

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

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18
19 **Abstract**

20
21 A combination of measurement and modelling was used to find a pragmatic solution to estimate the
22 annual total evaporation from the rare and indigenous Nkazana Peat Swamp Forest (PSF) on the east
23 coast of Southern Africa to improve the water balance estimates within the area. Actual total
24 evaporation (ET_a) was measured during three window periods (between seven and nine days each) using
25 an eddy covariance (EC) system on a telescopic mast above the forest canopy. Sapflows of an
26 understory tree and an emergent tree were measured using a low maintenance heat pulse velocity system
27 for an entire hydrological year (Oct 2009 to Sept 2010). An empirical model was derived, describing
28 the relationship between ET_a from the Nkazana PSF and sapflow measurements. These overlapped
29 during two of the window periods ($R^2 = 0.92$ and 0.90), providing hourly estimates of ET_a from the
30 Nkazana PSF for a year, totalling 1125mm (while rainfall was 650 mm). In building the empirical
31 model, it was found that to include the understory tree sapflow provided no benefit to the model
32 performance. In addition, the relationship between the emergent tree sapflow with ET_a between the two
33 field campaigns was consistent and could be represented by a single empirical model ($R^2=0.90$;
34 RMSE=0.08 mm h⁻¹).

35 During the window periods of EC measurement, no single meteorological variable was found to
36 describe the Nkazana PSF ET_a satisfactorily. However, in terms of evaporation models, the hourly FAO
37 Penman-Monteith reference evaporation (ET_o) best described ET_a during the August 2009 ($R^2 = 0.75$),
38 November 2009 ($R^2 = 0.85$) and March 2010 ($R^2 = 0.76$) field campaigns, compared to the Priestley-
39 Taylor potential evaporation (ET_p) model ($R^2 = 0.54, 0.74$ and 0.62 during the respective field
40 campaigns). From the extended record of ET_a (derived in this study from sapflow) and ET_o , a monthly
41 crop factor (K_c) was derived for the Nkazana PSF, providing a method of estimating long-term swamp
42 forest water-use from meteorological data. The monthly K_c indicated two distinct periods. From
43 February to May, it was between 1.2 and 1.4 compared with June to January, when the crop factor was

1 0.8 to 1.0. The derived monthly K_c values were verified as accurate (to one significant digit) using
2 historical data measured at the same site, also using EC, from a previous study.

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

9
10 Keywords: indigenous forest, tree water-use, tree transpiration, evaporation modelling, energy balance
11 closure, crop factor, groundwater dependent
12

13 **1 Introduction**

14
15 Severe water scarcity in parts of South Africa has threatened the health of internationally recognised
16 environmental areas such as the iSimangaliso Wetland Park, a UNESCO world heritage site. To
17 optimise the management of the water balance and understand the functioning of the area, there has
18 been a need to quantify the water-use of the dominant vegetation types of the Park such as the
19 endangered Peat Swamp Forests (Grundling et al., 1998; Clulow et al., 2012), a dominant plant type of
20 the Mfabeni Mire. However, little is known about the water-use characteristics of the species diverse
21 Peat Swamp Forests (PSFs) both locally and internationally in terms of model parameterisation. Despite
22 significant improvements to measurement techniques over vegetated surfaces (Savage et al., 1997),
23 these have not been of benefit for PSFs due to their remote and inaccessible nature. In addition, well
24 documented extreme events (such as the Demoina floods in 1987) pose a real threat in the area.
25 Sophisticated instruments are unfortunately vulnerable to damage and malfunction in such
26 environments and PSFs are therefore not good locations for long-term deployment of sensitive
27 equipment, a challenge facing researchers internationally and particularly in developing countries.

28
29 There are numerous, complex evaporation sources, which interact and contribute to actual total
30 evaporation (ET_a) in the Nkazana PSF. The areas of open water fluctuate, depending on groundwater
31 levels. Open water evaporation is well described from the early work of Penman (1948) to the more
32 recent work of Finch (2001) and Rosenberry et al. (2007) but none accounts for the effects of dense
33 vegetation cover on radiative shading and the prevention of convection over the water surface by a tall
34 and dense canopy. There are surface evaporation studies of peat (Nichols and Brown, 1980; Koerselman
35 and Beltman, 1988; Lafleur and Roulet, 1992; Thompson et al., 1999, Clulow et al., 2012), but none in
36 the context of a subtropical swamp forest. In addition the vegetated canopy is complex. There is a dense
37 cover of ferns, of which little is understood in terms of transpiration (Andrade and Nobel, 1997). Above
38 the ferns, the tree canopy consists of two levels described below (understory and emergent trees) and
39 there are tree-climbing vines. Estimating ET_a of the Nkazana PSF is clearly multifaceted due to its
40 diversity and our lack of understanding of the water-use of the specific plants, together with the potential
41 variation in the evaporative demand within and above the canopy.

42
43 Within South Africa, the only comparable study took place over an evergreen indigenous mixed forest
44 in the Southern Cape near the coast. Dye et al. (2008) measured ET_a using eddy covariance (EC),
45 scintillometry and Bowen ratio over 18 days in total, during three different field campaigns,
46 representing three different seasons within the year. The periods in-between were modelled using the
47 FAO56 Penman-Monteith reference equation of Allen et al. (1998), which generally underestimated

1 ET_a under high evaporative conditions and overestimated under low evaporative conditions. This was
2 attributed to the assumption of a constant surface resistance. The Penman-Monteith equation (Monteith,
3 1965) was found to give the best match of modelled to observed daily ET_a , but required measurements
4 or a sub-model, accounting for variable canopy conductance. The more complex WAVES (CSIRO,
5 Canberra, Australia) process-based model simulated canopy growth and water-use processes in much
6 more detail. However, successful parameterisation of the many model inputs was a significant challenge
7 and despite their best efforts, the WAVES output revealed an overestimation of daily ET_a under
8 conditions of low evaporative demand, which could not be corrected. They concluded that the best
9 technique for interpolating the periods between the three field campaigns would be the Penman-
10 Monteith equation despite the problem of the variable canopy conductance and the recommended that
11 further research into understanding the most appropriate techniques for interpolating measured data
12 would be necessary.

13
14 Internationally, no studies were found with measurements over a comparable subtropical peat swamp
15 forest. However, Vourlitis et al. (2002) provide a valuable study in which they attempted to measure
16 the long-term ET_a with an EC system over a tropical forest in Brazil. Despite the proximity to the city
17 of Sinop (offering a nearby base from which maintenance could be conducted), power issues hampered
18 the data collection and EC data was only collected 26% of the time. Meteorological data was therefore
19 used to estimate the latent energy flux (LE) using the Priestley-Taylor expression.

20
21 Since the beginning of the FLUXNET project, which was established to compile long-term
22 measurements of water vapour, carbon dioxide and energy exchanges from a global network of EC
23 systems, the problem of complete EC data sets and gap filling of records was recognised and is still an
24 ongoing challenge (Baldocchi et al., 1996; Baldocchi et al. 2001). Falge et al. (2001) found the average
25 data coverage for long-term EC systems to be only 65% due to system failure or data rejection with
26 most of these located in developed countries. Clearly, despite the benefit of EC systems, long-term,
27 continuous records of observed ET_a data over indigenous subtropical and tropical forests are improbable
28 without significant research budgets allowing daily maintenance, gap filling and processing of data
29 including complex spectral corrections, 3D corrections and coordinate rotation amongst others
30 (Massman and Lee, 2002; Finnigan et al., 2003; Hui et al., 2004). Intensive, short-term field campaigns,
31 offering reliable, continuous records, during different seasons seem to provide an appropriate strategy
32 to determine the annual cycle of ET_a . This is particularly the case in South Africa, where theft of
33 equipment and especially batteries from the foot of visible towers is a severe limitation, although this
34 is overcome by employing 24-hour security guarding services to protect the equipment during the short-
35 term measurement periods (Dye et al., 2008). However, this strategy only provides a viable solution if
36 the ET_a during the in-between periods, can be adequately estimated.

37
38 Wilson et al. (2001) applied EC and sapflow techniques in a deciduous forest of the south-eastern United
39 States, and found that there was a qualitative similarity between ET_a , derived using the EC technique,
40 and tree transpiration. With the recent advances in sapflow measurement techniques and upscaling of
41 individual tree transpiration measurements to canopy ET_a , it is believed that sapflow techniques offer a
42 reliable, standalone, long-term solution to estimating ET_a in uniform tree stands (Hatton and Wu, 1995;
43 Meiresonne et al., 1999; Crosbie et al., 2007). There are however, numerous complexities, bringing
44 some doubt as to the accuracy of the absolute sapflow results, such as the anisotropic properties of
45 sapwood (Vandegehuchte et al., 2012), species composition effects (Wullschleger et al., 2001), tree
46 symmetry (Vertessy et al., 1997), radial patterns of sapflow (Čermák and Nadezhdina, 1998) and
47 changes in spatial patterns of transpiration (Traver et al., 2010). In heterogeneous and complex canopies
48 such as the Nkazana PSF described above, sapflow systems alone are impractical for the prediction of

1 stand ET_a even with the recent advances in process based models of vegetation function such as the
2 Measpa model (Duursma and Medlyn, 2012). However, whether it is possible to use the qualitative
3 relationship of sapflow with measured ET_a , as found by Wilson et al. (2001), remains unknown.

4
5 For these reasons, a strategy to provide a measurement and modelling framework was developed and
6 tested, in which detailed water flux measurements were recorded using EC instruments in an
7 indigenous, heterogeneous forest over three window periods in August 2009, November 2009 and
8 March 2010 (Fig. 1). This minimised the cost and risk of damage to these expensive systems, and
9 provided continuous and reliable data from well-maintained instruments, operated by a team of
10 scientists, but were limited to three window periods. Two of these window periods overlapped with
11 long-term sapflow measurements, and a nearby weather station provided climatic data during the full
12 period (Fig. 1). The sapflow and weather station systems had lower maintenance and power
13 requirements, were less delicate, less visible, able to withstand the harsh environment, and operated for
14 longer periods unattended (one to two months) without compromising data quality. The aims were
15 therefore to (1) establish whether the long-term ET_a of the Nkazana PSF could be determined by this
16 combination of EC window periods and long-term sapflow measurements, (2) to provide a means of
17 modelling the ET_a of surrounding PSFs from nearby meteorological data and (3) to investigate the
18 controlling climatic variables and their influence on sapflow as well as the energy fluxes and
19 microclimate within the swamp forest (Fig. 1).

20 21 *1.1 The Study Area*

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

28
29 The Eastern Shores area has a subtropical climate and lies in a summer rainfall area (Schulze et al.,
30 2008). It has been reported that “the rainfall gradient westwards from the coast is strong, with a
31 precipitation at Mission Rocks on the Indian Ocean coastal barrier dune exceeding 1200 mm year⁻¹ and
32 decreasing to around 900 mm year⁻¹ at Fanies Island on the western shoreline of the estuary” (Taylor et
33 al., 2006). However, Lynch (2004) provides mean annual precipitation values of 1056 mm a⁻¹, 844 mm
34 a⁻¹ and 910 mm a⁻¹ from the nearby Fanies Island, Charters Creek and St. Lucia respectively from a 125
35 year raster database, and the Agricultural Research Council measured an average annual rainfall at St.
36 Lucia over a 22 year period of 975 mm a⁻¹ (ARC-ISCW, 2011). Clearly rainfall in the area is variable
37 and figures depend on the length of the period in years over which the rainfall was measured and the
38 particular location. During this study there was a well reported drought in the region (Grundling et al.,
39 2014).

40
41 The Eastern Shores area is flanked by the Indian Ocean to the east and Lake St. Lucia to the west (Fig.
42 2a). It includes coastal dunes (dune forest) to the east, the Embomveni Dunes (grassland) to the west
43 and the Mfabeni Mire as an interdunal drainage line through the middle. The perennial Nkazana stream
44 drains from the Mfabeni Mire, providing freshwater to Lake St. Lucia. This stream was recognised by
45 Vrdoljak and Hart (2007) as an ecologically important source of freshwater to Lake St. Lucia during
46 droughts. Clulow et al. (2013) state that “Organic matter and sediment have accumulated in the Mfabeni
47 Mire over the past 45 000 years, forming one of South Africa’s largest peatlands and one of the oldest
48 active peatlands in the world (Grundling et al., 1998)”. The Mfabeni Mire is approximately 8 km long

1 (north-south direction) and 4 km wide in places (east-west direction). It comprises of subtropical
2 freshwater wetland (SFW) with vegetation described by Vaeret and Sokolic (2008) and with a variable
3 canopy height averaging approximately 0.8 m (Clulow et al., 2012). The Nkazana PSF is the other
4 dominant vegetation type that runs down the western side of the Mfabeni Mire (Fig. 2b). The Nkazana
5 PSF falls within the Indian Ocean Coastal Belt Biome, and is described as being a “mixed, seasonal
6 grassland community” (Mucina and Rutherford, 2006). The Nkazana PSF is further classified by von
7 Maltitz et al., (2003) and Mucina and Rutherford (2006) as an Azonal Forest, indicating its presence
8 due to, and reliance on, the ground water surface within the Mfabeni Mire.

9

10 1.2 Site description

11 The Swamp Forest site (28° 10.176’S, 32° 30.070’E) posed significant logistical challenges due to the
12 20 m high tree canopy, thick undergrowth, soft ground, dangerous animals and general inaccessibility
13 by road. The measurements were concentrated at its widest point (approximately 1 km) to maximize
14 the fetch for the flux measurements above the tree canopy. Clulow et al. (2013) described previous
15 botanical research, explaining the structure of the Nkazana PSF and the vegetation in the vicinity of the
16 research site as follows:

17
18 “Wessels (1997) classified the swamp forests of the area into three logical subgroupings based
19 on dominant species, stand density and basal areas. The *Syzygium cordatum* subgroup is
20 characterised by an irregular, broken canopy of predominantly *Syzygium cordatum* trees (known
21 locally as the Water Berry) of up to 30 m, emerging above an intermediate canopy of
22 approximately 6-15 m. Dominant tree species found in the Swamp Forest and in the vicinity of
23 the site included: *Macaranga capensis*, *Bridelia macrantha*, *Tarenna pavettoides* and
24 *Stenochlaena tenuifolia*. An impenetrable fern (*Nephrolepis biserrata*) covers the forest floor
25 with a height of approximately 2.5 m and the *Stenochlaena tenuifolia* (Blechnaceae) fern grows
26 up the tree stems to a height of approximately 10 m.”

27
28 The layer of peat at the Nkazana PSF site was approximately 2 m thick and underlain by sand. The
29 water table depth was < 1.0 m but at the surface in low lying areas of the forest. The leaf area index
30 (LAI-2200, LI-COR Inc., Lincoln, Nebraska, USA) beneath the ferns and trees was approximately 7.2
31 and below the trees approximately 3.3.

32

33 2 Materials and methods

34 2.1 Micrometeorological measurements

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

47

48 2.2 Measurement of energy fluxes and actual total evaporation

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

$$6 \quad R_n = G + H + LE \quad (1)$$

7
8 where, R_n is the net irradiance, H is the sensible heat flux, G is the ground heat flux and LE is the latent
9 energy flux, which is the energy equivalent of evaporation by conversion (Savage et al., 2004).

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

$$11 \quad F = \rho_d \overline{w's'} \quad (2)$$

12
13 where, ρ_d is the density of dry air, w is vertical wind speed (measured with the sonic anemometer
14 described below) and s is the concentration of the scalar of interest (water vapour in this case). The
15 primes indicate fluctuation from a temporal average (i.e. $w' = w - \bar{w}$; $s' = s - \bar{s}$) and the over-bar
16 represents a time average. The averaging period of the instantaneous fluctuations, of w' and s' should
17 be long enough (30 to 60 minutes) to capture all of the eddy motions that contribute to the flux and
18 fulfil the assumption of stationarity (Meyers and Baldocchi, 2005).

19
20 The vertical flux densities of H (ET_a derived indirectly by the shortened energy balance equation) and
21 LE (ET_a derived directly) were estimated by calculating the mean covariance of sensible (Eq. 2) and
22 water vapour fluctuations respectively, with fluctuating vertical velocity (Baldocchi et al., 1988).

23
24 Soil heat flux was measured using two soil heat flux plates (HFT-3, REBS, Seattle, WA, USA) and a
25 system of parallel thermocouples (Type E). The plates were placed at a depth of 0.08 m below the peat
26 surface. The thermocouples were buried at 0.02 and 0.06 m and were used together with volumetric
27 water content (CS615, Campbell Scientific Inc., Logan, Utah, USA) in the upper 0.06 m to estimate the
28 heat stored above the soil heat flux plates. The measurements were stored every 10 s on a datalogger
29 (CR23X, Campbell Scientific Inc., Logan, Utah, USA) and 30-minute averages were computed. During
30 the measurements at the Nkazana Swamp Forest, the groundwater level was deeper than 0.1 m below
31 the surface and therefore, the total G was determined using the calorimetric methodology described by
32 Tanner (1960).

33
34 Over the corresponding time period, R_n was measured above the forest canopy, using a 21.3 m telescopic
35 mast (WT6, Clark Masts Systems Limited, Isle of Wight, England). It was erected within the forest, on
36 a fallen tree stump approximately 2.5 m high (Fig. 3a). This formed a firm base for the 90 kg mast
37 which was carried into the forest from the nearest road approximately 1 km away. The computer box
38 for the EC system (InSitu Flux Systems AB, Ockelbo, Sweden) was installed near the base of the mast
39 (Fig. 3b) and a generator that automatically charged a bank of four 100 Ah deep-cycle lead-acid batteries
40 (accumulators) was positioned approximately 50 m from the site in a predominantly downwind
41 direction (the North-West) to minimise any possible influence from the exhaust fumes on the flux
42 measurements. The generator was controlled by a logger (CR10X, Campbell Scientific Inc., Logan,
43 Utah, USA) which was set to activate the charging system (220VAC petrol generator and 40 A 12 V
44 charger) when the accumulators dropped below 12.4 V.

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

12
 13 The Bowen ratio (β) has historical significance in evaporation studies and is defined as:

$$15 \quad \beta = \frac{H}{LE} \quad (3)$$

16
 17 for a specified time period (Bowen, 1926). It informs on the dominance of H or LE and was calculated
 18 at a daily time interval in this study, providing a useful means of showing changes in the distribution
 19 and weighting of the energy balance components within and between field campaigns.

20 21 2.3 Energy balance closure

22 If each component of the energy balance is measured accurately and independently, then Eq. (1) should
 23 be satisfied, and closure is considered satisfied. However, energy balance closure could still be achieved
 24 if two or more terms have incorrect values and the terms in Eq. (1) still sum to zero (Savage et al.,
 25 2004). If the components of the shortened energy balance equation are measured independently then R_n
 26 $- G - H - LE = c$, where c is termed the energy balance closure (W m^{-2}), and closure is satisfied if $c =$
 27 0 W m^{-2} . By rearranging Eq. (1), closure is not achieved if the available energy $R_n - G$ does not equal
 28 the turbulent fluxes $H + LE$. Another measure of the lack of closure is the closure ratio or the energy
 29 balance closure discrepancy D defined by Twine et al. (2000) as:

$$30 \quad D = \frac{H+LE}{R_n-G} \quad (4)$$

31
 32 in which, a D of 1 indicates perfect closure. Several studies using numerous techniques over various
 33 surfaces have failed to achieve closure by up to 20 or 30 % (Wilson et al., 2001; Wilson et al., 2002;
 34 Barr et al., 2006). The vast majority have found higher energy input by radiation fluxes than loss by
 35 turbulent fluxes (H and LE) and G (Oncley et al., 2007). Therefore, the measured fluxes should be
 36 corrected or the uncertainties in the measured fluxes accepted (Twine et al., 2000). Several reasons for
 37 lack of energy balance closure have been discussed by Twine et al. (2000), Wilson et al. (2002), and
 38 Cava et al. (2008). These reasons include: (1) sampling errors associated with different measurement
 39 source areas for the terms in Eq. 1, (2) a systematic bias in instrumentation, (3) neglected energy sinks,
 40 (4) the loss of low and/or high frequency contributions to the turbulent flux, (5) neglected advection of
 41 scalars, (6) measurement errors related to sensor separation, alignment problems, interference from
 42 tower or instrument-mounting structure, and (7) errors in the measurement of R_n and/or G . Despite
 43 concerns that the direct method of determining total evaporation (ET_{ec}) by measuring water vapour
 44 concentrations using an Infrared Gas Analyzer may result in underestimates or overestimates of LE , in
 45 this study, it was considered that some of the closure pitfalls of the shortened energy balance method,
 46 such as (3) and (7) in particular, could be significant due to the tall canopy at the site (3) and point
 47

1 measurement location (7). Therefore, all ET_a results reported in this paper were calculated by the direct
2 method. Energy balance closure discrepancy was determined during the day-time period ($R_n > 0$) due to
3 the potentially large nocturnal influences reported by Wilson et al. (2002).

4 5 2.4 Measurement of tree sapflow

6 A heat pulse velocity system based on the heat ratio method (Burgess et al., 2001), was used to measure
7 sapflow at various depths across the sapwood of two trees over 20 months from September 2009 to
8 early May 2011 which overlapped with the November 2009 and March 2010 field campaigns (Fig. 1).
9 The trees measured were located approximately 40 m from the mast where the EC and energy balance
10 sensors were installed. Representative trees, in terms of species, stem diameter, canopy height and
11 proximity to each other, were selected given the cable length limitations of the HPV system. The
12 *Syzigium cordatum* tree selected was approximately 22.5 m tall and had a breast height stem diameter
13 of 0.430 m. Sapflow was measured at four depths across the sapwood on both the eastern and western
14 sides of the stem to account for differences in the sapwood depth around the tree. Sapflow was also
15 measured in a nearby understory tree (*Shirakiopsis elliptica*) with a smaller stem diameter (0.081 m) at
16 four depths within the sapwood. Air temperature and relative humidity (HMP45C, Vaisala Inc.,
17 Helsinki, Finland) within the canopy, at a height of 2 m above the ground, and soil volumetric water
18 content (θ) at the *Syzigium cordatum* tree (where the roots were most dense at a depth of 0.075 m), were
19 also measured. These were recorded hourly to coincide with the sapflow measurements. Further details
20 of the installation, equipment used, wounding corrections applied and calculations to derive the tree
21 sapflow are documented in Clulow et al. (2013). In this manuscript, following Dye et al. (2008), sapflow
22 is assumed to equate with tree transpiration and tree water-use.

23 24 2.5 Modelling actual total evaporation from sapflow

25 Polynomial regression (second order) analysis in the Genstat software (VSN International, 2011) was
26 used to describe the relationship between measured ET_a and sapflow of the emergent and understory
27 trees during the overlapping periods of the November 2009 and March 2010 field campaigns in order
28 to understand the possibility of extending the record of ET_a from the Nkazana PSF using the long-term
29 sapflow records. The hourly ET_a and sapflow data was checked for homoscedasticity and required a
30 square root transformation to correct the variance distribution. The model derived was applied over a
31 full year of sapflow data (October 2009 to September 2010) to obtain an annual ET_a (Fig. 1).

32 33 2.6 Evaporation models assessed

34 Two well recognised evaporation models were tested for applicability of modelling ET_a from the
35 Nkazana PSF (Fig. 1). After assessment of the models at an hourly temporal resolution, over the three
36 field campaigns, the most applicable model was applied to the long-term ET_a discussed above and
37 verified using historic data collected by the Council for Scientific and Industrial Research (CSIR) during
38 a preliminary study over the Nkazana PSF from 8 to 12 August 2008 and 12 to 20 November 2008
39 (unpublished). The CSIR measured ET_a with the identical EC equipment used during the field
40 campaigns in 2009 and 2010 described above and at the same site in the Nkazana PSF making the data
41 ideal for verification of the models.

42 FAO Penman-Monteith reference evaporation: The original Penman evaporation model (Penman,
43 1948), assumed an absence of any control on evaporation at the earth's surface, in effect, an open water
44 or wet surface situation. This was extended by Monteith (1965) to incorporate surface and aerodynamic
45 resistance functions applicable to vegetated surfaces and was widely used in this form as the Penman-
46 Monteith model. It is however, highly data intensive (Mao et al., 2002; Drexler et al., 2004) and the
47 model was therefore standardised by the Food and Agriculture Organisation in Irrigation and Drainage
48 Paper No. 56 (Allen et al., 1998) into a form known as the FAO56 Penman-Monteith model that could

1 be applied at both hourly and daily time intervals. The model received favourable acceptance
 2 internationally in establishing a reference evaporation (ET_o) index (atmospheric evaporative demand)
 3 as a function of weather variables measured at most standard weather station systems. The definition of
 4 a reference crop over which the weather variables should be measured was a “hypothetical crop with
 5 an assumed height of 0.12 m having a surface resistance of 70 s m^{-1} and an albedo of 0.23, closely
 6 resembling the evaporation of an extensive surface of green grass of uniform height, actively growing
 7 and adequately watered” (Allen et al., 1998). A nearby crop ET_a is calculated by adjusting ET_o by a
 8 crop factor (K_c) in the form:

$$9 \quad ET_a = ET_o \cdot K_c \quad (5)$$

10
 11
 12 where, the crop is not water-stressed. In Allen et al. (1998), values of K_c have been compiled for
 13 different vegetation types at different stages in crop development. Since recommendations by the
 14 American Society of Civil Engineers Evapotranspiration in Irrigation and Hydrology Committee (Allen,
 15 et al., 2000) and the work by Irmak et al. (2005) and Allen et al. (2006) amongst others, the tall crop
 16 reference (alfalfa height=0.5 m) and separate daytime ($r=50 \text{ s m}^{-1}$) and night-time ($r=200 \text{ s m}^{-1}$)
 17 resistances for hourly calculations were introduced. It was this most recent form of the equation, now
 18 referred to as the FAO Penman-Monteith ET_o , that was applied in the current study.

19
 20 Using the FAO Penman-Monteith ET_o in combination with the long-term ET_a (from sapflow), K_c was
 21 calculated (Eq. 5) for the Nkazana PSF at an hourly interval (while $R_n > 0$ and $ET_a > 0.1 \text{ mm h}^{-1}$) and
 22 summed to daily totals as recommended by Irmak et al. (2005). The reference evaporation approach has
 23 been successful internationally, partly due to technological advances leading to improvements in
 24 temporal and spatial data availability, but also because it provides a method for estimating ET_a , which
 25 is transferrable and can be applied to different vegetation types and locations across the world.

26
 27 Priestley-Taylor potential evaporation: Priestley and Taylor (Priestley and Taylor, 1972) simplified the
 28 theoretical Penman equation for specific conditions. They reasoned that, as an air mass moves over an
 29 expansive, short, well-watered canopy, evaporation would eventually reach a rate of equilibrium. In
 30 this case, where humid air moves over a wet surface, the aerodynamic resistances become negligible,
 31 while irradiance dominates, and the rate of evaporation would be equal to the potential evaporation
 32 (ET_p) which is written as:

$$33 \quad ET_p = \frac{\alpha}{L_v} \cdot \frac{\Delta}{\Delta + \gamma} \cdot (R_n - G) \quad (6)$$

34
 35 where, α is a constant, L_v is the specific latent heat of vaporisation of water (2.45 MJ kg^{-1}), Δ is the
 36 slope of the saturation water vapour pressure versus T_{air} , and γ is the psychometric constant.

37
 38
 39 The definition of the Priestley-Taylor model makes it suitable for estimation of evaporation from open
 40 water areas and wetlands (Price, 1992; Souch et al., 1996; Mao et al., 2002) but it has been applied over
 41 numerous other surfaces such as forests (Shuttleworth and Calder, 1979), cropped surfaces (Davies and
 42 Allen, 1973, Utset et al. 2004), pastures (Sumner and Jacobs, 2005) and even soil water limited
 43 conditions in forest clearcuts (Flint and Childs, 1991) with varied success and deviations from the
 44 originally proposed estimate for α of 1.26. In this study it was applied in the form described by Savage
 45 et al. (1997) where $(\Delta / (\Delta + \gamma))$ was estimated by:

$$\frac{\Delta}{\Delta+\gamma} = 0.413188419 + 0.0157973 \cdot T_{\text{air}} - 0.00011505 \cdot T_{\text{air}}^2 \quad (7)$$

where, T_{air} is average air temperature over the interval of calculation (hourly in this study). By rearranging Eq. (6), and substituting ET_a for ET_p , α 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 (Fig. 1) 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_{air} , RH and soil volumetric water content (θ) to determine individual and combined drivers of sapflow. The log of the sapflow measurements was modelled, as the variance of the measurements themselves was not homoscedastic and therefore required a variance stabilizing transformation. Significance of variables, with up to four-way interactions were considered. In addition, the predictor variables (I_s , RH, T_{air} and θ) were broken up into sets of data with different ranges using regression tree analysis. In regression tree analysis a different model is applied to individual ranges of data rather than a global model (such as in regression analysis), in which a single model is applied to the entire range of each variable. The relationship for a linear regression model is assumed to be the same no matter what the value of any of the predictor variables is. The consequence thereof, is that a good midday relationship between sapflow and a variable (such as I_s) for example, may be missed due to a poor relationship during the early morning and late afternoon periods. The regression tree analysis provides an alternative approach, in which the predictor variables are broken up into different sets, and a different model applied to each individual set. In the regression tree analysis output, which is represented by a hierarchical tree diagram, the longer the line, the greater the difference between the two subsets and the higher in the hierarchy a split occurs, the more significant is the split.

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 (Table 1). 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. 4). A difference in dawn T_{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-minute 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 (Figs. 5a, 5b and 5c). Even the

1 August 2009 data, despite being in the middle of the dry season, was influenced by cloud during six out
2 of the seven days of measurement (not shown). During the August 2009 and March 2010 field
3 campaigns, there was a noticeable dip in the average R_n at approximately 11:00 a.m. LT. with large
4 standard errors ($> 90 \text{ W m}^{-2}$) due to cloud cover. In November, the dip occurred at approximately 1:00
5 p.m. LT., also accompanied by large standard errors ($> 90 \text{ W m}^{-2}$). The cloud affected pattern of R_n was
6 translated through to H and LE, which were positive during the day, and with largest standard errors
7 coinciding with those of the R_n except for the early morning observed LE in August 2009, which was
8 attributed to the evaporation of dew on some days. The maximum rates of LE were approximately 400
9 W m^{-2} in August 2009, 600 W m^{-2} in November 2009 and 700 W m^{-2} in March 2010 (not shown). The
10 pattern of G fluctuated diurnally but due to attenuation (sensors were below the soil surface) the pattern
11 was smoother than the other fluxes during the course of the day.

12
13 During the August 2009 field campaign the daily net radiant density, between 10.2 and 11.8 MJ m^{-2} ,
14 was reasonably consistent at a daily level (Fig. 6a), despite the irregularity observed from the 30-minute
15 data. 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
16 campaigns, the daily net radiant density was more variable (Figs. 6b and 6c). This variability at a daily
17 level was translated through to the H and LE results, which during August 2009 were fairly consistent,
18 but irregular during November 2009 and March 2010. The average daily net radiant density was lowest
19 in August 2009 (11.2 MJ m^{-2}), highest in November 2009 (15.1 MJ m^{-2}) and in-between during March
20 2010 (12.7 MJ m^{-2}). The average daily soil heat flux did not mirror the pattern of R_n and was highest in
21 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
22 m^{-2}) and lowest in November 2009 at 1% of R_n (up to 0.3 MJ m^{-2}).

23
24 The daily total LE was higher than H in August 2009 (Fig. 6a), with a daily average β ratio of 0.7 (0.4
25 to 0.9). In November 2009 (Fig. 6b) the daily average β ratio was higher with a daily average of 0.9
26 (0.5 to 1.3) but in March (Fig. 6c) however, LE dominated the energy balance with an average β ratio
27 of 0.4 (0.1 to 0.6).

28
29 Closure discrepancy was different for each field campaign. In August 2009 the D was 0.98 indicating
30 exceedingly good closure. However, the second and third field campaigns produced D 's of 1.18 and
31 1.33 in November 2009 and March 2010, respectively, indicating (1) either an overestimation of LE
32 and/or H , and/or an underestimation of the available energy ($R_n - G$) and/or (2) unaccounted energy
33 such as advection or storage in the canopy biomass.

34

35 3.3 Measured actual total evaporation

36 The mean daily ET_a over the three field campaigns was significantly different (based on their 95%
37 confidence interval). The daily ET_a (Fig. 7) was lowest in the August 2009 (winter) and increased
38 progressively through November 2009 (early summer) to March 2010 (late summer). The standard
39 deviation (SD) for all field campaigns was similar (0.3 to 0.4 mm) but the coefficient of variation (not
40 shown) differed with the highest in November 2009 (12.0) and August 2009 (11.0) and lowest in March
41 2010 (8.8).

42

43 3.4 Relationship between sapflow and actual total evaporation measured during two field campaigns

44 The diurnal courses of the sapflow from the emergent and understory trees were surprisingly smooth in
45 comparison to the ET_a results (Figs. 8a-d). The ET_a is an integrated measure of soil evaporation and
46 transpiration from numerous plants at different levels within the canopy over the contributing area
47 described by the footprint, whereas the transpiration measurements (assumed to equal sapflow) describe
48 the physiology of a single tree. The R_n , frequently considered a significant driver of tree physiology,

1 fluctuated due to cloud cover (Figs. 5a, 5b and 5c). These fluctuations were not translated into
2 fluctuations in tree sapflow but are evident in the ET_a results particularly over the midday period. A
3 similar pattern was observed in the March 2010 ET_a data (Figs. 8c and 8d). It is recommended that G
4 be measured at numerous positions under swamp forest canopies in order to capture the variability in
5 G and a representative average.

6
7 Despite the greater midday variability of the ET_a data, the polynomial regression (least squares) between
8 hourly ET_a and tree sapflow showed a strong relationship in November 2009 for the emergent tree
9 (RMSE=0.05 mm h⁻¹) as well as the understory tree (RMSE=0.06 mm h⁻¹). The polynomial regression
10 was convex ($R^2=0.89$) rather than linear ($R^2=0.87$) in the case of the emergent tree (Fig. 9a) and concave
11 ($R^2=0.92$) rather than linear ($R^2=0.90$) in the case of the understory tree (Fig. 9b). The increase in the
12 rate of sapflow of the emergent tree was exponential for lower values of ET_a (morning and evening) but
13 the rate of sapflow versus ET_a for higher values of ET_a slowed down as the tree reached its peak
14 transpiration rate. In contrast the understory sapflow rate increased gradually per unit increase in ET_a
15 at lower values but at higher values of ET_a the increase in sapflow was exponential. In March 2010 the
16 results were similar with RMSE's of 0.07 mm h⁻¹ and 0.08 mm h⁻¹ for the emergent and understory
17 trees, respectively. Convex and concave trendlines again fitted the data best (Fig. 9c and 9d). Lagging
18 the sapflow by one hour as suggested by Granier et al. (2000) did not improve the regression of sapflow
19 on ET_a .

21 3.5 Comparison of the FAO Penman-Monteith reference evaporation versus the Priestley-Taylor 22 potential evaporation during the three field campaigns

23 The linear regression (least squares) of the hourly ET_o against hourly ET_a , explained 75%, 85% and
24 76% of the fluctuations in ET_a during the August 2009, November 2009 and March 2010 field
25 campaigns, respectively (Table 2). The Priestley-Taylor model did not perform as well, accounting for
26 54%, 74% and 62% of the variation in ET_a during the August 2009, November 2009 and March 2010
27 field campaigns, respectively (Table 2).

28
29 The slope of the linear regression (K_c) varied between field campaigns (Table 2) and was highest in
30 March (1.3), and lower in November 2009 (1.1) and August 2009 (0.8). The α , also estimated by the
31 slope of the linear regression, was similar during the August (1.0) and November 2009 (1.0) field
32 campaigns (Table 2) while during March 2010, α was slightly higher (1.1). The standard deviations and
33 root mean square errors of α were higher than those of the K_c (Table 2). Therefore, the FAO Penman-
34 Monteith ET_o model was adopted as most suitable for use over the Nkazana PSF in this study.

35
36 The time interval (hourly and daily) at which the FAO Penman-Monteith ET_o and Priestley-Taylor
37 models were computed resulted in different K_c and α estimates. Daily computations used average
38 daytime T_{air} , typically derived from an average of maximum and minimum daily T_{air} . In this research
39 the models were run hourly and the average T_{air} derived from 10 s measurements of T_{air} (while $R_n > 0$)
40 accurately representing that hour. However, using hourly data produced outliers in the calculation of K_c
41 and α at the beginning or end of a day where the measured or modelled results are very small numbers,
42 producing, from division, erroneous estimates of K_c and α (Eqs. 5 and 6). These typically occurred near
43 sunset or sunrise and were filtered out of the data as they represented outliers. In addition, due to the
44 vastly different canopy structures and heights within the Mfabeni Mire, of the SFW (~0.8 m) and
45 Nkazana PSF (~20 m), climatic data from above the forest was used as an input to the models to
46 determine whether the SDs of K_c and α could be minimised, but no significant improvement was found.
47 This indicated that the nearby weather station data (from within the SFW) was a suitable input for both

1 models supporting the application of these models using the standard FAO Penman-Monteith ET_o
2 weather station sensor heights of two meters (Allen et al., 2006).

3 4 *3.6 Modelling long-term actual total evaporation and monthly crop factors*

5 The long-term ET_a (October 2009 to September 2010) was modelled through the relationship between
6 the observed ET_a and observed sapflow over the November 2009 and March 2010 field campaigns. In
7 regressions of the emergent tree sapflow with ET_a over the two field campaigns (Figs. 9a and 9c), it
8 was found that there was little gain in using separate linear models for the two periods ($R^2 = 0.92$ and
9 0.89 ; $RMSE = 0.05 \text{ mm h}^{-1}$ and 0.06 mm h^{-1}) as a single, combined model, described ET_a equally well
10 ($R^2=0.90$; $RMSE = 0.07 \text{ mm h}^{-1}$). A similar result was found for the understory tree, indicating that for
11 both trees a single relationship between ET_a and sapflow represented both field campaigns.

12
13 In addition, a multiple regression, including the emergent and understory trees as predictors of ET_a ($R^2=$
14 0.91 ; $RMSE=0.08 \text{ mm h}^{-1}$), provided insufficient benefit over the use of the single model based on only
15 the emergent tree ($R^2=0.90$; $RMSE=0.08 \text{ mm h}^{-1}$). The understory tree sapflow was considerably less
16 (by 85%) than that of the emergent tree and the density of the understory trees within the Nkazana PSF
17 is much lower than the emergent trees. These results support the omission of the understory tree from
18 the prediction of ET_a , and the use of the following model to estimate the ET_a of the Nkazana PSF from
19 hourly sapflow data:

$$20 \quad \quad \quad 21 \quad \quad \quad ET_a = (0.16341 \cdot T_r + 0.06)^2 \quad \quad \quad (8)$$

22
23 where, ET_a is the actual total evaporation (mm hr^{-1}) and T_r the emergent tree sapflow (L hr^{-1}).

24
25 The annual ET_a (October 2009 to September 2010) from the Nkazana PSF was 1125 mm, over which
26 period the rainfall was 650 mm (well below the long-term average, reported to be between 844 and
27 1200 mm a^{-1} for the area). Finally, K_c was calculated at a daily interval from the extended ET_a and ET_o
28 (Eq. 5), and averaged for each month of the year (Fig. 10). These results equated well with the results
29 of K_c calculated during the field campaigns (Table 2) which were 0.8, 1.0 and 1.3 in August, November
30 and March, respectively. During a distinct period from February to May, K_c was between 1.2 and 1.4
31 while for the rest of the year it was 0.8 to 1.0. When $K_c = 1$ the Nkazana PSF ET_a equals the evaporative
32 demand, or in other words ET_o . However, a K_c of <1 or >1 indicates that the PSF ET_a is $<$ than or $>$
33 than the ET_o , respectively. Fig. 10 shows that the Nkazana PSF ET_a is at or just less than ET_o for eight
34 months of the year (June to January) and greater than ET_o for four months of the year (February to
35 May).

36
37 The derived crop factors were verified using independent measurements of ET_a over the Nkazana PSF
38 collected during window periods at the same site from 8 to 12 August 2008 and 12 to 20 November
39 2008 in an experimental unpublished study conducted by the CSIR. The surface conditions during 2008
40 within the Nkazana PSF were much wetter as the water table was close to the surface with open water
41 in low lying areas whereas in 2009 and 2010 the dry period had caused the water level to drop resulting
42 in only a few areas of open water within the forest. Despite this difference in groundwater level, the K_c
43 was 0.8 in August of 2008 and 0.9 during November 2008, validating the results derived for K_c (from
44 the extended record of ET_a modelled from sapflow of the emergent tree), thus confirming that the K_c
45 derived was applicable across wetter and drier years.

46 47 *3.7 Response of sapflow to climatic variables*

1 Equation 8 enabled the derivation of ET_a over the period during which there were sapflow
2 measurements (October 2009 to September 2010). The purpose for this was to better understand the
3 relationship between important climatic variables and ET_a . Three statistical approaches were used to
4 determine these relationships with sapflow, which were directly related to ET_a and the climatic
5 variables. The simple linear regressions of daily sapflow were considered with radiant flux density and
6 VPD and it was found that these were poor, with coefficients of determination of only 0.51 and 0.52
7 respectively (not shown). Clearly the relationship between climatic conditions and sapflow is more
8 complex. By applying multiple regression analysis I_s , RH, T_{air} and θ at 0.075 m were found to be
9 significant ($p < 0.001$) with up to four-way interactions. Finally a regression tree analysis was applied of
10 hourly log-transformed sapflow with the meteorological variables I_s , RH, T_{air} and θ (Fig. 11). This
11 showed again that the relationships are complex but that T_{air} and θ were not required for the optimal
12 split for the Nkazana PSF emergent tree sapflow. The most important split was between data with I_s of
13 less than 55.7 W m^{-2} and data with I_s greater than 55.7 W m^{-2} . Solar irradiance was clearly a key variable
14 to include and the first split observed, essentially separates day and night-time data. Solar irradiance
15 was also highly correlated with T_{air} , which may be the reason T_{air} was not found to be an additionally
16 required variable. The next important splits were for RH above and below 93.2% for the night time data
17 (essentially when it is raining and when it is not) and an additional split for I_s above and below 279.2 W
18 m^{-2} for the day time data; therefore splitting day time data during high and low irradiance periods. At
19 night the logged sapflow was found to be negative, with the greatest negative average logged sapflow
20 when the RH was less than 96.4%. The greatest average positive logged sapflow was found to be when
21 I_s was greater than 279.2 W m^{-2} , and this occurred 28% percent of the time.

22

23 **4 Discussion**

24

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

38

39 The challenge remained in interpolating and extrapolating the ET_a results from the EC system to annual
40 ET_a . In long-term evaporation studies where gaps occur or where window periods have been used, and
41 interpolation of the ET_a record is required, meteorological models are typically used. Total evaporation
42 has been estimated using models that are computationally simple such as the Priestley-Taylor model
43 (Priestley and Taylor, 1972; Shuttleworth and Calder, 1979) to more complex models using multi-layer
44 approaches within the canopy, but still based on the Penman-Monteith approach (Roberts et al., 1993;
45 Harding et al., 1992), with significant deviations between measurements and modelled results. These
46 meteorological models are, however, uncoupled from the transpiring vegetation and therefore the
47 pattern of actual tree sapflow was considered in this study as a predictor of ET_a .

48

1 External regulation of sapflow has been described by numerous variables including the readily available
2 soil water of the rooting area (Oren and Pataki, 2001), the micrometeorological conditions of the
3 atmosphere (Lundblad and Lindroth, 2002), leaf area (Granier et al., 2000), canopy conductance
4 (Granier et al., 2000), aerodynamic resistance (Jacobs and De Bruin, 1992; Hall, 2002), shading of
5 lower leaves (Cienciala et al., 2000) and wind stress (Kim et al., 2014). However, it has been found that
6 trees can have several mechanisms of internal regulation related to species specific morphology and
7 physiology that is partially uncoupled from the external conditions (Zweifel et al., 2005). Nevertheless,
8 in most trees with actively transpiring leaves and some readily available soil water, a diurnal pattern of
9 sapflow results from a combination of internal and external conditions, which determines how a tree
10 contributes to the ET_a of a forest stand.

11

12 With advances in sapflow measurement techniques, long-term forest ET_a has been estimated by up-
13 scaling from tree transpiration to forest ET_a using various techniques generally based on sapwood area
14 (Čermák et al., 2004). However, the large majority of these studies, especially where tree transpiration
15 has been up-scaled, have been conducted in uniform forest stands (Oren et al., 1999; Wilson et al.,
16 2001) and much of the work has taken place in temperate boreal stands (Lundblad and Lindroth, 2002;
17 Launiainen et al., 2011) and their applicability to other climatic zones needs consideration. In addition,
18 it has also been recognised that transpiration often varies amongst species (Oren and Pataki, 2001;
19 Ewers et al., 2002, Bowden and Bauerle, 2008) and up-scaling to forest transpiration in species-rich
20 indigenous forests is complex.

21

22 The results from this study showed that the hourly sapflow of a single emergent tree, selected as a
23 dominant species, correlated well with the hourly ET_a measured over two window periods. In species
24 rich forests, measuring the sapflow of the different vegetation types (including the ferns, vines,
25 understory and emergent trees) would be challenging, and up-scaling questionable, due to the variety
26 of plant structures within the canopy and our lack of information on the plant physiologies. Therefore,
27 the empirical relationship between the single tree and ET_a provided an ideal opportunity to derive the
28 annual ET_a of a vegetation type for which there is no information of the water-use characteristics. This
29 relationship indicated that the emergent canopy trees are the main contributors to ET_a . The other
30 contributors to ET_a , including open water, peat, ferns, vines and understory trees were either (1)
31 insignificant contributors due to the low irradiance and VPD below the emergent tree canopy (supported
32 by the low measured sapflow rate of the understory tree), or (2) follow similar diurnal trends in
33 evaporation and sapflow as the emergent tree (also supported by the diurnal trend in the sapflow rate of
34 the understory tree) and are therefore captured in the empirical model of the emergent tree.

35

36 Variation of the energy balance closure discrepancy (D) occurred between field campaigns, despite
37 replication of the same instrumentation at the same site and with the same data processing procedures.
38 Only the placement of the soil heat flux sensors changed slightly within the vicinity of the site between
39 field campaigns. However, the soil heat fluxes (as a percent of net irradiance) fluctuated from 1% to
40 11%, likely due to the specific placement of the sensors within the Nkazana PSF in a predominantly
41 shaded area in contrast to a sunlit location due to gaps in the canopy. During August 2009 when $D = 1$
42 (i.e. perfect closure of the energy balance), the soil heat flux was approximately 5% of R_n and was likely
43 to be the most representative result for G for a forested area, so agreeing with Dye et al. (2008). In
44 March, G was 11% of R_n and may have contributed to the poorest result of $D = 1.33$. Wilson et al.
45 (2002) found that energy balance closure, especially over forests, is seldom achieved. However, in most
46 cases the magnitude of the long-term turbulent fluxes is lower than the available energy (Twine et al.,
47 2000; Oliphant et al., 2004), which was not the case in the Nkazana PSF study where D increased with
48 increasing ET_a from August 2009 through November 2009 to March 2010.

1
2 An important observation, made over the three field campaigns, was that the average ET_a measured
3 during March 2010 (4.4 mm day^{-1}) did not correspond to the period of highest R_n (November 2009),
4 which is commonly accepted to be one of the main driving variables in the process of ET_a (Albaugh et
5 al., 2013). This may indicate a lag in the ET_a of the Nkazana PSF in relation to the maximum R_n , possibly
6 explaining the poor relationships observed between tree sapflow and climatic variables (such as R_n).
7 This lag was also observed in the high K_c values from February to May, where the ET_a of the PSF was
8 higher relative to ET_o . Typically, K_c is higher while vegetation is more actively transpiring and is
9 associated with higher I_s and water availability, which in the Nkazana PSF would coincide with the
10 summer period (October to March). However, the period of higher K_c values in the Nkazana PSF
11 occurred quite late (February to May) in the summer season (Fig. 10). Clulow et al. (2013) showed the
12 Nkazana PSF sapflow to be relatively consistent between seasons but that ET_o rapidly decreased from
13 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
14 ET_o measured at the meteorological station, while transpiration rates in the Nkazana PSF were
15 maintained into the late autumn period. A number of reasons may be attributed to this including the
16 microclimate of the Nkazana PSF. For example, the lower energy loss at night from the ground and
17 within the canopy, due to the combined effect of high water vapour levels (a greenhouse gas) and
18 reduced infrared emission as a result of canopy absorbance, reflectance and re-emission downwards,
19 compared to areas outside the PSF with shorter canopies, resulted in higher minimum daily
20 temperatures (Fig. 4). The area adjacent to the Nkazana PSF where the automatic weather station was
21 located (with a shorter canopy of approximately 1 m in height) experienced lower daily minimum
22 temperatures (Fig. 4). The importance of this result is that T_{air} affects biochemical processes such as
23 photosynthesis and senescence. This T_{air} difference, although greatest in winter, starts to build in
24 January and could play a role in influencing the ET_a in relation to the summer season as well as the
25 period of higher K_c values in the latter half of summer.

26
27 Two important points regarding the weather station data and model calculations were noted. Firstly,
28 where possible, hourly model time intervals should be used, which concurs with Irmak et al. (2005).
29 However, this frequently resulted in outliers in K_c and α at the beginning or end of a day where the
30 measured or modelled results were small numbers, producing, through division, erroneous estimates. It
31 was therefore favourable to sum the hourly ET_o and ET_a data for each day (while $R_n > 0$) and calculate
32 the daily K_c (which was then averaged for each month). Secondly, when calculating the K_c and α
33 coefficient's, there was no benefit in using the climatic data from above the tree canopy rather than the
34 climatic data from the adjacent SFW of the Mfabeni Mire, which sufficiently represented the
35 microclimate for the model calculations. This showed that data from nearby weather stations can be
36 used with the K_c to estimate the ET_a although this may only hold in humid environments where there is
37 little difference in the VPD of the boundary layer conditions over the Nkazana PSF and the surrounding
38 wetland areas, which are likely to all be at, or near, equilibrium evaporation.

39
40 The shape of the regressions of hourly ET_a versus hourly sapflow (Fig. 9) during the window periods
41 showed that sapflow of the emergent tree responded rapidly for low conditions of ET_a . These conditions
42 occurred most frequently in the early morning and late afternoon when the angle of the I_s was low but
43 still incident upon the emergent tree leaves. At higher rates of ET_a the sapflow peaked as the physiology
44 of the tree limited the sapflow rates. In contrast, the understory tree sapflow rate increased slowly
45 relative to ET_a , while ET_a was low, and exponentially for higher values of ET_a . This is likely due to
46 shading of the understory trees for low sun angles (early morning and late afternoon) with I_s limiting
47 transpiration (together with the low VPD discussed above) with maximum rates occurring when shading
48 by the emergent trees was at a minimum (noon) and ET_a was at a maximum. These different responses

1 of the trees indicated that a model to derive ET_a from sapflow would require the inclusion of both
2 emergent and understory trees. However, the sapflow and density of the understory trees was much
3 lower than the emergent trees and therefore its inclusion in the empirical model was not found to
4 significantly improve the relationship between entire canopy ET_a and sapflow. This conclusion applies
5 specifically to the Nkazana PSF. Some models such as the WAVES model permits two canopy
6 simulations due to the importance of the understory canopy in some forest sites (Dye et al., 2008).

7
8 Within South Africa, the study by Dye et al. (2008) measured daily ET_a of between 2 mm and 6 mm on
9 clear days over three field campaigns during February, June and October 2004, which are comparable
10 with the results from the Nkazana PSF of between 2.2 mm (August 2009) and 5.1 mm (March 2010).
11 Internationally, no results of ET_a or modelling guidelines for peat swamp forests were found, signifying
12 the unique contribution of this study.

13
14 The comparison of meteorological variables with sapflow revealed that it is unlikely that a single
15 climatic variable is able to determine sapflow, and in turn ET_a . The relationships were revealed to be
16 non-linear and that to model sapflow accurately, data needs to be sub-set into different periods; at least
17 into day and night.

18 **5 Conclusions and opportunities for further research**

19
20
21 This study portrayed the difficulties of using the most advanced systems available to measure ET_a , such
22 as EC, in remote and difficult-to-access areas. It showed that intensive window period measurements
23 using high maintenance EC systems provide reliable and continuous measurements of ET_a but require
24 a method to determine the ET_a during the in-between periods to be able to estimate long-term ET_a . This
25 was overcome by measuring the long-term sapflow of an emergent canopy tree and deriving a
26 qualitative model for ET_a based on sapflow measurements. Further research on the benefit of measuring
27 multiple emergent trees and the possible variability of transpiration within different species and the
28 extent to which this could improve the long-term estimate of forest ET_a together with window periods
29 of EC data would be beneficial.

30
31 Energy balance closure discrepancy (D) remains an unresolved matter which affects flux measurements
32 such as ET_a and CO_2 . Corrections suggested in research studies can be applied but without conclusively
33 identifying the source of the error in the observations. In contrast to most studies reported, the closure
34 discrepancy of the energy balance over the Nkazana PSF was greater than 1 for two of the field
35 campaigns. Although attributed in part to unrepresentative G measurements, D increased as ET_a
36 increased.

37
38 The model used to derive the annual ET_a from sapflow (Eq. 8), and then monthly crop factors was
39 verified with data from two independent field campaigns in 2008, when conditions were much wetter
40 and there were larger areas of open water within the forest. The much wetter conditions in 2008 did not
41 alter the K_c thus indicating that the relationship between ET_a and ET_o remained constant and that the K_c
42 derived can be applied over a range of climatic conditions. In addition, it indicates that the humid, low
43 VPD environment within the forest canopy minimises the contribution of open water evaporation within
44 the forest to ET_a . However, the general dearth of information on the ET_a of subtropical indigenous
45 forests internationally allows little comparison of the results obtained from the Nkazana PSF and similar
46 forest types and knowledge of the extent to which these crop factors can be extrapolated geographically
47 and to similar forests would benefit from further comparisons.

1 The Mfabeni Mire is actively managed by the iSimangaliso Wetland Park. These results provide the
2 basis for improved estimates of the ET_a component of the Nkazana PSF water balance and the
3 environmental water requirements. Water is critical to the functioning of this ecosystem for biotic and
4 abiotic life, the sequestration or release of carbon from the Mire and also to the spread of fires. The
5 annual ET_a estimated in this study (1125 mm) was even higher than the range (844 to 1200 mm yr⁻¹) of
6 reported estimates of mean annual precipitation for the area (Lynch, 2004; Taylor et al., 2006; ARC-
7 ISCW, 2011).

8
9 The difference between ET_a and rainfall highlights the importance of the groundwater contributions and
10 the critical role it plays in assuring the survival of this groundwater dependant ecosystem. The
11 groundwater available to the Mfabeni Mire is in part determined by the management of the upstream
12 catchments and the groundwater levels of the greater Zululand Coastal Aquifer, emphasising the need
13 for an integrated catchment management approach to the area.

14
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1 Table 1. Summary of weather conditions during the August 2009, November 2009 and March 2010
 2 field campaigns

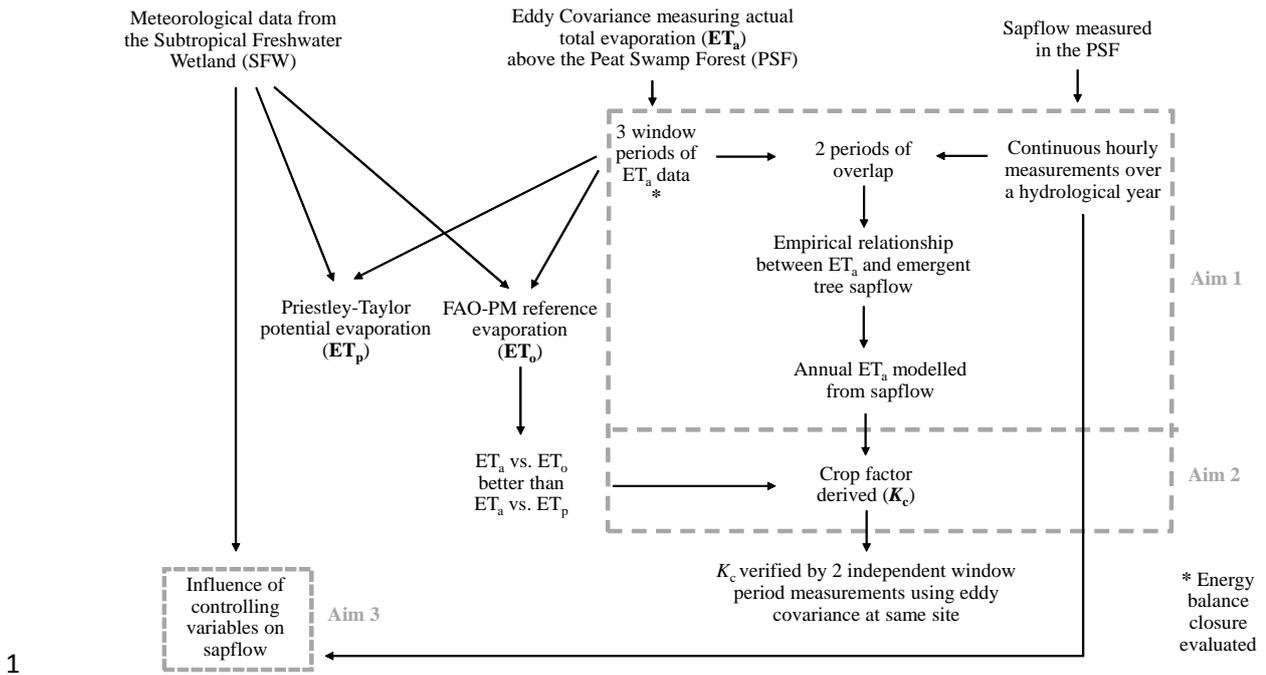
Date	Solar radiant density (MJ m ⁻²)	Wind speed (m s ⁻¹)	Wind direction (°)	VPD (kPa)	Air temperature (°C)		RH (%)		Rain (mm)
					Max	Min	Max	Min	
13/8/2009	14.2	4.6	199	1.5	21.9	13.0	88.7	33.7	
14/8/2009	15.4	2.4	194	1.2	21.8	9.8	95.5	43.3	
15/8/2009	15.3	1.9	83	1.1	22.4	7.0	98.9	50.7	
16/8/2009	14.9	2.5	74	1.1	23.3	11.2	97.0	46.5	
17/8/2009	15.9	3.2	38	1.0	22.6	10.4	98.1	48.9	0.8
18/8/2009	14.8	4.3	33	1.0	23.0	14.1	95.0	52.5	
19/8/2009	15.5	5.5	28	1.2	24.5	13.2	96.3	45.8	
Average	15.1	3.5		1.2	22.8	11.1	95.6	45.9	
4/11/2009	22.0	2.1	115	0.9	23.8	13.2	96.4	51.9	
5/11/2009	16.7	4.2	40	0.6	25.2	17.7	93.3	66.7	
6/11/2009	18.1	5.3	37	0.6	25.8	19.9	93.3	73.0	
7/11/2009	19.0	6.9	36	0.6	25.9	21.6	93.2	70.5	
8/11/2009	25.3	7.4	37	0.7	26.4	21.3	90.5	69.0	
9/11/2009	21.1	6.2	34	0.7	25.3	21.0	94.7	69.2	5.3
10/11/2009	16.2	4.4	223	0.6	24.7	19.4	95.6	61.6	0.3
11/11/2009	21.2	2.3	47	0.5	26.1	15.5	97.1	64.6	
Average	22.8	3.1		0.8	28.3	18.4	94.9	58.2	
16/3/2010	19.6	3.5	66	1.0	27.9	20.0	92.8	65.2	1.5
17/3/2010	14.6	2.3	245	0.8	28.7	18.4	96.0	60.3	4.6
18/3/2010	17.2	2.2	214	0.8	26.4	17.7	94.1	65.5	
19/3/2010	20.2	3.0	58	0.8	28.6	16.1	96.8	62.6	
20/3/2010	20.6	3.8	66	1.0	30.0	22.1	93.2	59.0	
21/3/2010	16.2	1.9	90	1.0	28.6	21.9	92.8	59.9	
22/3/2010	22.5	1.7	97	0.8	28.9	16.4	96.7	59.6	0.3
23/3/2010	19.9	2.3	234	1.0	28.8	16.8	97.3	63.9	
24/3/2010	21.2	2.1	79	1.0	30.2	18.6	96.7	56.6	
Average	19.1	2.5		1.1	28.7	18.7	95.2	61.4	

3

1 Table 2. Summary of the hourly crop co-efficient K_c and advective term α with standard deviation and
 2 root mean square error (RMSE) for each of the three field campaigns.

	K_c	K_c			α	α		
		Coefficient of determination	Standard deviation	RMSE		Coefficient of determination	Standard deviation	RMSE
Aug 2009	0.8	0.75	0.22	0.07	1.0	0.54	0.35	0.08
Nov 2009	1.0	0.85	0.17	0.07	1.0	0.74	0.34	0.11
Mar 2010	1.3	0.76	0.39	0.11	1.1	0.62	0.46	0.13

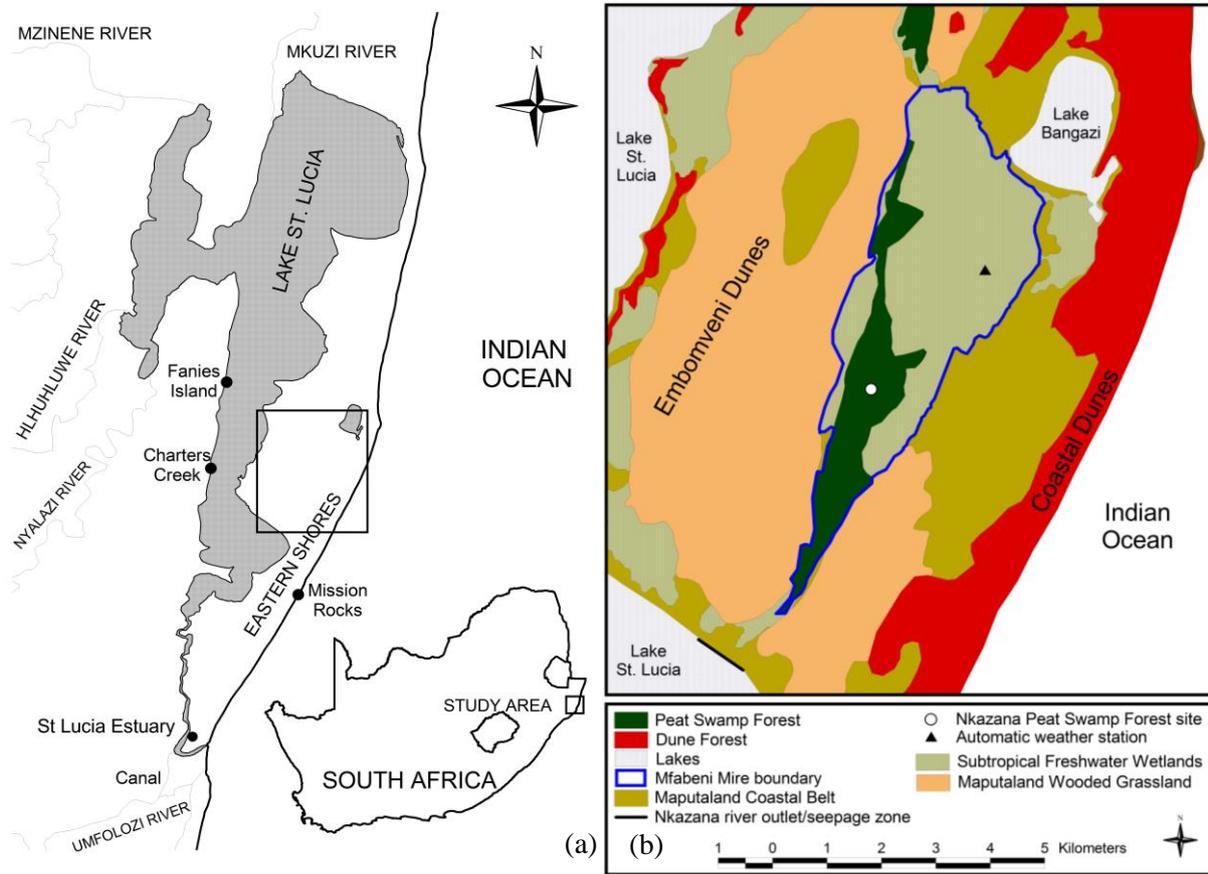
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2 Fig. 1. Flow diagram of the research strategy indicating the different measurement techniques, their
 3 combination, and link to the aims of the research.

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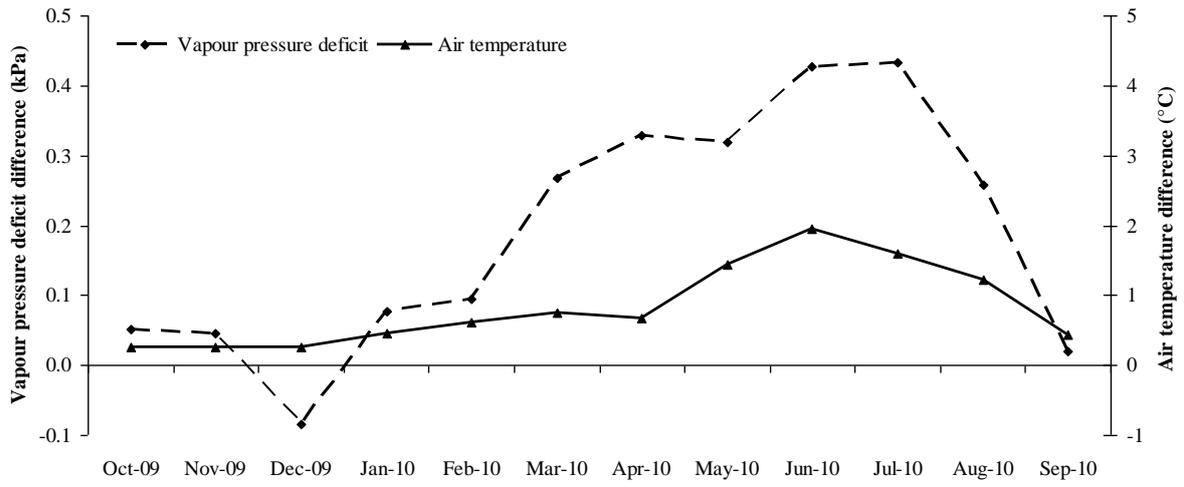


2 Fig. 2. (a) Location of the Eastern Shores within South Africa, (b) the Nkazana Peat Swamp Forest site
 3 (where the EC and sapflow systems were located) and the automatic weather station within the Mfabeni
 4 Mire on the Eastern Shores (data from Mucina and Rutherford, 2006).
 5



1
2 Fig. 3. (a) Telescopic mast (21.3 m) erected in the swamp forest to raise the eddy covariance instruments
3 above the forest canopy, (b) the computer installed at the swamp forest, housed in a temperature
4 controlled enclosure and (c) the instruments attached to the head of the mast.
5

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2

3

4 Fig. 4. The difference in the monthly daytime (9am to 3pm) vapour pressure deficit and difference
5 between the monthly average dawn air temperatures measured in the subtropical freshwater wetland
6 area of the Mfabeni Mire (reeds, sedges and grasses) and within the canopy of the Nkazana Peat Swamp
7 Forest site.

8

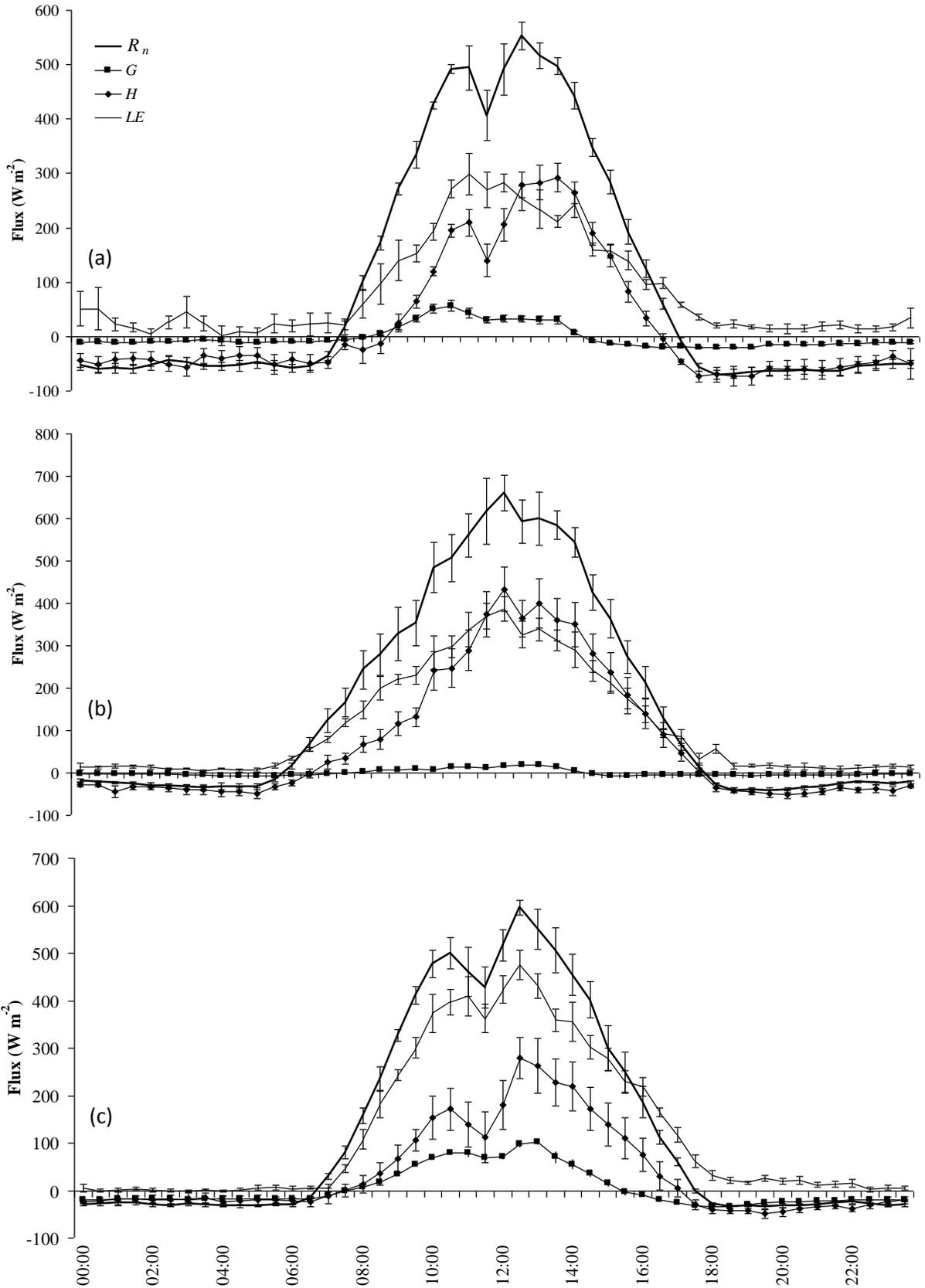
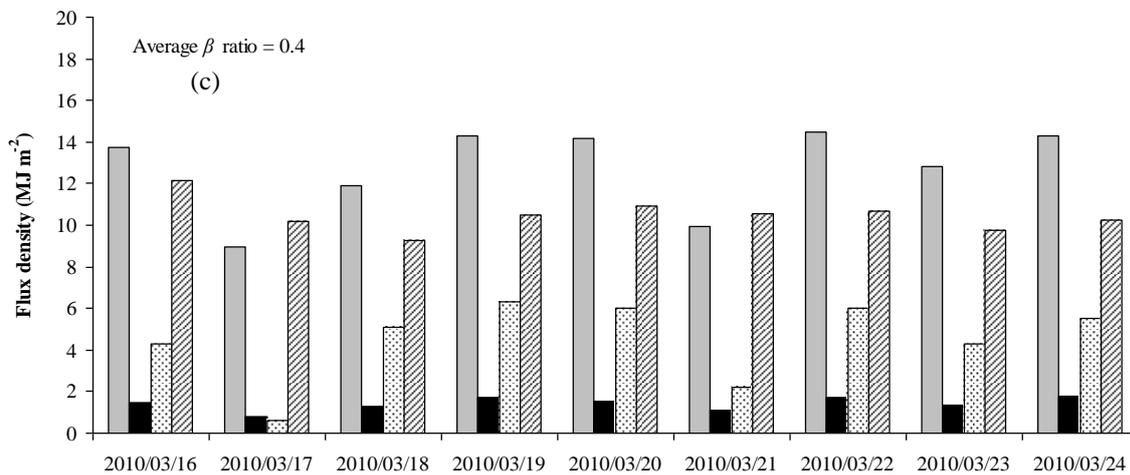
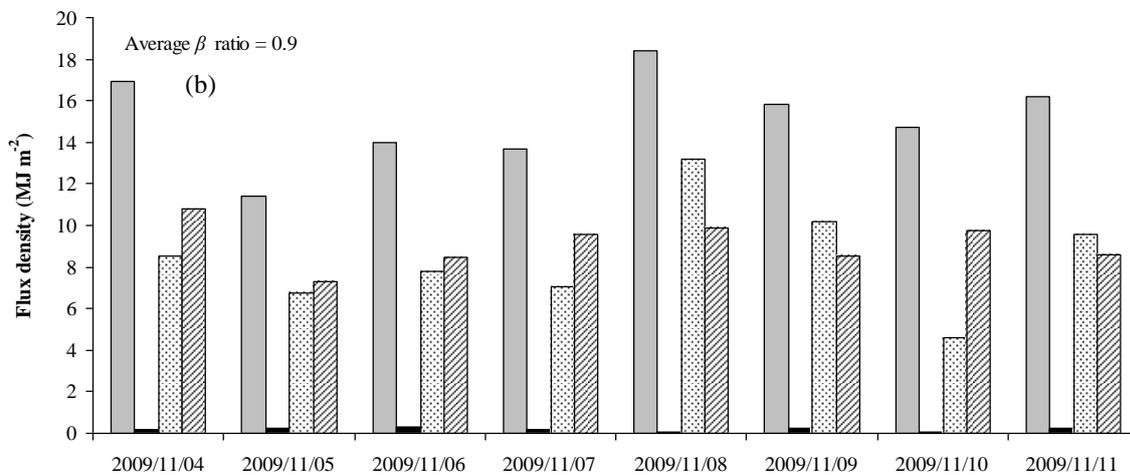
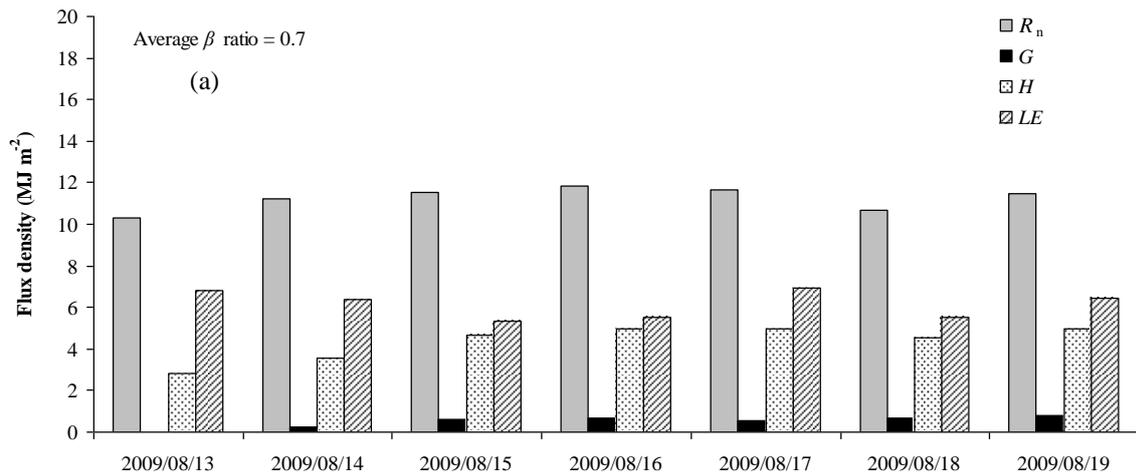
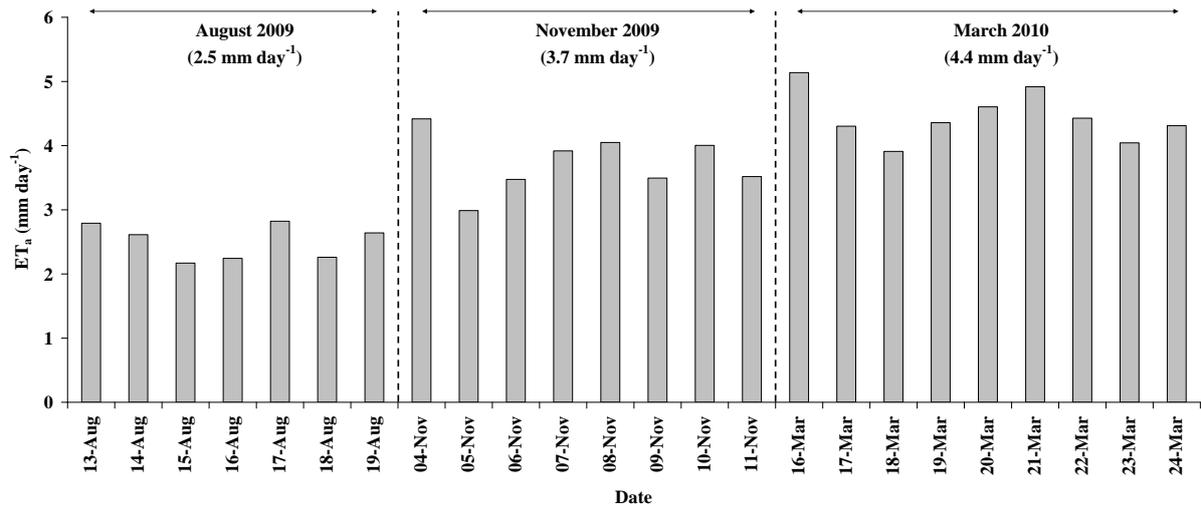


Fig. 5. The average of the half-hourly energy fluxes, with error bars indicating the standard error, measured at the Nkazana Swamp Forest in (a) August 2009, (b) November 2009 and (c) March 2010.



4 Fig. 6. Daily total energy densities (while $R_n > 0$) measured at the Nkazana Swamp Forest in (a) August
 5 2009, (b) November 2009 and (c) March 2010.

6



1
 2 Fig. 7. Daily actual total evaporation (ET_a) measured over the Nkazana Swamp Forest during three
 3 representative periods.
 4

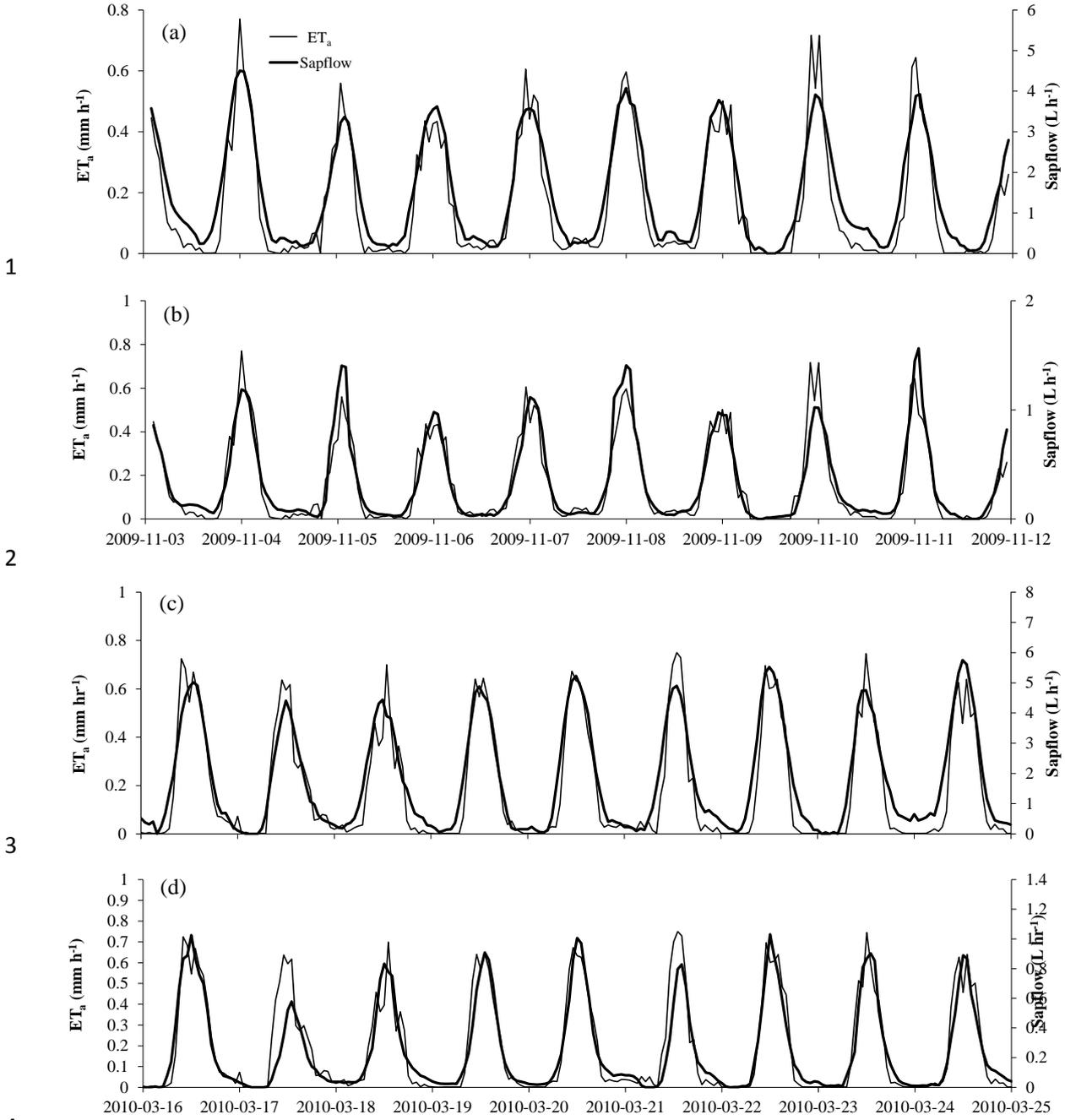


Fig. 8. Diurnal course of the hourly actual total evaporation (ET_a) and sapflow in November 2009 (a and b) and March 2010 (c and d) for the emergent and understory trees, respectively.

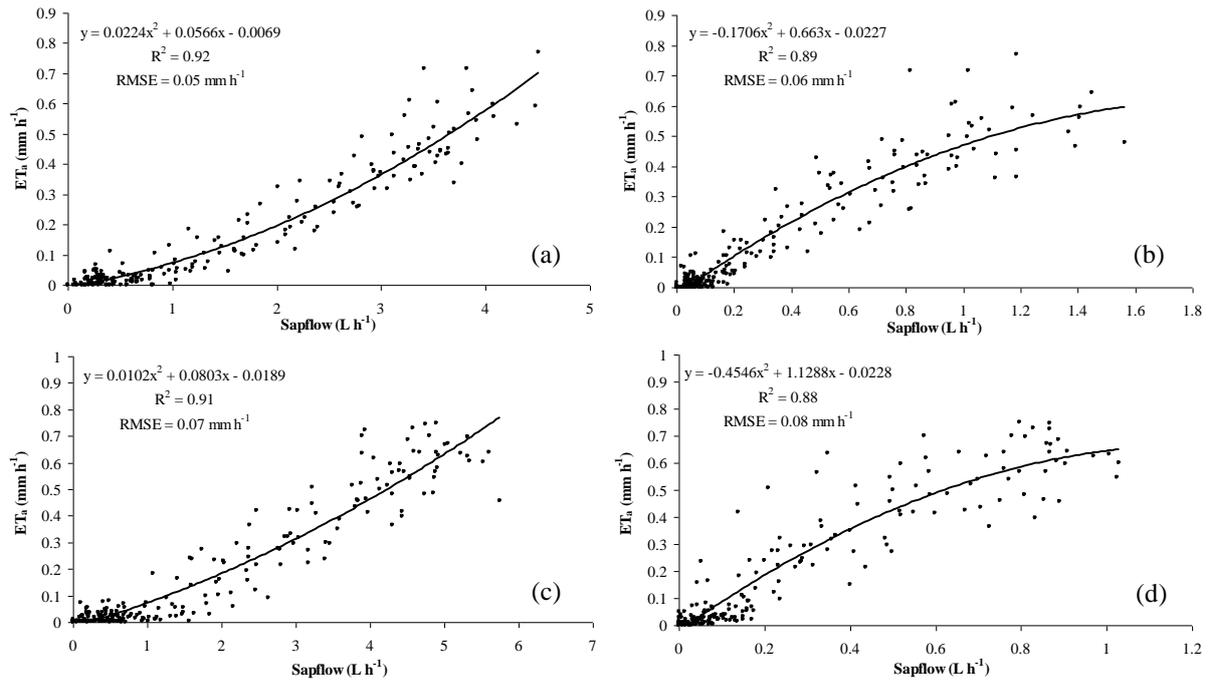
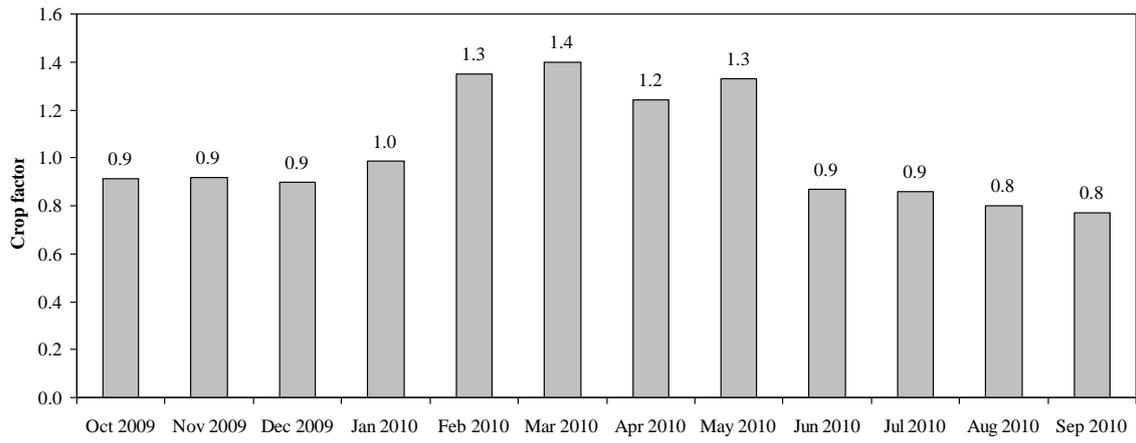
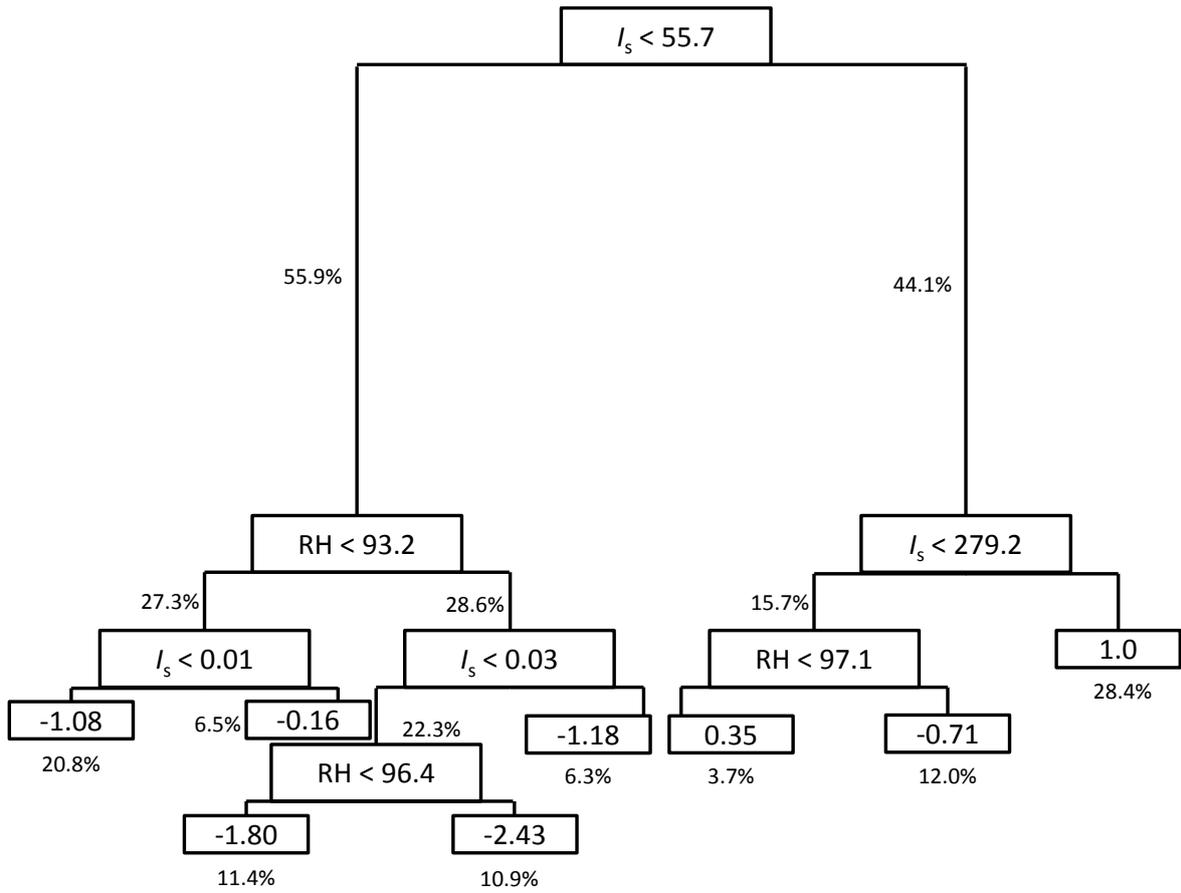


Fig. 9. Polynomial regressions of actual total evaporation (ET_a) against the hourly sapflow for the (a) emergent and (b) understory trees during November 2009, and the (c) emergent and (d) understory trees in March 2010.



1
2 Fig. 10. Monthly crop factors K_c for the Nkazana Peat Swamp Forest.
3



1
 2 Fig. 11. A regression tree analysis for sapflow showing the optimal splits of solar irradiance (W m^{-2})
 3 and relative humidity (%). Air temperature and volumetric water content were included, but these
 4 variables were not required for the optimal splits. The percentage of the total data at each split is also
 5 shown.