET estimates using available potato crop coefficient values [Allen et al., 1998].

Parameter Estimation

Moisture retention parameters of Ellzey fine sand were estimated by fitting soil moisture retention data to the modified van Genuchten function (equation 2-4) which showed a very good fit for both the original and modified equations (Figure 2-6). Acharya and Mylavarapu [2011] reported a lab-determined average $K_s = 10.0$ cm/hr for Ellzey fine sand, while Rosa [2000] reported $K_s = 7.0$ cm/hr below the WT of the same soil using a, slug-test method. In this study we used the latter value since in-situ observations are more likely to represent field conditions. Table 2-1 shows the values of the hydraulic parameters for Ellzey series.

The most rigorous method for estimating Gardner’s parameter $a_G$ would be to fit the model to experimental $\psi$-$K$ data. However, due to the constraints in collecting such experimental data, it is often necessary to revert to other methods. One such method is estimating $a_G$ by establishing correspondence with the parameters of the soil moisture retention model [e.g., Zhu et al., 2004; Rucker et al., 2005; Ghezzehei et al., 2007]. Ghezzehei et al. [2007] developed a simple formula to calculate $a_G$ from the original van Genuchten model parameters $\alpha$ and $n$ by

\[ a_G = \frac{n}{\alpha} \]
defining two $\psi$ values at which $\alpha$ and $a_G$ match. They reported that, for soils with $n > 2$, $a_G$ can be estimated reliably by

$$a_G \approx 1.3 \times n \times a$$  \hspace{1cm} (2-29)

Equation 2-29 was therefore used to estimate $a_G$ for Ellzey sand since the value of $n$ for Ellzey series was 2.62 (Table 2-1).

Parameters $\beta$ and $d'$ of the $ET\text{-}WT$ relationship in 2-17 and 2-18 were adapted from Shah et al. [2007] who determined the parameter values for the major soil textural classes under bare soil, grass and forest type vegetation. The soil in our field site was dominated by fine sand with clay content up to 11% at 70-120cm depth [Acharya and Mylavarapu, 2011]. Based on these results, the soil was considered close to the loamy sand texture. The parameter values were then determined by taking the average of the values for bare soil and grass since the root system of potato is normally shallow. The estimated values of $\beta$ and $d'$ for Ellzey sand were 0.08 and 45cm respectively. In order to estimate the parameters of the recharge function given by equation 2-19, several hypothetical simulations of rainfall infiltration with different initial WT depths and rainfall rates were carried out using HYDRUS 1D. The magnitudes of the soil moisture flux at the saturated-unsaturated zone interface (i.e. the phreatic surface) were recorded at each time step which were then fitted with the recharge function in 2-20, which provided the estimates of $\beta_r = 0.15$ and $d_r = 60$ cm for Ellzey fine sand.

**Boundary Conditions and Numerical Simulation**

The governing equations 2-27 and 2-28 are nonlinear partial differential equations whose analytical solutions are not always readily available. For our study, the equations were solved numerically by implicit finite-difference method for one-dimensional lateral flow. In the field, ditches and water furrows are used to control the WT during irrigation and drainage. The furrows
thus act as internal boundaries in additions to the ditches during irrigation and early stage of drainage. Therefore, in the numerical solution, the locations of the water furrows were treated as internal boundaries in addition to the external boundary conditions imposed at the main ditches. The numerical solution was obtained using the tri-diagonal matrix algorithm in the statistical programming language R [R Development Core Team, 2011]. The simulated WT dynamics were then compared with the field data collected from the study site during the spring of 2010 and 2011. The Nash and Sutcliffe [1970] efficiency coefficient, \( C_{\text{eff}} \), and root mean square error (RMSE) were used as goodness-of-fit measures between the model implemented only with \( \lambda_d \) and the model with both \( \lambda_d \) and \( \lambda_f \).

**Results**

**Drainable Porosity and Fillable Porosity under ET and Recharge**

Figure 2-7 shows \( \lambda_d \) and \( \lambda_f \) as functions of WT for clay, loam, and sand with typical grass type vegetation at different atmospheric ET demands. Increasing ET demand reduces \( \lambda_d \) to values smaller than hydrostatic (\( HS-\lambda \)), while at the same WT depths and ET demands, \( \lambda_f \) values are larger than \( HS-\lambda \). This deviation from the \( HS-\lambda \) is different for different soils and is maximum for WT depths near the ET ‘transition depth’ (see Figure 2-3). For example, in loam soil at ET = 0.02 cm/hr, when the WT is near 80cm from the surface, \( \lambda_d \) value is 22% less than the \( HS-\lambda \) while at the same WT depth \( \lambda_f \) is 51% more than \( HS-\lambda \). For clay soil, this difference is not as extreme as in loam while for sandy soils: the difference seems negligible even at ET rate of 0.02cm/hr. As the WT depth increases beyond the ET transition depth, both \( \lambda_d \) and \( \lambda_f \) converge towards \( HS-\lambda \) since the pressure distribution in the soil profile moves towards hydrostatic. When the WT is very close to the surface, both \( \lambda_d \) and \( \lambda_f \) are close to zero irrespective of the ET demand. For sandy soils, which normally have low capillary retention, \( \lambda_d \) and \( \lambda_f \) hardly differ from \( HS-\lambda \) even when the ET demand is high. On the other hand, clay and loam soils show
significant differences from HS-λ even under small ET demand because of their high moisture retention capacity.

The soil in our study site also did not show as large a shift in λ_d and λ_f from HS-λ as would clay or loam soils under ET (Figure 2-8). This can be attributed to the sandy nature of the profile and shallow root system of potato plants. Nonetheless, there is still a discernible downward shift in λ_d as ET demand increases. Figure 2-8 also suggests that the difference between λ_d and λ_f reaches as much as 18% of the effective porosity (θ_s - θ_r) at 45cm WT depth for ET = 0.03cm/hr. Hourly ET rates in the study area can reach as high as 0.07 cm/hr during mid-day [FAWN, 2011] which would result in significantly lower λ_d or higher λ_f values. The shift in the values of λ_d and λ_f under shallow water table conditions was more pronounced for recharge (Figure 2-9). Under recharge, for the same WT depth, λ_f becomes smaller than HS-λ as R_e increases while the opposite is true for λ_d. Since R_e tends to saturate the soil profile, extra water needs to be drained in order to achieve the same drop in WT that would have occurred under hydrostatic conditions, resulting in larger λ_d. On the other hand, λ_f decreases sharply since recharge reduces the storage deficit [Chapman and Dressler, 1984; Chapman, 1995] consequently reducing the amount of water required to achieve a unit rise in the water table. It should be noted that using different ET or R_e functions may result in slightly different relations between λ_d and λ_f WT for the same soil and vegetation types.

Water Table Dynamics

The observed and simulated hourly WT dynamics in Well 1 are shown in Figure 2-10 for 50-day periods in spring 2010 and 2011. Similar patterns were also observed in the other two wells (data not shown). Model results are compared for the single-parameter model (SPM) with HS-λ (equation 2-26) and the dual-parameter models (DPM) with separate steady-state λ_d and λ_f expressions developed in this study (equations 2-27 and 2-28). Note that the ET and R_e values
required for the estimation of $\lambda_d$ and $\lambda_f$ from equations 2-23 and 2-24 were determined \textit{a priori},
which enabled analytical calculation of these storage parameters as functions of the predicted
WT elevations during each simulation. In both years, the WT data clearly show the timing of
irrigation and drainage in the field. During irrigation periods, the WT was consistently
maintained around 40-60cm depth from the surface by raising water levels in the lateral ditches
and irrigation furrows accordingly. Diurnal fluctuations in the WT are also clearly discernible
during the days with irrigation. This diurnal fluctuation, despite the continuous subsurface lateral
inflow of water, indicates that the WT supports most of the ET demand of potato crop under this
water management system. The observed WT dynamics were matched better by the DPM
simulations than the SPM. Note that during the early part of the 2011 simulations, the observed
diurnal WT fluctuations were larger than predicted for both SPM and DPM. This indicates that
ET for the potato crop during this period was under-estimated, likely because the crop-
coefficient for the mid-growth stage of potato in this area may be higher than the general FAO
[1998] value. The average DPM $C_{eff}$ values in both 2010 and 2011 (0.56, 0.55) were higher
compared to those for SPM (0.34, -0.18), indicating improved performance for the DPM, with an
especially marked difference in 2011 (Table 2-2). The average RMSE (cm) values in both years
were correspondingly smaller for the DPM (0.34, 0.85) than the SPM (3.14, 0.86). Note that the
$C_{eff}$ values for Well 3 were significantly smaller than the values for Well 1 and Well 2. This was
likely due to the proximity of this well to the ditch (12 m) where the assumption of purely
horizontal flow may not have been supported due to flow convergence.

The SPM-simulated WT fluctuated at higher elevations than the observed WT with
narrower amplitudes both in 2010 and 2011. The observed diurnal amplitudes of WT depth were
well matched by the DPM. This discrepancy between DPM and SPM was more pronounced
towards the end of the simulation period both in 2010 and 2011. This period coincided with later stages of the potato crop when the water level in the ditches and furrows were brought to the shallowest elevation (approximately 30-35cm below ground surface). Rising WT during rainfall events was also predicted by DPM better than SPM, which underestimated WT rise during smaller events. These results highlight the contribution of the recharge-dependent $\lambda_f$ parameter in improving the simulation results.

**Discussions**

When flux-dependent $\lambda_d$ and $\lambda_f$ are considered in groundwater studies, it is also important to understand the relationship between these parameters and the magnitude of flux. Of particular interest is the relationship between $R_e$ and $\lambda_f$ because it determines the rate of WT rise during rainfall events. The relationship between $\lambda_f$ (equation 2-24) and $R_e$ (scaled to $K_s$) under Ellzey fine sand for four different WT depths is shown in Figure 2-11. These results suggest that when the WT is deep, even a small $R_e$ corresponds to a $\lambda_f$ value very close to zero. That is, by the time the wetting front has reached a deep WT, most of the soil profile is already saturated. Whereas under the same $R_e$ in a shallow WT, some portion of the soil pores may still be empty (i.e. $\lambda_f > 0$). However, for larger recharge rates $\lambda_f$ approaches zero irrespective of the WT depth. *Chapman and Dressler* [1984] showed a similar relationship between $R_e$ and an effective storage coefficient.

An important advantage a method to estimate $\lambda_f$ in the presence of vertical fluxes is that it enables simple estimation of WT rise during rainfall events. In shallow phreatic aquifers, the WT normally shows a quick response to the rainfall events. Using $\lambda_f$ and neglecting the contribution of groundwater flow and ET in equation 2-27, the water table position at time $t$ during rainfall be estimated simply as
\[
\frac{dh}{dt} = \frac{1}{\lambda_f} R_e
\]  

(2-30)

and

\[
h_t = h_{t-\Delta t} + \Delta t \frac{R_e(h_{t-\Delta t}, P_{(t-\Delta t), t})}{\lambda_f(h_{t-\Delta t}, R_e)}
\]  

(2-31)

where \(\Delta t\) is the time step increment, and \(P_{(t-\Delta t), t}\) is the rainfall during the period of each time step. Note that \(R_e\) is estimated from \(P\) using equations 2-19 and 2-20, which is then used to estimate \(\lambda_f\) using equation 2-24. The WT position is then estimated by simple Euler integration of equation 2-30 as shown in equation 2-31. Application of equation 2-30 to the field site using \(\lambda_f\) is compared in Figure 2-12 to the estimates based on \(HS-\lambda\). The WT rise is significantly underestimated by \(HS-\lambda\) as compared to our \(\lambda_f\) expression, especially when the rainfall rates are smaller. During high-intensity rainfall events, however, the error of estimation is relatively large for both expressions. Comparison of equation 2-31 with HYDRUS 1D for an initial WT depth of 120cm (Figure 2-13) shows good agreement for all rainfall rates (0.05cm/hour – 2.0cm/hr), although equation 2-26 estimates faster WT rise during the initial phase. This is an anticipated pattern because in deeper water tables \(R_e\) will not occur instantaneously, since the unsaturated zone captures most of the infiltrated moisture [Hilberts et al., 2007]. Application of \(\lambda_f\) (equation 2-24) in deeper initial WT conditions therefore may require additional modification of the recharge function to account for the lag between rainfall and recharge.

**Conclusions**

Drainable and fillable porosity parameters are critical in modeling groundwater flow in unconfined aquifers. They not only determine the water table dynamics but also influence the subsurface discharge at the aquifer outlet. In this paper we have presented a method to incorporate effects of unsaturated zone moisture fluxes while estimating these two parameters. By assuming successive steady state ET and \(R_e\) fluxes, closed-form expressions for \(\lambda_d\) and \(\lambda_f\)
were developed. The main advantage of these expressions is that the dynamic behavior of soil water is accounted for in the estimation of storage parameters. These expressions therefore relax the limitations of previously available methods that relied on a hydrostatic assumption to incorporate the effect of soil moisture retention. The strongest merit of these expressions, however, is the ability to estimate and \( \lambda_d \) and \( \lambda_f \) separately for WT drawdown and rise scenarios. This study shows that depending on the magnitude of vertical flux and soil hydraulic properties, the difference between \( \lambda_f \) and \( \lambda_d \) can be substantial. With the help of a hydraulic groundwater model implemented with flux-dependent \( \lambda_d \) and \( \lambda_f \), it is shown that the simulation of WT dynamics can be strongly improved.

The values of drainable and fillable porosity are different under hydrodynamic conditions, but they collapse into a single equation under a static moisture profile. The analytical expressions revealed that under the influence of ET, dynamic-state \( \lambda_d \) and \( \lambda_f \) of sandy soils hardly differ from \( HS-\lambda \) whereas they can be significantly different in clay and loam. In general, the results suggest at a given WT depth, the finer the soil, the more pronounced the difference between \( HS-\lambda \) and \( \lambda_d \) or \( \lambda_f \) will be for the same ET rate. Recharge also had strong effects on the magnitudes of both \( \lambda_d \) and \( \lambda_f \). It was found that the same recharge rates correspond to a much smaller \( \lambda_f \) values (very close to zero) under deeper WT than under shallower WT. This study highlights the importance of using fillable porosity for modeling water table dynamics. The inequality between \( \lambda_d \) and \( \lambda_f \) that develops during recharge or ET in shallow WT environments is significant and should not be neglected in most soil types. Fillable porosity is normally regarded to be smaller than \( \lambda_d \) due to hysteresis in soil moisture retention [Bouwer and Jackson, 1974; Bouwer, 1978; Sophocleous, 1991]. We found differences between \( \lambda_d \) and \( \lambda_f \) even when
hysteresis is not considered. Extending these analyses to incorporate hysteretic effects precludes analytical solutions and requires numerical methods [e.g., Stauffer et al., 1992].

Note that we examined shallow water table dynamics, but the expressions developed in this study may be effectively used in modeling shallow subsurface flow in hillslopes [e.g., Troch et al., 2003; Hilberts et al., 2004, 2005, 2007], land drainage [e.g. Skaggs 1980], and estimating ET from diurnal water table fluctuations [Loheide et al., 2005; Nachabe et al., 2005; Gribovszki et al., 2007].
Table 2-1. Original and Modified van Genuchten (VG) soil moisture retention model parameters and Gardner’s alpha ($a_G$) for Clay, Loam and Sand, and Ellzey fine sand series. Original VG parameters or Sand, Clay and Loam are from Carsel and Parris [1988]; $a_G$ parameters for Clay, Loam and Sand are from Reynolds and Elrick [1987, 1985].

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Clay Original</th>
<th>Clay Modified</th>
<th>Loam Original</th>
<th>Loam Modified</th>
<th>Sand Original</th>
<th>Sand Modified</th>
<th>Ellzey Original</th>
<th>Ellzey Modified</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\theta_s$</td>
<td>0.380</td>
<td>0.398</td>
<td>0.430</td>
<td>0.542</td>
<td>0.430</td>
<td>0.447</td>
<td>0.395</td>
<td>0.398</td>
</tr>
<tr>
<td>$\theta_r$</td>
<td>0.060</td>
<td>0.239</td>
<td>0.078</td>
<td>0.087</td>
<td>0.045</td>
<td>0.045</td>
<td>0.066</td>
<td>0.075</td>
</tr>
<tr>
<td>$N$</td>
<td>1.09</td>
<td>0.43</td>
<td>1.56</td>
<td>0.47</td>
<td>2.69</td>
<td>1.67</td>
<td>2.63</td>
<td>2.07</td>
</tr>
<tr>
<td>$K_s$ (cm/hr)</td>
<td>0.2</td>
<td>1.04</td>
<td>29.7</td>
<td>7.0</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$a_G$</td>
<td>0.001</td>
<td>-</td>
<td>0.0335</td>
<td>-</td>
<td>0.25</td>
<td>-</td>
<td>0.068</td>
<td>-</td>
</tr>
</tbody>
</table>

Table 2-2. Nash-Sutcliffe Coefficient of Efficiency ($C_{eff}$) and Root Mean Square Error ($RMSE$) of the model implemented with hydrostatic (common $\lambda_d$ and $\lambda_f$) and dynamic (different $\lambda_d$ and $\lambda_f$) expressions

<table>
<thead>
<tr>
<th>Year</th>
<th>Hydrostatic- $\lambda_d$</th>
<th>Dynamic- $\lambda_d/\lambda_f$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$C_{eff}$</td>
<td>RMSE</td>
</tr>
<tr>
<td>2010</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Well 1</td>
<td>0.28</td>
<td>3.30</td>
</tr>
<tr>
<td>Well 2</td>
<td>0.37</td>
<td>3.14</td>
</tr>
<tr>
<td>Well 3</td>
<td>-0.02</td>
<td>2.99</td>
</tr>
<tr>
<td>2011</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Well 1</td>
<td>0.21</td>
<td>0.59</td>
</tr>
<tr>
<td>Well 2</td>
<td>-0.10</td>
<td>1.51</td>
</tr>
<tr>
<td>Well 3</td>
<td>-0.65</td>
<td>2.50</td>
</tr>
</tbody>
</table>
Figure 2-1. Conceptual representation of the water table profile and water balance components in a typical unconfined aquifer with shallow phreatic surface, and bounded by a ditch or stream.

Figure 2-2. Conceptual representations of change storage-deficit ($D_s$) due to a unit rise in WT elevation (left), and changes in unsaturated storage ($U_s$) and saturated storage ($S_s$) due to a unit decline in WT(right) in an unconfined aquifer system.
Figure 2-3. Relationship between ET rate and WT depth for clay, loam and sand soils under grass type vegetation, based on equations 2-17 and 2-18. The parameter values for the soils were clay: $d' = 95\text{cm}$, $\beta = 0.011$; loam: $d' = 85\text{cm}$, $\beta = 0.028$; and sand: $d' = 30\text{cm}$, $\beta = 0.043$ [Shah et al., 2007].

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CHAPTER 3
ESTIMATION OF EVAPOTRANSPERSION FROM DIURNAL WATER TABLE FLUCTUATIONS USING DRAINABLE AND FILLABLE POROSITY

Background

In shallow groundwater environments such as riparian zone and wetlands, evapotranspiration is a significant component of the water balance [Dolan et al., 1984; Loheide et al., 2005; Luo and Sophocleous, 2010] where the shallow WT normally supports most of the plant water requirement [Shah et al., 2007; Gribovszki et al., 2007; Lowry and Loheide, 2010]. Consequently a distinct diurnal fluctuation pattern in the water table (WT) hydrograph is usually observed [Schilling, 2007; Butler et al., 2007; Nachabe et al., 2005; Loheide, 2008]. This diurnal water table fluctuation (DWTF) corresponds to the 24-hour variation cycles in the climatic variables, solar radiation, temperature and humidity [Gribovszki et al., 2007; Gribovszki et al., 2010]. A typical DWTF pattern observed in a crop field in northeast Florida is illustrated in Figure 3-1. Note that the observed WT elevation decreases more rapidly during daytime and slowly resumes during the nighttime.

Several different methods are available for estimating the ET from vegetated surfaces which normally involve the measurements of various weather parameters [e.g. Monteith, 1965; Priestley and Taylor, 1972, etc]. The measurements are used to solve different relationships describing the energy balance and aerodynamic vapor transfer phenomena to estimate the evaporative loss. The Penman-Monteith method [Monteith, 1965] is based on the combination of both energy balance and aerodynamic method [Chow et al., 1964]. One of the main drawbacks of the weather based ET estimation methods is that they usually involve complex calculation methods and require large number of parameters.
In unconfined aquifers shallow phreatic surface, there is normally a direct link between the vegetation water uptake and the water table (WT). Therefore, ET loss from these shallow WT environments may also be estimated from the soil water balance which can be expressed as

\[ \lambda \frac{dh}{dt} = q + R_e - ET \]  

(3-1)

where, \( h \) is the water table elevation, \( \lambda \) is the storage parameter of the aquifer, \( q \) is the groundwater flux, \( R_e \) is the rainfall recharge and \( ET \) is the evapotranspiration loss. The term in the left hand side of equation 3-1 gives the total change in soil moisture storage per unit area expressed in term of WT elevation. equation 1 can be reformulated to estimate the \( ET_g \) as

\[ ET_g = q + R - \lambda \frac{dh}{dt} \]  

(3-2)

From equation 3-1 it is therefore possible to obtain the \( ET_g \) if major portion of the ET is directly contributed by the water table. \textit{White} [1932] was probably the first to analyze and report such direct link between ET and WT fluctuations in shallow WT areas. Based on the observations DWTF in a shallow riparian zone, he used the soil water balance to develop a method to estimate the groundwater water consumption by the phreatophytic vegetation. His method however was based on different simplifying assumptions. He assumed that rate of groundwater flow to the area was constant over the 24 hour period and could be estimated by the rate of water table rise between midnight and 4:00 am when \( ET_g \) is negligible or zero. Based on these assumptions and ignoring the precipitation component, \textit{White’s} equation is expressed as

\[ ET_g = \lambda_d (24r + \Delta S) \]  

(3-3)

where, \( \lambda_d \) is the drainable porosity (or specific yield) of the aquifer, \( r \) is the rate of WT rise between midnight and 4:00 AM when the \( ET_g \) is considered zero, and \( \Delta S \) is the rise or fall of WT during the 24-hour period. In addition to these assumptions, the drainable porosity parameter, \( \lambda_d \) was also assumed as constant in equation 3-3.
Following White, many have investigated the DWTF phenomenon to estimate the shallow groundwater consumption by vegetation [e.g., Dolan et al., 1984; Gerla, 1992; Loheide et al., 2005; Nachabe et al., 2005, Lautz, 2008, Butler et al., 2007, Gribovszki et al., 2007; Loheide, 2008; Mould et al., 2010, etc.]. Some of these studies also evaluated the underlying assumptions in equation 3-3 and found that the assumptions of constant groundwater flow as well as constant $\lambda_d$ are normally not met in field conditions [Gerla, 1992; Loheide et al., 2005; Gribovszki et al., 2007]. Gerla [1992] reported that assuming a constant $\lambda_d$ in equation resulted in excessively large $ET_g$ estimation in a wetland environment. According to Loheide et al. [2005], the assumption of constant $\lambda_d$ creates problems in the White method and showed with the help of soil moisture flow simulations that $\lambda_d$ was critically important in accurately estimating $ET_g$ from DWTF. Loheide [2008] proposed a modified White method to estimate $ET_g$ at subdaily time scales considering the variable groundwater flow rates. Gribovszki et al. [2007] presented a similar modification to Loheide [2008] by using variable, successive steady-state groundwater flow and a WT dependent $\lambda_d$ expression developed by Nachabe [2005].

Although the implementation of variable groundwater flow and variable $\lambda_d$ have generally shown to resolve the inherent limitations of the White method and allow for subdaily (e.g. hourly) scale $ET_g$ estimations [Gribovszki et al., 2007; Loheide, 2008], there are still some difficulties associated with this method. Firstly, they are based on a $\lambda_d$ parameter estimated either from the hydrostatic soil moisture profile [Gribovszki et al., 2007] or from the analysis of rainfall induced WT rise [e.g. Dolan et al., 2007; Schiling, 2007] while this parameter dynamically depends on both WT elevation and soil moisture flux above the phreatic surface [Childs, 1960; Tritscher et al., 2000]. Since $\lambda_d$ estimated from hydrostatic moisture profile (hydrostatic-$\lambda_d$) does not account for the unsaturated zone flux directly induced by both ET and rainfall infiltration, its
value may not accurately represent the actual $\lambda_d$ in the field under hydrodynamic conditions [e.g. Tritscher et al., 2000]. This discrepancy therefore can introduce potential errors in the $ETg$ calculations from DWTF.

Besides neglecting the effect of soil moisture flux, hydrostatic-$\lambda_d$ expressions also have another drawback; these methods result in the same parameter value for both rising and falling water table conditions while in reality the rise in WT elevation is dictated by a distinct storage parameter called the fillable porosity ($\lambda_f$). Fillable porosity is an analogous concept to $\lambda_d$ which is defined as the amount of water absorbed by the aquifer per unit rise in WT elevation [Bouwer, 1978; Sophocleous, 1991; Park and Parker, 2008]. This parameter is therefore a measure of the volume of pore spaces that remain to be filled during the rise of the phreatic surface [Chapman and Dressler, 1984; Chapman, 1995]. In Chapter 1 it is shown that when vertical flux from and to the water table are considered, $\lambda_d$ and $\lambda_f$ tend to differ significantly during ET and recharge thus limiting the use of hydrostatic-$\lambda_d$ in either groundwater flow models or in $ETg$ estimation from DWTF.

Since the hydrostatic-$\lambda_d$ ignores the unsaturated zone soil moisture flux, they cannot accurately reflect rapid WT fluctuation shown by shallow phreatic aquifers during ET and precipitation. Especially during and after rainfall events, the assumption of hydrostatic moisture profile is severely disturbed because of rapid redistribution of soil moisture. Therefore, the actual $\lambda_d$ may differ significantly from the estimated hydrostatic-$\lambda_d$ [Tritscher et al., 2000] consequently resulting in potentially erroneous $ETg$ estimations. In order to avoid these complications during rainfall, most of the White based $ETg$ estimation methods presented in the literature typically omitted the days with precipitation in their calculations [e.g. Dolan et al., 1984; Gribovszki et al., 2007; Schilling, 2007]. As a result, these estimations are usually
discontinuous at smaller time scales (i.e. hourly or daily) although fairly well estimates may be obtained on larger time scales (e.g. monthly average) if the rainfall frequency is low.

While the potential impact of rainfall on the underlying assumptions of White based ETg estimation methods results due to the disturbance in the DWTF signal of the WT, the literature provides very little work on the extent to which the ETg estimates are affected when days with rainfall are included in the calculation. Since the method is based on soil water balance, it can be argued that this method should work satisfactorily even during the rainfall events, if the effect of rainfall on WT fluctuations can be quantified reliably. In one study, Gerla [1992] included the days with precipitation to estimate ET using the White method. He modified the original White method by accounting for the effect of infiltration on water table rise thus extending the use of the method even during precipitation events. His study however lacked an appropriate analytical expression to estimate the storage parameters and used \( \lambda_d \) values estimated from the WT rise during infiltration. In addition, Gerla’s method was based on daily time scale where the hourly dynamics of recharge and WT rise are lumped thus concealing the potential impacts of rainfall infiltration on the hourly ETg estimates. Nonetheless, Gerla [1992] showed a significant improvement over other the original White method, since it could be used during the rainfall events to obtain the ETg estimates that correlated well with the potential ET (PET) estimates from weather measurements.

The main purposes of this paper is therefore twofold: (1) To use a \( \lambda_d \) parameter estimated from the soil moisture profiles under successive steady state soil moisture flux (dynamic-\( \lambda_d \)), instead of the hydrostatic-\( \lambda_d \), to estimate ETg from DWTF and compare the results from the two methods, (2) To include the days with rainfall also in the calculations by modifying the White method to account for the effect of precipitation on the WT fluctuation using the \( \lambda_f \) parameter.
also estimated under successive steady soil moisture flux. The dynamic-$\lambda_d$ and $\lambda_f$ partially incorporate the effect of the soil moisture flux on the respective changes in aquifer storage and deficit during WT movement due to ET and/or groundwater flux. They may provide potentially more reliable approximation of the actual storage parameters of the aquifer than given by the hydrostatic-$\lambda_d$ while also enabling the application of appropriate parameters to WT rise and drawdown respectively.

Theoretically, use of these distinct $\lambda_d$ and $\lambda_f$ parameters should enable the estimation of $ET_g$ from soil moisture balance continuously regardless of the rainfall events because it is possible to estimate the contribution of rainfall to change in WT. Therefore, an attempt is made in this study to obtain the continuous estimates of $ET_g$ without avoiding rainy days in order to comparatively assess the impact of rainfall on $ET_g$ estimates. Firstly, the governing equations to estimate $ET_g$ implemented with corresponding $\lambda_d$ or $\lambda_f$ parameter are presented based on whether the WT is being driven downward or upwards due to the groundwater flow. The method of estimating the two storage parameters are also briefly discussed. A simple method of attributing the WT rise to recharge during rainfall based on the WT hydrograph and $\lambda_f$ is also illustrated. This step provides a scheme to extract the exact WT change due to the groundwater flow and $ET_g$ if there was no precipitation during a particular time step. Finally, the equations are applied to estimate $ET_g$ from a potato field under ditch drainage and irrigation system in northeast Florida. Both hydrostatic-$\lambda_d$ parameter method and the new method are used to estimate hourly and daily $ET_g$ over a 50-day period during spring 2010 and 2011 seasons. The results from the two methods are then compared with each other with respect to the standard Penman-Monteith method [Monteith, 1965] as a benchmark, estimated from the measurements of weather parameters in the study site.
Theory

Estimating ETg from DWTF using $\lambda_d$ and $\lambda_f$

The soil water balance given by equation 3-2 provides the total change in storage expressed in terms of WT elevation measured from a reference datum. The $\lambda$ parameter in equation 3-2 can be either $\lambda_d$ or $\lambda_f$ depending on whether the WT is receding or rising as suggested by their respective definitions [Bouwer, 1978; Sophocleous, 1991]. In order to implement these two parameters in the water balance equation it is therefore necessary to understand the respective effect of each water balance component on the water table. For example, ET drives the water table downwards from the soil surface. So, the total change in aquifer storage caused by ET can be estimated by using $\lambda_d$. Recharge ($R_e$) on the other hand, normally causes the phreatic surface to rise, which is dictated by the storage-deficit of the aquifer and hence by $\lambda_f$. The groundwater flow ($q$) can result in either rise or drawdown WT depending on the boundary conditions thus requiring appropriate storage parameter to determine the total change in storage. Therefore, based on whether the water table is rising or falling due to groundwater flow, the water balance equation can be reformulated appropriately by implementing $\lambda_d$ and $\lambda_f$ as

$$\frac{dh}{dt} = \frac{1}{\lambda_f} (q + R_e) - \frac{ET}{\lambda_d}$$

(3-4)

during groundwater inflow with recharge ET and

$$\frac{dh}{dt} = \frac{q}{\lambda_d} + R_e \left( \frac{1}{\lambda_f} - \frac{1}{\lambda_d} \right) - \frac{ET}{\lambda_d}$$

(3-5)

during groundwater outflow with recharge and ET

Note that in equations 3-5, the $R_e$ term is associated with both $\lambda_d$ and $\lambda_f$. This is because in a declining WT scenario, the total amount of recharge at the phreatic surface is partitioned into two fractions. A fraction that contributes to recharge and cause the water table to rise, and
another fraction that is drained with the groundwater outflow and lost from the profile. Therefore the effect on WT is determined by the difference between the reciprocal of \( \lambda_f \) and \( \lambda_d \).

Equations 3-4 and 3-5 can now be reorganized to give the governing expressions implemented with distinct \( \lambda_d \) and \( \lambda_f \) to estimate \( ET_g \) as

\[
ET_g = \left( \frac{\lambda_d}{\lambda_f} \right) q - \lambda_d \left( \frac{dh}{dt} - \frac{R_e}{\lambda_f} \right)
\]

(3-6)

for \( ET_g \) from WT during groundwater inflow (rising WT) and recharge.

\[
ET_g = q - \lambda_d \left( \frac{dh}{dt} - \frac{R_e}{\lambda_f} \right) - R_e
\]

(3-7)

for \( ET_g \) from WT during groundwater outflow (draining WT) and recharge

In equation 3-6 and 3-7, the term \( \frac{R_e}{\lambda_f} \) represents the contribution of recharge towards the rise in WT elevation, so it needs to be accounted for and removed from the observed change in WT elevation \( \left( \frac{dh}{dt} \right) \), if a particular day with recharge is to be included in \( ET_g \) calculations. This implies that the resulting equation is not applicable to days with recharge when the water table is declining. Implementing separate \( \lambda_d \) and \( \lambda_f \) parameters thus allows to include days with rainfall with improved \( ET_g \) estimations and generate continuous subdaily \( ET_g \) calculations. In addition, immediately after the rainfall event which causes the water table rise, ET loss normally occurs at the atmospheric potential rate. This results in \( \lambda_d \) values that are smaller than the hydrostatic-\( \lambda_d \) consequently estimating the WT downward more quickly than the latter. This eventually produces smaller ET estimates from equation 3-7 than from equation 3-2.

Estimation of \( \lambda_d \) and \( \lambda_f \)

Application of equations 3-6 and 3-7 to estimate \( ET_g \) necessitates a method to estimate \( \lambda_d \) and \( \lambda_f \) as the function of WT elevation and unsaturated zone moisture fluxes which was presented in Chapter 2. These \( \lambda_d \) and \( \lambda_f \) expressions partially account for these unsaturated zone water flow
by assuming steady state flow at successive time steps. The generalized expressions for $\lambda_d$ and $\lambda_f$ under hydrodynamic conditions in the soil profile given are given as

$$\lambda_d = (\theta_s - \theta_r) \left[ 1 - \left( \frac{d(\psi_T)}{dh} \right) \left( 1 + (\alpha \psi_T)^n \right)^{-\frac{n+1}{n}} \right]$$  \hspace{1cm} (3-8)

$$\lambda_f = (\theta_s - \theta_r) \left( \frac{d(\psi_T)}{dh} \right) \left[ 1 - \left( 1 + (\alpha \psi_T)^n \right)^{-\frac{n+1}{n}} \right]$$  \hspace{1cm} (3-9)

where, $\theta_s =$ saturated water content; $\theta_r =$ residual water content; $\psi_T =$ matric suction at the soil surface; and $\alpha$ and $n$ are the parameters of a modified van Genuchten [1980] soil moisture retention model [Troch, 1992; Bierkens, 1998; Hilberts et al., 2005]. The value of $\psi_T$ can be estimated by using Gardner’s hydraulic conductivity function [Gardner, 1958] assuming steady state flows. Note that, both equation 3-8 and 3-9 collapse into the same expression when hydrostatic conditions are assumed in the soil profile (i.e., $\frac{d(\psi_T)}{dh} = 1$). The magnitudes of $\lambda_d$ and $\lambda_f$ given by equation 3-8 and 3-9 therefore depend on the soil hydraulic parameters, the magnitudes of ET or $R_e$ and the vegetation types. Figure 3-2 shows the drainable and fillable porosity of Ellzey sand series soil from northeast Florida under potato during different rates of ET and recharge respectively. The plots suggest that the difference between $\lambda_d$ and $\lambda_f$ increases as the magnitude of the fluxes increase. Note $\lambda_d$ and $\lambda_f$ for ET are plotted under relatively large hourly ET values because it has been found that the ET loss in the area may reach as high as 0.07 cm/hr during the midday [FAWN, 2011]. Although the differences in $\lambda_d$ and $\lambda_f$ for the Ellzey soil during ET are not as high as they would be in finer textured soils, it is still substantial to be replaced by a single, hydrostatic-$\lambda_d$ parameter.

**Hourly Water Table Slopes ($\frac{dh}{dt}$)**

Normally during rainless days, hourly changes in the WT elevation ($\frac{dh}{dt}$) at each time step is simply estimated by subtracting the observed WT elevation at a previous time step from the
observation at the current time step. However, whenever there is rainfall, $\frac{dh}{dt}$ becomes large due to WT recharge, which eventually results in highly negative ETg estimations. It is due to this complication that most of the previous studies on ETg have typically avoided the days with substantial precipitation [e.g. Dolan et al., 1984; Gribovszki et al., 2007; Schilling, 2007]. However, if estimates of $R_e$ and $\lambda_f$ are available, the contribution of recharge on the observed time series of $\left(\frac{dh}{dt}\right)$ can be approximately quantified and removed from the WT time series [Healy and Cook, 2002]. Removal of the recharge induced WT rise at each time step $\left(\frac{dh_R}{dt}\right)$ then yields the estimate of the actual change in WT $\left(\frac{dh}{dt}\right)$ that would have been observed if there was no rainfall during that period. This approach is fundamentally similar to the methods that estimate recharge from WT time series [e.g., Crosbie et al., 2005] by extracting the WT level rise that can be attributed solely to recharge [Healy and Cook, 2002; Crosbie et al., 2005]. However, in our study, instead of estimating recharge from the $\frac{dh}{dt}$ time series as done by Crosbie et al.[2005), we use a simple exponential function to estimate $R_e$ based on the WT elevation and rainfall, which upon division by the $\lambda_f$ yields $\frac{dh_R}{dt}$ time series.

Figure 3-3 shows the schematic illustration of the sudden rise of an initially declining WT during a single time step (dt) whose magnitude is given by line BD. The observed value of $\frac{dh}{dt}$ is the ratio AB/BD which reflects the net effect of both recharge and the groundwater outflow. Note that in this particular case the groundwater flow drives the water table downward while the recharge drives it upward. Line DC in Figure 3-3 represents the anticipated locus of the WT elevation if there was no recharge [Healy and Cook, 2002]. This anticipated path of WT can be determined by estimating the groundwater flux at point D and dividing this value by $\lambda_d$. The ratio
AC/BD is the contribution of recharge on the WT rise. The exact change in WT \( \left( \frac{dh_E}{dt} \right) \), which is the ratio BC/BD, is then simply obtained as the difference between AC/BD and AB/BD. Note that when the water table recession is very slow during the time period from point D to C, i.e., if the groundwater flow rate is small, BC \( \approx 0 \), and AC \( \approx AB \). Such scenario may be present in shallow unconfined aquifers of large lateral extent where the groundwater flow is much slower as compared to the vertical recharge during rainfall.

Extraction of \( \frac{dh_E}{dt} \) from \( \frac{dh}{dt} \) requires the reliable estimation \( \lambda_f \) parameter [Healy and Cook, 2002] which changes dynamically during the infiltration process [Chapman and Dressler, 1984; Chapman, 1995]. The method employed in this study to extract \( \frac{dh_E}{dt} \) can be summarized as follows: Firstly, the observed \( \frac{dh}{dt} \) values are calculated from the entire dataset which is the total effect of \( ET_g \), \( q \), and \( R_e \). The next step requires estimation of \( R_e \) from any appropriate method which then enables the calculation of \( \lambda_f \) as the function of the WT elevation and \( R_e \). During recharge, the amount of water that accretes on the WT may be much larger than the amount removed/added via groundwater flow. Therefore, neglecting the groundwater flow during a particular time step with recharge, \( \frac{dh_R}{dt} \) can be estimated as

\[
\frac{dh_R}{dt} = \frac{R_e}{\lambda_f} \tag{3-10}
\]

Note that in equation 3-10, in addition to neglecting the groundwater flow, instantaneous recharge is assumed, which is reasonable for shallow water tables [Hilberts et al., 2007; Laio et al., 2009]. However, this assumptions may not hold in deeper WT environments where a lag in WT response is likely to occur during precipitation events as the infiltrated water need longer time to reach the water table. This lag in WT response, if likely to occur, can be determined by comparing the time series of rainfall and the \( \frac{dh}{dt} \) values. If a lag in \( \frac{dh}{dt} \) relative to the time of
rainfall is observed, i.e., if perceptible response of WT to \( R_e \) is not detected in that particular time step, then the \( \frac{dh_d}{dt} \) calculated from equation 3-10 is applied to the following time step that shows the effect of recharge (by sudden rise in WT). After this process is repeated over the entire time series of WT elevation, \( \frac{dh_d}{dt} \) can be simply calculated as

\[
\frac{dh_d}{dt} = \frac{dh}{dt} - \frac{dh_R}{dt}
\]  

(3-11)

During intensive rainfall events, estimated \( \frac{dh_R}{dt} \) values become very high, even yielding infinite values since large recharge rates drive \( \lambda_f \) towards zero [Chapman and Dressler, 1984]. In such cases, \( \frac{dh_d}{dt} \) can be replaced by the observed \( \frac{dh}{dt} \) since \( \frac{dh_R}{dt} \) will not exceed the observed \( \frac{dh}{dt} \) by a large magnitude at a given time step, unless the groundwater flux rate is very high. Figure 3-4 shows the observed \( \frac{dh}{dt} \), and the actual \( \frac{dh_d}{dt} \) calculated during three separate rainfall events at a field site in northeast Florida.

**ETg Calculations**

Once the effect of precipitation have been removed from the observed \( \frac{dh}{dt} \) time series, equations 3-6 and 3-7 can be solved to obtain the ETg estimates as follows:

during groundwater inflow,

\[
ETg_t = q_t - \lambda_d(h_{t-1}, R_{e_t}) \left( \frac{dh_d}{dt} \right)_t \hspace{1cm} R_e > 0
\]

(3-12)

\[
ETg_t = q_t - \lambda_d(h_{t-1}, ETg_{t-1}) \left( \frac{dh_d}{dt} \right)_t \hspace{1cm} R_e = 0
\]

(3-13)

and during groundwater outflow,

\[
ETg_t = q_t - \lambda_d(h_{t-1}, R_{e_t}) \left( \frac{dh_d}{dt} \right)_t - \lambda_f(h_{t-1}, R_t) \left( \frac{dh_R}{dt} \right)_t \hspace{1cm} R_e > 0
\]

(3-14)

\[
ETg_t = q_t - \lambda_d(h_{t-1}, ETg_{t-1}) \left( \frac{dh_d}{dt} \right)_t \hspace{1cm} R_e = 0
\]

(3-15)
where, \( t \) denotes the time step (hourly in this study) and all other symbols are as previously defined. Note that in equation 3-12 the term \( \frac{\lambda_d}{\lambda_f} \) is set equal to unity. This was because the original equation produced highly unstable \( ET_g \) estimates as compared to equations 3-12 and 3-13. This was possibly because of the dynamic nature of equations 3-12 and 3-13 since they require estimations of \( \lambda_d \) and \( \lambda_f \) using the \( ET_g \) value calculated at the previous time step to compute \( ET_g \) on the current time step. Moreover, if the magnitude of \( q \) in the aquifer is small, the first term on the right in equation 3-12 and 3-13 will not change by much even when the \( \frac{\lambda_d}{\lambda_f} \) is not treated as unity. Also, in equation 3-14 the recharge term is replaced by the product of \( \lambda_f \) and \( \frac{dh_E}{dt} \), which gives the actual value of recharge that was estimated during the calculation of \( \frac{dh_E}{dt} \), if the method involved accounting for the lag in WT response. This procedure therefore will ensure consistency in the overall calculation procedures.

**Application of the Modified Method**

**Site Description and Water Table Data**

The field site chosen for the application of modified White method developed in this study is located in northeast Florida. This area is typically characterized by low, flat landscape with a shallow water table predominantly planted to potato. The fields in the area are irrigated by a conventional water table management system known as “seepage irrigation”. In this system, shallow water table is maintained near the crop root-zone during the entire cropping season [Campbell et al., 1978] by simultaneously supplying water from individual water furrows and raising the water level in the ditches. Water is then supplied to the crop root-zone by capillarity and subsurface lateral flow [Pitts and Smajstrla, 1989; Smajstrla et al., 2000]. Figure 3-5 shows location of the study sites near the Atlantic coastline in northeast Florida. Each
seepage irrigated field is divided into approximately 18-20m wide crop beds separated by smaller, shallow drains (or furrows) approximately 0.30-0.50m wide. During irrigation the drains are continuously supplied with water which brings WT near the root-zone. The level of water in the drains and ditches thus determines the ultimate height of WT in the field. It can be hypothesized that almost all of the ET demand of the crops is contributed by the shallow WT under this water management system. It is based on the results of several previous studies on shallow groundwater areas [e.g., Nachabe et al., 2005; Shah et al., 2007; Lowry and Loheide, 2010 etc.]. In our study site, WT is very close to the surface which implies a more plausible scenario of full contribution of WT to plant ET. These field sites therefore also provide a setting to test this hypothesis by estimating $ET_g$ from observed diurnal water table fluctuations.

The soil at the sites is classified as Ellzey fine sand series (sandy, siliceous, hyperthermic, arenic endoaqualf; 90-95% sand, <2.5 % clay, <5% silt) [NRCS, 1999]. Although this series is predominated by sand, Acharya and Mylavarapu [2011] studied different physical properties of the soil at the study site at varying depths and reported that the clay content in the profile increased from 2% at 0-22cm to 11% at 90-120 cm. They also reported that the lab-determined average saturated hydraulic conductivity of the soil profile from 0-120 cm depth was approximately 10 cm/hr while in an in-situ measurement performed by Rosa [2000], it was found that $K_s$ of the soil was approximately 7 cm/hour. The latter was used in this study because in-situ measurement is more likely to represent the real world situation as compared to the lab-determined values.

Two monitoring wells, 1.1 m deep were installed in two crop beds near the center of a potato field of approximately 15ha (400m x 400m) area in the spring of 2010. Hourly water table elevation in the wells was recorded using the pressure transducers for 50 days from March to
May. In spring 2011, in addition to the field-1, two wells were installed in a second potato field (field-2) of similar dimensions and the water table elevations were recorded for approximately 50 days. Water table elevations recorded by the pressure transducers were then adjusted for the variations in barometric pressure recorded at the center wells of each field to obtain the corrected WT elevations. The approximation positions of the the wells in the two field sites is shown in Figure 3-5.

**Estimation of Recharge**

The recharge values were estimated using an empirical $R_e$ Vs.WT function which depends both on rainfall rate and WT elevation and is determined by the soil type. As shown in Chapter 2, for estimating the rise in shallow WT during rainfall events, this functions approximated the WT rise resonably well in the study site. The empirical function, which depends on the soil type can be expressed as

$$R_e = \alpha P = P \cdot \exp(-\beta_r (d_r - h)) \quad h \leq d_r$$

$$R_e = P \quad h > d_r$$

where, $P$ is the rainfall amount during the given time step, and $\beta_r$ and $d_r$ are the parameters that determine what fraction of rainfall is converted to recharge depending on the soil type. Equation 3-16 and 3-17 thus provide a method of estimating recharge and the consequent WT rise during rainfall without explicitly considering the unsaturated zone transport processes [Woods et al., 1997]. The values of $d_r$ and $\beta_r$, which were determined by fitting the equations with the WT rise obtained from hypothetical infiltration simulations by Hydrus 1D [Simunek et al., 1998], were estimated as 60 cm and 0.15 respectively. All other parameters required and used in this study were adopted from the earlier study conducted in the field sites.
Estimation of Subsurface Flux ($q$)

In areas with WT control systems, ditches and drains are appropriately constructed to raise or lower the WT elevations as needed during the crop growth period. Subsurface flow in these fields is dictated by the water level in the drains which can be treated as entirely horizontal, adopting the Dupuit-Forcheimmer (D-F) theory [van Schilfgaarde, 1974; Ritzema, 1994]. Figure 3-6 shows a conceptual representation of the WT profile in the study field with main lateral ditches and smaller inner ditches (or, furrows) during an irrigation period. During the drainage phase the water table profiles are exactly opposite (i.e. convex WT profile) than during irrigation phase. From the D-F theory, if an instantaneous steady state flow at each time step is assumed, the subsurface lateral flux to and from the ditches which fully penetrate to the impervious layer can be estimated as (Ritzema, 1994)

$$q_D = \frac{4K_s(h_m^2 - h_D^2)}{L_D^2} \quad (3-18)$$

where, $q_D =$ flux towards the outer ditches; $K_s =$ lateral saturated hydraulic conductivity of the soil; $h_m =$ height of WT at midpoint between the ditches, measured from impermeable layer; $h_D =$ height of the water level in the ditches; and $L_D =$ the distance between the main ditches.

However, if the drainage ditches are shallower, the flow converges near the drains which needs to be accounted for while estimating the subsurface flux. Flow to and from the shallow and partially penetrating drains can be estimated by the Houghoudt’s equation as

$$q_d = \frac{4K_s m(2d_e + m)}{L_d^2} \quad (3-19)$$

where, $m =$ difference between the height of WT at the drains and WT at midpoint between the drains; $L_d =$ the distance between the smaller drains which is normally 18m in the study fields. and $d_e =$ equivalent depth of the impermeable layer introduced to account for the convergence of
flow near the shallow drains [Moody, 1966; van Schilfgaarde, 1974; Ritzema, 1994]. Moody [1966] derived a simple formula to calculate the equivalent depth of the impermeable layer from drain radius, drain, depth of the impervious layer, and distance between the drains, which was used in this study. However, any other appropriate method [e.g., van Beers, 1979; van der Molen and Wesseling, 1991] may be used to estimate the equivalent depth. Once both $q_D$ and $q_d$ are estimated, the total flux can be simply obtained as $q = q_d + q_D$.

**Penman-Monteith ET**

Estimation of ET from Penman-Monteith (PM) method [Monteith, 1965] requires the data on solar radiation, temperature, relative humidity, and wind speed. These weather data were obtained from a Florida Automated Weather Network (FAWN) weather station adjacent to the field sites [FAWN, 2011]. The standardize PM method for hourly reference ET [Allen et al., 1998] is expressed as

$$ET_0 = \frac{\Delta (R_n - G)}{\lambda} + \frac{\gamma}{\frac{T}{273.16} + 37} \left( U_2 \cdot \frac{e_s - e_a}{\Delta + \gamma (U_2 + C_d U_2)} \right)$$

(3-20)

where $ET_0$ = reference evapotranspiration (mm/hr); $R_n$ = net solar radiation (MJ/m$^2$h); $G$ is the soil heat flux (MJ/m$^2$hr); $U_2$ = wind speed at 2m height (m/s); $\Delta$ = slope of the saturation vapor pressure curve (kPa/°C) at mean air temperature (T); $e_s$ = saturation vapor pressure at the mean hourly temperature (kPa); $e_a$ = actual vapor pressure at mean hourly temperature; $\lambda$ = latent heat of vaporization (MJ/kg); $C_d$ = bulk surface and aerodynamic resistance coefficient. Equation 3-20 was used to calculate the hourly reference ET ($ET_0$) for the field site during the spring of 2010 and 2011. The reference evapotranspiration was then converted to the actual $ET$ for potato using the potato crop coefficient reported by Allen et al. [1998]. The estimations were obtained for approximately 50 days in both years starting from the mid-growth stage of potato until one or two weeks before harvesting.
Results and Discussions

Hourly $ETg$ Estimates

Hourly estimates of $ETg$ from the governing equation implemented only with hydrostatic-$\lambda_d$ (HSM, hereafter) and dynamic-$\lambda_d$ and $\lambda_f$ (DM hereafter) compared with the standardized PM estimates during the entire study periods in 2010 and 2011 are presented in Figure 3-7. The data presented in Figure 3-7 are from the well located at the center of field-1 in 2010 and field-2 in 2011. However, the data from the other wells also showed highly similar results in both fields (data not shown). Note that during both seasons the days with rainfall were included in the calculations of $ETg$ by both HSM and DM. Generally, during rainless days, the HSM significantly overestimated ET as compared to the DM. Overestimation of $ETg$ by the HSM was more prominent in field-1 during 2010 than in field-2 during 2011. The generally higher amplitudes of the DWTF in 2010 than in 2011 (Figure 3-7) might have caused such higher degree of overestimation in 2010. Note that in 2010, during majority of the study period, average amplitude of diurnal WT fluctuations was more than 10cm while in 2011 it was less than 10cm. This was because of the generally low potential ET during the 2011 study period than during 2010. The weather data from the weather station revealed that, although the daily average temperatures during 2010 and 2011 in the site did not differ significantly (data not shown), 2010 was relatively drier than 2011 during the study period. This generally higher relative humidity might have played a role in producing higher PET values during 2010 which, on average, was 0.1cm more than in 2011 study period (March-May). The average RMSE of hourly $ETg$ estimates in 2010 and 2011 seasons were 0.012 cm and 0.0008cm for the HSM and DM respectively suggesting that the DM was able to vastly improve the $ETg$ estimates from the field sites during the study periods.
Since one of our main objectives was to study how the White method based hourly and daily ETg estimates are affected during and immediately after precipitation events, the results from HSM and DM models were compared specifically during rainy and clear days. Figures 3-8 and 3-9 show the hourly ETg estimates from HSM, DM and PM ET superimposed on the associated WT elevation curve during and immediately after two rainfall events and two rainless periods in 2010 and 2011. Note that even on the days with precipitation, hourly pattern of ETg estimates did not deviated greatly from the normal diurnal pattern (i.e. as shown by PM ET). It was found that during the rainfall events also, the ETg estimates from DM were fairly close to the PM ET while HSM tended to overestimate.

In general, there was a significant difference in the ETg estimates form HSM and DM immediately after rainfall when the WT was rapidly receding due to drainage and ET. However, for the rainfall event on 05/05 in 2010 and during the WT drawdown period following the event (Figure 3-8), the two method were mostly similar to each other and to the PM ET. The PM ET, on the other hand, showed relatively smaller values during this 3 day period as compared to the days before rainfall (05/04) and the times after WT recession stopped (05/09). This was also reflected by the relatively less-steep WT recession curve during that period as compared to the other drawdown events (e.g. 4/26-4/27 in 2010 and 5/16-5/18 in 2011). These observations are consistent with the theory of flux-dependent λd which suggests that as the flux rate decrease, the difference between hydrostatic-λd and dynamic-λd decreases and vice versa. As ET rates become smaller, the corresponding soil moisture profile above WT moves towards the hydrostatic profile and the two parameters become essentially the same. As a result DM also behaves as the HSM producing similar ETg estimates. These results suggest that the method used in our study to
account for contribution of recharge in observed $\frac{dh}{dt}$ performed reasonably well, improving the $ETg$ estimates during and after most rainfall events throughout the two study periods.

A closer evaluation of hourly $ETg$ estimates during the WT drawdown periods after each rainfall show that the result from DM model tend to fluctuate substantially near the end of each drawdown event, i.e., when the WT recession pattern halts as the irrigation is resumed. For example in Figures 3-8 and 3-9, a significant fluctuations in the daytime $ETg$ estimates from DM is observed on 4/15 and 4/28 in 2010 and 4/10 and 4/19 in 2011. This was due to the associated lag in the WT response to the irrigation from the smaller ditches (furrows). In the field, as the irrigation is resumed, the WT doesn’t necessarily begin to rise immediately but it requires a few hours in order for the water in the furrows to build up, infiltrate towards the WT, and begin lateral movement. However, in our calculations, this lag in WT response to irrigation was not considered since the infiltration from the smaller ditches (furrow) was not modeled. Therefore, during these particular times, there is a discrepancy between the observed WT slope and the associated groundwater flux in the calculations, resulting in the fluctuating $ETg$ values.

**Daily $ETg$ Estimates**

The daily values of ET and estimated $ETg$ also followed similar patterns observed in the hourly estimates. Average daily PM ET for the potatoes in 2010 and 2011 were approximately 0.65 and 0.55 cm while the average daily $ETg$ estimates were 0.95 & 0.82 for HSM and 0.64 and 0.56 for DM respectively. Daily $ETg$ estimates from HSM methods were also significantly higher than both DM and PM estimates during the clear days which is presented in Figure 3-10. Average RMSE of daily estimates were 0.28 cm and 0.009 cm for the two methods in 2010 and 2011 respectively, indicating a massive improvement in $ETg$ estimation by DM over the HSM method.
The potential PM ET values estimated in our study fall within the range of the previously reported potential ET values for the region [e.g., Jones et al., 1984; Smajstrla et al., 1985]. However, data on continuous crop coefficient (\(K_c\)) of potato in the area is still lacking which implies some degree of uncertainty on the actual PM ET values estimated for the potato crop. Majority of our study period included the mid growth stage of potatoes when the water demand is at the highest level. During the mid season growth, the FAO reported \(K_c\) for potato [Allen et al., 1998] is 1.10, which was used in our study. Other studies have suggested that this value may be higher than 1.1 for the area. Singleton [1990] reported monthly \(K_c\) values as high as 1.35 for the region based on the time of planting which would have produced actual PM ET values in the range of 0.65-0.75 for our study period. However, due to the lack of continuous data on \(K_c\), we used the FAO values which are generally regarded as the standard crop coefficients. Nevertheless, it should be noted that even if the actual potato \(K_c\) values were higher, the average daily ETg estimates from HSM method would still be much higher than the PM ET as compared to the DM suggesting the better performance by DM. Overall, the DM significantly improved the ETg estimates by HSM method during majority of the study periods in 2010 as well as 2011.

Interestingly, the daily ETg estimates on the rainy days were essentially similar for both HSM and DM at almost all instances (Figure 3-10). One of the primary reasons behind this may be the occurrence of mostly nighttime precipitation. During nighttime highly negative ETg estimates were converted to zero since during this time period ET may be negligibly small in most cases. Another cause may be that even the daytime rainfalls generally lasted only for a few hours in most precipitation events. As shown in Figures 3-7, 3-8 and 3-9, the longest rainfall durations were only a few hours during both 2010 and 2011 study periods. After each rainfall, the WT recession curves show a clear effect of diurnal ET pattern. That means the disturbances
to diurnal WT patterns due to rainfall lasted only briefly. Since the HSM method generally overestimated \( ET_g \) during immediate hours after the rainfall events, the 24-hour sum during those days was still similar to the DM estimates as well as the PM ET.

On the contrary to the exact days with rainfall, the HSM method significantly overestimated both DM an PM ET which was not surprising since the hourly estimates showed clear overestimation on most of the rainless days. In both 2010 and 2011, there were only a few days on which the DM method overestimated PM ET. The daily sums therefore also corroborate the hourly results suggesting that our method performed satisfactorily to handle the rainy days during \( ET_g \) estimations from the diurnal WT fluctuations.

The most critical aspect of estimating \( ET_g \) from DWTF, attributing the observed \( \frac{dh}{dt} \) appropriately to \( ET_g \) which is controlled by the \( \lambda_d \). Our study showed that during clear days, the \( ET_g \) dependent drainable porosity \( (\text{dynamic-} \lambda_d) \) predicted hourly estimates significantly better than the \( \text{hydrostatic-} \lambda_d \), which consistently produced higher values. Because the unsaturated zone soil moisture flux has a substantial effect in the magnitude of drainable porosity \([\text{Childs}, 1960; \text{Tritscher et al., 2000}]\), the WT drawdown is more appropriately attributed to \( ET_g \) when \( \text{dynamic-} \lambda_d \). This effect is not incorporated in \( \text{hydrostatic-} \lambda_d \), so it attains a greater value than the \( \text{dynamic-} \lambda_d \) for the same WT elevation, resulting in larger \( ET_g \) values. This is also explained by the observed dynamics of the two parameters during the \( ET_g \) computation process. Figure 3-11 shows the hourly variations in \( \text{hydrostatic-} \lambda_d \) and \( \text{dynamic-} \lambda_d \) due to the WT fluctuation during WT recession after rainfall and during clear days in 2010. Note that at daytime, \( \text{dynamic-} \lambda_d \) values decrease by as much as 50% of \( \text{hydrostatic-} \lambda_d \) values while at the nighttime, ET falls close to zero and \( \text{dynamic-} \lambda_d \) collapses into the \( \text{hydrostatic-} \lambda_d \).
An important aspect of the method presented in this study is that $ET_g$ and the storage parameters are estimated dynamically by using estimated $ET_g$ values from the the previous time step. This process sometimes may cause some fluctuations in the $\lambda_d$ and $\lambda_f$ and hence in the hourly $ET_g$ estimates. However, the overall results were generally consistent throughout the study periods in both 2010 and 2011 which strongly indicate that the method can be used effectively to estimate $ET_g$ form the diurnal water table fluctuations.

**Conclusions**

In this study we have presented an application of distinct, flux-dependent $\lambda_d$ and $\lambda_f$ parameters to estimate subdaily as well as daily $ET_g$ using a modified form of the White method. Most of the previous studied based on this method used either constant $\lambda_d$ [Lautz, 2008; Mould et al., 2010] or a single $\lambda_d$ estimated either from static soil moisture profile [Gribovszki et al., 2007] or from analysis of WT rise during rainfall events [Dolan et al., 1984; Schilling, 2007]. Use of these separate, flux-dependent storage parameters enable the modification of the White method by accounting for the effect of rainfall on the observed WT fluctuations. This modification therefore allows for the reasonable application of the White method even during rainy days. In order to include the rainy days in the calculations, a simple method to extract the recharge induced WT fluctuation from the observed hourly changes in WT was also employed in our study. A critical aspect of this method is the appropriate estimation of WT recharge from rainfall which also affects the magnitude of $\lambda_f$. The empirical recharge function used in the study produced satisfactory results for the shallow WT in northeast Florida potato fields providing good estimates of the recharge induced WT rise. For deeper WT, however, the method may need slight modifications to account for the lag in the WT response. Our method showed good estimations of the Penman-Monteith ET for potato in during spring of 2010 and 2011 even when precipitation events were included in the calculations.
Drainable porosity plays a critical role in estimating ET_{g} from White based methods since this method attempts to attribute the observed WT fluctuations to direct ET loss from groundwater via this \( \lambda_{d} \) parameter. Our study also showed that the \( \lambda_{d} \) estimated from hydrostatic soil moisture profile normally failed to appropriately attribute the WT fluctuations to the groundwater ET, producing significant overestimations. However, when the effect of ET is incorporated into the \( \lambda_{d} \), its value become smaller than hydrostatic-\( \lambda_{d} \) since this dynamic-moisture-profile based parameter also accounts for the soil moisture that is being lost from the profile via ET, in addition to the water lost by drainage. These smaller \( \lambda_{d} \) values therefore resulted in more reliable estimates of ET_{g} at hourly as well as daily time scales.

Our study was based on the assumptions of non-hysteretic soil moisture retention properties. It is well known that hysteresis also tends to cause significantly different \( \lambda_{d} \) and \( \lambda_{f} \) values [Nachabe, 2002; Nachabe et al., 2004; Loheide et al., 2005]. Since these hysteretic parameters are applicable in falling and rising water tables respectively, the resulting equation of ET_{g} will also need modification if hysteresis is considered [e.g. Loheide et al., 2005]. However, in field conditions the hysteretic effects may not always be significant due to other sources of variations. The results from our study further accentuate the critical role of \( \lambda_{d} \) and \( \lambda_{f} \) in White based ET_{g} estimation methods while providing a general framework of applying the method continuously regardless of the precipitation events using these two storage parameters. Although our conclusions are based on relatively short term results from a managed crop-field, close agreement between the results from two consecutive years suggest that the method is consistent and can be equally applied to other field conditions.
Table 3-1. Root Mean Square Error (RMSE) of hourly and daily $ET_g$ estimations using single, hydrostatic-$\lambda_d$ and distinct, dynamic-$\lambda_d$ and -$\lambda_f$.  

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<thead>
<tr>
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<th>RMSE (cm)</th>
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<tr>
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<td>Hourly</td>
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<tr>
<td>2010</td>
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<tr>
<td>Eq-$\lambda_d$</td>
<td>0.013</td>
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<tr>
<td>NonEq $\lambda_d$ and $\lambda_f$</td>
<td>0.0007</td>
</tr>
<tr>
<td>2011</td>
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<tr>
<td>Eq-$\lambda_d$</td>
<td>0.012</td>
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<tr>
<td>NonEq $\lambda_d$ and $\lambda_f$</td>
<td>0.001</td>
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Figure 3-1. Diurnal fluctuations in the solar radiation (red) and observed water table depth (blue) at a northeast Florida site.

Figure 3-2. Drainable and fillable porosity of Ellzey sand series under potato in northeast Florida under different potential-ET and precipitation rates.
Figure 3-3. Illustration of contribution of recharge induced sudden WT rise during a rainfall event and extraction of the exact change \( \left( \frac{dh_R}{dt} \right) \) from the observed \( \left( \frac{dh}{dt} \right) \) by subtracting the contribution of recharge \( \left( \frac{dh_R}{dt} \right) \).

Figure 3-4. Calculated hourly actual change in water table elevation \( \left( \frac{dh_E}{dt} \right) \) from the observed data after removing the contribution of recharge \( \left( \frac{dh_R}{dt} \right) \) during four rainfall events in spring 2010.
Figure 3-5. Land use of the three counties in northeast Florida and the two field sites with well positions, chosen for the study.

Figure 3-6. Conceptual representation of the watertable profiles and the flow towards the ditches in the field.
Figure 3-7. Time series of the hourly rainfall, Penman-Monteith and groundwater ET predicted by using a single drainable porosity (middle plots) and distinct drainable and fillable porosity (bottom plots) from field site 1 in 2010 (top), and field site 2 in 2011 (bottom) during the spring potato seasons.
Figure 3-8. Hourly $ET_g$ from the HSM method (black dashed line), DM method (red solid line), and PM ET (blue dots) during clear periods (top) and two rainfall events in 2010.

Figure 3-9. Same as Figure 8, during 2011 in field site 2.
Figure 3-10. Scatter plots of daily Penman-Monteith ET and predicted $ET_g$ from field site 1 in 2010 (left) and field site 2 in 2011 (right) during spring potato seasons.

Figure 3-11. Plots showing the variations in hydrostatic-$\lambda_d$ (open circles) and dynamic-$\lambda_d$ (filled diamonds) during WT drawdown after rainfall (top) and dirunal fluctuations in the WT elevation in a field site in 2010.
CHAPTER 4
MODELING SHALLOW WATER TABLE DYNAMICS UNDER A SUBSURFACE IRRIGATION AND DRAINAGE SYSTEM IN FLORIDA

Background

Shallow water table (WT) commonly exists in many areas of the world where field hydrology of such differs significantly from well drained soils with deep water table. The hydrology as well as growth and development of plants in such SWT areas are controlled by the dynamics of the phreatic surface [Bierkens, 1998; Nachabe et al., 2002]. Agricultural crop production in areas with shallow WT therefore requires an efficient drainage system to create optimum environment from crop growth and field operations [Schilfgaarde, 1974; Tang and Skaggs, 1977]. Drainage is used to remove excess soil moisture from the root-zone [Fipps and Skaggs, 1989]; to control potential salinity problems in irrigated arid areas [Fipps and Skaggs, 1989; Ritzema, 1994]; and to discreetly manage the water resources. In humid areas such as Florida, since the groundwater level is close to the surface, even a small amount of precipitation may bring WT close to the surface [Gillham, 1984; Novakowski and Gillham, 1988; Hilberts et al., 2007]. This requires rapid drainage to avoid excessive moisture stress to plants. Such rapid WT drawdown is normally enabled by constructing ditches and installing drains in the field. However, since the drains and ditches hasten the drainage process, the soil around the crop root-zone is likely to dry quickly. Therefore many of these subsurface drainage systems are operated as WT management systems which can be used for controlled, subsurface drainage as well as irrigation whenever necessary.

In Florida, groundwater level is normally close to the surface in majority of the agricultural areas. Crop production in these areas has been conventionally performed by managing the WT throughout the growing seasons. Water table is maintained at a desired elevation near the crop root-zone using open ditches and appropriately spaced surface water-
furrows, conventionally known as seepage irrigation system. Moisture supply to the plants in this system occurs by subsurface lateral flow and capillarity [Pitts and Smajstrla, 1989; Smajstrla et al., 2000]. Since the WT is very close to the root-zone, majority of the crop evapotranspiration (ET) demand is contributed by the direct upward flow from the water table [Nachabe et al., 2005].

The knowledge of WT dynamics due to diurnal ET forcing, frequent rainfall events, and water level changes in field ditches is therefore important for maintaining optimum soil moisture around the crop root-zone. The process of WT movement due to subsurface drainage/irrigation or rainfall/ET from the soil profile is governed by the principles of water movement through soils. Therefore, the design and analysis of such systems can typically be done with the help of Richard’s equation since it is the most rigorous approach to describe the water movement phenomenon in soils [Pikul et al., 1974; Tang and Skaggs, 1977; Hilberts et al., 2005]. However due to the inherent difficulties in numerical solutions of this equation, approximate methods that describe the drainage process are normally adopted [Tang and Skaggs, 1977]. Most of the research on the subsurface drainage of agricultural lands and hillslopes are based on the Dupuit-Forcheimemer (D-F) assumptions to describe groundwater flow in unconfined aquifers [e.g, van Schilfgaarde, 1963; Moody, 1966: Skaggs, 1991; Verhoest and Troch, 2000; Rupp and Selker, 2005; Singh and Jaiswal, 2006; etc]. The D-F assumption states that in shallow, unconfined aquifers with small WT inclinations, the vertical hydraulic gradient can be neglected and the flow can be assumed to occur entirely in the horizontal direction. Adoption of these assumptions results in Boussinesq-type groundwater flow models, also known as the hydraulic groundwater model [Brutsaert, 2005].
The Boussinesq-type groundwater flow models consist of a storage parameter which enables the expression of aquifer storage in terms of water table elevation. The response of WT to groundwater inflow and outflow is therefore controlled by this parameter. When the water table is receding, the parameter is called as the drainable porosity ($\lambda_d$) which gives the change in aquifer storage per unit decline in WT depth [Bouwer, 1978; Freeze and Cherry, 1979]. In case of WT rise, the parameter is termed as fillable porosity ($\lambda_f$) which is controlled by the moisture-deficit of the aquifer [Bouwer, 1978; Healy and Cook, 2002; Park and Parker, 2008].

It is now well known that the moisture storage, and hence WT dynamics of unconfined aquifers is significantly affected by the capillary properties of the soils [Gillham, 1984; Nielsen and Perrochet, 2000; Healy and Cook, 2002, etc.]. Some authors have incorporated the effect of soil moisture retention in estimation of aquifer storage parameters to study subsurface flow and water table dynamics in hillslopes and wetland environments [Hilberts et al., 2005, 2007; Laio et al., 2009]. However, all of the previous studies on the water table response to subsurface drainage and irrigation in managed agricultural lands treated this parameter as constant [Skaggs, 1991, Singh and Jaiswal, 2006 etc.]. This can introduce substantial errors in estimated WT rise or drawdown rates since the effect of unsaturated flow due to ET and recharge is not incorporated in parameter estimation. As shown in Chapter 1, although treated as a single common parameter estimated from static soil moisture profiles in recent studies [Bierkens, 1998; Hilberts et al., 2005], these two parameters are affected by the unsaturated fluxes above the phreatic surface and can assume substantially different values depending on the magnitude of ET or recharge fluxes.

In this study we numerically solved the 1D Boussinesq equation implemented with separate, flux-dependent $\lambda_d$ and $\lambda_f$ parameters in contrast to the conventional Boussinesq model.
with only one, hydrostatic- $\lambda_d$ parameter. The dual parameter model is therefore theoretically more robust since it appropriately implements the storage parameters depending on the effect of various water budget components on the direction of WT movement. The main objective of the study was to evaluate the effectiveness of the WT management system to maintain the desired WT elevation throughout the cropping period to create favorable soil moisture environment around-the crop root zone. The second objective was to evaluate the response of WT to potential alternative management scenarios. Firstly conceptual models of WT management system under subsurface irrigation and drainage system are presented. These WT management systems may consist of a typical ditch-drainage design as well as more complex systems with smaller, surface or subsurface drains in between the two main ditches. Appropriate boundary conditions are then applied to the system during irrigations and drainage to solve the governing equations by implicit finite difference method. Simulated water table elevations are then compared with the field data collected from a potato fields in northeast Florida managed under a subsurface irrigation/drainage system. The sensitivity of the water table movement to changes in soil parameters such as $K_s$, and the field characteristics, and potential alternative water management system specific to northeast Florida region is briefly discussed.

**Materials and Methods**

**Study Site**

The field site chosen for this study is located in northeast Florida (29.694’N, 81.446’W) near the Atlantic coastline (Figure 4-1). The flat and low (<10 masl) landscape with with a shallow groundwater level in the area therefore requires appropriate water table management measures. This is achieved by constructing larger (1-2 m deep) ditches around the field as well as smaller and shallow surface drains conventionally known as “seepage irrigation”. Each seepage irrigated field is divided into approximately 18-20 m wide crop beds separated by smaller drains
(or furrows) approximately 0.30-0.50 m wide. The shallow water table is maintained near the crop root-zone during the entire cropping season [Campbell et al., 1978] by supplying water from the inner drains and raising the water level in the ditches simultaneously. Moisture is supplied to the crop root-zone by capillary action and subsurface lateral flow [Pitts and Smajstrla, 1989; Smajstrla et al., 2000, Munoz et al., 2008; Acharya and Mylavarapu, 2011]. The water supply in the smaller drains is cut-off during and after rainfall vents and the water level in the ditches is lowered thus allowing for quicker water table drawdown. The ultimate water table elevation in the crop beds is thus determined by the water level in the ditches and the drains which change during irrigation and drainage. The soil at the field sites is classified as Ellzey fine sand series (sandy, siliceous, hyperthermic, arenic endoaqualf; 90-95% sand, <2.5 % clay, <5% silt) [NRCS, 1999]. Although this series is predominated by fine sand, soil contains some clay content which ranged from 2% at shallower depth to 11% at 120 cm [Acharya and Mylavarapu, 2011]. Lab determined average vertical saturated hydraulic conductivity (K) of the soil was 10 cm/hr while the insitu determined K values of the soil was found to be 7cm/hour [Rosa, 2000].

Model Development

Conceptual models: irrigation and drainage

Figures 4-2 and 4-3 show the conceptual representation of the water table profile during irrigation and drainage respectively in the study fields. At the start of an irrigation event, water is supplied to the inner drains while the water level is raised in the ditches from an initial height ($h_0$). Note that in Figure 4-2, initial water table profile is horizontal while in drained lands the water table normally tends to occur as curved profile [Skaggs, 1973; van Schilfgaarde, 1974]. We assumed a horizontal profile because of the large distance between the two ditches (normally
> 250 m) due to which the water table profile is usually close to horizontal near the end of a drainage event.

Whenever there is an anticipated rainfall event, the water supply to the field is cut-off, and water level in the ditches is lowered thus triggering subsurface drainage of the profile. Normally after rainfall, the drainage occurs in two distinct phases as shown in Figure 4-3. If the water table is raised close to the surface (i.e., above the depth of the furrows), both the outer ditches as well as the furrows contribute towards the drainage. This helps is rapid drawdown of the water table during the initial hours after cessation of rainfall. This is termed as the phase-1 drainage in Figure 4-3. As the water table moves below the level of the depth of the furrows, they cannot contribute to drainage anymore and the system is converted to a typical ditch drainage system with only the outer ditch removing the subsurface water from the profile. This is termed as phase two drainage in Figure 4-3. Phase-2 drainage is normally much slower than the phase-1 drainage due to large distance between the two ditches.

**Governing equations**

Since the subsurface drainage phenomena occur largely by the lateral movement of soil moisture below the water table, it can be described by the one dimensional Boussinesq equation subject to recharge and evapotranspiration. Specific governing equations implemented with the drainable and fillable porosity for the dynamics of the water table are discussed in Chapter 1. The equations for the drainage and irrigation phases differ only on the use of the storage parameter and can be expressed as

\[
\frac{dh}{dt} = \frac{1}{\lambda_d} \left[ K_s \left( \frac{d}{dx} \left( h \frac{dh}{dx} \right) \right) \right] - R_e \left( \frac{1}{\lambda_d} - \frac{1}{\lambda_f} \right) - \frac{ET}{\lambda_d} \tag{4-1}
\]
during drainage, when the water table is receding

\[
\frac{dh}{dt} = \frac{1}{\lambda_f} \left[ K_s \left( \frac{d}{dx} \left( h \frac{dh}{dx} \right) \right) \right] - \frac{ET}{\lambda_d} + \frac{R_e}{\lambda_f} \tag{4-2}
\]
during irrigation, when the water table is rising, and

where,

\[ h = \text{the elevation of water table measured from a reference datum} \]

\[ K_s = \text{saturated hydraulic conductivity of the profile} \]

\[ \lambda_d = \text{drainable porosity} \]

\[ \lambda_f = \text{fillable porosity} \]

\[ \text{ET} = \text{evapotranspiration} \]

\[ R_e = \text{Recharge} \]

Note that in equation 4-1 and 4-2, the both \( \lambda_d \) and \( \lambda_f \) appear instead of one common parameter as used in the previous studies. The processes that are governed by these equations are discussed in detail in Chapter 1. In order to solve 4-1, and 4-2, \( \lambda_d \) and \( \lambda_f \) should be estimated as the function of \( h \) by incorporating the effect of unsaturated zone flux since \( \text{ET} \) and \( R \) can largely affect their magnitudes [Childs, 1960, Chapman and Dressler, 1984, Chapman, 1995; Tritscher et al., 2000]. As derived in Chapter 1, specific expressions for \( \lambda_f \) and \( \lambda_d \) during \( \text{ET} \) and \( R_e \) are given as

\[
\lambda_d = (\theta_s - \theta_r) \left\{ 1 - \left( \frac{d(\psi_T)}{dh} \right) \left[ (1 + (\alpha \psi_T)^n)^{-\frac{n+1}{n}} \right] \right\} \tag{4-3}
\]

\[
\lambda_f = (\theta_s - \theta_r) \left( \frac{d(\psi_T)}{dh} \right) \left[ 1 - (1 + (\alpha \psi_T)^n)^{-\frac{n+1}{n}} \right] \tag{4-4}
\]

Where

\[ \psi_T = \text{the matric suction at the soil surface} \]

\[ \theta_s = \text{saturated water content profile} \]

\[ \theta_r = \text{residual soil water content of the profile} \]

\[ \alpha \text{ and } n = \text{parameters of the modified van Genuchten [1980] soil moisture retention equation [Troch, 1992; Hilberts et al.,2005]} \]
In equations 4-3 and 4-4, the soil matric suction at the surface is estimated from the water flow equation by using exponential K (ψ) function [Gardner, 1958] and assuming successive steady state vertical flow to and from the water table [Raats, 1974; Zhu and Mohanty, 2004] as shown in Chapter 1.

**Initial and boundary conditions**

**Phase 1 Drainage:** During this phase, the water furrows act as internal boundaries which need to be superimposed on the boundary conditions imposed at the field ditches. Following initial and boundary conditions are specified for the phase-1 drainage.

\[
\begin{align*}
    h &= h_0, \quad t = 0, \quad 0 < x < L \\
    h &= H_1, \quad t \geq 0, \quad x = 0, \quad L \\
    h &= H_2, \quad t \geq 0, \quad x = \text{Furrow Location} \ (L_F) \\
    h &= h_{x,t}, \quad t > 0, \quad 0 < x < L, \quad x \neq L_F
\end{align*}
\]

(4-5)

**Phase 2 Drainage:** Once the water table midpoint between the water furrows reaches the depth of the water furrows, subsurface flow occurs solely to the outer ditches since the furrows are hydraulically disconnected from the water table. At this phase following initial and boundary conditions can be specified

\[
\begin{align*}
    h &= h_0, \quad t = 0, \quad 0 < x < L \\
    h &= H_1, \quad t \geq 0, \quad x = 0, L \\
    h &= h_{x,t}, \quad t > 0, \quad 0 < x < L
\end{align*}
\]

(4-6)

**Irrigation:** The initial and boundary condition for the irrigation are similar to those for the phase-1 drainage (equation 4-5). Once the water supply to the furrows is initiated, water infiltrates vertically towards the WT. Since the width of the furrows is very small as compared to the field width, they can be considered as line sources [Gill, 1984]. After the wetting front
reaches the WT, the soil profile below the furrows is completely saturated and therefore can be considered as narrow vertical ditches from which water seeps laterally. If the WT is deep (e.g., > 2m), it may require a long time for the wetting front to reach the water table and contribute to WT rise. In our study fields, however, the wetting front quickly reaches the phreatic surface due to high saturated hydraulic conductivity of the soil and shallow WT depth. Therefore, it was assumed during the solution that the soil profile below the furrow is saturated immediately after the onset of irrigation.

**Numerical solution**

Equations 4-1 and 4-2 are nonlinear partial differential equation which cannot be solved analytically for many field geometries and numerical methods are generally adopted. In this study we solved the governing equations numerically by implicit finite-difference method for the initial and boundary conditions expressed in equations 4-5 and 4-6. Equation 4-1 can be expressed in finite difference form as

\[
\frac{h_x^{t+1} - h_x^t}{\Delta t} = \frac{K_s}{\lambda_d} \left[ h_{x-1}^t \left( \frac{h_{x-1}^{t+1} - h_{x}^{t+1}}{\Delta x^2} \right) - h_x^t \left( \frac{h_x^{t+1} - h_{x+1}^{t+1}}{\Delta x^2} \right) \right] - \frac{ET_t}{\lambda_d} - R_t \left( \frac{1}{\lambda_d} - \frac{1}{\lambda_f} \right) \tag{4-7}
\]

where, \( t = \) time; \( t = 0, 1, 2, 3, \ldots, N; \) \( x = \) distance: \( x = 0, 1, 2, 3, \ldots, L \)

Setting \( \Gamma = \frac{K \Delta t}{\lambda_d \Delta x^2} \), equation 4-7 can be rearranged to get a system of linear equations with three unknowns, expressed as

\[
A \cdot h_{x-1}^{t+1} + B \cdot h_x^{t+1} + C \cdot h_{x+1}^{t+1} = X_{x,t} \tag{4-8}
\]

where, \( A = -h_x^t; \) \( B = \frac{1}{\gamma} + h_x^t - h_{x-1}^t; \) \( C = h_x^t; \) & \( X = \frac{h_x^t - ET_t - R_t \left( \frac{1}{\lambda_d} - \frac{1}{\lambda_f} \right)}{\gamma} \)

Equation 4-8 was solved by using the Tri-diagonal matrix algorithm using R programming environment [R Development Core Team, 2012] to estimate hourly water table profiles within the field.
Water Table Data and Soil Parameters

Water table monitoring wells each approximately 1.2m deep, were installed at different locations in the field site in 2010. Figure 4-1 shows the location of the wells in the field which are located between the mid-point of the field and one of the ditches. Hourly water table data was recorded from the wells during the spring potato season of 2010 using pressure transducers. In the following year, hourly water table elevations were recorded in field site 1 as well as field site 2, which is approximately of same dimensions (400 m x 400 m). The data recorded by the pressure transducers were compensated to remove the effect of barometric pressure in the observed water table elevations. Water levels in the water furrows and ditches, which provide the boundary conditions for the water flow in the field, were treated as constants during the irrigation and drainage events thus creating constant head-type boundary conditions. The height of the water in the furrows during irrigation and drainage phases normally differed by 5-10cm. The beginnings of irrigation and drainage phases were obtained by recording the time of turning on/off of the water supply in the field.

Hourly rainfalls as well as other weather parameters (solar radiation, temperature, relative humidity, wind speed, and dew point temperature) were obtained from the weather station install adjacent to the study fields [FAWN, 2011]. Evapotranspiration from the potato fields were calculated at hourly time steps from the Penman-Monteith method [Monteith, 1965] using the hourly weather data. To calculate the recharge rate from rainfall data, a parametric relationship was used which is discussed in Chapter 1. All the parameter required for the simulations, along with the procedures of their estimation, are presented in Chapter 1.

Simulation of water table dynamics during irrigation and drainage events were performed on hourly time scales for 50 days period during the mid-growth season of potatoes in 2010. The simulated water table elevations at the monitoring wells were then compared to the field water
table data. In 2011 potato season, the model was tested against the two independent datasets from field 1 as well as field 2. The two statistical measures, Root Mean Square Error (RMSE) and *Nash-Sutcliffe* Coefficient of efficiency (*C*$_{eff}$) [Nash and Sutcliffe, 1972] were used to assess the performance of the simulation.

Once the performance of the WT dynamics simulation model was evaluated from the two year potato season data, the sensitivity of the model to few key parameters and alternative irrigation regimes was studied. One of the major drawbacks of the seepage irrigation system is that it is highly inefficient in terms of water use since it requires continuous pumping of water to the furrows during irrigation. Therefore, a water-saving irrigation regime, with only 12 hour water supply (7:00 am-7:00 pm) was formulated the resultant water table dynamics was studied with respect to the dynamics observed under conventional irrigation regime.

**Results and Discussions**

**Model Application**

The statistical measures used to compare the simulated and observed water table elevations during the 50 day simulation periods in 2010 and 2011 are presented in Table 4-1. Figure 4-4 shows the water table depths against time (hours) in Well-1 during two drainage/rainfall events and two irrigation events in 2010. Note that there is a significant difference in the water table movement rates during irrigation and drainage phases which are due to the effect of ET which affects drainable and fillable porosity resulting in a hysteretic behavior of the two storage parameters (Chapter 2), which eventually causes significant discrepancy in water table movement. During the irrigation phase, even though the water level in the furrows (internal boundary condition) was normally maintained around 40 cm, the WT fluctuated diurnally at 50-60cm depth showing the effect of direct ET from WT. The depth of the root-zone of potato crop in the area is reported to be approximately 36cm [Munoz et al., 2006]. These
observations therefore indicate that the conventional system is effective in consistently
maintaining WT close the bottom of the root-zone thus avoiding any moisture stress during
irrigation.

Simulated rates of WT rise due to rainfall as well as the subsequent recession rates
matched closely with the observed data in all three locations. The best agreement between the
observed and simulated water table dynamics over the 50 day period was found in Well-1 which
was located at the center of the field. The average $C_{eff}$ and RMSE (cm) of the simulations in 2010
were 0.56 and 0.51 respectively with Well-3 showing the lowest $C_{eff}$. This suggests that the
model was not able to capture the water table dynamics as well in the periphery of the field as it
did near the center. One of the primary reasons for this is the underlying assumption of the model
itself, where exclusively horizontal flow is assumed following the D-F theory. Near the ditches,
the streamlines tend to converge instead remaining parallel [van Schilfgaarde, 1974; Ritzema,
1994] which can overestimate the WT movement. Figure 4-5 shows the scatter plots of observed
of simulated WT depth in the three wells in 2010 which corroborate the results of Table 4-1. The
disagreement in observed and simulated WT depth occurred mainly during heavy rainfall events.
The model overestimated the WT drawdown especially after larger rainfall events potentially
because of the effect of surface storage. The surface storage can infiltrate towards the WT after
the rainfall ceases thus slowing down the drawdown process. Such slower drawdown rates can
also be detected in Figure 4-4 during the two precipitation events.

**Model Evaluation and Sensitivity Analysis**

The performance of the WT dynamics model was also tested against two independent WT
datasets collected from field-1 and field-2 during the 2011 potato season. Figure 4-6 shows the
plots of observed versus simulated water table depths in field-2 during the 50 day period in 2011.
The statistical measures of comparison between the simulated and observed data for both fields
are presented in Table 4-1, which strongly support the results from 2010 potato season. The average $C_{eff}$ and RMSE (cm) of simulations in 2011 were 0.55 and 0.85 for field-1, and 0.71 and 0.56 respectively. These comparisons suggest that the numerical model developed in this study reliably described the water flow processes that govern the drainage and irrigation in the field sites.

Although the numerical Boussinesq model presented in our study was able to simulate hourly water table dynamics during both irrigation as well as drainage phases in the field sites, it is critical to understand the sensitivity of the model outcomes to various parameters. One of the most critical parameters that determine the rate of WT rise and drawdown in the field is the saturated hydraulic conductivity ($K_s$). Figure 4-7 presents the effect of different $K_s$ rates on WT drawdown and rise after the onset of the drainage and irrigation respectively. Water table movement highly depends on the $K_s$ of the soil since this parameter determines the rate of soil water flow. In our particular fields, when the $K_s$ value increased from 2 to 7 cm/hr, the time required to bring the WT, initially at 20cm depth, below 40cm decreased dramatically to < 24 hours form approximately 50 hours. Similarly, when the $K_s$ value increased to 15 cm/hr the drawdown time further decreased approximately to 12 hours. Water table rise during irrigation event, like the drawdown phase, also differed significantly when the $K_s$ values changed. The time required for the WT, initially at 80cm below the surface, to reach near the bottom of the root-zone (40cm) was <24 hours for $K_s = 15$cm/hr while it was $\approx$100 hours for $K_s = 7.0$ cm/hr. The WT hydrograph observed in the field after the onset of the irrigation events also showed the same pattern, taking approximately 100 hours to reach the maximum WT elevation where it fluctuated diurnally due to the influence of ET. The maximum WT elevation reached during the irrigation event was also higher with greater $K_s$ values. These results can have an important
implication in irrigation scheduling under conventional WT management system in Florida since
the irrigation onset times can be optimized based on the anticipated WT movement rates.

In addition to the $K_s$ of the soil, a critical component of subsurface drainage system is the
depth and spacing of drains. In most parts of the Florida including our field sites, narrow water
furrows are created approximately at 18 m distance between the main lateral ditches which
facilitate both irrigation and drainage. Therefore, it is important to understand their contribution
to water table drawdown. Figure 4-8 presents the effect of furrow depth on the water table
recession below the potato crop root-zone in a field that resembles the field sites used in this
study. Note that increasing the drain depth just by 10 cm can significantly reduce the time
required to bring WT below the root-zone. When the furrow depth is increased to 65cm from
45cm, which is the normal depth of the furrows practiced in northeast Florida, time required to
bring water table below 40cm was reduced by more than 20 hours. Also note that after the water
table passes the depth of the furrow, the drawdown curve shows a nearly linear relationship to
time suggesting that the contribution of ditches to water table drawdown is small compared to
the furrows in these fields. The recession of WT after it passes the depth of the furrows therefore
mostly might have occurred due to the direct ET from the phreatic surface [e.g., Loheide et al.,
2005; Shah et al., 2007]. This is mainly because the large distance between the ditches (≈ 400 m)
which attenuates the effect of the ditch water level on subsurface lateral flow towards the field
edges.

A critical limitation of continuously maintaining the WT near the root-zone is that it
requires continuous pumping of water, a substantial portion of which is lost to the ditches via
runoff from furrow. Due to diminishing groundwater resources, it is necessary to optimize the
water management system to reduce water loss and increase the efficiency. A potential
alternative to the conventional system would be an intermittent system in which water is supplied only during the day time instead of the night time. It is important to know how the water table responds to the intermittent water supply system in order to understand the potential impact on crop growth. Figure 4-9 shows the simulated water table dynamics under a hypothetical intermittent irrigation system as compared to the observed dynamics during two irrigation events in 2010 potato season. Note that under the intermittent system, the diurnal pattern of the water table is still similar to the continuous system while the maximum elevation of the water table is decreased by as high as >15cm. This suggests that the intermittent system cannot maintain the water table as close to the root-zone as the continuous system does, which may potentially induce moisture stress to the root near the soil surface. However, for crops with relatively deeper root system, this system may be effective enough. Since the intermittent system save 50% of water application, its potential as an alternative water management system can be justified. However, extensive research on the root-zone soil moisture dynamics is necessary to understand the feasibility of such water management regimes in the area.

**Conclusions**

In this study we used numerical technique to solve the one dimensional Boussinesq equation to model the shallow water table dynamics in unconfined aquifers and applied to in a subsurface irrigation and drainage system typically practiced in Florida. Our model was developed by implementing two distinct drainable and fillable porosity parameters which respectively control the water table drawdown and rise in unconfined aquifers. This approach is significantly different from all the previous studies in which only a single drainable porosity was used. Our model was able to simulate the recession and rise of water table during respective drainage and irrigation events in northeast Florida. Direct ET loss from the phreatic surface, which commonly occurs in shallow water table environment, was captured very well by the numerical model.
Application of the model to the field sites in northeast Florida also suggested that, in addition to the soil hydraulic properties, the water table movement during irrigation and drainage is mostly dictated by the depth of the water furrows and ET while the effect of ditches at the field boundaries was comparatively negligible. This was further supported by the simulated water table responses at varying depths of the water furrows. Simulated water table dynamics during a water-saving, potential alternative to the conventional water table management revealed that the water table remained at approximately 15cm lower depth as compared to the current system. Although this alternative saves 50% of water pumping to the field, the possibility of lower water table elevations may induce moisture stress to the plant roots at shallow depths. Nonetheless, for relatively deeper rooted crops, the results indicate the feasibility of implementing such intermittent system in the area.

Subsurface irrigation and drainage system are necessary in shallow water table environments to facilitate crop growth and field traffic while they can also reduce the pollutant loads to surface waters. Numerical models not only help understand various soil water phenomena in the field but they also help in testing various alternative management scenarios that may be potentially adopted to improve the efficacy of current management systems. Since majority of the agricultural lands in Florida are under subsurface irrigation and drainage system, the numerical model presented in this study can be helpful in improving efficiency of the water management system as well as reduce intensive groundwater withdrawal in the region through optimization of irrigation scheduling.
Table 4-1. Comparison measures: Root Mean Square Error (RMSE) and *Nash-Sutcliffe*, Coefficient of Efficiency of water table dynamics simulation for the three wells in site-1 in 2010 and 2011.

<table>
<thead>
<tr>
<th>Distance From the Main Ditch</th>
<th>Ceff</th>
<th>RMSE(cm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Site 1</td>
<td></td>
<td></td>
</tr>
<tr>
<td>2010, Field 1</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Well 1</td>
<td>200</td>
<td>0.80</td>
</tr>
<tr>
<td>Well 2</td>
<td>125</td>
<td>0.75</td>
</tr>
<tr>
<td>Well 3</td>
<td>13</td>
<td>0.12</td>
</tr>
<tr>
<td>2011, Field 1</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Well 1</td>
<td>200</td>
<td>0.75</td>
</tr>
<tr>
<td>Well 2</td>
<td>125</td>
<td>0.63</td>
</tr>
<tr>
<td>Well 3</td>
<td>13</td>
<td>0.28</td>
</tr>
<tr>
<td>2011, Field 2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Well 1</td>
<td>200</td>
<td>0.78</td>
</tr>
<tr>
<td>Well 2</td>
<td>125</td>
<td>0.76</td>
</tr>
<tr>
<td>Well 3</td>
<td>13</td>
<td>0.59</td>
</tr>
</tbody>
</table>
Figure 4-1. Land use in the St. Johns, Putnam and Flagler counties in northeast Florida and a field site, planted to potato, chosen for the study.
Figure 4-2. Conceptual representation of water table profiles during irrigation phase. Note that only 2 water furrows are depicted here; the actual number of furrows in the fields may be 10-30.

Figure 4-3. Conceptual representation of water table profile during the early drainage phase (top) and late drainage phase (bottom). The dashed line represents the initial water table elevation after a precipitation event.
Figure 4-4. Observed (filled circles) and simulated (solid lines) water table rise due to rainfall (vertical bars) and subsequent drawdown (top), and water table rise after onset of irrigation (bottom) in Well-1 located approximately at the center of the field site-1 in 2010 potato season.
Figure 4-5. Scatterplots of the observed and simulated hourly water table depths in the three wells in 2010. The red line indicated 1:1 relationship.

Figure 4-6. Scatterplots of observed and simulated hourly water table depths in field site 2 in 2011 potato season. The red line indicates 1:1 relationship.
Figure 4-7. Effect of $K_s$ on the time required to lower the water table below 50 cm depth from an initial WT depth of 10 cm (left); and time required to raise the WT to 50 cm from an initial depth of 80 cm (right), in a field of 15 ha area (400m x 400 m).

Figure 4-8. Effect of furrow depths on the water table drawdown rate during subsurface drainage.
Figure 4-9. Hourly water table depths during irrigation events of an alternative, 50% water saving irrigation regime (12 hour irrigation/drainage cycle).
CHAPTER 5
SUMMARY AND CONCLUSIONS

In Chapter 2, a method to incorporate effects of unsaturated zone moisture fluxes while estimating $\lambda_d$ and $\lambda_f$ was developed. By assuming successive steady state ET and $R_e$ fluxes, closed-form expressions for $\lambda_d$ and $\lambda_f$ were derived. The major advantage of the method is that the dynamic behavior of soil water is accounted for in the estimation of storage parameters thus relaxing the limitations of previously available methods that relied on a hydrostatic assumption to incorporate the effect of soil moisture retention. Most important advancement, however, is the ability to estimate and $\lambda_d$ and $\lambda_f$ separately for WT drawdown and rise scenarios. With the help of a hydraulic groundwater model implemented with flux-dependent $\lambda_d$ and $\lambda_f$, it is shown that the simulation of WT dynamics can be strongly improved. Our study therefore has highlighted the importance of using fillable porosity which can be critical in modeling WT dynamics in land drainage systems [e.g. Skaggs 1980], hillslopes [e.g., Troch et al., 2003; Hilberts et al., 2004, 2005, 2007] and wetlands [e.g., Laio et al., 2009; Pumo et al., 2010].

In Chapter 3, an application of the two analytical expressions $\lambda_d$ and $\lambda_f$ developed in Chapter 2 to $ET_g$ from observed diurnal water table fluctuations observed is presented. In doing so, the original method developed by White [1932] and later modified by others [e.g., Gribovszki et al., 2007], was further modified to account for the effect of precipitation in observed WT fluctuations. This modification therefore allowed for the reasonable application of the White method even during rainy days. Our study also showed that the $\lambda_d$ estimated from hydrostatic soil moisture profile normally failed to appropriately attribute the WT fluctuations to the groundwater ET, producing significant overestimations. However, when the effect of ET is incorporated into the $\lambda_d$, its value becomes smaller than hydrostatic-$\lambda_d$ since this non-equilibrium parameter also accounts for the soil moisture that is being lost from the profile via
ET, in addition to the water lost by drainage. These smaller \( \lambda_d \) values therefore resulted in more reliable estimates of \( ET_g \) at hourly as well as daily time scales. The results from our study further accentuate the critical role of \( \lambda_d \) and \( \lambda_f \) in White based \( ET_g \) estimation methods while providing a general framework of applying the method continuously regardless of the precipitation events using these two parameters. Although our conclusions were based on relatively short term results from a managed crop-field, close agreement between the results from two consecutive years suggest that the method is consistent and can be equally applied to other field conditions.

In Chapter 4, a numerical model of WT fluctuation under a subsurface drainage and irrigation system is developed by numerically solving the 1D Boussinesq equations. The governing equations resulted by implementing separate \( \lambda_d \) and \( \lambda_f \) as shown in Chapter 2 were applied in the study. The model was then applied to a subsurface irrigation and drainage system typically practiced in northeast Florida. The model simulated the drawdown and rise of water table during respective drainage and irrigation events in northeast Florida. Direct ET loss from the phreatic surface, which commonly occurs in shallow water table environment, was captured very well by the numerical model. Application of the model to the field sites in northeast Florida also suggested that, in addition to the soil hydraulic properties, the water table movement during irrigation and drainage is mostly dictated by the depth of the water furrows and ET while the effect of ditches at the field boundaries was comparatively negligible. Subsurface irrigation and drainage system are necessary in shallow WT environments like northeast Florida to facilitate crop growth and field traffic while they can also reduce the pollutant loads to surface waters. Numerical models not only help understand various soil water phenomena in the field but they also help in testing various alternative management scenarios that may be potentially adopted to improve the efficacy of current management systems. Therefore, the model presented in this
study can be potentially helpful in improving efficiency of the water management system as well as reduce intensive groundwater withdrawal in the region through optimization of irrigation scheduling.

The dynamics of shallow WT in unconfined aquifers largely control the hydrology and vegetation growth development in natural areas while it presents a great challenge to efficient water management in agricultural lands. Such dynamic behavior is in turn controlled by the physical characteristics of the aquifer material as well as the vegetation. Drainable and fillable porosity are the two primary aquifer parameters that control the WT fluctuation in shallow phreatic aquifers. They are critical in modeling groundwater flow in unconfined aquifers and not only determine the water table dynamics but also influence the subsurface discharge at the aquifer outlet. This research primarily focused on investigating the influence of unsaturated zone flux above the WT on the behavior of $\lambda_d$ and $\lambda_f$ which was not done until now. The results revealed that their relationship with the WT is manifested differently under hydrodynamic conditions. These results thus paved a way to formulate more robust expressions of the hydraulic groundwater model that can be applied in agricultural WT management systems, shallow hillslope subsurface flow, or WT fluctuations in wetlands and riparian zones. Furthermore, the results enabled the investigation of close interactions between shallow WT and plant ET through the development of more detailed and reliable calculation methods. Collectively, these improvements aid to better understanding of the complex dynamics of soil-plant-water interactions in shallow WT environments which are present all around the globe and help solve more problems in the future.
seepageModel <- function(pars) {
  N = constants$Time
  RL = constants$Z.eq
  delt = constants$delt
  delz = constants$delz
  z = seq(0, pars$zmax, delz)
  L = length(z)
  Bed.Len = pars$zmax

  #### Dupuit Forcheimmer Flux Function
  flux.DF <- function(wt, ditch.wl) {
    return(pars$Ksat * wt * (wt - ditch.wl)/delz)
  }

  ### Determine number of Furrow/Drains
  Bed.num <- function(pars, constants) {
    Rd1.wd = constants$Rd1.wd
    Rd2.wd = constants$Rd2.wd
    Bed.wd = pars$Bed.wd
    Fur.wd = pars$Fur.wd
    Fur.no <- (pars$zmax - (Rd1.wd + Rd2.wd) + pars$Bed.wd)/(pars$Bed.wd + pars$Fur.wd)
    Bed.no <- as.integer(Fur.no - 1)
    return(Bed.no)
  }

  #### Determine Furrow locations
  FurLoc <- function(pars, constants) {
    Fur.no = Bed.num(pars, constants) + 1
    xx <- rep(pars$Bed.wd/delz, Fur.no)
    xy <- 1:Fur.no - 1
    Ist.fur = constants$Rd1.wd/delz + 1
    xxz <- xx * xy
    zz <- Ist.fur + xxz
    yy <- rep(pars$Fur.wd/delz, length(xxz))
    fur <- recur(xx, yy)
    Furrow.Loc = fur + xy
    return(Furrow.Loc)
  }

  ####### Drainable Porosity under ET
  sdFor.ET <- function(pars, constants, Ht, ET) {
    e = min(ET, ET * exp(-pars$b * (pars$Td - Ht)))
    qv = e
    psiT <- function(Ht, qv) {
      return(1/pars$a * log(((qv + pars$Ksat) * exp(pars$a * (Ht - RL)) - qv)/pars$Ksat))
    }
    A <- function(Ht, qv) {
      return((qv + pars$Ksat) * exp(pars$a * (Ht - RL))/(qv + pars$Ksat)*
      exp(pars$a * (Ht - RL)) - qv))
    }
}
sdP <- function(psiTop, gama, pars) {
    return(pmax((pars$WCs - pars$WCr) * (1 - gama * (1 + (-pars$alpha * psiTop)^pars$n)^(-(1 + 1/pars$n))), 0))
}

dpor = sdP(psiT(Ht, qv), A(Ht, qv), pars)
return(dpor)
}

dPor.ET <- cmpfun(sdPor.ET)

### Fillable Porosity Under ET
sfPor.ET <- function(pars, constants, Ht, ET) {
    e = min(ET, ET * exp(-pars$b * (pars$Td - Ht)))
    qv = e
    psiT <- function(Ht, qv) {
        return(1/pars$a * log(((qv + pars$Ksat) * exp(pars$a * (Ht - RL)) - qv)/pars$Ksat))
    }
    A <- function(Ht, qv) {
        return((qv + pars$Ksat) * exp(pars$a * (Ht - RL))/((qv + pars$Ksat) * exp(pars$a * (Ht - RL)) - qv))
    }
    sfPor <- function(psiTop, gama, pars) {
        fill.por = (pars$WCs - pars$WCr) * gama * (1 - (1 + (-pars$alpha * psiTop)^pars$n)^(-(1 + 1/pars$n)))
        return(fill.por)
    }
    fpor = sfPor(psiT(Ht, qv), A(Ht, qv), pars)
    return(fpor)
}

fPor.ET <- cmpfun(sfPor.ET)

### Drainable Porosity under Rainfall/Recharge
sdPor.Rf <- function(pars, constants, Ht, RF) {
    qv = min(RF, RF * exp(-pars$br * (pars$Tr - Ht)))
    psiT <- function(Ht, qv) {
        return(1/pars$a * log(((qv + pars$Ksat) * exp(pars$a * (Ht - RL)) - qv)/pars$Ksat))
    }
    A <- function(Ht, qv) {
        return((qv + pars$Ksat) * exp(pars$a * (Ht - RL))/((qv + pars$Ksat) * exp(pars$a * (Ht - RL)) - qv))
    }
    sdP <- function(psiTop, gama, pars) {
        return(pmax((pars$WCs - pars$WCr) * (1 - gama * (1 + (-pars$alpha * psiTop)^pars$n)^(-(1 + 1/pars$n))), 0))
    }
\[ dpor = sdP(\psiT(Ht, -qv), A(Ht, -qv), pars) \]

\[ return(dpor) \]

\[ dPor.Rf <- cmpfun(sdPor.Rf) \]

#### Fillable Porosity under Rainfall/Recharge

\[ sfPor.Rf <- function(pars, constants, Ht, Rf) \]

\[ qv = min(Rf, Rf * \exp(-pars$br * (pars$Tr - Ht))) \]

\[ psiT <- function(Ht, qv) \]

\[ return(1/pars$a * \log(((qv + pars$Ksat) * \exp(pars$a * (Ht - RL)) - qv)/pars$Ksat)) \]

\[ A <- function(Ht, qv) \]

\[ return((qv + pars$Ksat) * \exp(pars$a * (Ht - RL))/((qv + pars$Ksat) * \exp(pars$a * (Ht - RL)) - qv)) \]

\[ sfpor = function(psiTop, gama, pars) \]

\[ fill.por = (pars$WCs - pars$WCr) * (gama) * (1 - (1 + (-pars$alpha * psiTop)^pars$n)^(-1 + 1/pars$n))) \]

\[ return(fill.por) \]

\[ fpor = sfpor(psiT(Ht, -qv), A(Ht, -qv), pars) \]

\[ return(fpor) \]

\[ fPor.Rf <- cmpfun(sfPor.Rf) \]

#### Thomas Algorithm For Tridiagonal Matrix

\[ Tridia <- function(A, B, C, R) \]

\[ A = as.vector(A) \]  # First Diag
\[ B = as.vector(B) \]  # True Diagonal
\[ C = as.vector(C) \]  # 3rd Diag
\[ R = as.vector(R) \]
\[ nn = length(A) \]
\[ BETA <- rep(0, nn) \]
\[ GAMMA <- rep(0, nn) \]

\[ for (i in 2:nn) \]

\[ BETA[i] = C[i]/(B[i] - BETA[i - 1] * A[i]) \]

\[ Y <- as.vector(rep(0, nn)) \]
\[ Y[nn] = GAMMA[nn] \]

\[ for (j in 1:(nn - 1)) \]

return(Y)
}

Tridiag <- cmpfun(Tridia)

#### Coefficient Vector Functions: Second Phase Drainage (Ditch Only Scenario)
AA.Drain <- function(Ht) {
  A <- as.vector(rep(0, length(Ht)))
  nn = length(A)
  for (i in 2:(nn - 1)) {
    A[i] = -Ht[i - 1]
    A[1] = 0
    A[nn] = 0
  }
  return(A)
}

AA.D <- cmpfun(AA.Drain)

####
BB.Drain <- function(lamda, Ht) {
  B <- as.vector(rep(0, length(Ht)))
  nn = length(B)
  for (i in 2:(nn - 1)) {
    B[i] = 1/lamda[i] + Ht[i - 1] + Ht[i]
  }
  B[1] = 1
  B[nn] = 1
  return(B)
}

BB.D <- cmpfun(BB.Drain)

####
CC.Drain <- function(Ht) {
  C <- as.vector(rep(0, length(Ht)))
  nn = length(C)
  for (i in 2:(nn - 1)) {
    C[i] = -Ht[i]
  }
  C[1] = 0
  C[nn] = 0
  return(C)
}

CC.D <- cmpfun(CC.Drain)

####
RR.Drain <- function(pars, BC1.D, BC2.D, Ht, dp, fp, lamda, ET, Rf) {
  e = ET
  Re = min(Rf, Rf * exp(-pars$br * (pars$Tr - Ht)))
  R <- as.vector(rep(0, length(Ht)))
  nn = length(R)
  for (i in 1:nn - 1) {
    R[i] = (Ht[i] - delt * e/dp[i] + delt * Re * (1/fp[i] - 1/dp[i]))/lamda[i]
  }
  R[1] = BC1.D
  R[nn] = BC2.D
return(R)

RR.D <- cmpfun(RR.Drain)

#### Coefficient Vectors: First Phase Drainage: Ditches and Furrow/drains
AA.Drain2 <- function(Ht) {
    fur.loc <- FurLoc(pars, constants)
    A <- as.vector(rep(0, length(Ht)))
    nn = length(A)
    for (i in 2:(nn - 1)) {
        A[i] = -Ht[i - 1]
        A[1] = 0
        A[fur.loc] = 0
        A[nn] = 0
    }
    return(A)
}

AA.D2 <- cmpfun(AA.Drain2)

####
BB.Drain2 <- function(lamda, Ht) {
    fur.loc <- FurLoc(pars, constants)
    B <- as.vector(rep(0, length(Ht)))
    nn = length(B)
    for (i in 2:(nn - 1)) {
        B[i] = 1/lamda[i] + Ht[i - 1] + Ht[i]
    }
    B[fur.loc] = 1
    B[1] = 1
    B[nn] = 1
    return(B)
}

BB.D2 <- cmpfun(BB.Drain2)

####
CC.Drain2 <- function(Ht) {
    fur.loc <- FurLoc(pars, constants)
    C <- as.vector(rep(0, length(Ht)))
    nn = length(C)
    for (i in 2:(nn - 1)) {
        C[i] = -Ht[i]
    }
    C[fur.loc] = 0
    C[1] = 0
    C[nn] = 0
    return(C)
}

CC.D2 <- cmpfun(CC.Drain2)

####
    e = ET
    Re = min(Rf, Rf * exp(-pars$br * (pars$Tr - Ht)))
    fur.loc <- FurLoc(pars, constants)
    R <- as.vector(rep(0, length(Ht)))
}
nn = length(R)
for (i in 2:nn - 1) {
    R[i] = (Ht[i] - delt * e/dp[i] + delt * Re * (1/fp[i] - 1/dp[i]))/lamda[i]
}
R[1] = BC1.D
R[nn] = BC2.D
R[fur.loc] = BC3.D
return(R)
}

RR.D2 <- cmpfun(RR.Drain2)

#### Coefficient Vector Functions: Irrigation Case: Same as Ist Phase Drainage

AA.Irrig <- function(Ht) {
    fur.loc <- FurLoc(pars, constants)
    A <- as.vector(rep(0, length(Ht)))
    nn = length(A)
    for (i in 2:(nn - 1)) {
        A[i] = -Ht[i - 1]
        A[1] = 0
        A[fur.loc] = 0
        A[nn] = 0
    }
    return(A)
}

AA.I <- cmpfun(AA.Irrig)

####

BB.Irrig <- function(lamda, Ht) {
    fur.loc <- FurLoc(pars, constants)
    B <- as.vector(rep(0, length(Ht)))
    nn = length(B)
    for (i in 2:(nn - 1)) {
        B[i] = 1/lamda[i] + Ht[i - 1] + Ht[i]
    }
    B[1] = 1
    B[nn] = 1
    B[fur.loc] = 1
    return(B)
}

BB.I <- cmpfun(BB.Irrig)

####

CC.Irrig <- function(Ht) {
    fur.loc <- FurLoc(pars, constants)
    C <- as.vector(rep(0, length(Ht)))
    nn = length(C)
    for (i in 2:(nn - 1)) {
        C[i] = -Ht[i]
    }
    C[1] = 0
    C[nn] = 0
    C[fur.loc] = 0
    return(C)
}
### Subroutine for Irrigation Phase

Iterate.Irrig <- function(pars, constants, BC1.I, BC2.I, BC3.I, OldtHt, ET, Rf) {
  OldmHt = OldtHt
  if (Rf > 0 && Rf > ET) {
    fp = fPor.Rf(pars, constants, OldmHt, (Rf - ET))
    dp = dPor.Rf(pars, constants, OldmHt, (Rf - ET))
  } else {
    fp = fPor.ET(pars, constants, OldmHt, (ET - Rf))
    dp = dPor.ET(pars, constants, OldmHt, (ET - Rf))
  }

  lamda <- pars$Ksat * delt / (fp * delz^2)
  A <- AA.I(OldmHt)
  B <- BB.I(lamda, OldmHt)
  C <- CC.I(OldmHt)
  R <- RR.I(pars, BC1.I, BC2.I, BC3.I, OldtHt, dp, fp, lamda, ET, Rf)

  NewmHt <- Tridiag(A, B, C, R)
  return(NewmHt)
}

### Subroutine for Drainage

  OldmHt = OldtHt

  fp = fPor.Rf(pars, constants, OldmHt, (Rf - ET))
  dp = dPor.Rf(pars, constants, OldmHt, (Rf - ET))
  if (Rf == 0 && ET == 0) {
    fp = fPor.ET(pars, constants, OldmHt, ET)
    dp = dPor.ET(pars, constants, OldmHt, ET)
  } else if (Rf > 0) {
    fp = fPor.Rf(pars, constants, OldmHt, (Rf))
    dp = dPor.Rf(pars, constants, OldmHt, (Rf))
  } else {
    fp = fPor.ET(pars, constants, OldmHt, (ET))
  }
}

CC.I <- cmpfun(CC.Irrig)

### Subroutine for Drainage

RR.Irrig <- function(pars, BC1.I, BC2.I, BC3.I, Ht, dp, fp, lamda, ET, Rf) {
  e = ET
  Re = min(Rf, Rf * exp(-pars$br * (pars$Tr - Ht)))
  fur.loc <- FurLoc(pars, constants)
  R <- as.vector(rep(0, length(Ht)))
  for (i in 1:(length(R) - 1)) {
    R[i] = (Ht[i] - delt * e/dp[i] + delt * Re/fp[i])/lamda[i]
  }
  R[1] = BC1.I
  R[length(R)] = BC2.I
  R[fur.loc] = BC3.I

  return(R)
}

RR.I <- cmpfun(RR.Irrig)
Define Initial Conditions

# Initialize variables
lamda <- pars$Ksat * delt/(dp * delz^2)

mHt = OldmHt[(L + 1)/2]

if (mHt > BC3.D) {
  A <- AA.D2(OldmHt)
  B <- BB.D2(lamda, OldmHt)
  C <- CC.D2(OldmHt)
} else {
  A <- AA.D(OldmHt)
  B <- BB.D(lamda, OldmHt)
  C <- CC.D(OldmHt)
  R <- RR.D(pars, BC1.D, BC2.D, OldtHt, dp, fp, lamda, ET, Rf)
}

NewmHt <- Tridiag(A, B, C, R)

return(NewmHt)

Iter.Drain <- cmpfun(Iterate.Drain)

### Define Initial Conditions
Ht.Ini = RL - constants$Wt.Ini
Ht.data <- matrix(0, N, L)
Ht.data[1,] = Ht.Ini

BC1.D = RL - (pars$dw1 - rowht)
BC2.D = RL - (pars$dw1 - rowht)
BC3.D = RL - (pars$BC3.D - rowht)

BC1.I = RL - (pars$dw1 - rowht)
BC2.I = RL - (pars$dw1 - rowht)
BC3.I = RL - (pars$BC3.I - rowht)

subsur.flux <- rep(0, N)

pb <- tkProgressBar(title = "progress bar", min = 0, max = N, width = 300)

for (i in 2:N) {
  if (Irrig.Data$Status[i] == 0) {
    Ht.data[i, 1] = BC1.D
    Ht.data[i, L] = BC2.D
    BC3.D = BC3.D
    Ht.data[i, ] <- Iter.Drain(pars, constants, BC1.D, BC2.D, BC3.D, OldtHt = Ht.data[i - 1, ], ET[i - 1], Rf[i - 1])
    Ht.data[i, ] <- ifelse(Ht.data[i, ] >= (RL - 20), (RL - 20), Ht.data[i, ])
    subsur.flux[i] = flux.DF(Ht.data[i, L - 1], (BC1.D))
  }

  if (Irrig.Data$Status[i] == 1) {
    Ht.data[i, 1] = BC1.I
    Ht.data[i, L] = BC2.I
    BC3.I = BC3.I
    Ht.data[i, ] <- Iter.Irrig(pars, constants, BC1.I, BC2.I, BC3.I, Ht.data[i - 1, ], ET[i - 1], Rf[i - 1])
    subsur.flux[i] = flux.DF(Ht.data[i, L - 1], (BC1.I))
  }
}

125
} 

print(i) 

setTkProgressBar(pb, i, title = paste(round(i/N * 100, 0), "% Progress")) 

} 

close(pb) 

Wt.data <- RL - Ht.data 
# write.table(Wt.data, file = 'Wtdata.out', sep = ',', col.names = NA, 
# qmethod = 'double') ## For Writing the simulation result to a output 
# file 
# return(Wt.data[N, (L+1)/2]) ##For mid point WT elevation 

return(list(Wt.data = Wt.data, slf = subsur.flux)) 

}
LIST OF REFERENCES


BIOGRAPHICAL SKETCH

Subodh Acharya was born in 1980 in Nepal, in a small village called Shankhu, approximately 36 kilometers from the capital city Kathmandu. He earned his bachelor’s degree in Agriculture from the Tribhuvan University, Nepal. In fall 2006, he joined the graduate program at the University of Florida, Soil and Water Science Department (SWSD) as a master’s student. After earning his master’s degree in soil and water science in summer 2008, he joined the PhD program in the SWSD department in fall 2008 and received the Gradute Alumni Fellowship to support his study. This thesis is the culmination of the studies he conducted during 2008-2011 as part of his PhD dissertation research. His primary research interests include the study of soil-water-plant interactions in areas with shallow water table through field as well as modeling based approaches. He is also highly interested in the surface and subsurface hydrology, especially in shallow water table areas, both natural and managed. In May 2012, he successfully defended his dissertation and earned the doctorate degree in soil and water science in August 2012.