MODELLING TRENDS IN EVAPOTRANSPIRATION USING THE MODIS LAI FOR SELECTED EASTERN CAPE CATCHMENTS

BY

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Submitted in fulfillment of the requirements for the degree of Masters in Science at the Nelson Mandela Metropolitan University

December 2011

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DECLARATION

I, Andiswa Finca (209202653), hereby declare that the treatise/ dissertation/ thesis for Masters of Science is my own work and that it has not previously been submitted for assessment or completion of any postgraduate qualification to another University or for another qualification.

Andiswa Finca
ACKNOWLEDGEMENTS

Every journey has its own ups and down and this was no different, and it wouldn’t have been possible without the guidance, encouragements and prayers of a number of important people. Firstly I would like to thank God for being my constant help and my strength through my academic journey and for the people He has brought my way. My sincere gratitude goes to Dr. Palmer, who was not only my supervisor but also my mentor and sometimes even a parent to me. Doc thank you for your guidance, assistance, encouragement and patience, you have given me the push I needed to complete this study. I would also like to thank my second supervisor Prof. Kakembo for affording me the opportunity to do my M.Sc with your department and for constantly assuring me that I am on the right track.

Thanks are also due to the Agricultural Research Council (ARC); Red Meat Research and Development-SA; National Research Fund; and NMMU’s Research Capacity Development for funding my studies.

I would also like to thank my colleagues, Mrs. Verwey and Mr. Hintsa and an ex-colleague Nokubonga Mgqatsa for their assistance with field work. Then I would like to thank my family and friends for being there for me and for believing in me.
Grassland is the dominant vegetation cover of many of the 19 Water Catchment Areas within South Africa. The inappropriate management of some of these grassland catchments by the communities that depend on them for their livelihoods, often results in overgrazed lands with low biomass or invasive alien species. The short grass maintained by grazing policies of many communities results in high storm flows that have an adverse effect on the quantity and quality of runoff and recharge. Catchment-scale water balances depend on accurate estimates of run-off, recharge and evapotranspiration (ET). This study focuses on the ET component of the catchment scale water balance and explores the effect of two different grazing strategies on ET. To achieve this, two contrasting but adjacent quaternary catchments namely: P10A (a high biomass site) and Q91C (a low biomass site) were selected within the Bushman’s River Primary catchment as primary study sites. Within each catchment, a relatively homogenous pixel of 1 km was selected, representing contrasting example of high and low intensity grazing. From an eleven year MODIS leaf area index (LAI) data stack (March 2000 – 2010), 8-day LAI values was extracted for each pixel in each catchment. Using the Penman-Monteith equation, potential evapotranspiration (ET\textsubscript{0}) was calculated using data from a nearly automatic weather station. Actual evapotranspiration was estimated by adjusting ET\textsubscript{0} using the values extracted from the MODIS LAI product. The MODIS LAI ET (ET\textsubscript{MODIS}) obtained for the eleven year period for both 1 km pixels decreased consistently, reflecting a general trend in declining LAI throughout the Eastern Cape. The highest ET\textsubscript{MODIS} obtained from P10A was 610.3 mm (2001) and the lowest was 333.1 mm (2009). Then from Q91C the highest ET obtained was 534.7 mm (2006) and the lowest was 266.2 mm (2009). The ET\textsubscript{MODIS} results were validated for each catchment using the Open Top Chamber (OTC) which sums the water lost from vegetation and soil within the chamber. This validation was conducted during the growing season of 2010–11. Wind speed; relative humidity and temperature were measured both at the inlet and the outlet of the chamber on five clear sunny days for each 1 km pixel. ET\textsubscript{a} for the same period was compared to the OTC ET (ET\textsubscript{OTC}) using the regression analysis and a good relationship was observed with the \( r^2 \) of 0.7065. The relationship observed confirmed that ET\textsubscript{OTC} closely approximates ET\textsubscript{MODIS} and that the OTC can be used as a tool to validate MODIS LAI ET on clear, low winds and sunny days.

In order to demonstrate proof-of-concept for the use of this modeling of ET\textsubscript{MODIS} within a Payment for Ecosystem Services framework, the approach was applied to two other quaternary catchments under communal tenure. Within each catchment, three land use
scenarios were created for each catchment to reflect potential changes in the standing above-ground biomass. For Scenario 1, the *status quo* was maintained; for Scenario 2, MODIS pixels representing 28 km in each catchment were selected and the LAI of these pixels was doubled; and for scenario 3, LAI was halved. ET\textsubscript{MODIS} was calculated for each scenario by adjusting the ET\textsubscript{0} data from a nearby automatic weather station with the MODIS LAI product. The results showed that the estimated annual ET\textsubscript{MODIS} obtained from the high biomass catchment was 111 mm greater than that obtained from the low biomass catchment. When comparing between the scenarios, the annual ET\textsubscript{MODIS} obtained from scenario 2 was the highest of the 3 scenarios for both sites. These results confirm that increased leaf area results in higher annual ET\textsubscript{MODIS}. This has a positive long term impact on stream flow, as high grass biomass allows the rainfall to infiltrate the soil and be gradually released to the dams with reduced magnitude of storm flows. This approach has the potential to quantify the benefits to down-stream water users of improving above-ground biomass in catchments.
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CHAPTER 1
INTRODUCTION

1.1. WATER AVAILABILITY IN SOUTH AFRICA

Water is one of the most important sources of life. It is a basic need for all humans, plants and animals and is vital for sustainable social and economic development (Muller 2009). However over the past 100 years population growth coupled with human activities such as intensive agriculture, urbanization and industrialization have increased demand for water, and increased the impact that humans have on water quality. South Africa receives a mean annual rainfall of ~450 mm, which is much less than the world’s average of 860 mm, and is categorized as a water stressed country (Ashton 2002; DWAF 2002; Otieno and Ochieng 2004; Scholes 2001). Moreover the impact of climate changes could lead to an overall rainfall reduction of about 20%, decreasing the mean annual runoff and recharge from several rivers in the country (Mugabi 2010; Turpie, Winkler et al. 2002). The current mean annual runoff is about 49 200 x 10^6 m^3 including the inflow from Lesotho and Swaziland and the usable yield is about 13 911 x 10^6 m^3 (DWAF 2002).

FAO estimates water available to be 1 154 m^3 per capita y^-1 and according to International Water Management Institute (IWMI) by 2025 this could be reduced to less than 1000 m^3 per capita y^-1 (Otieno and Ochieng 2004; Seckler, Amarasinghe et al. 1996). Malin Falkenmark, a Swedish hydrologist who developed the concept of ‘water scarcity index’ on the basis of minimum water required per capita, describes countries that have <1700 m^3 per capita y^-1 as being water stressed. At <1000 m^3 the country begins to experience chronic water shortage that could harm economic development and lead to degradation of the environment (Falkenmark 1994; Falkenmark and Widstrand 1992). This is a situation faced by South Africa and at the current population rates water requirement is exceeding water availability. The current population is 49 991 300 (StatisticsSA 2010) and the daily minimum water requirement per person was estimated to be 100 l for drinking, bathing and cooking and five to 20 times of this amount is required for agricultural, afforestation, energy and mining and industrial sectors (Falkenmark and Widstrand 1992). At the moment South Africa is ranking water as the most urgent development constraint that needs to be addressed and it is closely
linked to poverty, hunger and disease (Ashton and Haasbroek 2002; Ashton and Seetal 2002; Falkenmark 1994)

Water supply in South Africa relies greatly on rivers, dams and under ground water (Turpie, Winkler et al. 2002). There are 320 major dams in South Africa with the total capacity of 32 400 x 10^6 m^3 (DEAT 2007). The details of the five major dams including the vegetation types surrounding them named according to an up to date and complete overview of the vegetation of South Africa edited by Mucina and Rutherford (2006) are listed in Table 1.

**Table 1: Details of the five major dams of South Africa**

<table>
<thead>
<tr>
<th>Dams</th>
<th>Capacity (m^3)</th>
<th>Located Between</th>
<th>Rivers</th>
<th>Co-ordinates</th>
<th>VEGETATION TYPES</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gariep dam</td>
<td>5341 x 10^6</td>
<td>South of EC</td>
<td>Orange River</td>
<td>30.40°S, 25.30°E</td>
<td>Xhariep Karroid Grassland; Eastern Upper Karoo; Besemkaree Koppies Shrubland (Mucina and Rutherford 2006)</td>
</tr>
<tr>
<td>Vanderkloof dam</td>
<td>3200 x 10^6</td>
<td>NC, NW of Orania</td>
<td>Vaal River</td>
<td>29.59°S, 24.44°E</td>
<td>Xhariep Karroid Grassland (Mucina and Rutherford 2006)</td>
</tr>
<tr>
<td>Vaal dam</td>
<td>2536 x 10^6</td>
<td>Gauteng and FS</td>
<td>Vaal River</td>
<td>26.53°S, 28.7°E</td>
<td>Frankford Highveld Grassland (Mucina and Rutherford 2006)</td>
</tr>
<tr>
<td>Bloemhof dam</td>
<td>1269 x 10^6</td>
<td>NW and Gauteng</td>
<td>Vaal River</td>
<td>27.40°S, 25.37°E</td>
<td>Soweto Highveld Grassland; Besemkaree Koppies Shrubland; Eastern Upper Karoo (Mucina and Rutherford 2006)</td>
</tr>
<tr>
<td>Grootdraai dam</td>
<td>364 x 10^5</td>
<td>Mpumalanga</td>
<td>Orange River</td>
<td>26.55°S, 29.17°E</td>
<td>Kimberly Thornveld; Highveld Alluvial Vegetation; Vaal-Vet Sandy Grassland; Saldanha Limestone Strandveld; Swartland Granite Renosterveld (Mucina and Rutherford 2006)</td>
</tr>
</tbody>
</table>

Abbreviations: Eastern Cape (EC); Free State (FS); Northern Cape (NC); North West (NW)
Due to water loss caused by high evaporation from dams, only about 62% of the average annual runoff can be used cost-effectively (DWAF 2004). However high evaporation rates from the dams is not the only factor affecting water availability. Factors such as changes in land use activities, climate change, and policy and regulation also affect the availability of water:

- **Changes in land use activities** - these are mostly driven by humans and are one of the major factors that affect water availability. Four main ways that the changing patterns of land use have affected water flow and availability as recorded by Department of Environmental Affairs and Tourism – DEAT (2007) are (i) urbanization which is responsible for the reduced volumes of runoff that can recharge groundwater, because of the solidification of most surfaces, (ii) hydrological patterns that are greatly affected by the construction of dams, weirs and bridges and mining within the watercourse, (iii) the misuse of land (e.g. overgrazing and inappropriate burning regimes) which leads to soil erosion that can result in siltation of dams and rivers, and (iv) invasion of alien vegetation that uses more water than the indigenous vegetation and reduces stream flow by up to 10% (DEAT 2007). According to DWAF (2004) the agricultural sector uses about 60% of the total water requirement and other sectoral allocations are afforestation (4%), urban and rural population (25% and 4% respectively) mining and bulk industrial (6%) and power generation (2%) (DWAF 2004).

- **Climate Changes** - Water availability is greatly affected by low rainfall and high evaporation rates which are due to the changes in climatic patterns, resulting in low run-off (DEAT 2007). Temperature changes may lead to a 10 – 40% increase in average river flows in some regions and a 10 – 30% decrease in others, which could greatly affect the supply of water. A sector such as agriculture, which is the major source of livelihoods for many households in the Eastern Cape, will suffer the most (Sadoff and Muller 2009) as a result of increases in temperature. The Second World Climate Conference recognized “…that among the most important impacts of climate change were its effects on the hydrologic cycle and on water management systems and, through these, on socio-economic systems.” (SWCC 1990)

- **Policy and regulations** - National management of water resources is influenced by the National Water Policy 1997; the National Water Act No. 36 of 1998 (NWA), and the Water Services Act No. 108 of 1997, amongst others. Decisions taken in respect
to water management and distribution by relevant authorities of government are
determined by these policies. Land use practice policies also have an indirect impact
on water quality and availability (DEAT 2000; DEAT 2007).

1.1.2 Water Supply in South Africa

Only 8.6% of SA’s annual rainfall is converted to annual runoff, most of which is received by
the Eastern region with the Western region being the drier part of the country. This
conversion of annual rainfall to annual runoff is the lowest in the world; the rest is lost
through evaporation and to groundwater (Walmsley, Walmsley et al. 1999). The hard rock
base in most parts of this country makes it difficult to have major groundwater aquifers
(DWAF 2002). Thus water is provided through the conservation of catchment areas
(watersheds), riparian zones and wetlands (Turpie, Marais et al. 2008). South Africa has 19
Water Catchment Areas that supply water to the dams (Blignaut and De Wit 2004) and these
are predominantly grasslands. Grassland catchments have an ability to act as a sponge that
catches the summer rainfall and then gradually release the infiltrated water in the catchment
during the dry season (Turpie, Marais et al. 2008). However in the Eastern Cape Province
these grasslands catchments are often inappropriately managed by the communities that live
in them (Hoffman and Ashwell 2001). Many of these communities that occupy catchments in
the province have high livestock numbers and there is inconsistent application of a rotational
grazing system, and this leads to overgrazing, soil erosion and invasion of alien plants (Fraser
1989; Fraser 2004). These three factors have an adverse effect on water supply. Firstly, the
low biomass condition associated with intensive continuous grazing results in high storm
flows and rapid erosion, with the resultant siltation of the dams (Mander, Blignaut et al.
2007). Secondly, the short green grass that is maintained by the continuous grazing policies
of many communities affects the rate and quantity of evapotranspiration (ET). This study set
out to examine the effect of two grazing strategies on ET and to quantify the effect on run-
off. At present, livestock owners are still reluctant to reduce the number of their livestock and
the answer has not been found in legislative instruments (e.g. to enforce stock reduction
through the Conservation of Agricultural Resources Act 23 of 1984).
1.2. EFFECTS OF HIGH LIVESTOCK NUMBERS ON WATER QUANTITY

Dense rural settlements, cultivation, overgrazing and excessive burning have led to severe erosion, particularly in the former homeland areas. Overgrazing and soil erosion has resulted in reduced ground water supplies and the silting of dams and rivers (Ongwenyi, Kithiia et al. 1993). High livestock numbers in particular have had a great impact on the degradation of natural rangeland. Communal farmers are often reluctant to part with their animals for cultural and traditional obligations, resulting in the continuous heavy use of natural resources. Reasons for keeping livestock include traditional uses, as a source of wealth, provision of milk, draught power, manure and meat (Cousins 1997). Livestock ownership also gives the owner prestige and status in the community (Ainslie 2002). Since most communal grazing is open access, individual livestock owners can continuously increase their herd size without any consideration of the damage to the grazing land because they do not bear the full cost (Vink 1986). Each livestock owner would consider only their own private costs and benefits (Fraser 1989). This in turn affects the land surface condition and ultimately quality and quantity of runoff because overgrazed lands have less grass cover.

Some of the consequences of overgrazing as recorded in African Adrenalin (2007) are:

1. “Overgrazed agricultural land is permitted to heat up under the sun and as a consequence, water evaporates quickly without being afforded an opportunity to permeate the soil for slow release on a sustainable basis.

2. Overgrazed lands also permit any breeze to speed up evaporation and similarly reduce permeation of water into the soil.

3. Thirdly, overgrazing removes grass cover on the ground, which permits water to flow in volume over the ground after heavy rains, causing wash away of the topsoil.” (African Adrenalin 2007)

Also overgrazed lands are more prone to invasion by IAPS and indigenous woody shrubs which in most cases utilize more water and have higher transpiration rates than the native species. Moran et al (2009) have shown that soil evaporation contribution to ET was doubled where Eragrostis lehmanniana, a southern African grass, had invaded in the United States (Moran, Scott et al. 2009).

One of the ways that could be used to meet the challenge of overgrazed lands is for water users to pay livestock owners directly where they can show that reduction of their livestock numbers coupled with good management practices has lead to an improvement of the grass cover within the catchments. This could be achieved by developing a Payment for Ecosystem
Service (PES) model which is being adopted as a way to solve problems with environmental conservation in many parts of the world (Proctor, Köllner et al. 2008). The economic impact of environmental degradation can only be recognized when the ecosystem services have a price. For instance, the availability of objective price can lead to reduction of degradation through the use of economically efficient market based mechanisms (Millenium Ecosystem Assessment 2005). Richmond et al (2007) attempted to answer two questions relating to ecosystems: firstly they wanted to find out if ecosystems contribute economic gross domestic production and secondly if it does contribute can the contributions be used to calculate the shadow price for environmental services? To answer these questions they estimated a Cobb Douglas production function using the net primary production, capital and labour. Then the positive output elasticity was used to calculate the marginal product of the net primary production which is the shadow price for the ecosystem service (Richmond, Kaufmann et al. 2007). More details on payments for ecosystem service are discussed below.

1.3. PAYMENTS FOR ECOSYSTEM SERVICES (PES)

Degradation of natural resources has had a negative impact on development (Boyd and Banzhaf 2007). One system that has been suggested to counteract this impact is “Payments for Ecosystem Services” (PES). Daily (1997) describes ecosystem services as “the benefits of nature to households, communities, and economies” and promotes the idea that ecosystems are socially valuable (Daily 1997). Payment for ecosystem services is a mechanism to translate external, non-market values of the environment into real financial incentives that encourage land users to provide certain services (Engel, Pagiola et al. 2008). Pagiola and Platais (2007) provide the logical framework for the benefits that can be derived from PES (Figure 1).
PES came about in South Africa through establishment of the WfW programme in 1995 (Turpie, Marais et al. 2008) and has focused on payments for services such as carbon sequestration, improved water supply, biodiversity and soil conservation (Blignaut, Marais et al. 2007). In a PES program, a payment for a certain ecosystem service is offered by an ecosystem service buyer to an ecosystem service seller that has a service that will benefit the buyer (Engel, Pagiola et al. 2008). Most ecosystem service sellers are landowners, of either the state; private; or communal land (Turpie, Marais et al. 2008). Private and communal landowners are mainly farmers that use some of these ecosystem services for either commercial or survival purposes. They also use the affected areas as rangelands, and therefore reduction of alien invasive species and increase in grass biomass will also benefit them and increases their production (Turpie and Heydenrych 2000) and (Turpie, Heydenrych et al. 2003). In cases where the ecosystem services paid for involve changing land use activities, the costs are higher as opposed to when they are made for retaining them (Engel, Pagiola et al. 2008). In the Eastern Cape there is need for land use activities to change within the grassland catchments that are also used as rangelands. Currently these grassland catchments have less grass cover due to inappropriate land practices such as overgrazing. This leads to rapid runoff that silts the dam and also removes the top soil.

Application of effective management regimes is required in order to improve the grass cover within the grassland catchments as this can ensure that there is gradual supply of water to the

**Figure 1:** The logic of PES. Source (Pagiola and Platais 2007)
catchments. Also this will increase the amount of water available to meet the country’s water demand. In the previous years water demands were met through a complex system of engineering supply side solutions which included major inter-basin transfer and water pumping schemes (Smakhtin, Ashton et al. 2001). However these have become costly and the water available for that has reduced because 12 of South Africa's 19 water catchment areas receive insufficient rainfall (Blignaut and De Wit 2004). This has increased the need to explore other options for increasing and conserving water supplies (Ashton and Seetal 2002). Therefore through the application of a payment for ecosystem services (PES) approach livestock owners can be encourage to improve the vegetative cover in the catchments they manage and receive direct compensation from the water user (e.g. dam owners, local authorities, irrigation agriculture) (Engel, Pagiola et al. 2008).

This approach requires a thorough understanding of the quantity of water that can be guaranteed if the livestock numbers are reduced and the vegetative cover changes. Hydrologists have traditionally tried to predict water supply by using models which rely on data from gauging weirs and a network of rain gauges (ACRU (Schulze 1990); SPATSIM (Hughes 2002) SWAP (Kroes and Van Dam 2003). However, these models need to be parameterized for each catchment and this is problematic in un-gauged catchments, which is the predominant situation in southern Africa. The Working for Water Programme in South Africa has demonstrated the importance of understanding the role of plant water use within a catchment (Blignaut, Marais et al. 2007). New techniques have been developed for improving the understanding of the role of evapotranspiration in water balances within catchments, and these can be used to drive management interventions in gauged as well as un-gauged catchments. Several examples of improving estimates of catchment-scale ET have been developed and are strongly driven by input from satellite-borne sensors and programmes (SPOT, MODIS). One of these, the MODerate Resolution Imaging Spectroradiometer (MODIS), produces the leaf area index (LAI) which provides an opportunity (Wang, Woodcock et al. 2004) to model actual ET ($ET_a$) when combined with ground-based meteorological data.

1.5. LEAF AREA INDEX (LAI)

LAI is the total one-sided area of leaf tissue per unit ground surface (Watson 1947). LAI is frequently used to estimate ET and it the component of crop growth analysis that is
responsible for the ability of the crop to capture light energy. LAI also plays a key role in canopy microclimate, water interception, radiation extinction, and water and carbon gas exchange between global ecosystems and the atmosphere (Fan, Gao et al. 2009). LAI estimates can be obtained by using either the direct or indirect methods.

One of the most direct methods involved physical removal of the leaves from the plants (He, Guo et al. 2007) and then the total leaf area of the plant per unit ground surface can be measured using instrumentation (de Jesus, do Vale et al. 2001). This approach is however destructive and is time consuming. Also its accuracy is limited to small areas (Breda 2003). Indirect methods that are based on light transmission through plant canopies, making use of the radiative transfer theory (Ross 1981) have been developed. These include commercial canopy analysers such as the AccuPAR Ceptometer (Decagon Devices, Pullman, WA) and Li-Cor LAI 2000 (Li-Cor, Lincoln, NE, USA) and the analysis of photographs taken upward with fish eye lenses (Schleppi, Conedera et al. 2007) which include Hemispherical Photography, Digital Cover photography. These methods are usually based on the gap fraction theory at different zenith angles (Nilson 1971). The zenith angle (θ) has an influence on both the distance travelled by the light through the canopy and the angle at which the leaves are seen at the measuring point; hence it is used twice in calculations (Schleppi, Conedera et al. 2007). In this study ET was modeled by adjusting ET₀ using the LAI value obtained from MODIS LAI.

1.4. EVAPOTRANSPIRATION (ET)

ET is a process whereby water flows from the soil, through plants to the atmosphere. It is the essential process that enables plants to capture CO₂ from the atmosphere (Teixeira 2010). Because the process is driven largely by incoming solar energy, ET contributes significantly to water loss in arid and semi-arid regions (Warren, Hakonson et al. 1996). The available energy and rainfall amounts drive the rates of the annual mean ET, runoff and groundwater recharge for a specific area. Runoff is usually higher than ET when the available energy and potential evaporation rates are low. When available energy and potential evaporation rates are high (e.g. in arid and semi-arid regions), ET may exceed runoff, except in exceptionally wet years. Observations show that annual ET is almost the same as the annual rainfall in regions where the available energy is much greater than the amount required to evaporate annual
rainfall. While in regions where available energy is smaller than the amount required to evaporate the annual rainfall, annual ET is closer to potential evaporation (Arora 2002).

An accurate estimate of temporal and spatial distribution of ET is an important component of the water budget in and around a region (Dai, Yosuke et al. 2004). Accurate calculations of ET can be obtained using methods such as the Bowen ratio (BR), eddy covariance (EC) techniques and scintillometry (Teixeira 2010). The widely accepted standard method that can be used for estimating potential or reference ET ($ET_0$) is the Penman-Monteith equation which was derived from the Penman equation (Allen, Pereira et al. 1998; Monteith 1965; Penman 1948). $ET_0$ is the evaporation (E) or transpiration (T) that could occur if the surface is well watered (Allen, Pereira et al. 1998). When coupled together, methods for estimating $ET_0$ and ET provide an understanding of what the regional water budget will be after loss of water through the ET process (Briney 2010). This study set out to explore the effects of the different grazing management systems on ET and ultimately on run-off within the grassland catchments of the Eastern Cape and it was envisaged that the outcomes will pave a way for the establishment of a PES model.

1.6. AIMS OF THE STUDY

1) To examine the possibility of using a PES system in rural rangelands as a possible solution to degradation and water issues (quantity and quality). Over the previous years new rangelands management approaches have been introduced to reduce rangeland degradation, however these have not been as successful as envisaged. It is believed that land users usually over exploit the rangeland because it is free and they are not paid to conserve it or simply for survival. Therefore the application of the PES can offer a solution to this issue by ensuring that the livestock owners receive direct compensation for reducing their livestock numbers and improving the condition of their rangelands. However this approach will require a thorough understanding of how to quantify the improvements made by the livestock owners and how much water they can guarantee in order to receive the compensation.

2) To identify suitable models for predicting ET using MODIS LAI and other possible products (SPOT), MODIS ET, from the literature. ET describes the soil-water-plant interactions and is one of the key components in the catchment-scale water balance and accurately modelling ET can help in the determination of the water budget in and around the
Over the years many different techniques for estimating ET based on ground meteorological parameters have been developed and these include the Penman Monteith (P-M), Bowen Ratio (BR), and Eddy Covariance (EC), Large Aperture Scintillometer (LAS). However these techniques are expensive and can only estimate ET on homogenous surfaces at micro-scale (Tsouni, Kontoes et al. 2008). Recently, remote sensing techniques have been developed, and though these cannot measure ET directly they provide synoptic surface information, that makes it possible to estimate surface ET at large scale (Carlson, Capehart et al. 1995; Kustas and Norman 1996). The MODIS data can be used with other ET models to predict actual ET and the reason it is a preferred technique in this study other than the fact that it is free, is because it has high temporal resolution, moderate spatial resolution and high radiometric sensitivity (Zhang and Wegehenkel 2006).

3) To explore and compare ET between two land use sites with contrasting biomass states.

In order to understand the effects that the changes in land-use activity have on the ET rates and ultimately on runoff, it was necessary to explore and compare ET estimates from two different biomass states (high and low biomass). It is argued that high biomass cover is better for water conservation because water from the summer rainfalls is allowed to infiltrate the soil (Turpie, Marais et al. 2008). Whereas low biomass cover which is usually associated with intensive continuous grazing may result in high storm flows leading to soil erosion and dam siltation (Hoffman and Ashwell 2001). Dam siltation has a negative effect on both the quality and the quantity of water. The land-use scenarios show how an increase or decrease in leaf area affects the rates of ET.

4) To apply the selected model for predicting ET to these sites. Applying the model to the selected sites would allow for accurate estimation of ET.

5) To validate the predictions of the model using the open top chamber (OTC). Due to some uncertainties in remote sensing ET data, it is always necessary to develop validation techniques that will give ground truthed measurements of ET. The open top chamber is one such instrument and the advantage of using it as opposed to other ground based techniques is the fact that it is easy to construct, portable, replicable and inexpensive (Marion, Henry et al. 1997).
CHAPTER 2

REVIEW OF TECHNIQUES USED TO MEASURE AND ESTIMATE EVAPOTRANSPIRATION

This chapter reviews some of the methods that are used for estimating ET and provides an explanation of their characteristics, applications, limitations and successes. It has been necessary to prepare this review because various instruments and approaches used to measure ET have been sourced to provide validation data for this project. Although it strives for a comprehensive treatment of the methods, it does not claim any degree of completeness and aims to provide the reader with general background material for the other components of this study. Estimation of ET began a little more than two centuries ago when Dalton in 1802 proposed a mass transport equation for estimating ET (Kairu 1991; Savage, Everson et al. 2004). Since then different methods for estimating ET have been developed. Factors such as the weather parameters, soil water availability and type of plants affect ET and also influence the quality of ET obtained from different techniques (Allen, Pereira et al. 1998). For the purpose of providing background, some comments on the role of various factors in the control of ET are presented. The processes controlling the passage of water from the soil to the atmosphere are complex and their comprehensive review is not the aim of this study.

Weather Parameters: The main weather parameters are temperature, relative humidity (RH), wind speed (U) and solar radiation.

i) Temperature - Generally, when the air temperature is warm, T rates increase and this causes the stomata (the openings where water is released) of the plants to open releasing more water to the atmosphere. However if the air temperature becomes too hot the stomata will close. Cooler air temperature also causes the stomata to close, releasing less water and decreasing the T rates (Briney 2010; USGS 2010).

ii) Relative humidity (RH) - At high RH, T is expected to decrease. This is because as the RH increases, the air becomes saturated and this makes it difficult for water to evaporate. Consequently when RH is low T rates increase.

iii) Wind and air movement - When the wind speed is increased, the, rate of T by plants also increases and this is because when the air is moving, it is less saturated. As the wind moves, the saturated air around the plants is replaced by the drier air.
Soil-moisture availability - Lack of soil moisture causes the plants to transpire less water and they may even lose some leaves.

Type of plant: T by different plants can occur at different rates and things that count for this may be differences in resistance to T, crop height, crop roughness, reflection, ground cover and crop rooting characteristics. For example succulents and cacti transpire less water than other plants so that they can conserve some (USGS 2010). Management and environmental conditions: ET rates may decrease under these conditions – soil salinity, poor land fertility, and inadequate application of fertilizers, hard soil horizons, lack of control of diseases and pests and poor soil management. All of the above factors must be considered when one is assessing or developing procedures to measure ET.

Methods that estimate ET have been grouped in three and they are: water balance methods, micrometeorological (energy balance) methods and plant physiology methods. The most commonly used water balance method is the weighing lysimeter. Methods such as Penman-Monteith equation, Bowen ratio method and Eddy covariance method are widely applied micrometeorological methods. Chamber method, tracer technique and cut-tree method belong to the plant physiology method group (Xiong, Qiu et al. 2008). Recently, remote sensing measurements have been developed, though these cannot measure ET directly they extend the above methods which can only estimate ET from homogenous surfaces at micro scale to larger areas (Tsouni, Kontoes et al. 2008).

2.1. WATER BALANCE METHODS

Water balance methods determine ET by measuring different aspect of the soil water balance. These measure the incoming and outgoing water flux into the crop root zone and give ET estimates over a week or ten days (Allen, Pereira et al. 1998). The most commonly used water balance method is the lysimeter (weighing lysimeter and microlysimeter).

Weighing lysimeters are large containers filled with soil, water, other chemicals and entire plants. The weighing lysimeter provides direct measures of ET fluxes from the vegetation surface and it is regarded as the standard method for measuring latent heat flux (Aboukhaled, Alfaro et al. 1982). Weight measurements are made at regular time intervals. The evaporation rates are obtained by dividing the weight difference per unit time difference by the density of water and by the cross-sectional area of the lysimeter. The rate of water lost from the
containers can be measured either for very short or long time intervals. The results obtained from the lysimeter can be used as a standard to validate other methods; however these results cannot be used to confirm ET estimates from other methods at regional scale because they only provide point data. Also the weight lysimeter methods are expensive to construct, they are destructive to the soil and not easy to maintain (Kairu 1991; Savage, Everson et al. 2004).

2.1.2. Microlysimeter

A microlysimeter is relatively inexpensive to construct, and mostly used to measure soil evaporation rates over a short time interval. This is due to the fact that they have a much small surface area than the weighing lysimeter and often cannot contain the whole plant (Savage, Odhiambo et al. 2010).

2.2. MICROMETEOROLOGICAL METHODS

Obtaining accuracy when estimating ET in the field is difficult and the instruments that are constructed for that are either expensive, labour intensive, disruptive to the natural environment or only limited to specific conditions. This has lead to the development of micrometeorological methods which compute ET estimates from weather data and these methods include: Penman Monteith (P-M), optical scintillation methods, eddy covariance (EC), Bowen ratio (BR) energy balance, surface renewal, and flux variance. The advantage of the meteorological methods is that there are no disturbances to the microenvironment and they incorporate fluxes over a big area thereby reducing sampling errors (Mengistu and Savage 2010).

2.2.1. Penman Monteith Method

The P-M method (Monteith 1965) has been adopted by the United Food and Agriculture Organization (FAO) as the standard method for estimating ET. ET is what would be evaporated from the surface if it is well watered and it was introduced to study the evaporative demand of the atmosphere without considering the crop type, development and management regimes. ET is not affected by the soil factors because it a climatic parameter that is estimated from the weather data. The crop ET (ETc) under standard conditions is determined by the crop coefficients (Kc) that relates it to ET while ETc under non-standard conditions is adjusted by a water stress coefficient (Ks) and/or by adjusting the Kc (Allen,
The FAO P-M, referred to as FAO-56, has been selected for use in this study because of its close estimation of grass ET₀ at the site being estimated, and it integrates both physiological and aerodynamic parameters (Allen, Pereira et al. 1998). The FAO-56 standards for application of the P-M model require daily maximum (Tₘₐₓ) and minimum temperature (Tₘᵟᵦₜ), solar radiation (Rₛ), maximum (RHₘₐₓ) and minimum (RHₘᵟᵦₜ) relative humidity or dew point temperature, and wind speed (U). However in cases where only daily temperature data is available, generalized temperature-based procedures to estimate the other meteorological data required, can be provided by FAO-56 through ClimGen (Castellvi and Stockle 2002; Stockle and Nelson 1999). ClimGen is a weather generation program that can give estimates of several climate parameters based on local input data. ClimGen can give daily estimates of Rₛ and vapour pressure deficit (VPD) from Tₘₐₓ and Tₘᵟᵦₜ data, and produce statistically correct U values though it is not linked to other weather parameters. Therefore, P-M ET₀ can be calculated using only the temperature records (Stockle, Kjelgaard et al. 2004). In this study FAO Penman-Monteith was computed using weather data obtained from the Automatic Weather Stations close to the selected sites which provide all the required parameters.

2.2.2. Scintillometer

The scintillometer is described as an optical instrument that measures the intensity of fluctuations of visible or infrared radiation, after propagation above the plant canopy of interest through a turbulent medium, to work out different meteorological parameters (Hill, 1992; Savage, Odhiambo et al. 2010). The scintillometer methods rely on Monin-Obukhov similarity theory (MOST) and when they are combined they provide path-averaged measurements of sensible heat (H) and momentum fluxes over distances of between 50 m and 250 m up to 350 m (Savage, Odhiambo et al. 2010; Thiermann and Grassl 1992).

There are three different types of scintillometers that are used to measure water fluxes. The surface layer scintillometers (SLS) comprises two laser beams and either two or four detectors and has a wavelength of 670 nm. The SLS measures sensible heat flux density (H) and momentum τ flux density over a path distance of about 50 and 350 m and areas of about 0.25 to 5 ha. They have the receiver aperture size that is less than the Fresnel zone (F). The large aperture scintillometers (LAS) and extra large aperture scintillometer (XLAS) are for measuring refractive index structure constant, Cₙ² only and they use a near infrared beam.
wavelength together with horizontal wind-speed measurements required for the estimation of $H$ using MOST. The LAS operate over distances of about 250 m up to 3 km and areas that are over 6 ha with receiver aperture size that is greater than $F$ (Savage, Odhiambo et al. 2010) while the XLAS operate over longer distances of up to 10 km and have a receiver aperture size that is nearly double that of the LAS (Odhiambo and Savage 2009). One of the limitations of using a scintillometer is its high price.

### 2.2.3. Bowen Ratio Energy Balance Method

The Bowen ratio/energy balance (BREB) method presents a simple and inexpensive alternative to weighing lysimeter which give accurate measures of ET. BREB is widely used for estimating latent heat flux ($\lambda E$) over a wide range of natural surfaces such as irrigated wheat, forest, bare soils and some non-ideal sites (Liu and Foken 2001). The BREB method is based on Bowen ratio (Bowen 1926) and the energy balance equation and determines ET using only temperature and humidity measurements, net radiation, and soil heat flux (Rosenberg, Blad et al. 1983). The Bowen ratio is the ratio of the sensible heat flux ($H$) to water vapor flux ($H/\lambda E$) and it can be estimated from temperature and vapor pressure gradients by:

$$B = \gamma \left( \frac{\Delta T}{\Delta e} \right)$$

Where: $\gamma$ is the psychrometric constant, $\Delta T$ is the air temperature gradient and $\Delta e$ is the vapor pressure gradient. The basic energy balance model is then calculated from:

$$R_n + G + H + \lambda E = 0$$

Where: $R_n$ is net radiation, $G$ is soil heat flux, $H$ is sensible heat to the atmosphere, and $\lambda E$ is the evaporative flux.

BREB method was found to be accurate and can be used under semi-arid conditions, however due to recent observations of an imbalance of the surface energy budgets, this accuracy has been questioned (Foken, Kukharets et al. 1999). To avoid these imbalances Liu and Foken (2001) proposed a modified Bowen ratio method (MBR) for measuring $H$ and $\lambda E$ using a sonic anemometer together with temperature and humidity measurements at two levels (Liu and Foken 2001). Their results showed that humidity and velocity transformations with errors of < 10%, and the latent heat flux with errors of <20% are acceptable when the buoyancy flux is measured with a sonic anemometer (Liu and Foken 2001).
2.2.4. Eddy Covariance

The EC method is applied in many micrometeorological sites to measure CO₂ and H₂O fluxes (Mammarella, Werle et al. 2010) and depends on the turbulence of the airflow above and through a canopy. Air travelling over the canopy has a large number of eddies which are small chaotic currents of wind, and these can move at any direction; at any time. The EC system is usually located several meters above canopies and it is made up of an open path infra-red gas analyser (measures CO₂ and H₂O concentrations); a 3-dimension sonic anemometer (determines the velocity in 3-dimensions) and a temperature sensor (measures the air temperature) (Eamus, Hatton et al. 2006). EC technique is still the only technique that directly measure gas fluxes to and from the canopy; however like the BREB method they do not discriminate flux from the soil. One of the major limitations of this technique is the complexity of its design, the difficulty to implement and the large volumes of data that it processes (Burba and Anderson 2007). However EC technique has had success in addressing issues like the water and CO₂ budgets of the catchments (Eamus, Hatton et al. 2006).

2.3. PLANT PHYSIOLOGY METHODS

2.3.1. Open Top Chamber (OTC)

The initial aim for designing the OTC was to use it in the field for studying the effects of air pollutants, and then later it was modified for use in elevated CO₂ researches (Heagles et al., 1973; Drake et al., 1989). Eamus et al., (2006) have shown how the OTC can also be used for measuring the actual ET from low lying vegetation by summing up the water lost from the vegetation and soil within the chamber. The OTCs are inexpensive, easy to construct (making them easily replicable and replaceable) and transport. The design of the OTC has a cylindrical base and the cone shaped top which are covered using Melinex® that is basically polyethylene terephthalate film with high mechanical strength, good heat resistance, good flexibility, excellent electrical insulating and light transmission properties (http://www.alibaba.com/product-tp/105582310/Melinex_Film.html). A centrifugal fan (size: 100mm; air volume flow rate varies from 0.009 – 0.061 depending on the pressure; the sound and electrical data is stated at 230V 50Hz) blows air into the chamber and this air is monitored when entering and leaving the chamber using probes that can measure wind speed, relative humidity, temperature and dew point (Eamus, Hatton et al. 2006). More
specifications of the OTC design are detailed in Chapter 3. Then transpiration from the vegetation within the chamber is calculated from the following equation:

$$E = \frac{\Delta e - Fa}{Al}$$  \hspace{1cm} (1)

Where $E$ is transpiration rate (mol m$^{-2}$ s$^{-1}$); $e$ is the difference in water vapour concentration (mol m$^{-3}$) between the inlet and outlet of the OTC; $Fa$ is the rate of air flow through the chamber (m$^3$ s$^{-1}$) and $Al$ is the total leaf area (m$^2$) within the chamber at the time of measurement (Eamus, Hatton et al. 2006).

2.4. THE APPLICATION OF REMOTE SENSING TECHNIQUES FOR ESTIMATING ET

Most of the above-mentioned methods are better suited to homogenous surfaces at micro-scale, and remote sensing technology was developed in the late 1970’s which provided the option for spatial modeling of water use. Remote sensing technology provides synoptic surface information, making it possible to estimate surface ET at large areal extent through empirical/statistical relationships that use remotely sensed vegetation indices (VIs) (Carlson, Capehart et al. 1995; Kustas and Norman 1996). Also there are physical models that can estimate ET from remotely sensed data which are based on solving surface energy balance (SEB) equations through remotely sensed land surface temperature estimates (Kustas and Norman 1996; Overgaard, Rosbjerg et al. 2006). In general, remote sensing techniques can only measure E or ET indirectly (AGU, 1995). Some of the new remote sensing techniques that can assist in estimating ET include the Moderate Resolution Imaging Spectrometer (MODIS) sensor system on the Terra satellite (Glenn, Huete et al. 2007a).

2.4.1. Satellite Pour l’Observation de la Terre (SPOT)

SPOT is a high-resolution, optical imaging Earth observation satellite system that is operated by the SPOT Image in Toulouse, France. It was first initiated in the 1970’s by the Centre National d’’Etudes Spatiales- the French space agency (CNES) and developed in association with the Belgian Scientific, Technical and Cultural Services (SSTC) and the Swedish National Space Board (SNSB). The SPOT has a series of optical remote sensing satellites with the main objective to obtain Earth imagery for land-use, agriculture, forestry, geology, cartography, regional planning, water resources and GIS applications. A single SPOT satellite
flies over the same point within 26 days and provides global coverage between 87 degrees north latitude and 87 degrees south latitude. SPOT 1, 2 and 3 were launched on 22 February 1986, 22 January 1990 and 26 September 1993 respectively. SPOT 3 is no longer functioning.

The SPOT 1 & 2 have two identical High Resolution Visible (HRV) sensors with a 20 m spatial resolution and three spectral bands sensor resolution: two visible bands, the green (0.5-0.59 µm) and the red (0.61-0.68 µm) and one near infrared band (0.79-0.89 µm). SPOT 4 was launched on the 24 March 1996 and has a High Resolution Visible Infrared (HRVIR) sensor with 20 m spatial resolution and has four spectral bands, two visible bands, the green (0.50–0.59 µm) and the red (0.61–0.68 µm); one near infrared (0.78–0.89 µm) and the shortwave infrared (1.58–1.75 µm). The most recent one, SPOT 5 was launched on the 04 May 2002 and has two High Resolution Geometric (HRG) sensors that were inferred from SPOT 4’s HRVIR and has a 10 m spatial resolution (CRISP 2001; Khalaf 2010; Aboelghar, Arafat 2011).

Spot images provide more spatial details and have longer time series than other data sets such as MODIS since it was first launched in 1986. However some limitations in relation to their spatial resolution have been reported that they are not always suited for small plots (Khalaf 2010; Gers and Schmidt 2001).

2.4.2. **Moderate Resolution Imaging Spectrometer** (MODIS LAI)

The MODIS instrument was first developed after the Engineering Model was completed in mid-1995. Since then, it has become the key instrument aboard the Terra (EOS AM) satellite launched in 1999 and Aqua (EOS PM) satellite launched in 2004. The MODIS instrument has been built according to the specifications of NASA by the Santa Barbara Remote Sensing Centre (NASA 2009) and it makes it possible to expand ground truthed ET measurements to larger scale with satellite imagery (Glenn, Huete et al. 2007a). MODIS data is useful for the estimation of ET and surface soil water availability using vegetation index such as the normalized difference vegetation index (NDVI) and LAI (Nagler, Cleverly et al. 2005; Nishida, Nemani et al. 2003; Venturini, Bisht et al. 2004). These data can also be used to estimate canopy water stress index from the reflectance of MODIS multispectral bands (Fensholt and Sandholt 2003) and provide physical models with remote sensing based information such as LAI (Zhou, Liang et al. 2004). MODIS began delivering 1 km global scale LAI products at an 8 day interval after the launch of the EOS Terre satellite in March.
2000. The MODIS LAI product has been used in many validation studies and its accuracy is 0.66 LAI units RMSE when all plants are included and it is 0.5 LAI units RMSE when broadleaf forests are excluded (NASA 2009). The validation of MODIS LAI product have two parts, firstly the validation of LAI Radiative Transfer algorithm and secondly the accuracy of LAI product by using empirical method (Cohen, Maier-serger et al. 2003; Tian, Woodcocka et al. 2002). These are being analysed at several sites over the world (Kim and Lee 2003). It has been found that the earlier versions of MODIS LAI were over-estimating the LAI product in forest, whereas they were under-estimating them in grasslands and crop lands (Cohen, T.K. et al. 2006). These irregularities have been corrected in Collection 4 (Tan, Hu et al. 2005) which is an improvement on simple band products (Fuentes, Palmer et al. 2008). The MODIS LAI has since been improved to Collection 5 which gives better quality retrievals over broad leaf forests (Hill, Senarath et al. 2006).

According to Knyazikhin et al (1998) the retrievals are affected by three random variables which are: uncertainties in surface reflectance product (due to corrections for atmosphere effects); landcover identification (due to biome identification) and georegistration. Due to variations caused by uncertainties the algorithm does not guarantee an accurate LAI value in one pixel and this can be reduced by averaging over a homogenous area or take the best value of parameters to estimate the most probable value of LAI. Since the algorithm provides the distribution of LAI values as if they were derived from ground based measurements, it is therefore important to use statistical techniques that can compare field measured and satellite-derived LAIs to validate the MODIS LAI product. Ground based validation techniques play a vital role in measuring uncertainties in the MODIS LAI product. Fuentes et al. (2008) used ground-truthed measurements of canopy LAI obtained using gap fraction analysis of upward-looking digital photographs at eight sites in Australian Eucalyptus woodland to validate MODIS LAI (Fuentes, Palmer et al. 2008; Macfarlane, Hoffman et al. 2007).

Palmer et al (2010) validated the predictions by the MODIS LAI-Stand Water Use (MODIS LAI-SWU) model they developed using data obtained from two catchments. The one in Northern Territory (Howard River), Australia had ground truth measurements from EC, sapflow and OTC techniques which were compared with the MODIS LAI-SWU model predictions. From the other one which was a gauged quaternary catchment in South Africa, ET was calculated from the available rainfall and runoff data and also compared with the predictions by MODIS LAI-SWU model (Palmer, Fuentes et al. 2010). Their results showed an acceptable comparison between the MODIS LAI-SWU model predictions and the EC data.
which led to the conclusion that the model can improve catchment scale estimates especially in ungauged catchments (Cook, Hatton et al. 1998; Palmer, Fuentes et al. 2010).

Studies done by Leuning et al. (2005) showed that MODIS LAI has an ability to approximate the stomatal conductance component \( (g_s) \) of the Penman Monteith equation and it was used in this study as an input into the Penman–Monteith equation (Leuning, Cleugh et al. 2005). The MODIS LAI estimates obtained are ground truthed using the OTC that sums up the water lost from the vegetation inside the chamber. MODIS was selected in this study because it has a higher radiometric resolution than any other high temporal resolution imagery (Khalaf 2010).
CHAPTER 3
METHODS AND MATERIALS

3.1. SITE SELECTION:

3.1.1. Site selection Criteria

Sites considered suitable for this study were quaternary catchments that have dams that supply water to the urban areas and the rural water schemes. The selected sites were either of high or low biomass cover for comparison. Rivers, dams and quaternary catchments were downloaded from the Department of Water Affairs and Forestry website: http://www.dwaf.gov.za/iwqs/gis_data/river/rivs500k.html and the map of the Local Municipalities of the Eastern Cape (EC) from the Municipal Demarcation Board website: http://www.demarcation.org.za/. These were downloaded as ARCVIEW shapefiles and then imported to IDRISI (Version I32.11, Clark Labs, Clark University, Worcester, Massachusetts) where they were saved as vector files. Rivers, dams and quaternary catchments were displayed as overlays on the MODIS LAI product from the growing season in 2009 to select quaternary catchments suitable for further study. Images from Google Earth were also used to visually confirm the cover and biomass state of the selected sites. Four quaternary catchments were selected in the following Local Municipalities of the Eastern Cape, South Africa: Emalahleni, Sakhisizwe and Intsika Yethu (quaternary catchment S20C and S50E) and Makana Local Municipalities (quaternary catchment Q91C and P10A).

3.1.2. Site Description

3.1.2.1. Kei River Primary Catchment

Two quaternary catchments were selected within the Kei River Primary Catchment (figure 2 and 3), S50B (31.72°S, 27.58°E) which represents a high biomass (HB) condition and S20C (31.69°S, 27.39°E) which represents a low biomass (LB) condition (Appendix 1 provides physical details of these catchments). Quaternary catchment S50E, near Tsomo, contains the Emalahleni and Sakhisizwe Local Municipalities which are both under the Chris Hani District Municipality. This catchment supplies the Ncora dam on the Tsomo River. The Ncora dam was established in 1975; has a capacity of 150 x 10^6 m^3 and a surface area of 1392 ha. Quaternary catchment S20C, near Qamata in the Emalahleni Local Municipality, is within the Chris Hani District Municipality. The catchment supplies water to the Lubisi dam.
on the Indwe River. The Lubisi dam was established in 1968, it has a capacity of $158 \times 10^6$ m$^3$ and a surface area of 1129 ha. Both the Lubisi and Ncora dam are maintained by DWA.

### 3.1.2.2. Bushmans River Primary Catchment

Two quaternary catchments (P10A and Q91C) were selected in the Bushmans River Primary (figure 2 and 3) as they contained examples of judicious and high intensity grazing. The primary catchment has an area of 2675 km$^2$ and 4% of this catchment is degraded bushland, 90% is natural vegetation and comprises forest, succulent thicket, shrubland and grassland. One percent of the catchment is urban, including residential and industrial development. Two 1 km pixels were selected from P10A and Q91C quaternary catchments. The 1 km pixel (33.29°S, 26.49°E) selected from P10A represents a high biomass state that is lightly grazed by domestic livestock and wild herbivores and the 1 km pixel selected from Q91C (33.27°S, 26°.60°E) represents a low biomass state that is intensively grazed under communal management. Both these quaternary catchments are located near Grahamstown in the Makana Local Municipality (Cacadu District Municipality). There are two small dams that receive water from quaternary catchment P10A, namely Milner and Jameson dams. There is no dam in the Q91C quaternary catchment; however water is supplied by the Great Fish River.

![Figure 2: Map showing the Primary Catchments of the Eastern Cape](www.environment.gov.za/enviro-info/prov/ec/ecpcat.jpeg)
Figure 3(a): Map showing the selected quaternary catchments within the Chris Hani District Municipality with the dams and the surrounding towns.

Figure 3(b): Map showing the selected quaternary catchments within the Cacadu District Municipality with the dams and the surrounding towns.

3.2. EVAPOTRANSPIRATION ESTIMATES

3.2.1. Actual ET (ET$_a$) Quality Assessment

In order to ensure accuracy when estimating ET using remote sensing techniques, it is necessary to validate the remotely sensed ET using ground truthed measurements from
instruments such as Bowen-ratio energy balance (BR), eddy-covariance system (EC), Large Aperture Scintillometer (LAS) and Open Top Chambers (OTC). (Leuning, Zhang et al. 2008) (Tsouni, Kontoes et al. 2008). The strength of the quality of the remotely sensed ET depends on the availability of these validation instruments. Therefore a critical assessment of the quality levels of ETa obtained from each selected catchment was done based on the proximity of the AWS to the catchment and the availability of BR, EC, LAS, or OTC equipment to validate the modeled ET. The quality was rated:

- Very high where the AWS (inside the catchment) + EC or BR or LAS or OTC + MODIS LAI were available for validation.
- High where the AWS (inside the catchment) + OTC + MODIS LAI were available.
- Moderate where AWS (outside the catchment) + OTC + MODIS LAI were available and
- Low where AWS (outside the catchment) + MODIS LAI were available and
- Very Low where only MODIS LAI was available.

The quality levels of ETa calculated for the selected quaternary catchments in this study are shown in the in Table 2 below.

<table>
<thead>
<tr>
<th>Catchment</th>
<th>Year</th>
<th>AWS (in)</th>
<th>AWS (out)</th>
<th>EC</th>
<th>LAS</th>
<th>OTC</th>
<th>MODIS LAI</th>
<th>Rating</th>
</tr>
</thead>
<tbody>
<tr>
<td>S20C</td>
<td>2009</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>Low</td>
</tr>
<tr>
<td>S50B</td>
<td>2009</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>Low</td>
</tr>
<tr>
<td>Q91C</td>
<td>2000-2008</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>Low</td>
</tr>
<tr>
<td>Q91C</td>
<td>2009-2010</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>X</td>
<td>Moderate</td>
</tr>
<tr>
<td>P10A</td>
<td>2000-2008</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>Low</td>
</tr>
<tr>
<td>P10A</td>
<td>2009-2010</td>
<td>-</td>
<td>X</td>
<td>-</td>
<td>-</td>
<td>X</td>
<td>X</td>
<td>Moderate</td>
</tr>
</tbody>
</table>

### 3.2.2. Acquiring MODIS LAI Data

Eleven years of the MODIS 8-day 1 km Collection 5 (MOD15A2) composite LAI/FPAR product for the period March 2000 to December 2010 have been acquired from the NASA Distributed Active Archive Centre. Images were extracted from the archive using the MODIS reprojection tool (HegTool, HDF Group, University of Illinois Research Park, Champaign, Illinois) and imported into IDRISI (Version I32.11, Clark Labs, Clark University, Worcester, Massachusetts) image processing package to create a 11-year data stack with 451 layers.
MODIS LAI has been used in previous studies to estimate ET. Cleugh et al (2007) estimated E using the P-M equation (Monteith 1965) with a simple model for surface conductance to E given by:
\[ G_s = c_L L_{ai} + G_{s,min} \]  
(2)
Where \( L_{ai} \) is the leaf area index obtained from MODIS remote sensing and \( c_L \) is a constant and \( G_{s,min} \) is the surface conductance controlling soil E. They found that these results were more acceptable than the poor comparison they found when the multi-seasonal time series of E estimated using MODIS measurements of the radiative surface temperature was compared with the resistance surface energy budget approach (Cleugh, Leuning et al. 2007).

### 3.2.2.1. Selecting the most appropriate model for calculating ET<sub>a</sub>

Palmer et al (2010) examined the relationship amongst MODIS LAI (Collection 5), pre-dawn leaf water potential (a surrogate for plant water availability), vegetation water use (ET) and pan evaporation (\( E_0 \)) in forest, evergreen woodland, open shrubland and savanna in Australia (New South Wales, Western Australia and the Northern Territory) in the period of 2002 – 2008. A regression model was prepared from a maximum value composite of MODIS LAI and the pre-dawn water potential in the savanna of Northern Territory (Palmer, Fuentes et al. 2008). However as the present study is confined to homogenous grasslands, the regression model was deemed inappropriate as it only applied to Australian woodlands (Palmer, Fuentes et al. 2010).

A model was developed in order to improve ET estimates by exploring the potential of MODIS (MOD15A2) to factorize the surface conductance term in the P-M equation (Cleugh et al 2007). Palmer and Weideman (2011) proposed a simple surface conductance model that can be used to predict ET<sub>a</sub> in a semi-arid savanna site in the Kruger National Park (KNP) using MODIS LAI values to inform the P-M equation of the surface vegetation processes. To achieve, this hourly meteorological data from a nearby automatic weather station was obtained and \( E_0 \) was calculated using FAO56. Then maximum LAI (LAI\(_{max}\)) values were extracted for a single MODIS pixel retrieved in the KNP (25.05°S 31.51°E) for year 2007. An eddy covariance (EC) system (FLUXNET) that records fluxes at 30 min intervals at the KNP site was used to validate the predicted ET<sub>a</sub> by MODIS LAI. Clean EC data for 173 days of 2007 which were assumed to be the true reflection of actual ET were extracted for both the
wet and dry season. Then MODIS LAI was used to parameterize the P-M equation using the following function:

\[ \text{ET}_a = \text{ET}_0 \times (\text{LAI}/\text{LAI}_{\text{max}}) \]  

(3)

Where:

LAI is the MODIS LAI for the 8-day period, \( \text{LAI}_{\text{max}} \) is the maximum retrieved at the site over the whole data record (~500 records) and \( \text{ET}_0 \) was calculated from an hourly meteorological data from Malekuta AWS for 2007 using the revised Penman-Monteith equation (FAO56) (Allen et al. 1998).

During the dry season, ET obtained from Eq. (3) had to be further adjusted by an optimized scaling factor of 0.65 as it was observed that MODIS tends to over-predict LAI during this season. The results showed a close correlation between the ET obtained from the EC (\( \text{ET}_{\text{EC}} \)) (147.6 mm) and ET obtained from Eq. (3) (\( \text{ET}_{\text{MODIS}} \)) (139.2 mm) (Figure 4)

![Figure 4: Accumulated ET in 2007 for the site in KNP. Source: (Palmer & Weideman 2011)](image)

To further validate the model Palmer and Weideman (2011) used daily estimates of \( \text{ET}_a \) from a Large Apperture Scintillometer (LAS) for two week periods in 2005. The LAS \( \text{ET}_a \) (\( \text{ET}_{\text{LAS}} \)) estimates were obtained for a week in the growing season (February) and a week in the dry season (May) of 2005 and a strong correlation was observed when these were compared with the MODIS LAI \( \text{ET}_a \) (\( \text{ET}_{\text{MODIS}} \)) obtained from Eq. (3) (figure 5a and 5b).
As the ET\textsubscript{MODIS} model provided the most parsimonious result of the possible ET models evaluated, Eq. (3) was used to calculate daily ET for the quaternary catchments selected in the present study (Q91C; P10A; S20C and S50E). ET\textsubscript{0} was calculated for each catchment using the stations listed (Table 3).
Table 3: Locations of the weather stations and the years the data was available

<table>
<thead>
<tr>
<th>Weather Station</th>
<th>Period</th>
<th>Co-ordinates</th>
</tr>
</thead>
<tbody>
<tr>
<td>Qamata AWS</td>
<td>2006 – 2010</td>
<td>-31.98 27.44</td>
</tr>
<tr>
<td>Bucklands MWS</td>
<td>2000 – 2010</td>
<td>-33.1 26.72</td>
</tr>
<tr>
<td>Rhodes University AWS</td>
<td>07/2008 – 2011</td>
<td>-33.18 26.30</td>
</tr>
<tr>
<td>AWS</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Data from the AWS stations hourly data for: $T_{max}$ and $T_{min}$; $RH_{max}$ and $RH_{min}$; $U$; $R_s$; wind direction; rainfall data and $ET_0$. The accuracy of the $ET_0$ calculated internally by the data logger on these Campbell Scientific automatic weather stations was verified by manually computing $ET_0$ using the data and FAO56 computer program (Allen, Pereira et al. 1998). Daily data is provided by the manual weather station (MWS) at Bucklands, and the FAO56 for daily data was used to calculate daily $ET_0$.

The $ET_{MODIS}$ model (Palmer & Weideman 2011) was applied to all further estimates of $ET_a$. The resulting $ET_{MODIS}$ was further validated using ET obtained from the OTC ($ET_{OTC}$) (see details in 3.4.).

3.2.3. Mean annual ET for P10A (HB) and Q91C (LB) 1 km pixels

Estimation of $ET_{MODIS}$ for the 1 km pixels was done by firstly digitizing the boundaries of quaternary catchment Q91C and P10A as polygons. Then within each quaternary catchment the selected 1 km pixels were digitized as polygons (figure 6).

Figure 6: MODIS LAI Image 2009017 showing the catchment boundaries of Q91C and P10A and the 1 km pixels within each quaternary catchment.
LAI was then extracted from an 11 year data stack of MODIS LAI (March 2000-March 2011) for each 1 km pixel (high biomass and low biomass). The actual annual ET for each 1 km pixel was calculated by adjusting the ET$_0$ data from Bucklands MWS (2000-2008) and Rhodes University WS (2009-2011) with the MODIS LAI product using the Eq. (3). In Statistica version 9, ANOVA (Analysis of Variance) was used to check if there was any significant difference between the mean annual ET$_{MODIS}$ estimates over the 11 year period. Where mean annual ET$_{MODIS}$ was significantly different the Tukey-Honestly Significantly Different (HSD) Post-hoc test was used.

The resulting ET$_{MODIS}$ estimates for the growing season of September 2010 – March 2011 were validated using the OTC measurements obtained from the two 1 km pixels.

### 3.3. OPEN TOP CHAMBER

Ground truthed estimates of ET were obtained using an open top chamber (OTC) which is small and semi-portable, constructed following methods described by (Eamus, Hatton et al. 2006). The OTC sums the water lost from vegetation and soil within the chamber and was used in this study to validate the ET$_{MODIS}$.

#### 3.5.1. Design of the Open Top Chamber:

The OTC is designed to be portable (Figure 7). It consists of a round cross-section of clear Mellinex® which is used because of its good light transmission properties, with a tapered top.

The OTC has two sections – a lower cylindrical base of 0.77 m diameter and 1.23 m height, and a metal frame supporting a Mellinex® cone mounted on top of this base. Total chamber height is 2 m and the volume enclosed is 0.78 m$^3$. A 12 volt 100 Amp hour battery was used to power the fan supplying air to the chamber. The OTC had one input pipe at the cylindrical base and the cone shape top, and an exit pipe at the cone shaped top and each had a vent for instruments that measure the temperature, wind speed and relative humidity. The exit pipe at cylindrical base had an in-line valve which could be adjusted to increase or decrease the rate of air entering the chamber.
3.5.2. Method

The OTC was tested from the grasslands of the two 1 km pixels selected from the quaternary catchments P10A and Q91C and during the growing season of September 2010 – March 2011. Measurements were initially taken using a Kestrel 3000 hand-held vane anemometer only; however the quality of its relative humidity and temperature sensors raised some concerns. To improve the quality of relative humidity and temperature, a Vaisala HMP45C RH and TEMP probe was acquired and used together with the Kestrel 3000 anemometer. The parameter measured by the Kestrel 3000 anemometer was the wind speed and Vaisala HMP45C RH and TEMP probe was used to measure relative humidity and temperature. These parameters were measured both at the inlet and the outlet of the chamber at 30 minute interval and this was done on 5 clear, low wind (<6 m s\(^{-1}\)) and sunny days.

Transpiration was then calculated using Eq. (1):

\[
E = \frac{(\Delta e)(F_a)}{(A_l)}
\]

Where \(E\) is transpiration rate (mol m\(^{-2}\) s\(^{-1}\)); \(\Delta e\) is the difference in water vapour concentration (mol m\(^{-3}\)) between the inlet and outlet of the OTC; \(F_a\) is the rate of air flow through the chamber (m\(^3\) s\(^{-1}\)) and \(A_l\) is the total green leaf area (m\(^2\)) within the chamber at the time of measurement. Then a regression analysis was used to prepare a relationship between ET
estimates obtained from the OTC and those obtained from MODIS LAI for all days of recording under both treatments.

3.4. LAND-USE SCENARIOS FOR QUATERNARY CATCHMENTS S50E (HB) AND S20C (LB)

To show the impact that changes in the biomass cover can have on ET rates and ultimately runoff/recharge rates, three scenarios were created in IDRISI (Version I32.11, Clark Labs, Clark University, Worcester, Massachusetts) Scenario 1; 2; 3. Two quaternary catchments were selected for this purpose a high biomass catchment (S50E) and a low biomass catchment (S20C).

3.4.1. Scenario 1

In the first scenario, the state of the grazing camps within the selected quaternary catchments was left unchanged, and the status quo was maintained. This was achieved by extracting the 8 day MODIS LAI value (Appendix B) of each of the selected quaternary catchments (S20C and S50E) for the period of 2009. Quaternary catchments S20C and S50E had high values (40 to 255) that had to be re-classed as zeros before extracting the 8 day MODIS LAI values. According to the information available on (http://cybele.bu.edu/courses/estonia/modis-lab-03-ascii.pdf) MODIS LAI/FPAR products distributed in the HDF format (raw/unfiltered), do not only have LAI and FPAR data layers, but also several quality control layers, which include pixel-by-pixel data about cloud, aerosol, snow contamination, algorithm path, overall product quality, etc. Therefore it depends on users to design their custom data mask that will suit their application to filter out the undesired data (http://cybele.bu.edu/courses/estonia/modis-lab-03-ascii.pdf) accessed on 04.01.11.
The mean annual $ET_a$ for each selected quaternary catchment (S20C and S50E) was calculated using Eq. (3). $ET_0$ used in the calculations was obtained from Qamata AWS and confirmed using PM-FAO56 and the maximum LAI value of quaternary catchments S20C and S50E for the period of 2009 was used.

### 3.4.2. Scenario 2

For Scenario 2, MODIS pixels representing 28 km and 14 km were selected in catchments S20C and S50E respectively and the LAI of these pixels was doubled. This scenario represented a state where the standing above-ground biomass would be increased.

### 3.4.3. Scenario 3

For scenario 3, the LAI of the pixels that were selected for scenario 2 was halved. This scenario represented a state where the above-ground biomass would be decreased. Then actual annual ET for each scenario was calculated from the function used to calculate $ET_a$ for the 1km pixels.
CHAPTER 4

RESULTS

4.1. ET ESTIMATES IN TEST PIXELS IN CATCHMENTS Q91C AND P10A

4.1.1. The annual ET\textsubscript{MODIS} for 2000 – 2010

The mean annual ET over a ten year period (2000 – 2010) was higher by 78 mm in the high biomass km pixel (P10A) than that obtained in the low biomass km pixel (Q91C). In 2001 the P10A 1 km pixel had the highest annual ET estimate of 610.3 mm, while highest annual ET for the Q91C 1 km pixel was observed in 2006 (534.7 mm). Then the lowest annual ET estimates were observed in 2010 for both 1 km pixels (P10A: 333.1 mm and Q91C: 266.2 mm). \(ET_0\) was highest in 2004 with 1080.7 mm and the lowest \(ET_0\) was 724.5 mm in 2010 the same year both P10A and Q91C 1 km pixel had the lowest annual ET estimates (Table 4). A consistent decrease in annual ET estimates was noticed on both km pixels starting from 2007 to 2010. The Standard Deviation of the mean annual ET\textsubscript{MODIS} is shown in Figure 9a and 9b for the P10A and Q91C 1 km pixels for the period of 2000 -2010.

<table>
<thead>
<tr>
<th>Year</th>
<th>(ET_{\text{MODIS}}) (mm)</th>
<th>(ET_0) (mm)</th>
<th>Rain (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
<td>345.9</td>
<td>448.5</td>
<td>1039.1</td>
</tr>
<tr>
<td>2001</td>
<td>460.2</td>
<td>610.3</td>
<td>1046.8</td>
</tr>
<tr>
<td>2002</td>
<td>450.1</td>
<td>566.5</td>
<td>1072.5</td>
</tr>
<tr>
<td>2003</td>
<td>331.5</td>
<td>433.2</td>
<td>1076.4</td>
</tr>
<tr>
<td>2004</td>
<td>340</td>
<td>420.6</td>
<td>\textbf{318.2}</td>
</tr>
<tr>
<td>2005</td>
<td>501</td>
<td>517</td>
<td>1049.2</td>
</tr>
<tr>
<td>2006</td>
<td>\textbf{534.7}</td>
<td>545.6</td>
<td>979.8</td>
</tr>
<tr>
<td>2007</td>
<td>445.7</td>
<td>510.9</td>
<td>1072.2</td>
</tr>
<tr>
<td>2008</td>
<td>376.4</td>
<td>462.8</td>
<td>1065.0</td>
</tr>
<tr>
<td>2009</td>
<td>304.5</td>
<td>385.2</td>
<td>924.1</td>
</tr>
<tr>
<td>2010</td>
<td>\textbf{266.2}</td>
<td>333.1</td>
<td>\textbf{724.5}</td>
</tr>
<tr>
<td>Mean</td>
<td>\textbf{396.01}</td>
<td>\textbf{475.79}</td>
<td>\textbf{1011.84}</td>
</tr>
</tbody>
</table>
Figure 9(a): Standard Deviation of the mean annual ET of P10A (HB) 1 km pixel for the period of March 2000 – December 2010.

Figure 9(b): Standard Deviation of the mean annual ET of Q91C (LB) 1 km pixel for the period of March 2000 – December 2010.

The ANOVA indicated that the mean annual $\text{ET}_{\text{MODIS}}$ estimates for the period of 2000 – 2010 were significantly different ($p<0.05$) for both P10A and Q91C 1 km pixels. The results for the Post Tukey-HSD Post-hoc test are shown in Table 5(a) for P10A 1 km pixel and in Table 5(b) for Q91C 1 km pixel. The p values in bold show which years the mean annual $\text{ET}_{\text{MODIS}}$ estimates were significantly different for a particular year. For example in Table 5 (a), in year 2000 the mean annual $\text{ET}_{\text{MODIS}}$ was significantly different to the mean annual $\text{ET}_{\text{MODIS}}$ obtained for 2004, 2009 and 2010.
Table 5(a): p values showing significant difference for P10A (HB) 1 km pixel for the period of 2000 - 2010

<table>
<thead>
<tr>
<th>Years</th>
<th>2000</th>
<th>2001</th>
<th>2002</th>
<th>2003</th>
<th>2004</th>
<th>2005</th>
<th>2006</th>
<th>2007</th>
<th>2008</th>
<th>2009</th>
<th>2010</th>
</tr>
</thead>
<tbody>
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<td>2000</td>
<td>0.208511</td>
<td>0.976649</td>
<td>0.069979</td>
<td>0.016421</td>
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<td>0.999973</td>
<td>0.999972</td>
<td>0.518326</td>
<td>0.000164</td>
<td>0.000015</td>
<td></td>
</tr>
<tr>
<td>2001</td>
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</tr>
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<td>0.985781</td>
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</tr>
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<td>0.000015</td>
<td>0.000127</td>
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</tr>
</tbody>
</table>

Table 5(b): p values showing significant difference for Q91C (LB) 1 km pixel for the period of 2000 - 2010

<table>
<thead>
<tr>
<th>Years</th>
<th>2000</th>
<th>2001</th>
<th>2002</th>
<th>2003</th>
<th>2004</th>
<th>2005</th>
<th>2006</th>
<th>2007</th>
<th>2008</th>
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<th>2010</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
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<td>0.133685</td>
<td>0.260383</td>
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</tr>
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<td>0.095755</td>
<td>0.813318</td>
<td>0.938400</td>
<td>0.000039</td>
<td>0.000015</td>
<td>0.150156</td>
<td>0.149954</td>
<td>0.001369</td>
<td></td>
</tr>
<tr>
<td>2009</td>
<td>0.000164</td>
<td>0.000015</td>
<td>0.000015</td>
<td>0.992407</td>
<td>0.953027</td>
<td>0.000015</td>
<td>0.000015</td>
<td>0.000015</td>
<td>0.149954</td>
<td>0.954090</td>
<td></td>
</tr>
<tr>
<td>2010</td>
<td>0.000024</td>
<td>0.000015</td>
<td>0.000015</td>
<td>0.331827</td>
<td>0.173763</td>
<td>0.000015</td>
<td>0.000015</td>
<td>0.001369</td>
<td>0.954090</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

4.1.2. Validation of ET$_{MODIS}$ with ET$_{OTC}$

The ET$_{OTC}$ obtained using the Kestrel 3000 anemometer for the growing season of (September 2009 – March 2010) were negative for both HB (P10A) and LB (Q91C) 1 km pixels, on most occasions it was measured in the field. No positive correlation was observed when ET$_{OTC}$ was compared ET$_{MODIS}$ for those days, and these results were discarded. When using both the Kestrel 3000 anemometer and the Vaisala HMP45C RH and TEMP probe, positive results were obtained. Since the daily ET$_{OTC}$ estimates was a sum of ±10 hours of the day it was measured, it was deemed important to also calculate the hourly ET$_{MODIS}$ for the same hours as ET$_{OTC}$ on those same days. Figure 10(a - e) and 11 (a – e) show how hourly ET$_{OTC}$ (the actual) performed in the field compared to the hourly ET$_{MODIS}$ (the predicted) for the same days.
Figure 10(a - e): P10A (HB) hourly performance of $ET_{OTC}$ and $ET_{MODIS}$ on the days ET was measured in the field with the OTC in the growing season of 2010 – 2011. The solid line represents $ET_{MODIS}$ while the dotted line represents $ET_{OTC}$. 
Figure 11(a - e): Q91C (LB) hourly performance of ET_{OTC} and ET_{MODIS} on the days ET was measured in the field with the OTC in the growing season of 2010 – 2011. The solid line represents ET-MODIS while the dotted line represents ET-OTC.
A good relationship was observed between the daily $\text{ET}_{\text{OTC}}$ and $\text{ET}_{\text{MODIS}}$ as shown by the linear regression analysis in Figure 12 with the $R^2 = 0.7065$.

\[
y = 0.6601x + 0.8322
\]

\[R^2 = 0.7065\]

**Figure 12:** Regression Analysis showing the relationship between $\text{ET}_{\text{OTC}}$ and $\text{ET}_{\text{MODIS}}$ for all days.

The relationship observed confirms that $\text{ET}_{\text{OTC}}$ closely approximates $\text{ET}_{\text{MODIS}}$ and that the OTC can be used as a tool to validated MODIS LAI ET on clear, low winds and sunny days.

**Figure 13(a):** Daily estimates of $\text{ET}_{\text{MODIS}}$ and $\text{ET}_{\text{OTC}}$ and the rainfall measured from P10A (HB) 1 km pixel for the growing season of October 2010 – March 2011.
Figure 13(b): Daily estimates of ET$_{\text{MODIS}}$ and ET$_{\text{OTC}}$ and the rainfall measured from Q91C (LB) for the growing season of October 2010 – March 2011.

4.1.3. The actual leaf area index

Table 6: The actual LAI, ET-MODIS, ET-OTC and ET$_0$

<table>
<thead>
<tr>
<th>Dates</th>
<th>Sites</th>
<th>LAI</th>
<th>ET$_0$</th>
<th>ET-MOD</th>
<th>ET-OTC</th>
</tr>
</thead>
<tbody>
<tr>
<td>15/11/2010</td>
<td>Q91C</td>
<td>0.3576</td>
<td>3.09369</td>
<td>2.141785</td>
<td>2.17</td>
</tr>
<tr>
<td>26/11/2010</td>
<td>P10A</td>
<td>0.3029</td>
<td>3.20062</td>
<td>2.560499</td>
<td>2.09</td>
</tr>
<tr>
<td>27/11/2010</td>
<td>Q91C</td>
<td>0.2698</td>
<td>2.639897</td>
<td>2.233759</td>
<td>2.44</td>
</tr>
<tr>
<td>30/11/2010</td>
<td>P10A</td>
<td>0.3954</td>
<td>3.14436</td>
<td>2.575494</td>
<td>2.86</td>
</tr>
<tr>
<td>01/12/2010</td>
<td>Q91C</td>
<td>0.6423</td>
<td>3.12263</td>
<td>2.643071</td>
<td>2.35</td>
</tr>
<tr>
<td>15/02/2011</td>
<td>P10A</td>
<td>0.3868</td>
<td>3.165128</td>
<td>1.978205</td>
<td>2.36</td>
</tr>
<tr>
<td>20/02/2011</td>
<td>Q91C</td>
<td>0.4114</td>
<td>2.732352</td>
<td>2.146848</td>
<td>2.29</td>
</tr>
<tr>
<td>23/02/2011</td>
<td>P10A</td>
<td>0.56</td>
<td>2.989981</td>
<td>1.494991</td>
<td>1.66</td>
</tr>
<tr>
<td>01/03/2011</td>
<td>Q91C</td>
<td>0.3669</td>
<td>2.706708</td>
<td>1.353354</td>
<td>1.78</td>
</tr>
</tbody>
</table>
4.2. LAND USE SCENARIOS FOR HB and LB QUATERNARY CATCHMENTS

The mean annual ET\textsubscript{MODIS} estimates obtained in quaternary catchments S50E (high biomass site) were higher by 111mm than those obtained for quaternary catchment S20C (low biomass site). When comparing between the scenarios it was observed that scenario 2 which is the scenario in which the LAI values were doubled had higher ET\textsubscript{MODIS} estimates than both scenario 1 and 3. Table 7 shows the differences in the ET\textsubscript{MODIS} between the two quaternary catchments (S50E and S20C) and between the scenarios 1, 2 and 3.

Table 7: Scenario 1-3 for the S50E (HB) and S20C (LB) quaternary catchments in 2009

<table>
<thead>
<tr>
<th>Catchments</th>
<th>Scenario 1 ET(mm)</th>
<th>Scenario 2 ET(mm)</th>
<th>Scenario 3 ET(mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>S50E</td>
<td>426.83</td>
<td>436.36</td>
<td>420.77</td>
</tr>
<tr>
<td>S20C</td>
<td>315.81</td>
<td>324.84</td>
<td>314.40</td>
</tr>
</tbody>
</table>

4.2.1. Accumulated ET\textsubscript{MODIS} for scenario 1-3

The accumulated rates of MODIS LAI ET\textsubscript{a} for scenario 1 – 3 are shown in figures 14a and 15a. The differences between the scenarios for quaternary catchments S50E and S20C are clearly shown in Figure 14b and 15b.
**Figure 14(a):** Accumulated daily ET\(_{\text{MODIS}}\) for quaternary catchment S50E showing the difference in ET between scenario 1, 2 and 3.

**Figure 14(b):** Accumulated ET\(_{\text{MODIS}}\) showing only the 12\(^{th}\) month of 2009 for the high biomass 1 km pixel.

**Figure 15(a):** Accumulated daily ET\(_{\text{MODIS}}\) for quaternary catchment S20C showing the difference in ET between scenario 1, 2 and 3.
Figure 15(b): Accumulated ET$_{MODIS}$ showing only the 12$^{th}$ month of 2009 for the low biomass 1 km pixel.
CHAPTER 5
DISCUSSION AND CONCLUSION

In this study ET estimates of the grasslands of the Eastern Cape were obtained using MODIS LAI as an input into the Penman-Monteith equation (Palmer and Weideman 2011) and validated using an Open Top Chamber. The aim was to find out how the different management regimes or lack thereof of the grassland catchment affected the rates of ET and ultimately the runoff rates. It was assumed that the mismanagement of the grassland catchments of the Eastern Cape has had an adverse effect on the amount of runoff received annually as it leads to overgrazed land with little biomass cover and soil erosion. Also these overgrazed lands lead to invasion of alien species that usually utilize more water than the native species and lead to high ET rates (Blignaut, Marais et al. 2007). The effect of these alien species on ET rates was not explored in this study. Two different grazing strategies (High and low biomass cover) were explore from quaternary catchments selected from Bushmans River Primary Catchment and from Kei River Primary Catchment. First the trend observed for the annual ET estimates over a 11 year period (2000 – 2010) showed that ET has been decreasing in the last four years from both the high and low biomass sites. It was also observed that the high biomass sites had higher ET estimates than the low biomass sites. These results were in agreement with the observation made for the land-use scenarios. High ET rates have been previously associated with the decrease in water supply; however in this study we argue that if these ET estimates are obtained from good condition grasslands then an improvement to the runoff rates is expected.

5.1 THE ANNUAL ET_{MODIS} FOR 2000 – 2010

The results obtained showed that the mean annual ET_{MODIS} obtained for the high biomass 1 km pixel was higher than the mean annual ET_{MODIS} obtained for the low biomass 1 km pixel over the period of 2000 – 2010. When high ET estimates are obtained it is expected that the runoff rates will be reduced. However in cases of grassland catchments though high biomass cover results in high ET estimates which may reduce runoff rates as more water is lost to the atmosphere, high biomass cover also acts as a sponge that catches the summer rainfall and allow the water to be infiltrated in the soil and then be gradually released to the streams. Also high biomass cover reduces the magnitude of storm flows. Whereas in areas with low
biomass cover, high runoff rates may result, however infiltration is reduced and storm flows are increased. This also results in soil erosion and siltation of the dams which decreases the capacity of the dams to store water. (DWA 1986; Turpie, Marais et al. 2008).

It was also observed that from 2007 up to 2010 there was a consistent decrease of ET for both the low and high biomass 1 km pixel. In an ideal situation decrease in ET rates would mean that less water is lost to the atmosphere and therefore runoff and recharge rates are improved. However where grasslands are concerned this could be true in cases where the high biomass cover is moribund, and the dead grass still allow water to infiltrate the soil. Therefore in this case the decrease of the ET rates is not seen as positive since it was observed for both the high biomass (representing good condition grassland) and the low biomass (representing bad condition grassland).

Factors that could have contributed to this consistent decrease of ET$_{MODIS}$ are at this stage uncertain. However it was noted that for the same period (2007 – 2010), ET$_0$ was also consistently decreasing. The concept of ET$_0$ has been used in many studies as a method for estimating ET (Jenson full citation) and in this study ET was estimated by adjusting ET$_0$ using the 8 day MODIS LAI data (Palmer and Weideman 2011). Therefore the consistent decrease in ET$_0$ can be viewed as one of the contributing factors to the decrease observed for ET$_{MODIS}$. Hansen (1984) defined ET$_0$ as the maximum E and T from a surface with sufficient water available and the amount of energy available for ET is greatly influenced by the surface condition (i.e. whether the surface is well covered with vegetation and well watered or it is lightly covered and water stressed) (Hansen 1984). Therefore it could be argued that over the past four years both the high and the low biomass 1 km pixels have experienced a consistent change in the vegetation cover and soil water availability.

Reynolds et al., (2000) stated that in semi-arid regions where plant growth is limited by water availability, variations in ET rates can be controlled by variability in precipitation at daily to inter-annual timescales (Reynolds, Kemp et al. 2000). The lowest annual rainfall was obtained in 2003 and the ET$_0$ rates obtained for that year are higher than those obtained for 2007 – 2010 and this complicates the effects of rainfall on ET$_0$. However an answer may be found in the study done by Ferretti et al., (2003) that examined the effect of precipitation variations on the separate fluxes of E and T in short grass steppe. They estimated the average daily E and T fluxes for each time interval by multiplying the fraction of each flux by the total moisture for that interval. The results they obtained showed that for two consecutive years (2000 and 2001) that had similar annual rainfall, the rates for one year were almost
double the rates obtained for the other year. They attributed this to the timing of precipitation. The dry spring led a decrease in biomass and when the rain finally came in late August of 2000 it was not used by the short grass. Whereas the spring rains of 2001 increase the biomass leading to high T rates. This then indicates that vegetation growth is not only affected by the rainfall amounts but also by when it is occurring (Ferretti, Pendall et al. 2003; Overpeck, Bartlein et al. 1991).

5.2. VALIDATION OF ET\textsubscript{MODIS} ESTIMATES

5.2.1. The variations in the hourly ET\textsubscript{OTC}

It was observed that for most of the days ET was measured in the field with the OTC, the curve of ET\textsubscript{OTC} was not as smooth as the one plotted for ET\textsubscript{MODIS}. ET\textsubscript{MODIS} was obtained by adjusting ET\textsubscript{0} (a climatic parameter that is estimated from the weather data) by MODIS LAI data (a Remote Sensing technique that measures ET indirectly) and therefore it represented the predicted ET in this study. While ET\textsubscript{OTC} represents the actual ET as it was measured directly using a micro-meteorological method (Allen, Pereira et al. 1998; Breda 2003). ET is the sum of two processes, the direct E of water from open water bodies and moist soils and transpiration through the stomatal pores of leaves (Eamus, Hatton et al. 2006). According to Glenn et al., (2007) T of the plants accounts for over 80% of the terrestrial ET. Transpiration depends on the opening and the closing of the stomata which is affected by the increase or decrease in air temperature, RH and U. The decrease in transpiration means that less water is released through the stomata and this also means that the rate of ET will decrease (Eamus, Hatton et al. 2006; Glenn, Huete et al. 2007b). Therefore as the air temperature RH and U fluctuate throughout the day, the same occurs in the ET rates. This would explain the fluctuating rates of the hourly ET\textsubscript{OTC}.

5.2.1. Comparison between ET\textsubscript{OTC} and ET\textsubscript{MODIS}

ET estimations were obtained in this study using the MODIS LAI as an input into the Penman-Monteith equation (Palmer and Weideman 2011) and validated using the Open Top Chamber. The relationship between the ET\textsubscript{MODIS} and ET\textsubscript{OTC} observed from the linear regression analysis ($r^2 = 0.7065$) showed that ET\textsubscript{OTC} closely approximates ET\textsubscript{MODIS} and thus the OTC can be used to validate MODIS LAI ET. ET is a major component of water cycle and obtaining improved estimates of ET is important for determining water budget in and
around a region. (Leuning, Zhang et al. 2008; Tsouni, Kontois et al. 2008). Remotely sensed data is one of the best ways to determine spatially distributed E fluxes due to their spatial and temporal continuity.

The model of estimating ET using MODIS LAI data as an input into the P-M equation was first proposed by Cleugh et al., (2007) in a study where the 8-day evaporation at 1-km resolution was estimated using gridded meteorological fields and the P-M equation, in which the surface conductance ($G_s$) was a function in the MODIS LAI (Cleugh, Leuning et al. 2007). This was later modified by Mu et al., (2007) who revised the simple model for $G_s$ to account for the response of stomata to temperature and humidity deficit of the air and the results obtained were in good agreement between predictions of ET by the P-M equation and evaporation measurements from a number of flux stations in the USA (Mu, Heinsch et al. 2007). Both models for $G_s$ were replaced by Leuning et al., (2008) with a biophysical, two-parameter model for surface conductance that accounted for the sensitivity of the stomatal conductance to the shortage of humidity in the atmosphere and to light. Excellent estimates of evaporation were obtained from this new model for $G_s$ when tested against data sets from 15 flux station sites covering a wide range in climate and vegetation types globally. These results show that the P-M equation is a biophysically sound and robust framework for estimating daily E at regional to global scales using remotely sensed data (Leuning, Zhang et al. 2008).

Then Zhang et al., (2008) examined the potential of using remote sensing ET to predict long-term runoff by comparing the runoff values derived from water balance estimates with measured runoff at the 120 gauged catchments in the period of 2001–2005. Their results showed a good comparison between the mean annual run-off estimates calculated by calibrated remote sensing ET and the recorded values. These results lead to a conclusion that remotely sensed ET can be used with rainfall-runoff models to improve the accuracy of estimated runoff in ungauged catchments(Zhang, Chiew et al. 2008).

The model used in this study to estimate the actual ET (Palmer and Weideman 2011) which is also a modification of the model developed by Cleugh et al., (2007) was developed around two different data sets (EC and LAS) from different sites (KNP and Malekutu) and in both cases a good relationship was observed between the ET obtained from either data set and ET obtained using the model. The OTC used in this study is a third instrument to be used to ground truth the ET estimates obtained using the model and the advantage of OTC over the
other two instruments is that it is easy to construct, replicable and replaceable, easy to transport and inexpensive. Therefore obtaining $ET_{OTC}$ results that closely approximate $ET_{MODIS}$ presents a cheaper and uncomplicated alternative of determining the effects of the different grazing regimes on ET and ultimately on runoff rates using MODIS LAI products.

5.3. LAND-USE SCENARIOS

The results obtained for the land-use scenario showed that the high biomass quaternary catchment had higher ET than the low biomass quaternary catchment and also when comparing between the scenarios, ET estimates were higher in the scenario with the pixels that had doubled LAI values. These results are in agreement with results obtained for the 1 km pixels in which high ET estimates were obtained for the high biomass km pixel than the low biomass 1 km pixel.

The purpose of doing the land use scenarios was to show the impact that changes in land use activity can have on ET and ultimately on runoff rates. The first scenario was the original state of the quaternary catchment. The second scenario showed how ET would be affected if a small area within the quaternary catchment can have an increased LAI and as shown by the results ET would be increased as well. Then the third scenario is the opposite of scenario 2 and resulted in a decrease in ET. The ideal state would be one that has reduced rates of ET (i.e. scenario 3) as this would result in high rates of runoff according to the equation below:

$$\text{Runoff} = \text{Precipitation} - ET$$

However in this case the reduction of LAI in the selected area is assumed to mean that degradation in that area is increasing. This degradation might lead to loss of the above ground biomass which would increase storm flows leading to dam siltation and also reduce surface water infiltration. Therefore though in scenario 3 ET was reduced, increased water supply is not guaranteed instead there is a high possibility that not only would the runoff and recharge rates be decreased but the capacity of the dams would also be decreased due to siltation.

This then denotes that scenario 2 is an ideal state, though high ET rates were obtained as in the case of the high biomass 1 km pixel, the high grass cover can act as a sponge that catches the summer rainfall and allows it to infiltrate the soil and be gradually released to the dam. Also if this state is maintained over the years the above grass can be moribund which could reduce ET rates and continue to allow the infiltration of water (Turpie, Marais et al. 2008) (Hogan 2010).
Creating the land use scenarios using the MODIS LAI data was important in this study as this has presented an opportunity to demonstrate how MODIS LAI data can be used to quantify the improvements or further degradation of the vegetation cover within the catchments. This would greatly assist in the application of the payment for ecosystem services as it a step closer to knowing how much water can be expected when the grass cover is improved. The current state of the grasslands of South Africa is of great concern; in arid and semi arid climates overgrazing of rangelands by livestock amongst other things add to land degradation resulting in decline of the soil structure that reduced infiltration and water holding capacity (WMO 1998). This then denotes that there is a need for land use activities to change within the grassland catchments that are also used as rangelands. Livestock owners need to be advised to apply effective management regimes in order to improve the grass cover within the grassland catchments they inhabit to ensure gradual supply of water to the catchments. Then through the PES they can receive direct compensation from the downstream water users for improving the grass cover to improve water supply. In cases where the ecosystem services paid for involve changing land use activities, the costs are higher as opposed to when they are made for retaining them (Engel, Pagiola et al. 2008). If the PES can be successfully applied in areas that require change in land use activity in order to improve the above ground biomass which in turn will improve the rates of runoff and recharge then the livestock owner who in this case will be the service providers can expect to be paid more. Although the aim of developing the PES is firstly to meet the environmental objectives and to ensure that natural resources are well managed, the effect on income redistribution is also an important side objective. This is more suited to developing countries where most of the people providing the ecosystem services fall below the poverty line (Hengsdijk, Groot et al. 2009).

In the Eastern Cape there is need for land use activities to change within the grassland catchments that are also used as rangelands. Currently these grassland catchments have less grass cover due to inappropriate land practices such as overgrazing. This leads to rapid runoff that silts the dam and also removes the top soil. Livestock owners need to be advised to apply effective management regimes in order to improve the grass cover within the grassland catchments they inhabit to ensure gradual supply of water to the catchments and to reduce their livestock numbers. The PES program that had similar objectives (i.e. the improvement of water supply through change in land use activity) was successfully applied by the Maloti Drakensberg Transfrontier Project (MDTP) in the Thukela and Mzimvubu catchments where they managed to encourage the land owners to change certain land use activities in
order to improve water supply. Their results confirmed that application of proper managements system can increase infiltration, reduce summer storm flow and increase winter base flow. These results lead the MDTP to conclude that providing an incentive for land owners for their proper management of the grasslands catchment is efficient and equitable (Mander, Blignaut et al. 2007).

Though the PES programs has shown the potential for ‘double dividend’ payoffs in terms of biodiversity conservation and poverty reduction (van Wilgen, Le Maitre et al. 1998), studies done in Costa Rica and Mexico have indicated that participation in the program is not easy for most poor households (Miranda, Porraz et al. 2003; Ortiz and Kellenberg 2002; Zbinden and Lee 2005). This is due to the fact that participation in the PES program is limited to households that can provide the desired services and only these household are eligible to receive payment. Constraints such as the formal tenure requirements discriminate against poor farmers who would like to benefit from the program but have no formal rights to the land (Wunder 2005). For instance in Latin America, Southgate and Wunder (2009) found that the PES program was more successful where the service providers had secured land tenures and that cooperation and good compensation were ensured. An example of this is taken from the study by (Varga, Harwood et al. 2009). They found that in one of their case studies, knowing who owned the property rights of land upstream, enabled the service buyers to enter into an appropriate contract with the service providers (Varga, Harwood et al. 2009).

Another constraint is that in terms of payments, the costs to the ecosystem service buyer are much higher when dealing with many smallholder or a land that is collectively owned by communities compared to when paying only a few big landowners (Smith and Scherr 2002). In this study, where the land users are livestock owners, the above constraint could be experienced since in most cases livestock owners usually own land collectively. Secondly encouraging livestock owners to reduce their livestock numbers would require the ecosystem service buyer or downstream users to ensure that the benefits of participating in the program are higher than the current benefit they receive from their livestock (Pagiola and Platais 2007).

There is another aspect that needs to be looked at that of the differences between private PES programs and national government financed PES programs. In private PES programs land owners or ecosystem service providers can be paid directly by the users of the specific ecosystem service due to a limited geographic scope. For example in a single watershed the
downstream hydrological service users can directly pay the upstream service providers (Blackman and Woodward 2009). The advantage of private PES programs is that the ecosystem service buyer and the ecosystem service seller are directly involved meaning they can ensure the efficiency of the system and they are in a better position to renegotiate the agreements (Varga, Harwood et al. 2009). Then there are those that have a national scope, these are financed by government who represents the users of that specific ecosystem service throughout the country. This type of PES program has been applied in countries like China, Costa Rica, Mexico and South Africa, while some are still planning to implement them (Pagiola 2007; Wunder 2005). The main advantage of the government financed PES programs is economies of scale, where national programs are able to spread costs over a large number of agents, and reduce some costs. However these are very inefficient, firstly because as opposed to the private PES programs the national governments are not directly involved and do a poor job of identifying providers of important ecosystem services and monitoring the efficiency of the program. Secondly, they depend on national tax revenues and international funds they cannot guarantee long term financing (Blackman and Woodward 2009). One way that could correct this is if the individual service users of a specific ecosystem service would voluntarily contribute to the system by paying for them to the government administrators.

There is a lot in the literature about how the PES programs work for different scenarios and in this study the practical application of the program has not been initiated yet. Only the technique that demonstrate how to identify areas where overgrazing could lead to reduced water supply have been explored. It is not yet clear whether the private PES program or the national government financed PES program will be suitable in this kind of study.

5.4. CONCLUDING REMARKS

It is clear that South Africa has a major shortage of water and the country is at a stage where water demand is exceeding water availability due to population growth (DEAT 2000) (Mukheibhir 2007). Previously water demands were met through a complex system of engineering supply side solutions which included major inter-basin transfer and water pumping schemes (Smakhtin, Ashton et al. 2001). However these have become costly and the water available for that has reduced because 12 of South Africa's 19 water catchment areas receive insufficient rainfall (Blignaut and De Wit 2004). This has increased the need to
explore cheaper and less complicated alternatives for increasing and conserving water supplies in ungauged catchments (Ashton and Seetal 2002). This study set out to explore the effect of two different grazing strategies on ET rates using MODIS LAI data since changes in ET rates also affect the quantities of water supply. The results obtained showed that high biomass sites had higher ET rates and though high ET rates are associated with low runoff rates it was argued in this study that where grasslands are concerned the opposite is true. Since grasslands allow summer rainfall to infiltrate the soil and be gradually released to the dams, it is recommended that the grass cover must be high. As opposed to the previous ways of meeting water demand the use of the MODIS LAI product as an input into the Penman Monteith to estimate the ET rates is a much cheaper and less complicated alternative. Also it’s validation using an open top chamber that is inexpensive, easy to construct and transport, makes the process highly replicable. The findings of this study pave a way for the application of the PES model and the approach used has the potential to quantify the benefits to down-stream water users of improving above-ground biomass in catchments.

5.5 FUTURE RESEARCH

In using contrasting land-use scenarios, this study demonstrated that when LAI is decreased, ET estimates also decrease. Although the low biomass scenario can be viewed as a situation when less water is lost to the atmosphere, it was important to note that in most cases the reduction of LAI is associated with increased run-off, higher silt loads and rangeland degradation. On the other hand, increased LAI, which is accompanied by increased ET, could mean that the rangeland is in a healthy state and can lead to the reduction of storm flows and dam siltation since infiltration of the rainfall occurs. These results high-light the challenges associated with balancing biomass production, vegetation cover, soil conservation and water yield (quality and quantity). Determining the income benefit that livestock owners can receive through the PES model for the improvements made to their rangelands will require further research. This will require a clear understanding of how to quantify the annual biomass increments from the selected sites and the impact on water supply. This can be achieved through the development of techniques for modelling net primary production and then water use efficiency. Degradation or improvement of the rangeland can be determined by monitoring the changes in water use efficiency (Holm, Watson et al. 2003). Several authors (Palmer et al 2010, Snyman 1986, 1994 and Holm, Watson et al. 3003) have
prepared water use efficiency values for selected vegetation types in Southern Africa, and this research should be pursued further.
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APPENDICES:

Appendix 1: Differences between quaternary catchment S50E (HB) and S20C (LB)

Description of the veld condition of quaternary catchment S50E:
1. High biomass cover dominated by *Eragrostis plana*.
2. The average disc pasture meter reading from three sites was 9.5 cm.
3. Soil is red and deep
4. Fire is used
5. No camp resting
6. There are some Black wattles and Pine plantations

The summary of the point to turf distance:

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<p>| S50E-HB Count of Distance | 49 |
| S50E-HB Average of Distance | 0.897959184 |
| Total Count of Distance | 49 |
| Total Average of Distance | 0.897959184 |</p>
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Photos showing the biomass cover of Quaternary Catchment S50E (HB):
Description of the veld condition of quaternary catchment S20C:
1. Low biomass cover dominated by *Euryops annae* and some *Cynodon dactylon*.
2. No disc pasture meter readings were taken from this quaternary catchment due to its low biomass.
3. The soil is shale and shallow.
4. No fire used.
5. No camp resting.
6. No black wattle or pine plantations were observed.

The summary of the point to turf distance:

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### Co-ordinates: 27.44275 -31.74145

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Photos showing the biomass cover of Quaternary Catchment S50E (HB):
Appendix 2: A Macro File showing how Scenarios 1 – 3 were created in IDRISI 32

Scenario 1

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OVERLAY x
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Scenario 3

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