Aerodynamic Resistance and Penman–Monteith Evapotranspiration over a Seasonally Two-Layered Canopy in Semiarid Central Australia

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ABSTRACT

Accurate prediction of evapotranspiration E depends upon representative characterization of meteorological conditions in the boundary layer. Drag and bulk transfer coefficient schemes for estimating aerodynamic resistance to vapor transfer were compared over a semiarid natural woodland ecosystem in central Australia. Aerodynamic resistance was overestimated from the drag coefficient, resulting in limited E at intermediate values of vapor pressure deficit. Large vertical humidity gradients were present during the summer, causing divergence between momentum and vapor transport within and above the canopy surface. Because of intermittency in growth of the summer-active, rain-dependent understory and physiological responses of the canopy, leaf resistance varied from less than 50 s m⁻¹ to greater than 10⁶ s m⁻¹, in which the particularly large values were obtained from inversion of drag coefficient resistance. Soil moisture limitations further contributed to divergence between actual and reference E. Unsurprisingly, inclusion of site-specific meteorological (e.g., vertical humidity gradients) and hydrological (e.g., soil moisture content) information improved the accuracy of predicting E when applying Penman–Monteith analysis. These results apply regardless of canopy layering (i.e., even when the understory was not present) wherever atmospheric humidity gradients develop and are thus not restricted to two-layer canopies in semiarid regions.

1. Introduction

One-fifth of the global land area is arid or semiarid, where water scarcity can limit productivity in agricultural and native vegetation. Except where groundwater is present, evapotranspiration E in arid regions is second only to precipitation P as the largest component of the hydrometeorological cycle (Sheffield et al. 2010); thus,

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accurate knowledge of E is required for managing water resources (Er-Raki et al. 2010; Tabari et al. 2012). Australia is the driest permanently inhabited continent on Earth, with over 70% of the land surface characterized as (semi) arid (Eamus and Froend 2006), where the seasonality of moisture limitations exerts strong control over vegetation characteristics and E (Hutley et al. 2005; O'Grady et al. 2009).

To provide a reference of $E(E_0)$, the Food and Agricultural Organization adopted standard parameterization of the Penman–Monteith (PM) model for a reference grass crop (FAO56; Allen et al. 1998). This mechanistic model incorporates plant physiological constraints (Monteith 1965) into Penman's original model that

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combines available energy Q_A with atmospheric constraints (Penman 1948). FAO56 remains the most feasible and widely adopted method for predicting *E* across the globe (Steduto et al. 2003). The importance of FAO56 as a tool in water resources management is unequalled because of its simplicity and intercomparability among various crop and ecosystem types (Lemeur and Zhang 1990).

With the growth of worldwide eddy covariance (EC) networks, FAO56's aerodynamic resistance and surface conductance G_S terms can be evaluated by inverting the equation with measured values of actual E (Reichstein et al. 2002; Wohlfahrt et al. 2009). Numerous computational methods are used to estimate aerodynamic resistance to momentum flux r_{am} as a function of the drag coefficient C_D , either directly (i.e., from the friction coefficient u^* in moderately unstable conditions; Isaac et al. 2004) or inferentially from the logarithmic profile of wind speed U above a surface (Allen et al. 1998; Brutsaert 1982). Likewise, aerodynamic resistance to vapor transfer r_{av} is determined either directly from surface-to-air gradients of specific humidity (i.e., the bulk transfer coefficient C_E ; Brutsaert 1982; Stull 1988) or inferentially from C_D , in which case r_{av} and r_{am} are equivalent. While direct assessment of r_{av} is preferred because of the unambiguous relationship between C_E and r_{av} , measurements of vertical profiles in humidity are historically rare and r_{av} must be inferred from C_D (Brutsaert 1982; Cleugh et al. 2004).

Errors in Penman–Monteith (PM) predictions have been observed at low to moderate values of D (Whitley et al. 2009). Thus, we hypothesized that these errors are the result of differences between r_{av} (from C_E) and r_{am} (from C_D). This study compares r_{av} to r_{am} by inverting the PM equation at an OzFlux EC site in semiarid, central Australia (Cleverly 2011). To clarify variable designations, the generic subscript x will represent mfor momentum or v for vapor (e.g., G_{Sx} represents G_{Sm} when derived from r_{am}). Likewise, subscripts a and Srefer to aerodynamic and surface, respectively.

2. Measurements and methods

a. Aerodynamic resistance to momentum and vapor transfer

Application of r_{am} to vapor transport follows an inverse function of U and C_D , which can be estimated from canopy structural characteristics (Allen et al. 1998; Stull 1988):

$$r_{am} = (UC_D)^{-1} \approx \frac{\ln[(z_m - z_d)/z_{0m}]\ln[(z_m - z_d)/z_{0v}]}{k^2 U},$$
(1)



FIG. 1. Conditional three-circuit, two-layer aerodynamic resistance model. See text for variable descriptions.

where k is von Kármán's constant (0.41), z_m (m) is the measurement height, z_d (m) is the displacement height (²/₃ of the canopy height z_c), and z_{0m} and $z_{0\nu}$ (m) are the roughness lengths for momentum and vapor, respectively. The logarithmic profile that underpins r_{am} is not applicable below the height z_d ; thus, r_{am} was computed for a single layer representing the canopy surface.

Alternatively, aerodynamic resistance to vapor transport r_{av} was determined as a function of U and C_E , which is determined as a function of U, kinematic vapor flux $(\overline{w'q'})$, and the specific humidity $(q; \text{ kg kg}^{-1})$ gradient between the surface and air $(q_a - q_0)$ (Brutsaert 1982; Stull 1988):

$$r_{av} = (UC_E)^{-1} = -\frac{q_a - q_0}{w'q'}.$$
 (2)

Equation (2) is directly analogous to Ohm's law. Using profile measurements of q, r_{av} was solved during each half hour across two atmospheric layers during one of three scenarios (Fig. 1): 1) when r_{av} across the base layer was zero or undefined (base short circuit), 2) when r_{av} across the top layer was zero or undefined (top short circuit), or 3) when r_{av} across both layers was defined (parallel circuit). Layer boundaries were set to z_m , the effective canopy surface height ($z_{surface} = z_d + z_0$), and



FIG. 2. Specific humidity q profiles through the canopy and atmosphere, scaled by atmospheric layer thickness Δz . Subscripts represent measurement height. (a)–(c) The difference between q measured at two heights (subscripts: m, measurement; c, canopy; 2, 2 m). (d) Atmospheric q profile (\pm standard error) during the period of maximal gradients in the top layer (2100–2300 LST 22 Dec 2012; squares and broken line) and in the base layer (0130–0400 LST 5 Nov 2012; circles and solid line).

the soil surface (z_{soil} ; Fig. 1). Measurement heights for the finite difference $q_a - q_0$ were set, for example, to z_m and $z_{surface}$ in the top layer.

b. Meteorology and eddy covariance

E and $\overline{w'q'}$ were measured using the eddy covariance method as fully described for this site in Eamus et al. (2013). Flux measurements were made at a height of 11.7 m (z_m) for vapor and momentum (CSAT3, Campbell Scientific Inc., Logan, Utah; LI7500, Li-Cor Biosciences, Lincoln, Nebraska) and 12.6 m for radiation (CNR1, Kipp & Zonen, Delft, Netherlands). Surface soil moisture was measured across the top 10-cm depth (CS616, Campbell Scientific) and in profiles to 1-m depth (CS605, Campbell Scientific; Cleverly et al. 2013). Soil porosity was $0.36 \pm 0.005 \text{ m}^3 \text{ m}^{-3}$ above the variabledepth hardpan (Cleverly et al. 2013). Additionally, temperature and humidity measurements were made at heights of 11.7 m (z_m), 6.5 m (z_c), 4.25 m (z_d), and 2 m (HMP45C, Vaisala, Helsinki, Finland). Given the separation required to resolve differences between HMP45C measurements, q_0 and q_a were obtained within each layer by linearly extrapolating paired measurements of q with height. That is, q_0 was extrapolated from the slope $[(z_m - z_c)/(q_m - q_c)]$ in the top layer and from the slope $[(z_d - 2)/(q_d - q_2)]$ in the base layer. All measurements were collected during the calendar year 2012 over a two-layer (canopy and soil plus intermittent understory) surface.



FIG. 3. Comparison of C_D -based (G_{Sm} and G_{am}) and C_E -based (G_{Sv} and G_{av}) conductances. Comparison of surface conductances G_{Sv} vs G_{Sm} at (a),(c) 30-min and (b) daily time scales. Dashed line shows 1:1 line. In (b), complete (squares and solid line) and top layer (circles and broken line) are shown. In (c), top layer (filled squares and solid line), full layer (open squares and broken line), and base layer (circles and dash-dotted line) are shown. (d) Aero-dynamic conductance to vapor transfer G_{av} vs momentum transfer G_{am} .

c. Penman–Monteith: Surface and canopy conductance

The PM equation was inverted to compare G_{Sm} to G_{Sv} . In both cases, G_{Sx} includes conductance from soil, understory, and canopy surfaces (Brutsaert 1982). Inversion of the PM equation was computed as

$$G_{Sx}^{-1} = r_{ax} \left\{ \left[\left(\frac{\Delta Q_A + \rho_a c_P D r_{ax}^{-1}}{\lambda \overline{w' q'}} - \Delta \right) \gamma^{-1} \right] - 1 \right\}, \quad (3)$$

where Q_A is the difference between net radiation flux and ground heat flux, ρ_a is the density of moist air, c_P is the heat capacity of moist air, D is the vapor pressure deficit, Δ is the slope of the saturation vapor pressure curve against temperature, and γ is the psychrometric coefficient. For comparison to physiological measures of conductance (G), units were converted as G (mmol m⁻² s⁻¹) = ρ_a (g m⁻³) G (m s⁻¹)/0.018 (g mmol⁻¹).

Additionally, FAO56 was applied using r_{ax} to estimate E_{0x} (Allen et al. 1998; Jensen et al. 1990):

$$\lambda E_{0x} = \frac{\Delta Q_A + \rho_a c_P D r_{ax}^{-1}}{\Delta + \gamma [1 + (G_C r_{ax})^{-1}]}.$$
 (4)



FIG. 4. G_{Sv} or G_{Sm} as a function of vapor pressure deficit D. (a) Complete model G_{Sv} (squares) and G_{Sm} (circles). (b) Top (squares) and base (circles) layers. The inset is the same as in (b) but with top layer superimposed over base layer.

Canopy conductance (G_C) was estimated in place of G_S as a function of leaf resistance (r_L) and leaf area index (LAI): $G_C^{-1} = r_L / (0.5 \text{ LAI})$, in which $r_L (100 \text{ sm}^{-1})$ is the leaf resistance of a well-illuminated grass crop (Allen et al. 1998; Jensen et al. 1990). E_{0m} is equal to FAO56 for this site's canopy characteristics (i.e., z_m , z_d , z_0). LAI was interpolated to daily values by applying a spline function to the Moderate Resolution Imaging Spectroradiometer (MODIS) LAI product (Cleverly et al. 2013). Mulga ecosystem r_L was determined by inverting the LAI relationship: $r_{Lx} = G_{Sx}^{-1}$ (0.5 LAI). Additional limitations of soil moisture (below air dry) and small D on evaporation (Choudhury and Monteith 1988) and large D on stomatal function (Ball et al. 1987; Leuning 1995) were included in the calculation of E_0 by using the Water, Atmosphere, Vegetation, Energy and Solutes (WAVES) model, which is a soil-vegetation-atmosphere transfer scheme that employs PM and maintains balance in model complexity among carbon, energy, and water processes (McCallum et al. 2010; Zhang and Dawes 1998).

3. Results

Surface soil moisture content had a strong influence over atmospheric moisture gradients across each layer (Figs. 2a–c). Short circuits (i.e., $q_a - q_0 > 0$) were restricted to periods when soil moisture content was low: <0.2 m³ m⁻³ for the top layer (Fig. 2a), <0.05 m³ m⁻³ in the base layer (Fig. 2b), and <0.12 m³ m⁻³ across the full layer (Fig. 2c). Atmospheric humidity gradients were larger and more consistently negative (i.e., $q_a < q_0$)

in the base layer than across the top and full layers (Fig. 2). The largest gradients in the top and base layers were observed in late spring (5 November) and early summer (22 December), respectively (Fig. 2d), which represent the start of the growing season in the canopy and understory, respectively (Cleverly et al. 2013; Eamus et al. 2013; Ma et al. 2013).

Figure 3 shows the comparison of G_{Sm} and G_{Sv} . Most values were smaller than 15 mmol m⁻² s⁻¹ (Figs. 3a,c). The slope between G_{Sm} and G_{Sv} was smaller than 1:1, indicating that values of G_{Sm} tended to be smaller than G_{Sv} (Figs. 3a–c). Notable exceptions in which $G_{Sm} > G_{Sv}$ occurred at intermediate values of G_{Sv} (Figs. 3a,c). Likewise, when integrated over a day, G_{Sm} exceeded G_{Sv} at low values, which was more pronounced in the top layer alone than when the complete model (Fig. 1) was used for computing G_{Sv} (Fig. 3b). When scaled to the same units as G_{Sx} , aerodynamic conductance to momentum G_{am} was much smaller than G_{av} and only marginally larger than G_{Sm} (Fig. 3d).

Maximal values of G_{Sv} and G_{Sm} declined exponentially in response to increasing D (Fig. 4a). At small (<0.5 kPa) and large (>4 kPa) values of D, G_{Sv} , and G_{Sm} were similar (Fig. 4a). However, at intermediate values of D, maximal values of G_{Sm} were smaller than maximal values of G_{Sv} (Fig. 4a). In the top layer, G_{Sv} was large and restricted to low and intermediate values of D(<4 kPa; Fig. 4b). In contrast, in the base layer, which is solvable only during the top-layer short circuit, G_{Sv} was generally smaller than 5 mmol m⁻² s⁻¹ and was restricted to intermediate and large values of D (>0.5 kPa; Fig. 4b).



FIG. 5. Daily patterns of (a) P and air temperature at a height of $2 \text{ m}(T_a)$, (b) E and D, (c) G_{Sv} , and (d) G_{Sv} and G_{Sm} .

Figure 5 illustrates daily totals of P, E, G_{Sv} , and G_{Sm} along with average air temperature (T_a) and D. Winter (June-August) was particularly dry; thus, E and resultant conductances were negligible (Fig. 5). The Din the winter was often small (minimal daily average 0.6 kPa) because of low temperature. Rainfall events resulted in further reductions in D (minimal daily average 0.3 kPa) during any season and at any temperature (Fig. 5b). During periods when E was large, G_{Sv} was larger in the top layer than in the full layer (Fig. 5c). The G_{Sv} from the base layer was small but prominent in the early autumn (March; Fig. 5c). The complete model showed a stronger conductance response following rainfall than the top layer alone (Fig. 5d); G_{Sm} followed the same general pattern as G_{Sv} , but with smaller values (Fig. 5d).

In general, r_{Lv} and r_{Lm} were in close agreement except during late summer (Fig. 6a). Consequently, E_{0v} and

 E_{0m} followed similar patterns (Fig. 6b). A phase shift was observed between EC-based actual $E(E_a)$ and all formulations of E_0 such that peaks in E_a preceded peaks in E_0 by approximately 2–8 days (Fig. 6b). Divergence between E_{0v} and E_{0m} occurred during the dry period of the late summer when T_a and D were large (cf. Figs. 5 and 6). Application of WAVES E_0 resulted in an improved fit to E_a (Fig. 6b), especially during the early autumn (Fig. 6b). During summer (e.g., mid-January to mid-February), discrepancy remained between E_a and WAVES but was of smaller magnitude than the difference between E_a and E_{0x} (Fig. 6b).

4. Discussion

The estimation of r_{am} as a function of the C_D leads to overprediction of G_{Sm} at small values of E and to underprediction of G_{Sm} at large values of E (Figs. 3b and



FIG. 6. Daily patterns of leaf resistance r_L and E. (a) The r_L determined from surface resistance inverted with r_{av} (r_{Lv} , solid line) or r_{am} (r_{Lm} , dashed line). Grass reference r_L shown as a horizontal broken line at 100 s m⁻¹ (b) Actual E from EC measurements (broken line) and E_0 parameterized with r_{av} (E_{0v} , dashed line), r_{am} (E_{0m} , solid line), or from the output of the WAVES model (dash-dotted line).

5b,d,e). Steduto et al. (2003) similarly observed overprediction of E_0 by FAO56 at low values of E_0 and underprediction at high values. These patterns of over- and underprediction were observed during the summer when atmospheric gradients in specific humidity q were large (cf. Figs. 2 and 6b), which was the time of year when r_{am} was expected to underpredict r_{av} . Surface conductance computed from r_{av} (G_{Sv}) was much smaller than aerodynamic conductance of vapor G_{av} (Fig. 3) and consequently reduced the influence of errors in estimation of G_{av} on the determination of G_{Sv} and E_0 (Fig. 3).

To accurately fit E_0 to prevailing conditions, a large amount of meteorological data is required (Dehbozorgi and Sepaskhah 2012; Tian and Martinez 2012). Failure to account for the influence of atmospheric gradients in q by C_D resulted in underestimation of G_{Sm} at small to moderate values of vapor pressure deficit (0.5 kPa < D < 3.5 kPa; Fig. 4a). This underestimation occurred because r_{am} was too large relative to r_{av} , thus reducing the strength of the aerodynamic term $\rho_a c_P D r_{ax}^{-1}$ [Eqs. (3) and (4)]. Because E is sensitive to small variations in G_{Sm} regardless of the C_D formulation that is used (Shahrokhnia and Sepaskhah 2012), measurement of atmospheric humidity profiles improved PM estimates of E_0 by more closely matching the true site-specific aerodynamic resistance (cf. Figs. 2, 5, and 6).

Leaf-level stomatal resistance r_L was variable and much larger than for the grass reference (i.e., 100 sm^{-1} ,

Fig. 6a). A constant r_L accounts for neither nonphysiological (Steduto et al. 2003) nor physiological contributions to canopy conductance G_C in vegetation that demonstrates large physiological responses to precipitation. The Mulga plants under investigation here are a good example of vegetation that experiences only partial stomatal closure at low water potential (<-5 MPa) and large D (O'Grady et al. 2009). Natural ecosystems are characterized by complex soil and atmospheric moisture gradients that change with soil moisture, growth responses in understory vegetation, and stress responses in the upper canopy (Cleverly et al. 2013).

Mismatch between E_a and E_0 occurred because of 1) the delay between precipitation and vegetation greening, 2) unaccounted soil moisture limitations on r_L , and 3) seasonal divergence between r_{av} and r_{am} . The lag of E_0 behind E_a following precipitation was the result of 1) the dependence of E_0 on LAI and 2) heterogeneously rapid drying of the soil surface, which leads to underestimation of E_0 by WAVES (cf. Fig. 6 and Cleverly et al. 2013). Thus, E_0 failed to capture the initial pulse of evaporation, which can be particularly important in semiarid areas with intermittent rainfall (Cleverly et al. 2013; Eamus et al. 2013). Application of WAVES improved the fit between E_0 and E_a , especially during seasons when $E_{0\nu}$ and E_{0m} were equivalent (Fig. 6). Altogether, accurate prediction of E_0 was dependent upon 1) characterization of r_{av} that was responsive to CLEVERLY ET AL.

atmospheric humidity profiles and 2) identification of leaf physiological and soil moisture (Choi et al. 2012) limitations on *E* that are not modeled by PM.

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