Extending periodic eddy covariance latent heat fluxes through tree sapflow measurements to estimate long-term total evaporation in a peat swamp forest

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Abstract

A combination of measurement and modelling was used to find a pragmatic solution to estimate the annual total evaporation (ET) from the rare and indigenous Nkazana Peat Swamp Forest (PSF) on the east coast of Southern Africa to improve the water balance estimates within the area. Total evaporation was measured during three window periods (between seven and nine days each) using an eddy covariance (EC) system on a telescopic mast above the forest canopy. Sapflow of an understory and an emergent tree was measured using a low maintenance heat pulse velocity system for an entire hydrological year (October 2009 to September 2010). An empirical model was derived, describing the relationship between the observed ET of the Nkazana PSF measured during two of the window periods ($R^2 = 0.92$ and $0.90$) which, overlapped with sapflow measurements, thereby providing hourly estimates of predicted ET of the Nkazana PSF for a year, totalling 1125 mm (while rainfall was 650 mm). In building the empirical model, it was found that including the understory tree sapflow provided no benefit to the model performance. In addition, the observed emergent tree sapflow relationship with observed ET between the two field campaigns was consistent and could be represented by a single empirical model ($R^2 = 0.90$; RMSE = 0.08 mm).

During the window periods of EC measurement, no single meteorological variable was found to describe the Nkazana PSF ET satisfactorily. However, in terms of evaporation models, the hourly FAO56 Penman–Monteith equation best described the observed ET from EC during the August 2009 ($R^2 = 0.75$), November 2009 ($R^2 = 0.85$) and March 2010 ($R^2 = 0.76$) field campaigns, compared to the Priestley–Taylor model ($R^2 = 0.54$, 0.74 and 0.62 during the respective field campaigns). From the empirical model of ET and the FAO56 Penman–Monteith equation, a monthly crop factor ($K_c$) was derived for the Nkazana PSF providing a method of estimating long-term swamp forest ET from meteorological data. The monthly crop factor indicated two distinct periods. From February to May, it was between 1.2 and 1.4 compared with June to January, when the crop factor was 0.8 to 1.0. The derived monthly $K_c$ values were verified as
accurate (to one significant digit) using historical data measured at the same site, also using EC, from a previous study.

The measurements provided insights into the microclimate within a subtropical peat swamp forest and the contrasting sapflow of emergent and understory trees. They showed that expensive, high maintenance equipment can be used during manageable window periods in conjunction with low maintenance systems, dedicated to individual trees, to derive a model to estimate long-term ET over remote heterogeneous forests. In addition, the contrast in ET and rainfall emphasises the reliance of the Nkazana PSF on groundwater.

1 Introduction

Severe water scarcity in parts of South Africa has threatened the health of internationally recognised environmental areas such as the iSimangaliso Wetland Park, a UNESCO world heritage site. To optimise the management of the water balance and understand the functioning of the area, there has been a need to quantify the water-use of the dominant vegetation types of the Park such as the endangered Peat Swamp Forests (Grundling et al., 1998; Clulow et al., 2012), a dominant plant type of the Mfabeni Mire. However, little is known about the water-use characteristics of the species diverse Peat Swamp Forests (PSFs) both locally and internationally in terms of model parameterisation. Despite significant improvements to measurement techniques over vegetated surfaces (Savage et al., 1997), these have not been of benefit for PSFs due to their remote and inaccessible nature. In addition, well documented extreme events (such as the Demoina floods in 1987) pose a real threat in the area. Sophisticated instruments are unfortunately vulnerable to damage and malfunction in such environments and PSFs are therefore not good locations for long-term deployment of sensitive equipment, a challenge facing researchers internationally and particularly in developing countries.
There are numerous, complex evaporation sources, which interact and contribute to total evaporation (ET) in the Nkazana PSF. The areas of open water fluctuate, depending on groundwater levels. Open water evaporation is well described from the early work of Penman (1948) to the more recent work of Finch (2001) and Rosenberry et al. (2007) but none accounts for the effects of dense vegetation cover on radiative shading and the prevention of convection over the water surface by a tall and dense canopy. There are surface evaporation studies of peat (Nichols and Brown, 1980; Koerselman and Beltman, 1988; Lafleur and Roulet, 1992; Thompson et al., 1999; Clulow et al., 2012), but none in the context of a subtropical swamp forest. In addition the vegetated canopy is complex. There is a dense cover of ferns, of which little is understood in terms of transpiration (Andrade and Nobel, 1997). Above the ferns, the tree canopy consists of two levels described below (understory and emergent trees) and there are tree-climbing vines. Estimating the ET of the Nkazana PSF is clearly multifaceted due to its diversity and our lack of understanding of the water-use of the specific plants, together with the potential variation in the evaporative demand within and above the canopy.

Within South Africa, the only comparable study took place over an evergreen indigenous mixed forest in the Southern Cape near the coast. Dye et al. (2008) measured ET using eddy covariance (EC), scintillometry and Bowen ratio over 18 days in total, during three different field campaigns, representing three different seasons within the year. The periods in-between were modelled using the FAO56 Penman–Monteith reference equation of Allen et al. (1998) which generally underestimated ET under high evaporative conditions and overestimated under low evaporative conditions. This was attributed to the assumption of a constant surface resistance. The Penman–Monteith equation (Monteith, 1965) was found to give the best match of predicted to observed daily ET, but required measurements or a sub-model accounting for variable canopy conductance. The more complex WAVES (CSIRO, Canberra, Australia) process-based model simulated canopy growth and water-use processes in much more detail. However, successful parameterisation of the many model inputs was a significant challenge.
and despite their best efforts, the WAVES output revealed an overestimation of daily ET under conditions of low evaporative demand which could not be corrected. They concluded that the best technique for interpolating the periods between the three field campaigns to be the Penman–Monteith equation despite the problem of the variable canopy conductance and recommended that further research into understanding the most appropriate techniques for interpolating measured data was necessary.

Internationally, no studies were found with measurements over a comparable subtropical peat swamp forest. However, Vourlitis et al. (2002) provide a valuable study in which they attempted to measure the long-term ET with an EC system over a tropical forest in Brazil. Despite the proximity to the city of Sinop (offering a nearby base from which maintenance could be conducted), power issues hampered the data collection and EC data was only collected 26% of the time. Meteorological data was therefore used to estimate the latent energy flux (LE) using the Priestley–Taylor expression.

Since the beginning of the FLUXNET project, which was established to compile long-term measurements of water vapour, carbon dioxide and energy exchanges from a global network of EC systems, the problem of complete EC data sets and gap filling of records was recognised and is still an ongoing challenge (Baldocchi et al., 1996, 2001). Farge et al. (2000) found the average data coverage for long-term EC systems to be only 65% due to system failure or data rejection with most of these located in developed countries. Clearly, despite the benefit of EC systems, long-term, continuous records of observed ET data over indigenous subtropical and tropical forests are improbable without significant research budgets allowing daily maintenance, gap filling and processing of data including complex spectral corrections, 3-D corrections and coordinate rotation amongst others (Massman and Lee, 2002; Finnigan et al., 2003; Hui et al., 2004). Intensive, short-term field campaigns, offering reliable, continuous records, during different seasons seems an appropriate strategy to determine the annual cycle of ET, above all at temporary sites. This is particularly the case in South Africa, where theft of equipment and especially batteries from the foot of visible towers is a severe limitation although this is overcome by employing 24 h security guarding.
services to protect the equipment during the short-term measurement periods (Dye et al., 2008). However, this strategy only provides a viable solution if the ET during the in-between periods can be adequately estimated.

Wilson et al. (2001) applied EC and sapflow techniques in a deciduous forest of the south-eastern United States, and found that there was a qualitative similarity between ET_{ec} and tree transpiration. With the recent advances in sapflow measurement techniques and upscaling of individual tree transpiration measurements to canopy ET, it is believed, that sapflow techniques offer a reliable, standalone, long-term solution to estimating ET in uniform tree stands (Hatton and Wu, 1995; Meiresonne et al., 1999; Crosbie et al., 2007). There are however, numerous complexities bringing some doubt as to the accuracy of the absolute sapflow results, such as the anisotropic properties of sapwood (Vandegehuchte et al., 2012), species composition effects (Wullschleger et al., 2001), tree symmetry (Vertessy et al., 1997), radial patterns of sapflow (Cermak and Nadezhdina, 1998) and changes in spatial patterns of transpiration (Traver et al., 2010). In heterogeneous and complex canopies such as the Nkazana PSF described above, sapflow systems alone are impractical for the prediction of stand ET even with the recent advances in process based models of vegetation function such as the Mea-spa model (Duursma and Medlyn, 2012). However, whether it is possible to use the qualitative relationship of sapflow with ET_{ec}, as found by Wilson et al. (2000), remains unknown.

For these reasons, a strategy to provide a measurement and modelling framework was developed and tested, in which detailed water flux measurements were recorded using EC instruments in an indigenous, heterogeneous forest over three window periods in August 2009, November 2009 and March 2010. This minimised the cost and risk of damage to these expensive systems, and provided continuous and reliable data from well-maintained instruments operated by a team of scientists, but were limited to three window periods. Two of these window periods overlapped in time and space with long-term sapflow measurements, and a nearby weather station provided measurements during the full period. The sapflow and weather station systems had lower
maintenance and power requirements, were less delicate, less visible, able to with-
stand the harsh environment, and operated for longer periods unattended (one to two
months) without compromising data quality. The aims were therefore to (1) establish
whether the long-term ET of the Nkazana PSF could be determined by this combination
of EC window periods and long-term sapflow measurements, (2) to provide a means
of modelling the ET of surrounding PSFs from nearby meteorological data and (3) to
investigate the controlling climatic variables and their influence on sapflow as well as
the energy fluxes and microclimate within the swamp forest.

1.1 The study area

The study area is located in Maputaland, South Africa, on the Eastern Shores of the
iSimangaliso Wetland Park. It has held international status as a UNESCO World Her-
itage Site since 1999 (Taylor et al., 2006) and falls within the St Lucia Ramsar Site
designated in 1986 (Taylor, 1991). It is one of the largest protected aquatic systems
in southern Africa and due to its biodiversity and natural beauty, has become an inter-
national tourist destination and now a “regional economic hub” (Whitfield and Taylor,
2009).

The Eastern Shores area has a subtropical climate and lies in a summer rainfall area
(Schulze et al., 2008). It has been reported that “the rainfall gradient westwards from
the coast is strong, with a precipitation at Mission Rocks on the Indian Ocean coastal
barrier dune exceeding 1200 mm yr$^{-1}$ and decreasing to around 900 mm yr$^{-1}$ at Fanies
Island on the western shoreline of the estuary” (Taylor et al., 2006). However, Lynch
(2004) provides mean annual precipitation values of 1056, 844 and 910 mm yr$^{-1}$ from
the nearby Fanies Island, Charters Creek and St. Lucia from a 125 year raster database
and the Agricultural Research Council measured an average annual rainfall at St. Lucia
over a 22 year period of 975 mm yr$^{-1}$ (ARC-IS CW, 2011). Clearly rainfall in the area is
variable and figures depend on the period over which the rainfall was measured and
the particular location. During this study there was a well reported drought in the region
(Grundling et al., 2014).
The Eastern Shores area is flanked by the Indian Ocean to the east and Lake St. Lucia to the west (Fig. 1a). It includes coastal dunes (dune forest) to the east, the Em-bomveni Dunes (grassland) to the west and the Mfabeni Mire as an interdunal drainage line through the middle. The perennial Nkazana stream drains from the Mfabeni Mire providing freshwater to Lake St. Lucia. This stream was recognised by Vrdoljak and Hart (2007) as an ecologically important source of freshwater to Lake St. Lucia during droughts. Clulow et al. (2013) state that “Organic matter and sediment have accumu-
lated in the Mfabeni Mire over the past 45 000 years, forming one of South Africa’s largest peatlands and one of the oldest active peatlands in the world (Grundling et al., 1998)”. The Mfabeni Mire is approximately 8 km long (north–south direction) and 4 km wide in places (east–west direction). It comprises of subtropical freshwater wetland (SFW) vegetation described by Vaeret and Sokolic (2008) and with a variable canopy height averaging approximately 0.8 m (Clulow et al., 2012). The Nkazana PSF is the other dominant vegetation type that runs down the western side of the Mfabeni Mire (Fig. 1b). The Nkazana PSF falls within the Indian Ocean Coastal Belt Biome, and is described as being a “mixed, seasonal grassland community” (Mucina and Rutherford, 2006). The Nkazana PSF is further classified by von Maltitz et al. (2003) and Mucina and Rutherford (2006) as an Azonal Forest indicating its presence due to, and reliance on, the groundwater surface within the Mfabeni Mire.

1.2 Site description

The Swamp Forest site (28°10.176’ S, 32°30.070’ E) posed significant logistical chal-
lenges due to the 20 m high tree canopy, thick undergrowth, soft ground, dangerous animals and general inaccessibility by road. The measurements were concentrated at its widest point (approximately 1 km) to maximize the fetch for the flux measurements above the tree canopy. Clulow et al. (2013) described previous botanical research ex-
plaining the structure of the Nkazana PSF and the vegetation in the vicinity of the research site:
“Wessels (1997) classified the swamp forests of the area into three logical subgroupings based on dominant species, stand density and basal areas. The Syzygium cordatum subgroup is characterised by an irregular, broken canopy of predominantly Syzygium cordatum trees (known locally as the Water Berry) of up to 30 m, emerging above an intermediate canopy of approximately 6–15 m. Dominant tree species found in the Swamp Forest and in the vicinity of the site included: Macaranga capensis, Bridelia macrantha, Tarenna pavettoides and Stenochlaena tenuifolia. An impenetrable fern (Nephrolepis biserrata) covers the forest floor with a height of approximately 2.5 m and the Stenochlaena tenuifolia (Blechnaceae) fern grows up the tree stems to a height of approximately 10 m.”

The layer of peat at the Nkazana PSF site was approximately 2 m thick and underlain by sand. The water table depth was < 1.0 m but at the surface in low lying areas of the forest. The leaf area index (LAI-2200, LI-COR Inc., Lincoln, Nebraska, USA) beneath the ferns and trees was approximately 7.2 and below the trees approximately 3.3.

2 Materials and methods

2.1 Micrometeorological measurements

An automatic weather station provided supporting meteorological data. It was located adjacent to the Nkazana PSF in the Mfabeni Mire over a reed, sedge and grass dominated vegetation, described broadly as SFW (Fig. 1b). Observations of rainfall (TE525, Texas Electronics Inc., Dallas, Texas, USA), air temperature and relative humidity (HMP45C, Vaisala Inc., Helsinki, Finland), solar irradiance (LI-200X, LI-COR, Lincoln, Nebraska, USA), net irradiance (NRLite, Kipp and Zonen, Delft, the Netherlands), windspeed and direction (Model 03002, R.M. Young, Traverse city, Michigan, USA) were made every 10 s. The appropriate statistical outputs were stored on a data-logger (CR1000, Campbell Scientific Inc., Logan, Utah, USA) at 30 min intervals. Sensors were installed according to recommendations of the World Meteorological Organ-
isation (WMO, 2008) with the raingauge orifice at 1.2 m and the remaining sensors 2 m above the ground. Vapour pressure deficit (VPD) was calculated on the datalogger from air temperature ($T_{air}$) and relative humidity (RH) measurements according to Savage et al. (1997).

2.2 Measurement of energy fluxes and total evaporation

The shortened energy balance equation is commonly used in evaporation studies (Drexler et al., 2004) to describe the partitioning of energy at the earth’s surface and provides an indirect method to determine ET (Eq. 1). The “shortened” version ignores those energies associated with photosynthesis, respiration and energy stored in plant canopies. However, these are considered small when compared with the other terms (Thom, 1975). The shortened energy balance equation is written as:

$$R_n = G + H + LE$$

where, $R_n$ is the net irradiance, $H$ is the sensible heat flux, $G$ is the ground heat flux and LE is the latent energy flux, which is the energy equivalent of ET by conversion (Savage et al., 2004).

Eddy covariance is based on the estimation of the eddy flux which is expressed as:

$$F = \rho_d \overline{w'}s'$$

where, $\rho_d$ is the density of dry air, $w$ is vertical wind speed (measured with the sonic anemometer described below) and $s$ is the concentration of the scalar of interest (water vapour in this case). The primes indicate fluctuation from a temporal average (i.e. $w' = w - \overline{w}$; $s' = s - \overline{s}$) and the over-bar represents a time average. The averaging period of the instantaneous fluctuations, of $w'$ and $s'$ should be long enough (30 to 60 min) to capture all of the eddy motions that contribute to the flux and fulfil the assumption of stationarity (Meyers and Baldocchi, 2005).

The vertical flux densities of $H$ (ET derived indirectly by the shortened energy balance equation) and LE (ET derived directly) were estimated by calculating the mean
covariance of sensible (Eq. 2) and water vapour fluctuations respectively, with fluctuating vertical velocity (Baldocchi et al., 1988).

Soil heat flux was measured using two soil heat flux plates (HFT-3, REBS, Seattle, WA, USA) and a system of parallel thermocouples (Type E). The plates were placed at a depth of 0.08 m below the peat surface. The thermocouples were buried at 0.02 and 0.06 m and were used together with volumetric water content (CS615, Campbell Scientific Inc., Logan, Utah, USA) in the upper 0.06 m to estimate the heat stored above the soil heat flux plates. The measurements were stored every 10 s on a datalogger (CR23X, Campbell Scientific Inc., Logan, Utah, USA) and 30 min averages were computed. During the measurements at the Nkazana Swamp Forest, the groundwater level was deeper than 0.1 m below the surface and therefore, the total $G$ was determined using the calorimetric methodology described by Tanner (1960).

Over the corresponding time period, $R_n$ was measured above the forest canopy, using a 21.3 m telescopic mast (WT6, Clark Masts Systems Limited, Isle of Wight, England). It was erected within the forest, on a fallen tree stump approximately 2.5 m high (Fig. 2a). This formed a firm base for the 90 kg mast which was carried into the forest from the nearest road approximately 1 km away. The computer box for the EC system (InSitu Flux Systems AB, Ockelbo, Sweden) was installed near the base of the mast (Fig. 2b) and a generator that automatically charged a bank of four 100 Ah deep-cycle lead-acid batteries (accumulators) was positioned approximately 50 m from the site in the predominantly downwind direction (the North-West) to minimise any possible influence from the exhaust fumes on the flux measurements. The generator was controlled by a logger (CR10X, Campbell Scientific Inc., Logan, Utah, USA) which was set to activate the charging system (220 V AC petrol generator and 40 A 12 V charger) when the accumulators dropped below 12.4 V.

A “SATI-3VX” style, three-dimensional (3-D) sonic anemometer (Applied Technologies, Inc., Longmont, CO, USA) and open-path infrared gas analyser (LI7500, LI-COR, Lincoln, NE, USA) were mounted on the head of the mast (0.089 m diameter) orientated to face the east (predominant wind direction) to avoid air-flow distortion from the mast.
(Fig. 2c). In addition, $T_{air}$ (PT-10, Peak Sensors Ltd., Chesterfield, UK) and $R_n$ (NR-Lite, Kipp and Zonen, Delft, the Netherlands) were measured at the head of the mast. Data collection and analyses of the system was made in real time by the ECOFLUX software fully described by Grelle and Lindroth (1996) using a Flux Computer (In Situ Flux Systems AB, Ockelbo, Sweden). The system operated with a sampling rate of 10 Hz and the average fluxes were calculated every 30 min. The raw data were also stored for further processing. All the necessary corrections for air-density effects and 3-D coordinate rotation were performed on the Flux Computer to determine $H$ (Grelle and Lindroth, 1996).

The Bowen ratio ($\beta$) has historical significance in evaporation studies and is defined as:

$$\beta = \frac{H}{LE}$$

for a specified time period. It informs on the dominance of $H$ or LE and was calculated at a daily time interval in this study providing a useful means of showing changes in the distribution and weighting of the energy balance components within and between field campaigns.

### 2.3 Energy balance closure

If each component of the energy balance is measured accurately and independently, then Eq. (1) should be satisfied, and closure is considered satisfied. However, energy balance closure could still be achieved if two or more terms have incorrect values and the terms in Eq. (1) still sum to zero (Savage et al., 2004). If the components of the shortened energy balance equation are measured independently then, $R_n - G - H - LE = c$ where $c$ is termed the energy balance closure ($Wm^{-2}$), and closure is satisfied if $c = 0Wm^{-2}$. By rearranging Eq. (1), closure is not achieved if the available energy $R_n - G$ does not equal the turbulent fluxes $H + LE$. Another measure of the lack of closure is the closure ratio or the energy balance closure discrepancy $D$ defined by
Twine et al. (2000) as:

\[ D = \frac{H + LE}{R_n - G} \]  

(4) 

in which, a \( D \) of 1 indicates perfect closure. Several studies using numerous techniques over various surfaces have failed to achieve closure by up to 20 or 30% (Wilson et al., 2001, 2002; Bar et al., 2006). The vast majority have found higher energy input by radiation fluxes than loss by turbulent fluxes (\( H \) and LE) and \( G \) (Oncley et al., 2007). Therefore, the measured fluxes should be corrected or adjusted or the uncertainties in the measured fluxes accepted (Twine et al., 2000). Several reasons for lack of energy balance closure have been discussed by Twine et al. (2000), Wilson et al. (2002), and Cava et al. (2008) including: (1) sampling errors associated with different measurement source areas for the terms in Eq. (1), (2) a systematic bias in instrumentation, (3) neglected energy sinks, (4) the loss of low and/or high frequency contributions to the turbulent flux, (5) neglected advection of scalars, (6) measurement errors related to sensor separation, alignment problems, interference from tower or instrument-mounting structure, and (7) errors in the measurement of \( R_n \) and/or \( G \). Despite concerns that the direct method of determining total evaporation (\( ET_{ec} \)) by measuring water vapour concentrations using an Infrared Gas Analyzer may result in underestimates or overestimates of LE, in this study, it was considered that some of the closure pitfalls of the shortened energy balance method, such as (3) and (7) in particular, could be significant due to the tall canopy at the site (3) and point measurement location (7). Therefore, all \( ET \) results reported in this paper were calculated by the direct method. Energy balance closure discrepancy was determined during the day-time period (\( R_n > 0 \)) due to the potentially large nocturnal influences reported by Wilson et al. (2002).

2.4 Measurement of tree sapflow

A heat pulse velocity system based on the heat ratio method (Burgess et al., 2001), was used to measure sapflow at various depths across the sapwood of two trees over
20 months from September 2009 to early May 2011 which overlapped with the November 2009 and March 2010 field campaigns. The trees measured were located approximately 40 m from the mast where the EC and energy balance sensors were installed. Representative trees, in terms of species, stem diameter, canopy height and proximity to each other, were selected given the cable length limitations of the HPV system. The *Syzigium cordatum* tree selected was approximately 22.5 m tall and had a breast height stem diameter of 0.430 m. Sapflow was measured at four depths across the sapwood on both the eastern and western sides of the stem to account for differences in the sapwood depth around the tree. Sapflow was also measured in a nearby understory tree (*Shirakiopsis elliptica*) with a smaller stem diameter (0.081 m) at four depths within the sapwood. Air temperature and relative humidity (HMP45C, Vaisala Inc., Helsinki, Finland) within the canopy at a height of 2 m above the ground was also measured hourly. Further details of the installation, equipment used, wounding corrections applied and calculations to derive the tree sapflow are documented in Clulow et al. (2013). In this manuscript, following Dye et al. (2008), sapflow is assumed to equate with tree transpiration and tree water-use.

### 2.5 Predicting annual total evaporation from sapflow

Polynomial regression (second order) analysis in the Genstat software (VSN International, 2011) was used to describe the relationship between observed ET$_{ec}$ and observed sapflow of the emergent and understory trees during the overlapping periods of the November 2009 and March 2010 field campaigns in order to understand the possibility of predicting ET of the Nkazana PSF using the long-term sapflow records. The hourly data was checked for homoscedasticity and square root transformed where necessary. The model derived was applied over a full year of sapflow data (October 2009 to September 2010) to obtain an annual predicted ET.
2.6 Evaporation models assessed

Two well recognised evaporation models were tested for applicability of estimating ET from the Nkazana PSF. After assessment of the models at an hourly temporal resolution, over the three field campaigns, the most applicable model was applied to the long-term predicted ET discussed above and verified using historic data collected by the Council for Scientific and Industrial Research (CSIR) during a preliminary study over the Nkazana PSF from 8 to 12 August 2008 and 12 to 20 November 2008 (unpublished). The CSIR measured ET with the identical EC equipment used during the field campaigns in 2009 and 2010 described above and at the same site in the Nkazana PSF making the data ideal for verification of the models.

**FAO56 Penman–Monteith:** The original Penman evaporation model (Penman, 1948), assumed an absence of any control on evaporation at the earth’s surface, in effect, an open water or wet surface situation. This was extended by Monteith (1965) to incorporate surface and aerodynamic resistance functions applicable to vegetated surfaces and was widely used in this form as the Penman–Monteith model. It is however, highly data intensive (Mao et al., 2002; Drexler et al., 2004) and the model was therefore standardised by the Food and Agriculture Organisation in Irrigation and Drainage Paper No. 56 (Allen et al., 1998) into a form known as the FAO56 Penman–Monteith (FAO-PM) model that could be applied at both hourly and daily time intervals. The model received favourable acceptance internationally in establishing a reference evapotranspiration (ET$_r$) index (atmospheric evaporative demand) as a function of weather variables measured at most standard weather station systems. The definition of a reference crop over which the weather variables should be measured was a “hypothetical crop with an assumed height of 0.12 m having a surface resistance of 70 s m$^{-1}$ and an albedo of 0.23, closely resembling the evaporation of an extensive surface of green grass of uniform height, actively growing and adequately watered” (Allen et al., 1998). A nearby crop ET (ET$_c$) is calculated by adjusting the ET$_r$ by a crop factor ($K_c$) in the...
form:

\[ \text{ET}_c = \text{ET}_r \cdot K_c \]  

(5)

where, the crop is not water-stressed. In Allen et al. (1998), values of \( K_c \) have been compiled for different vegetation types at different stages in crop development. Since recommendations by the American Society of Civil Engineers (ASCE) Evapotranspiration in Irrigation and Hydrology Committee (Allen et al., 2000) and the work by Irmak et al. (2005) and Allen et al. (2006) amongst others, the tall crop reference (alfalfa height = 0.5 m) and separate daytime (\( r = 50 \text{ s m}^{-1} \)) and night-time (\( r = 200 \text{ s m}^{-1} \)) resistances for hourly calculations were introduced.

By rearranging Eq. (5) to make \( K_c \) the subject of the equation, and using the long-term predicted ET as a surrogate for \( \text{ET}_c \), \( K_c \) was calculated for the Nkazana PSF at an hourly interval (while \( R_n > 0 \) and \( \text{ET}_{ec} > 0.1 \text{ mm h}^{-1} \)) and summed to daily totals as recommended by Irmak et al. (2005). The reference evaporation approach has been successful internationally, partly due to technological advances leading to improvements in temporal and spatial data availability but also because it provides a method for estimating \( \text{ET}_c \), which is transferrable and can be applied to different vegetation types and locations across the world.

**Priestley–Taylor**: Priestley and Taylor (Priestley and Taylor, 1972) simplified the theoretical Penman equation for specific conditions. They reasoned that, as an air mass moves over an expansive, short, well-watered canopy, ET would eventually reach a rate of equilibrium. In this case, where humid air moves over a wet surface, the aerodynamic resistances become negligible, while irradiance dominates, and the rate of ET would be equal to the potential evaporation (\( \text{ET}_p \)) which is written as:

\[ \text{ET}_p = \frac{\alpha \cdot \Delta}{L_v \cdot \Delta + \gamma} \]  

(6)

where, \( \alpha \) is a constant, \( L_v \) is the specific latent heat of vaporisation of water (2.45 MJ kg\(^{-1}\)), \( \Delta \) is the slope of the saturation water vapour pressure vs. \( T_{air} \), and \( \gamma \) is the psychrometric constant.
The definition of the Priestley–Taylor model makes it suitable for estimation of evaporation from open water areas and wetlands (Price, 1992; Souch et al., 1996; Mao et al., 2002) but it has been applied over numerous other surfaces such as forests (Shuttleworth and Calder, 1979), cropped surfaces (Davies and Allen, 1973; Utset et al., 2004), pastures (Sumner and Jacobs, 2005) and even soil water limited conditions in forest clearcuts (Flint and Childs, 1991) with varied success and deviations from the originally proposed estimate for $\alpha$ of 1.26. In this study it was applied in the form described by Savage et al. (1997) where $\Delta/\left(\Delta + \gamma\right)$ was estimated by:

$$\frac{\Delta}{\Delta + \gamma} = 0.413188419 + 0.0157973 \cdot T_{\text{air}} - 0.00011505 \cdot T_{\text{air}}^2$$

(7)

where, $T_{\text{air}}$ is average air temperature over the interval of calculation (hourly in this study). By rearranging Eq. (6), and substituting $ET_{ec}$ for $ET_p$, $\alpha$ was estimated in the same way as $K_c$ above.

### 2.7 Investigating climatic controls and drivers of sapflow

Sapflow was compared by simple linear regression to climatic variables generally considered to control sapflow in trees such as solar irradiance ($I_s$) and VPD (Albaugh et al., 2013). Sapflow was also compared by multiple regression analysis to the micrometeorological parameters including $I_s$, $T_{\text{air}}$, RH and soil volumetric water content ($\theta$) to determine individual and combined drivers of sapflow. The sapflow was logged as the variability in sapflow increased as the magnitude of the predictor variable increased. Significance of variables, with up to four-way interactions were considered. In addition, the predictor variables were broken up into sets of data with different ranges using regression tree analysis in which a different model was applied to individual sets of data rather than a global model (such as regression analysis) in which a single model is applied to the entire range of each variable.
3 Results

3.1 Weather conditions during the study

The daily radiant densities (integrated solar irradiance over a day) were lowest in August 2009 (~ 15 MJ m\(^{-2}\)) and most consistent (Table 1), whereas in November 2009 and March 2010 they were higher and more variable (between ~ 16 and ~ 25 MJ m\(^{-2}\)), particularly in November 2009 (Tables 2 and 3). The daily maximum temperatures were highest in March 2010 (~ 29 °C) and lowest in August 2009 (22.8 °C). Average minimum RH was lowest in August 2009 (~ 34 %) and the average daytime VPD was highest (1.2 kPa). Average daily windspeeds were notably high in November 2009 (> 7 m s\(^{-1}\)) and the dominant wind direction for the site was from the north-east and the south. Some rainfall (< 7 mm) occurred during the field campaigns but fortunately fell at night and did not affect the daytime flux measurements.

The microclimate within the Nkazana PSF was noticeably different to the adjacent SFW areas. The VPD within the Nkazana PSF canopy was consistently lower than the SFW where the automatic weather station was located approximately 3 km away, with the larger differences occurring from March to August, which is the winter period (Fig. 3). A difference in dawn \(T_{\text{air}}\) between the Nkazana PSF and the adjacent area was also noted. The difference was lowest in summer and highest in winter with the Nkazana PSF being up to 6 °C warmer on some mornings in June 2010.

3.2 Eddy covariance flux measurements

Despite the apparent consistency in the daily radiant density during August 2009 noted above (Table 1), the 30 min net irradiance flux data showed that all field campaigns were affected by cloud during the daytime, as indicated by the standard error bars of the net irradiance (Fig. 4a–c). Even the August 2009 data, despite being in the middle of the dry season, was influenced by cloud during six out of the seven days of measurement (not shown). During the August 2009 and March 2010 field campaigns, there was
a noticeable dip in the average $R_n$ at approximately 11.00 a.m. LT with large standard errors (> 90 W m$^{-2}$) due to cloud cover. In November a dip occurred at approximately 1.00 p.m. LT, also accompanied by large standard errors (> 90 W m$^{-2}$). The cloud affected pattern of $R_n$ was translated through to $H$ and LE, which were positive during the day, and with largest standard errors coinciding with those of the $R_n$ except for the early morning observed LE in August 2009 which was attributed to the evaporation of dew on some days. The maximum rates of LE were approximately 400 W m$^{-2}$ in August 2009, 600 W m$^{-2}$ in November 2009 and 700 W m$^{-2}$ in March 2010 (not shown). The pattern of $G$ fluctuated diurnally but due to attenuation (sensors were below the soil surface) the pattern was smoother than the other fluxes during the course of the day.

During the August 2009 field campaign the daily net radiant density was between 10.2 and 11.8 MJ m$^{-2}$ which, despite the irregularity observed from the 30 min data, were reasonably consistent at a daily level (Fig. 5a). During the November 2009 (11.4 to 18.3 MJ m$^{-2}$) and the March 2010 (9.0 to 14.4 MJ m$^{-2}$) field campaigns, the daily net radiant density was more variable (Fig. 5b and c). This variability at a daily level was translated through to the $H$ and LE results, which during August 2009 were fairly consistent, but irregular during November 2009 and March 2010. The average daily net radiant density was lowest in August 2009 (11.2 MJ m$^{-2}$), highest in November 2009 (15.1 MJ m$^{-2}$) and in-between during March 2010 (12.7 MJ m$^{-2}$). The average daily soil heat flux did not mirror the pattern of $R_n$ and was highest in March 2010 at approximately 11% of $R_n$ (up to 1.8 MJ m$^{-2}$), lower in August 2009 at 5% of $R_n$ (0.7 MJ m$^{-2}$) and lowest in November 2009 at 1% of $R_n$ (up to 0.3 MJ m$^{-2}$).

The daily total LE was higher than $H$ in August 2009 (Fig. 5a), with a daily average $\beta$ ratio of 0.7 (0.4 to 0.9). In November 2009 (Fig. 5b) the daily average $\beta$ ratio was higher with a daily average of 0.9 (0.5 to 1.3) but in March (Fig. 5c) however, LE dominated the energy balance with an average $\beta$ ratio of 0.4 (0.1 to 0.6).

Closure discrepancy was different for each field campaign. In August 2009 the $D$ was 0.98 indicating exceedingly good closure. However, the second and third field...
campaigns produced $D$’s of 1.18 and 1.33 in November 2009 and March 2010, respectively, indicating (1) either an overestimation of LE and/or $H$, and/or an underestimation of the available energy ($R_n - G$) and/or (2) unaccounted energy such as advection or storage in the canopy biomass.

3.3 Observed total evaporation

The mean daily ET over the three field campaigns was significantly different (based on their 95% confidence interval). The daily ET (Fig. 6) was lowest in the August 2009 (winter) and increased progressively through November 2009 (early summer) to March 2010 (late summer). The SD for all field campaigns was similar (0.3 to 0.4 mm) but the coefficient of variation (not shown) differed with the highest in November 2009 (12.0) and August 2009 (11.0) and lowest in March 2010 (8.8).

3.4 Relationship between sapflow and observed total evaporation during two field campaigns

The diurnal course of the sapflow from the emergent and understory trees were surprisingly smooth in comparison to the ET$_{ec}$ results (Figs. 7a, b and 8a, b). The ET$_{ec}$ is an integrated measure of soil evaporation and transpiration from numerous plants at different levels within the canopy over the contributing area described by the footprint whereas the transpiration measurements (assumed to equal sapflow) describe the physiology of a single tree. The $R_n$, frequently considered a significant driver of tree physiology, fluctuated due to cloud cover (Fig. 4a–c). These fluctuations were not translated into fluctuations in tree sapflow but are evident in the ET$_{ec}$ results particularly over the midday period. A similar pattern was observed in the March 2010 ET$_{ec}$ data (Figs. 7a, b and 8a, b).

Despite the greater midday variability of the ET$_{ec}$ data, the polynomial regression (least squares) between hourly ET$_{ec}$ and tree sapflow showed a strong relationship in November 2009 for the emergent tree ($\text{RMSE} = 0.05 \text{ mm}$) as well as the under-
story tree (RMSE = 0.06 mm). The polynomial regression was convex in the case of the emergent tree (Fig. 9a) and concave in the case of the understory tree (Fig. 9b), indicating that the increase in the rate of sapflow of the emergent tree was exponential for lower values of \( \text{ET}_{\text{ec}} \) (morning and evening) but that this rate of sapflow for higher values of \( \text{ET}_{\text{ec}} \) slowed down as the tree reached its peak transpiration rate. In contrast the understory sapflow rate increased gradually per unit increase in \( \text{ET}_{\text{ec}} \) at lower values but at higher values of \( \text{ET}_{\text{ec}} \) the increase in sapflow was exponential. In March 2010 the results were similar with RMSE’s of 0.07 and 0.08 mm for the emergent and understory trees, respectively (Fig. 10a and b). Lagging the sapflow by one hour as suggested by Granier et al. (2000) did not improve the regression of sapflow on \( \text{ET}_{\text{ec}} \).

3.5 Comparison of the FAO56 Penman–Monteith vs. the Priestley–Taylor model during the three field campaigns

The linear regression (least squares) of the hourly FAO56 Penman–Monteith modelled \( \text{ET} \) (\( \text{ET}_r \)), against hourly \( \text{ET}_{\text{ec}} \), explained 75, 85 and 76 % of the fluctuations in \( \text{ET}_{\text{ec}} \) during the August 2009, November 2009 and March 2010 field campaigns, respectively (Table 4). The Priestley–Taylor model did not perform as well, accounting for 54, 74 and 62 % of the variation in \( \text{ET}_{\text{ec}} \) during the August 2009, November 2009 and March 2010 field campaigns, respectively (Table 4).

The slope of the linear regression (\( K_c \)) varied between field campaigns (Table 4) and was highest in March (1.3), and lower in November 2009 (1.1) and August 2009 (0.8). The \( \alpha \), also estimated by the slope of the linear regression, was similar during the August (1.0) and November 2009 (1.0) field campaigns (Table 4) while during March 2010, \( \alpha \) was slightly higher (1.1). The SDs and root means square errors (RMSE) of \( \alpha \) were higher than those of the \( K_c \) (Table 4). Therefore, the FAO56 Penman–Monteith model was adopted as most suitable for use over the Nkazana PSF in this study.

The time interval (hourly and daily) at which the FAO56 Penman–Monteith and Priestley–Taylor models were computed resulted in different \( K_c \) and \( \alpha \) estimates. Daily computations used average daytime \( T_{\text{air}} \), typically derived from an average of maximum
and minimum daily $T_{\text{air}}$. In this research the models were run hourly and the average $T_{\text{air}}$ derived from 10 s measurements of $T_{\text{air}}$ (while $R_n > 0$) accurately representing that hour. However, using hourly data produced outliers in the calculation of $K_c$ and $\alpha$ at the beginning or end of a day where the measured or modelled results are very small numbers, producing, from division, erroneous estimates of $K_c$ and $\alpha$ (Eqs. 5 and 6). These typically occurred near sunset or sunrise and were filtered out of the data as they represented outliers. In addition, due to the vastly different canopy structures and heights within the Mfabeni Mire, of the SFW ($\sim 0.8$ m) and Nkazana PSF ($\sim 20$ m), climatic data from above the forest was used as an input to the models to determine whether the SDs of $K_c$ and $\alpha$ could be minimised, but no significant improvement was found. This indicated that the nearby weather station data (from within the SFW) was a suitable input for both models supporting the application of these models using the standard FAO56 weather station sensor heights of two meters (Allen et al., 2006).

### 3.6 Predicted long-term total evaporation and monthly crop factors

The long-term ET (October 2009 to September 2010) was predicted by modelling the relationship between the observed $ET_{\text{ec}}$ and observed sapflow over the November 2009 and March 2010 field campaigns. In regressions of the emergent tree sapflow with $ET_{\text{ec}}$ over the two field campaigns (Figs. 9a and 10a), it was found that there was little gain in using separate linear models for the two periods ($R^2 = 0.90$ and 0.89; RMSE = 0.05 mm and 0.06 mm) as a single, combined model, described $ET_{\text{ec}}$ equally well ($R^2 = 0.90$; RMSE = 0.07 mm). A similar result was found for the understory tree, indicating that for both trees a single relationship between $ET_{\text{ec}}$ and sapflow represented both field campaigns.

In addition, a multiple regression, including the emergent and understory trees as predictors of $ET_{\text{ec}}$ ($R^2 = 0.91; \text{RMSE} = 0.08 \text{mm}$), provided insufficient benefit over the use of the single model based on only the emergent tree ($R^2 = 0.90; \text{RMSE} = 0.08 \text{mm}$). The understory tree sapflow was considerably less (by 85%) than that of
the emergent tree and the density of the understory trees within the Nkazana PSF is much lower than the emergent trees. These results support the omission of the understory tree from the prediction of ET_{ec}, and the use of the following model to estimate the ET of the Nkazana PSF from hourly sapflow data:

\[
ET = (0.16341 \cdot T_r + 0.06)^2
\]  

where, ET is the estimated total evaporation (mm h\(^{-1}\)) and \(T_r\) the emergent tree sapflow (L h\(^{-1}\)).

The estimated total annual ET (October 2009 to September 2010) from the Nkazana PSF was 1125 mm, over-which period the rainfall was 650 mm (well below the long-term average for the area). Finally, \(K_c\) was calculated at a daily interval from the modelled ET and ET\(_r\) (Eq. 5) and averaged for each month of the year (Fig. 11). These results equated well with the results of \(K_c\) calculated during the field campaigns (Table 4) which were 0.8, 1.0 and 1.3 in August, November and March, respectively. During a distinct period from February to May, \(K_c\) was between 1.2 and 1.4 while for the rest of the year it was 0.8 to 1.0.

The derived crop factors were verified using independent measurements of ET_{ec} over the Nkazana PSF collected during window periods at the same site from 8 to 12 August 2008 and 12 to 20 November 2008 in an experimental unpublished study conducted by the CSIR. The surface conditions during 2008 within the Nkazana PSF were much wetter as the water table was close to the surface with open water in lowlying areas whereas in 2009 and 2010 the dry period had caused the water level to drop, resulting in only a few areas of open water within the forest. Despite this difference in groundwater level, the \(K_c\) was 0.8 in August of 2008 and 0.9 during November 2008, validating the results derived for \(K_c\) from the ET modelled from sapflow of the emergent tree and confirming the \(K_c\) for non-water stressed situations was applicable.
3.7 Response of sapflow to climatic variables

The equation developed between sapflow and total observed ET (Eq. 8) allowed us to predict ET over the period during which we had sap flow measurements (October 2009 to September 2010). The purpose for this was to better understand the relationship between important climatic variables and ET in order to understand the climate risks over a long-term period. Three statistical approaches were used to determine these relationships with sapflow, which were directly related to ET and the climatic variables. We considered the simple linear regressions of daily sapflow with radiant flux density and VPD and found that these were poor, with coefficients of determination of only 0.51 and 0.52 respectively (not shown). Clearly the relationship between climatic conditions and sapflow is complex. We applied multiple regression analysis and found $I_s$, RH, $T_{\text{air}}$, and $\theta$ at 0.075 m to be significant ($p < 0.001$) with up to four-way interactions. We finally applied a regression tree analysis of hourly log-transformed sapflow with the meteorological variables $I_s$, RH, $T_{\text{air}}$ and $\theta$ (Fig. 12). This showed again that the relationships are complex but that $T_{\text{air}}$ and $\theta$ were not required for the optimal split for the Nkazana PSF emergent tree sapflow. The most important split was between data with $I_s$ of less than 55.7 $\text{Wm}^{-2}$ and data with $I_s$ greater than 55.7 $\text{Wm}^{-2}$. Solar irradiance was clearly a key variable to include and the first split observed essentially separates day and night-time data. Solar irradiance was also highly correlated with $T_{\text{air}}$, which may be the reason $T_{\text{air}}$ was not found to be an additionally required variable. The next important splits are for RH above and below 93.2% for the night time data (essentially when it is raining and when it is not) and an additional split for $I_s$ above and below 279.2 $\text{Wm}^{-2}$ for the day time data; therefore splitting day time data during high and low irradiance periods. At night the logged sapflow was found to be negative, with the greatest negative average logged sapflow when the RH was less than 96.4%. The greatest average positive logged sapflow was found to be when $I_s$ was greater than 279.2 $\text{Wm}^{-2}$, and this occurred 28% percent of the time.
4 Discussion

The EC method is recognised internationally to be a suitable and accurate technique for estimating ET over vegetated surfaces and long-term EC measurements over the Nkazana PSF could provide the data required to understand the annual cycles of ET. However, EC systems have relatively high power requirements and need careful and frequent attendance as well as data checking, correction and analysis for complete records. The remote location of the Nkazana PSF, with no road access and difficult access on foot, high wind speeds and dangerous wild animals such as buffalo, rhinoceros, hippopotamus and crocodiles, prompted a research strategy to characterise the ET of the Nkazana PSF during field campaigns conducted in representative seasons, as it was impractical to maintain a full EC system over an extended period of time (such as a year). There was a risk that a period of unusual weather could have coincided with the window periods (between seven and nine consecutive days at a time), however the weather conditions during the field campaigns showed that a range of climatic conditions were captured that were representative of the seasons (Tables 1 to 3). With this approach, field campaigns could be extended if unusual weather conditions are encountered over the planned measurement period.

The challenge remained in interpolating and extrapolating the ET\textsubscript{ec} results from the EC system to annual ET. In long-term evaporation studies where gaps occur or where window periods have been used, and interpolation of the ET record is required, meteorological models are typically used. Total evaporation has been estimated using models that are computationally simple such as the Priestley–Taylor model (Priestley and Taylor, 1972; Shuttleworth and Calder, 1979) to more complex models using multi-layer approaches within the canopy, but still based on the Penman–Monteith approach (Roberts et al., 1993; Harding et al., 1994), with significant deviations between measurements and modelled results. These meteorological models are however, uncoupled from the transpiring vegetation and therefore the pattern of actual tree sapflow was considered in this study as a predictor of ET.
External regulation of sapflow has been described by numerous variables including the readily available soil water of the rooting area (Oren and Pataki, 2001), the micrometeorological conditions of the atmosphere (Lundblad and Lindroth, 2002), leaf area (Granier et al., 2000), canopy conductance (Granier et al., 2000), aerodynamic resistance, shading of lower leaves and wind stress. However, it has been found that trees can have several mechanisms of internal regulation related to species specific morphology and physiology that is partially uncoupled from the external conditions (Zweifel et al., 2002). Nevertheless, in most trees with actively transpiring leaves and some readily available soil water, a diurnal pattern of sapflow results from a combination of internal and external conditions, which determines how a tree contributes to the ET of a forest stand.

With advances in sapflow measurement techniques, long-term forest ET has been estimated by up-scaling from tree transpiration to forest ET using various techniques generally based on sapwood area (Čermák et al., 2004). However, the large majority of these studies, especially where tree transpiration has been up-scaled, have been conducted in uniform forest stands (Oren et al., 1999; Wilson et al., 2001) and much of the work has taken place in temperate boreal stands (Lundblad and Lindroth, 2002; Launiainen et al., 2011) and their applicability to other climatic zones needs consideration. In addition, it has also been recognised that transpiration often varies amongst species (Oren and Pataki, 2001; Ewers et al., 2002; Bowden and Bauerle, 2008) and up-scaling to forest transpiration in species rich indigenous forests is complex.

The results from this study showed that the hourly sapflow of a single emergent tree, selected as a dominant species, correlated well with the hourly ET measured over two window periods. In species rich forests, measuring the sapflow of the different vegetation types (including the ferns, vines, understory and emergent trees) would be challenging, and up scaling questionable, due to the variety of plant structures within the canopy and our lack of information on the plant physiologies. Therefore the empirical relationship between the single tree and ET provided an ideal opportunity to model the annual ET. This relationship indicated that the emergent canopy trees are the main
contributors to ET. The other contributors to ET, including open water, peat, ferns, vines and understory trees were either (1) insignificant contributors due to the low irradiance and VPD below the emergent tree canopy (supported by the low measured sapflow rate of the understory tree), or (2) follow similar diurnal trends in evaporation and sapflow as the emergent tree (also supported by the diurnal trend in the sapflow rate of the understory tree) and are therefore captured in the empirical model of the emergent tree.

Variation of the energy balance closure discrepancy ($D$) occurred between field campaigns, despite replication of the same instrumentation at the same site and with the same data processing procedures. Only the placement of the soil heat flux sensors changed slightly within the vicinity of the site between field campaigns. However, the soil heat fluxes (as a percent of net irradiance) fluctuated from 1 to 11%, likely due to the specific placement of the sensors within the Nkazana PSF in a predominantly shaded area in contrast to a sunlit location due to gaps in the canopy. During August 2009 when $D = 1$ (i.e. perfect closure of the energy balance), the soil heat flux was approximately 5% of $R_n$ and was likely to be the most representative result for $G$ for a forested area agreeing with Dye et al. (2008). In March, $G$ was 11% of $R_n$ and may have contributed to the poorest result of $D = 1.33$. Wilson et al. (2002) found that energy balance closure, especially over forests, is seldom achieved. However, in most cases the magnitude of the long-term turbulent fluxes is lower than the available energy (Twine et al., 2000; Oliphant et al., 2004), which was not the case in the Nkazana PSF study where $D$ increased with increasing ET from August 2009 through November 2009 to March 2010.

An important observation made over the three field campaigns, was that the average ET measured during March 2010 (4.4 mm day$^{-1}$) did not correspond to the period of highest $R_n$ (November 2009), which is commonly accepted to be one of the main driving variables in the process of ET (Albaugh et al., 2013). This indicated a lag in the ET of the Nkazana PSF in relation to the maximum $R_n$, possibly explaining the poor relationships observed between tree sapflow and climatic variables (such as $R_n$).
This lag was also observed in the high $K_c$ values from February to May, where the ET of the PSF was higher relative to $ET_r$. Typically, $K_c$ is higher while vegetation is more actively transpiring and is associated with higher $I_s$ and water availability, which in the Nkazana PSF would coincide with the summer period (October to March). However, the period of higher $K_c$ values in the Nkazana PSF occurred quite late (February to May) in the summer season (Fig. 11). Clulow et al. (2013) showed the Nkazana PSF sapflow to be relatively consistent between seasons but that $ET_r$ rapidly decreased from February to May (4.2 mm day$^{-1}$ to 2.4 mm day$^{-1}$). The high $K_c$ is therefore likely a result of decreasing $ET_r$ while transpiration rates in the Nkazana PSF were maintained into the late autumn period. A number of reasons may be attributed to this including the microclimate of the Nkazana PSF. For example, the lower energy loss at night from the ground and within the canopy due to the combined effect of high water vapour levels (a greenhouse gas) and reduced infrared emission as a result of canopy absorbance, reflectance and re-emission downwards, compared to areas outside the PSF with shorter canopies, resulted in higher minimum daily temperatures (Fig. 3). The area adjacent to the Nkazana PSF where the automatic weather station was located (with a shorter canopy of approximately 1 m in height) experienced lower daily minimum temperatures (Fig. 3). The importance of this result is that $T_{air}$ affects biochemical processes such as photosynthesis and senescence. This $T_{air}$ difference, although greatest in winter, starts to build in January and could play a role in influencing the ET in relation to the summer season as well as the period of higher $K_c$ values in the latter half of summer.

Two important points regarding the weather station data and model calculations were noted. Firstly, where possible, hourly model time intervals should be used which, concurs with Irmak et al. (2005). However, this frequently resulted in outliers in $K_c$ and $\alpha$ at the beginning or end of a day where the measured or modelled results were small numbers, producing, through division, erroneous estimates. It was therefore favourable to sum the hourly $ET_r$ and ET data for each day (while $R_n > 0$) and calculate the daily $K_c$ (which was then averaged for each month). Secondly, when calculating the $K_c$ and $\alpha$
Coefficient’s, there was no benefit in using the climatic data from above the tree canopy rather than the climatic data from the adjacent SFW of the Mfabeni Mire which, sufficiently represented the microclimate for the model calculations. This showed that data from nearby weather stations can be used with the $K_c$ to estimate the ET although this may only hold in humid environments where there is little difference in the VPD of the boundary layer conditions over the Nkazana PSF and the surrounding wetland areas which are likely to all be at, or near, equilibrium evaporation.

The shape of the regressions of hourly $ET_{ec}$ vs. hourly sapflow (Figs. 9 and 10) showed that sapflow of the emergent tree responded rapidly for low conditions of $ET_{ec}$. These conditions occur most frequently in the early morning and late afternoon when the angle of the $I_s$ is low but still incident upon the emergent tree leaves. At higher rates of $ET_{ec}$ the sapflow peaked as the physiology of the tree limited the sapflow rates. In contrast, the understory tree sapflow rate increased slowly relative to $ET_{ec}$, while $ET_{ec}$ was low, and exponentially for higher values of $ET_{ec}$. This is likely due to shading of the understory trees for low sun angles (early morning and late afternoon) with $I_s$ limiting transpiration (together with the low VPD discussed above) with maximum rates occurring when shading by the emergent trees was at a minimum (noon) and $ET_{ec}$ was at a maximum. These different responses of the trees indicated that a model to derive ET from sapflow would require the inclusion of both trees. However, the sapflow and density of the understory trees was much lower than the emergent trees and therefore its inclusion in the empirical model not found to significantly improve the relationship between entire canopy $ET_{ec}$ and sapflow. This conclusion applies specifically to the Nkazana PSF. Some models such as the WAVES model permits two canopy simulations due to the importance of the understory canopy in some forest sites (Dye et al., 2008).

Within South Africa, the study by Dye et al. (2008) measured daily ET’s of between 2 and 6 mm on clear days over three field campaigns during February, June and October 2004 which are comparable with the results from the Nkazana PSF of between 2.2 mm (August 2009) and 5.1 mm (March 2010). Internationally, no results of ET or
modelling guidelines for peat swamp forests were found, signifying the unique contribution of this study.

The comparison of meteorological variables with sapflow revealed that it is unlikely that a single climatic variable is able to determine sapflow, and in turn ET. The relationships were revealed to be non-linear and that to model sapflow accurately, data need to be sub-set into different periods; at least into day and night.

5 Conclusions and opportunities for further research

This study portrayed the difficulties of using the most advanced systems available to measure ET, such as EC, in remote and difficult to access areas. It showed that intensive window period measurements using high maintenance EC systems provide reliable and continuous measurements of ET but require a method to determine the ET during the in-between periods to be able to estimate long-term ET. This was overcome by measuring the long-term sapflow of an emergent canopy tree and deriving a qualitative model for ET based on sapflow measurements. Further research on the benefit of measuring multiple emergent trees and the possible variability of transpiration within different species and the extent to which this could improve the long-term estimate of forest ET together with window periods of EC data would be beneficial.

Energy balance closure discrepancy ($D$) remains an unresolved matter which affects flux measurements such as ET and CO$_2$. Corrections suggested in research studies can be applied but without conclusively identifying the source of the error in the observations. In contrast to most studies reported, the closure discrepancy of the energy balance over the Nkazana PSF was greater than 1 for two of the field campaigns. Although attributed in part to unrepresentative $G$ measurements, $D$ increased as ET increased.

The empirical model used to derive the annual ET from sapflow, and then monthly crop factors was verified with data from two independent field campaigns in 2008, when conditions were much wetter and there were larger areas of open water within
the forest. The crop factors derived during these two field campaigns in 2008 (August and November) verified those derived from the empirical model calculations. The much wetter conditions in 2008 however did not alter the $K_c$ indicating that the relationship between ET and $ET_r$ remained constant and that the $K_c$ derived can be applied over a range of climatic conditions. In addition, it indicates that the humid, low VPD environment within the forest canopy minimises the contribution of open water evaporation within the forest to ET. However, the general dearth of information on the ET of subtropical indigenous forests internationally allows little comparison of the results obtained from the Nkazana PSF and similar forest types and the extent to which these crop factors can be extrapolated geographically and to similar forests would benefit from further comparisons.

The Mfabeni Mire is actively managed by the iSimangaliso Wetland Park. These results provide the basis for improved estimates of the ET component of the Nkazana PSF water balance and the environmental water requirements. Water is critical to the functioning of this ecosystem for biotic and abiotic life, the sequestration or release of carbon from the Mire and the spread of fires. The annual ET estimated in this study (1125 mm) was significantly higher than the measured rainfall (650 mm) and even higher than the reported estimates of mean annual precipitation (∼950 mm). The difference between ET and rainfall highlights the importance of the groundwater contributions and the critical role it plays in assuring the survival of this groundwater dependent ecosystem. The groundwater available to the Mfabeni Mire is in part determined by the management of the upstream catchments and the groundwater levels of the greater Zululand Coastal Aquifer, emphasising the need for an integrated catchment management approach to the area.
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Table 1. Summary of weather conditions during the August 2009 field campaign.

<table>
<thead>
<tr>
<th>Date</th>
<th>Solar radiant density (MJ m$^{-2}$)</th>
<th>Wind speed (m s$^{-1}$)</th>
<th>Wind direction ($^\circ$)</th>
<th>VPD (kPa)</th>
<th>Air temperature ($^\circ$C) Max</th>
<th>Min</th>
<th>RH (%) Max</th>
<th>Min</th>
<th>Rain (mm) Max</th>
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Table 2. Summary of weather conditions during the November 2009 field campaign.

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<th>Date</th>
<th>Solar radiant density (MJ m(^{-2}))</th>
<th>Wind speed (m s(^{-1}))</th>
<th>Wind direction (°)</th>
<th>VPD (kPa)</th>
<th>Air temperature (°C)</th>
<th>RH (%)</th>
<th>Rain (mm)</th>
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### Table 3. Summary of weather conditions during the March 2010 field campaign.

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<th>Date</th>
<th>Solar radiant density (MJ m(^{-2}))</th>
<th>Wind speed (m s(^{-1}))</th>
<th>Wind direction (°)</th>
<th>VPD (kPa)</th>
<th>Air temperature (°C)</th>
<th>RH (%)</th>
<th>Rain (mm)</th>
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<td>28.6</td>
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<td><strong>28.7</strong></td>
<td><strong>18.7</strong></td>
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Table 4. Summary of the hourly crop co-efficient $K_c$ and advective term $\alpha$ with SD and root mean square error (RMSE) for each of the three field campaigns.

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<tr>
<th></th>
<th>$K_c$</th>
<th>Coefficient of determination</th>
<th>SD</th>
<th>RMSE</th>
<th>$\alpha$</th>
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<td>0.17</td>
<td>0.07</td>
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<td>0.74</td>
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<td>0.76</td>
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<td>0.11</td>
<td>1.1</td>
<td>0.62</td>
<td>0.46</td>
<td>0.13</td>
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Figure 1. (a) Location of the Eastern Shores within South Africa, (b) the Nkazana Peat Swamp Forest site (where the EC and sapflow systems were located) and the automatic weather station within the Mfabeni Mire on the Eastern Shores (data from Mucina and Rutherford, 2006).
Figure 2. (a) Telescopic mast (21.3 m) erected in the swamp forest to raise the eddy covariance instruments above the forest canopy, (b) the computer installed at the swamp forest, housed in a temperature controlled enclosure and (c) the instruments attached to the head of the mast.
Figure 3. The difference in the monthly daytime (9 a.m.–3 p.m.) vapour pressure deficit and difference between the monthly average dawn air temperatures measured in the subtropical freshwater wetland area of the Mfabeni Mire (reeds, sedges and grasses) and within the canopy of the Nkazana Peat Swamp Forest site.
Figure 4. The average of the half-hourly energy fluxes, with error bars indicating the standard error, measured at the Nkazana Swamp Forest in (a) August 2009, (b) November 2009 and (c) March 2010.
Figure 5. Daily total energy densities (while $R_n > 0$) measured at the Nkazana Swamp Forest in (a) August 2009, (b) November 2009 and (c) March 2010.
Figure 6. Daily total evaporation measured over the Nkazana Swamp Forest during three representative periods.
Figure 7. The diurnal course of the hourly sapflow of the (a) emergent tree and (b) understory tree in the Nkazana PSF together with the hourly total evaporation during the November 2009 field campaign.
Figure 8. The diurnal course of the hourly sapflow of the (a) emergent tree and (b) understory tree in the Nkazana PSF together with the hourly total evaporation during the March 2010 field campaign. Note the influence of rain on the early morning ET results of 21 March 2010.
Figure 9. Polynomial regressions of the total evaporation (ET) against the hourly sapflow for the (a) emergent tree and (b) understory tree in the Nkazana PSF during the November 2009 field campaign.
Figure 10. Polynomial regressions of the total evaporation (ET) against the hourly sapflow for the (a) emergent tree and (b) understory tree in the Nkazana PSF during the March 2010 field campaign.
Figure 11. Monthly crop factors $K_c$ for the Nkazana Peat Swamp Forest.
Figure 12. A regression tree analysis for sapflow showing the optimal splits of solar irradiance (Wm$^{-2}$) and relative humidity (%). Air temperature and volumetric water content were included, but these variables were not required for the optimal splits. The percentage of the total data at each split is also shown.