Long-term growth and water balance predictions for a mountain ash (*Eucalyptus regnans*) forest catchment subject to clear-felling and regeneration

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Summary We used a physically based ecohydrological model to predict the water balance and growth responses of a mountain ash (Eucalyptus regnans F. Muell.) forest catchment to clear-felling and regeneration. The model, Topog-IRM, was applied to a 0.53 km² catchment for a 3-year pretreatment period, and a 20-year period following clear-felling and reseeding of 78% of the catchment area. Simulations were evaluated by comparing observed and predicted streamflows, rainfall interception and soil water values. The model faithfully simulated observed temporal patterns of overstory live stem carbon gain and produced a leaf area trajectory consistent with field observations. Cumulative throughfall was predicted within 1% of observations over an 18-year period. Over a 4-year period, predicted soil water storage in the upper 1.5 m of soil agreed well with field observations. There was fair correspondence between observed and predicted daily streamflows, and the model explained 76% of the variation in monthly flows. Over the 23-year simulation period, the model overpredicted cumulative streamflow by 6%. We argue that there is a useful role for physically based ecohydrological models in the management of mountain ash forest catchments that cannot be satisfied by simple empirical approaches.

Keywords: ecohydrology, forest growth, modeling, water yield.

Introduction

Water for the city of Melbourne is harvested from 1550 km^2 of forested catchments in the Central Highlands of Victoria, Australia. Ash forests, principally mountain ash (*Eucalyptus regnans* F. Muell.), cover just under half this area, but yield 80% of the streamflow because they grow on the higher rainfall sites.

Regrowth mountain ash forests yield significantly less water than old growth stands because of differences in forest density and structure (Langford 1976, Jayasuriya et al. 1993). Kuczera (1985) developed an idealized curve describing the relationship between mean annual streamflow and mountain ash forest age (Figure 1). The curve is based on the measured hydrologic responses of eight large catchments to a fire in 1939 and is constructed for the hypothetical case of a pure mountain ash forest catchment. It assumes that old growth forests yield an average runoff of 1195 mm year⁻¹. It indicates that after disturbance caused by fire or logging and subsequent regrowth, runoff falls to 580 mm year⁻¹ until age 27 years, after which yield slowly recovers over a period of around 150 years. However, the curve has wide error bands associated with it, particularly for forests aged 50 to 120 years, making it difficult to predict accurately when water yields will recover after disturbance. Moreover, because the curve is highly generalized, variation that exists among mountain ash forest catchments with different site characteristics is masked. Despite these limitations, the curve has been used to predict water yield from eucalypt forest in morphoclimatic settings entirely different from the Central Highlands (Moran 1988, Wilby and Gell 1994).

Vertessy et al. (1993) demonstrated that water yield from an old growth mountain ash stand could be predicted using a topographically based catchment model. That study assumed that the vegetation was in steady state and no attempt was made to simulate how the carbon and water balances change after a



Figure 1. Empirical relationship between mountain ash forest age and average annual water yield, after Kuczera (1985). Dashed lines indicate 95% confidence limits.

forest is perturbed. Here we report an application of a variant of this model, Topog-IRM, to a mountain ash forest catchment where vegetation changed abruptly from old growth to regrowth. The Topog-IRM model simulates the spatial and temporal changes in the forest carbon balance and the effect that this has on the catchment water balance.

Materials and methods

Site description

We applied Topog-IRM to the 0.53 km² Picaninny catchment, which forms part of the Coranderrk experimental area established by the former Melbourne and Metropolitan Board of Works (now Melbourne Water Corporation) in 1955 (Howard and O'Shaughnessy 1971, Langford and O'Shaughnessy 1980*a*). Picaninny was originally covered by old growth mountain ash forest (about 200 years old), but between November 1971 and March 1972, 78% of the catchment was clear-felled and re-seeded with mountain ash. Germination of the seed started in July 1972. Buffer vegetation (accounting for the remaining 22% of the catchment area) was retained around the stream course (Figure 2A). The adjacent 0.62 km² Slip Creek catchment served as an untreated control catchment for Picaninny.

Picaninny has a mean slope of 37%, is south facing and has an elevation range of 230 to 780 m above mean sea level. Mean annual rainfall is 1180 mm (low for mountain ash forest) and is evenly distributed throughout the year with a mild spring maximum. Mean annual runoffs from the Picaninny and Slip Creek catchments for the 3 years before forest clearance were 332 and 406 mm, respectively. Hence, these catchments yield about one third of the annual streamflow assumed in the Kuczera (1985) model.

Vegetation consists mainly of mature (and eventually regrowth) mountain ash forest with a mixed hardwood understory of mountain hickory wattle (*Acacia frigrescens* J.H. Willis), blanket leaf (*Bedfordia arborescens* Hochr.) and musk (*Olearia argophylla* Labill. (Benth.)). In the buffer area around the stream, mature mountain ash vegetation is mixed with hazel (*Pomaderris aspera* DC.), myrtle beech (*Nothofagus cunninghamii* (Hook.f.) Ørst.) and sassafras (*Atherosperma* *moschatum* Labill.). The Coranderrk soils have been described in detail by Langford and O'Shaughnessy (1980*b*) who discriminate between upper, middle and lower slope soils (Figure 2B). All soil groups are high in clay content (up to 70%) but are well structured and therefore fairly transmissive. Porosities range from 30 to 70%, and depth to bedrock varies between 5 and 20 m.

Hydrologic monitoring

Streamflow has been continuously monitored at Picaninny and Slip Creek since 1958. Rainfall and temperature have been continuously recorded at the Lower Coranderrk meteorological station (Figure 3) since late 1968. Between 1975 and 1983, soil water measurements were made at approximately monthly intervals at 12 sites within Picaninny (Figure 3). At each site, soil water measurements were made at 30-cm depth intervals with a neutron moisture meter to depths of between 1.5 and 5.1 m depending on local soil state. At two sites, throughfall measurements (rainfall less interception and stemflow) were made every 2 weeks beginning in October 1974 (Figure 3).

Forest growth monitoring

Starting in 1980, growth of the regenerating mountain ash vegetation was periodically monitored in 45 randomly located, permanent sample plots, each 100 or 200 m² in area (Figure 3). Estimates of gross stem bole volumes, based on measurements of tree heights and stem diameters at 1.3 m above ground level, were recorded on eight occasions between December 1981 and January 1992. For each survey, total stem carbon (C_{st}) of the overstory was computed by:

$$C_{\rm st} = V_{\rm bt} F_{\rm w} D_{\rm w} F_{\rm c},\tag{1}$$

where C_{st} is total stem carbon in kg m⁻², V_{bt} is gross bole volume in 1 m⁻², F_w is the wood density fraction, assumed to be 0.32 (Dunn and Connor 1991), D_w is the density of cellulose, assumed to be 1.53 kg l⁻¹, and F_c is the carbon fraction by weight of oven-dry wood, assumed to be 0.5.

Live stem carbon was computed as:

$$C_{\rm sl} = V_{\rm bs} F_{\rm w} D_{\rm w} F_{\rm c}, \tag{2}$$



Figure 2. The Picaninny catchment: (A) vegetation map, (B) soils map, and (C) Topog flow net.



Figure 3. Picaninny catchment hydrologic and forest growth monitoring network.

where $C_{\rm sl}$ is live stem carbon in kg m⁻², and $V_{\rm bs}$ is stem sapwood volume in l m⁻². Bark and sapwood thicknesses used to calculate stem sapwood volume were based on an analysis of 263 mountain ash trees, selected randomly in stands of age 7, 20, 53 and 92 years.

Model description

Overview

Topog-IRM is a physically based ecohydrological model designed to predict the three dimensional water and plant carbon balances of heterogeneous catchments (Hatton et al. 1992, Dawes and Hatton 1993, Dawes et al. 1995). It links soil water processes, canopy microclimate, plant growth and catchment topography, and is composed of modules that solve energy and water balance and calculate plant growth. The model is driven by time-series inputs of daily rainfall, total daily solar radiation, mean daily temperature and humidity. It predicts daily rainfall interception, vertical and horizontal soil water distribution, the horizontal distribution of daily evapotranspiration from both the overstory and understory vegetation, daily evaporation from the soil-litter layer, and daily streamflow. The model also predicts daily changes in leaf, stem and root carbon of both the understory and overstory, and leaf fall and decomposition. The component algorithms used in the model are briefly described below.

The computational network

Topog-IRM uses a flow net defined by elevation contours and lines of steepest slope (O'Loughlin 1986, Dawes and Short 1994) (Figure 2C). An automated terrain analysis system has been designed to build the flow net and, for each element, to calculate terrain attributes (such as slope and aspect) that impact on the catchment water and carbon balance. Routines have also been developed to ascribe spatially variable catchment properties (such as soil or vegetation type) to specific elements.

Soil water dynamics

The soil water dynamics module in Topog-IRM is identical to that in the Topog-Yield model (Vertessy et al. 1993). Vertical soil water fluxes (unsaturated and saturated flows) are modeled in each element by a fast implicit solution of the Richards equation. In the lateral dimension, subsurface and overland flows run downslope within the flow net. Darcy's law is used to model saturated subsurface flow, but no allowance is made for unsaturated lateral flow.

Evapotranspiration

Evaporation from the overstory, understory and soil layers is modeled on a daily time-step. For each layer, the surface radiation balance consists of four components: incoming shortwave radiation (a model input), reflected shortwave radiation from the surface, incoming longwave radiation from the atmosphere, and emitted longwave radiation from the surface. Ignoring advection, for each catchment element, the radiation balance can be written as:

$$R_{\rm n} = R_{\rm sd} - R_{\rm su} + R_{\rm ld} - R_{\rm lu}, \tag{3}$$

where R_n is the net radiation, R_{sd} is the shortwave downward radiation, R_{su} is the shortwave upward radiation, R_{ld} is the longwave downward radiation, and R_{lu} is the longwave upward radiation. The shortwave downward radiation on a horizontal surface is modified according to the slope and aspect of the element (Klein 1977).

An approach similar to that of Monteith and Unsworth (1990) was adopted for modeling the radiation balance of each layer. By assuming all leaves are randomly distributed with horizontal inclination, the radiation balance for the overstory is written as follows:

$$R_{\rm sv1} \downarrow = R_{\rm sd} \left[1 - \exp(-K_{\rm e} {\rm LAI}_{\rm I})\right] \tag{4}$$

$$R_{\rm sv1}\uparrow = R_{\rm sd}\,\alpha_1[1 - \exp(-K_{\rm e}{\rm LAI_1})] \tag{5}$$

$$R_{\rm lv1}\downarrow = R_{\rm ld}[1 - \exp(-K_{\rm e}{\rm LAI_1})]$$
(6)

$$R_{\rm lv1}\uparrow = R_{\rm lu}[1 - \exp(-K_{\rm e}{\rm LAI_1})],\tag{7}$$

where the subscripts sv1 and lv1 indicate shortwave and longwave radiation, respectively, for vegetation layer 1 (overstory), the arrows indicate the flux directions, K_e is the radiation extinction coefficient for the overstory, LAI₁ is the cumulative leaf area index of the overstory, and α_1 is the albedo of the overstory. Following Brutsaert (1982), the longwave radiation is estimated as:

$$R_{\rm ld} = \varepsilon_{\rm a} \sigma T_{\rm a}^4 \tag{8}$$

$$R_{\rm lu} = \varepsilon_{\rm s} \sigma T_{\rm a}^4 \tag{9}$$

with

$$\varepsilon_{\rm a} = 1.24 (e_{\rm a}/T_{\rm a})^{1/7},$$
 (10)

where ε_a is the atmospheric emissivity, T_a is the air temperature in K, e_a is the vapor pressure in kPa, ε_s is the surface emissivity (assumed to be 0.97), and σ is the Stefan-Boltzman constant.

The radiation balances of the understory canopy and the soil surface are treated in a similar fashion. The energy balance for the canopy and soil layers is thus:

$$R_{\rm nv1} = R_{\rm sv1} \downarrow - R_{\rm sv1} \uparrow + R_{\rm lv1} \downarrow - R_{\rm lv1} \uparrow \tag{11}$$

$$R_{\rm nv2} = R_{\rm sv2} \downarrow - R_{\rm sv2} \uparrow + R_{\rm lv2} \downarrow - R_{\rm lv2} \uparrow \tag{12}$$

$$R_{\rm ng} = R_{\rm sg} \downarrow - R_{\rm sg} \uparrow + R_{\rm lg} \downarrow - R_{\rm lg} \uparrow, \qquad (13)$$

where R_{nvl} , R_{nv2} and R_{ng} are the net radiation for overstory, understory and ground surface, respectively.

Evaporation (E) from the overstory, understory and soil surface is calculated by the Penman-Monteith equation:

$$E = \frac{sR_{\rm n} + \rho c_{\rm p} D_{\rm a}/r_{\rm a}}{s + \gamma (1 + r_{\rm s}/r_{\rm a})},$$
(14)

where *s* is the slope of the saturation vapor pressure curve, R_n is net radiation, *r* is the air density, c_p is the specific heat of air, D_a is the vapor pressure deficit of the air, γ is the psychometric constant, r_a is the aerodynamic resistance, assumed to be constant (cf. Running and Coughlan 1988, Leuning et al. 1991, Hatton et al. 1992), and r_s is the surface resistance. On a daily time-step, soil heat flux was assumed to be negligible.

For vegetation transpiration, the canopy conductance is calculated by the empirical model of Ball et al. (1987) as modified by Leuning (1995):

$$r_{\rm s} = (g \text{LAI})^{-1} = \left(\left[g_0 + \frac{g_1 A}{(c_{\rm s} - \Gamma)(1 + D_{\rm c}/D_{\rm c0})} \right] \text{LAI} \right)^{-1}, \quad (15)$$

where g_0 is a residual stomatal conductance, g_1 is an empirical coefficient, A is assimilation rate, c_s is the CO₂ mol fraction at the leaf surface, Γ is the CO₂ compensation point, D_c is the imposed vapor pressure deficit at the canopy surface, and D_{c0} is an empirical coefficient.

In solving Equation 14 for the overstory, understory and soil layers, different values of D_a are assumed for each layer. The vapor pressure deficit at the overstory canopy (D_c) is modified to reflect the effective deficit at the understory canopy surface (*sensu* Grantz and Meinzer 1990); this modified deficit (D_{mod}) is calculated using the omega coefficient proposed by Jarvis and McNaughton (1986):

$$D_{\rm mod} = \Omega_{\rm c} D_{\rm eq} + (1 - \Omega_{\rm c}) D_{\rm c} \tag{16}$$

with

$$D_{\rm eq} = \frac{\gamma \epsilon R_{\rm n}(c_{\rm p}/r_{\rm s})}{\epsilon + 1} \tag{17}$$

$$\Omega = \frac{\varepsilon + 1}{\varepsilon + 1 + r_{\rm s}/r_{\rm a}},\tag{18}$$

where Ω_c is the decoupling coefficient, D_{eq} is the equilibrium saturation deficit, and ε is s/γ . A similar approach is used to reduce the vapor pressure deficit below the understory for the soil surface, except instead of using D_c in Equation 16, the understory canopy value of D_{mod} is used.

Soil evaporation is calculated by the Penman-Monteith equation with the surface resistance set to zero if the soil surface is not air-dry, or determined by the method of Choudhury and Monteith (1988) when in second phase drying:

$$r_{\rm s} = \tau l/pD_{\rm m},\tag{19}$$

where *p* is the porosity of the soil, D_m is the molecular diffusion coefficient for water vapor, τ is a tortuosity factor, and *l* is the depth of the air-dry soil layer as determined dynamically by the Richards equation solution of water content.

Plant growth

The plant growth model in Topog-IRM has been described in detail (Hatton et al. 1992, Dawes and Hatton 1993, Wu et al. 1994). The model assumes that the actual CO_2 assimilation rate is dependent on a maximum rate and the relative availabilities of light, water and nutrients. It computes the assimilation index (*r*) of Wu et al. (1994):

$$r = \frac{1 + w_{\rm H} + w_{\rm N}}{1/m_{\rm L}x_{\rm L} + w_{\rm H}/x_{\rm H} + w_{\rm N}/x_{\rm N}},$$
(20)

where $w_{\rm H}$ is the weighting of water relative to light, $w_{\rm N}$ is the weighting of nutrients relative to light, $x_{\rm H}$, $x_{\rm N}$ and $x_{\rm L}$ are the relative resource availabilities for water, nutrition and light, respectively, and $m_{\rm L}$ is the modifier of light availability due to temperature. The $m_{\rm L}$ modifier is based on a Gaussian distribution of temperature effects parameterized by the optimal temperature and the temperature at the half-saturation response. Finally, carbon assimilation (*A*) is calculated as:

$$A = rA_{\max},\tag{21}$$

where A_{max} is the maximum mean daily assimilation rate of CO₂.

Five carbon pools are defined: litter, leaf, root, live stem, and a soluble pool expressed as a fraction of leaf carbon. Maintenance respiration is calculated as a function of temperature (cf. Running and Coughlan 1988), and net photosynthesis is calculated as assimilation less total respiration. When daily respiration exceeds assimilation, carbon is taken from the soluble pool (if present), then from the leaf and root pools. Growth respiration is assumed to be a constant fraction of allocated carbon (Rauscher et al. 1990).

On days when assimilation exceeds respiration, a soluble carbohydrate pool is filled. When it reaches its maximum, the excess is assigned dynamically to leaves, roots and stems. Leaf carbon may not exceed a level where 95% of incoming light is intercepted, nor exceed stem carbon. These conditions recognize that once canopies capture all available light, there is little advantage in adding more leaves, and that leaves require some minimum mechanical and hydraulic support. Carbon allocation to leaves, roots and stems is dynamic and follows the approach of Running and Gower (1991) in that the leaf/root ratio is based on the relative availabilities of above- and belowground resources (*sensu* Chapin et al. 1987, Wilson 1988), in this case as calculated by the equation for r.

Increments to root carbon are distributed by dynamic allocation of carbon through the root zone. We assume that root carbon is continuous from the surface downward, that roots are added to the wettest (most favorable) region of soil, and that roots grow downward in search of water. The favorability of a soil region is a function of the availability of water.

Irrespective of daily assimilation, each of the three living carbon pools suffer daily decrements, representing the turnover of leaves, bark and roots. Shedding from the two aboveground carbon pools, as well as any carbon from net negative assimilation, contributes to the litter pool. The litter pool decays as a function of temperature and water availability (Hatton et al. 1992).

Model application

Topog-IRM was applied to the Picaninny catchment for two phases, the pretreatment phase (January 1, 1969, to July 22, 1972) and the recovery phase (July 23, 1972, to July 18, 1992); however, the growth modeling capabilities of the model were only employed in the recovery phase. In the pretreatment phase, the leaf area index (LAI) of the forest was fixed at 2.0 for both the overstory and the understory.

A flow net comprised of 690 elements was computed (Figure 2C). Elements varied in area between 188 and 3600 m² with a mean of 885 m². Soils and vegetation maps were overlain on the element network so that spatial variability of these properties was represented (Figures 2B and 2C).

Four soil complexes were represented in the simulations (Figure 2B) (see Langford and O'Shaughnessy 1980*b*). In each complex, three separate soil layers were defined, spanning the depths 0–0.7, 0.7–2.0 and 2.0–16.0 m. The soil hydraulic properties ascribed to these layers in each complex are given in Table 1.

The forest was represented as a three-layer system, comprised of overstory and understory vegetation and a litter layer. Eighteen parameters were used to define the physiological and energy balance properties of the vegetation (Table 2). Vegetation parameter values used in the simulation were based on a combination of published and measured values, although a few parameters (such as the respiration coefficients) were treated as calibration parameters. We assumed that nutrients were optimal (cf. Weston 1990).

The value used for the slope of the relationship between stomatal conductance and regrowth vegetation is similar to that reported by Leuning (1990) for *Eucalyptus grandis* W. Hill ex Maiden. Connor et al. (1977) reported a decrease in stomatal conductance in mature forest relative to regrowth forest of about 20% (derived from results in their Table 3). Given this observation and the fact that the stomatal conductance algorithm used in Topog-IRM is linear with respect to the slope parameter (g_1), the slope parameter was reduced by 20% for the pretreatment forest.

To evaluate the potential of the model to predict the consequences of forest disturbance on streamflow in ungauged catchments, simulations identical to those described above were performed on Picaninny but with modifications to catchment aspect and climate. In the first of these simulations, the catchment element network was rotated 180° to change the aspect from southerly to northerly. In the second of these simulations, the Picaninny catchment was modeled using a climate representative of a higher elevation site. The modified climate series was generated with MTCLIM (Running et al. 1987), a model that extrapolates meteorological data in mountainous terrain on the basis of lapse rates and an imposed rainfall and elevation gradient. Annual rainfall was approximately double that measured at the Picaninny catchment to represent known changes at sites 1000 m higher in elevation. The resulting climate had a mean annual air temperature 2.9 °C lower, a mean annual vapor pressure deficit 0.18 kPa lower, and a mean daily solar radiation 0.5 MJ m⁻² lower than the Lower Coranderrk meteorological station.

Results

Most results are plotted on a time axis that uses the time of forest clearance (July 1972) as an origin. Thus reference to Year 15 means 15 years after forest clearance.

Table 1. Soil parameters used in the Topog-IRM simulations for the Picaninny catchment. Foursoils, each with three layers, were distributed across the catchment as shown in Figure 2C.

Parameter	Unit	Layer 1	Layer 2	Layer 3
All soils				
Soil depth	m	0-0.7	0.7-1.5	1.516.0
Saturated soil water content	$cm^3 cm^{-3}$	0.60	0.60	0.4
Residual soil water content	$\rm cm^3 cm^{-3}$	0.05	0.05	0.05
Saturated hydraulic conductivity values				
Middle slopes, stony phase (42% of area)	$m day^{-1}$	10.0	5.0	0.1
Middle slopes, red phase (13% of area)	m day ^{-1}	10.0	5.0	2.5
Upper slopes (27% of area)	$m day^{-1}$	10.0	5.0	2.5
Lower slopes (18% of area)	m day ⁻¹	10.0	5.0	0.1

Table 2. Overstory and understory vegetation parameters used in the Topog-IRM simulations for the Picaninny catchment. Parameters used in the calibration of the growth model are indicated. Symbols: [o] = old growth forest, [r] = regrowth forest, and an asterisk denotes a dimensionless variable.

Parameter	Overstory	Understory	Reference/note
IRM weighting factor for water*	2.1	2.1	Hatton et al. 1992
IRM weighting factor for nutrients*	0.3	0.3	Hatton et al. 1992
Canopy albedo*	0.2	0.2	Lee 1980
Light extinction coefficient*	-0.42	-0.6	Pook 1985, Monsi and Saeki 1953
Slope in conductance model (g1)*	8.0[o], 11.0[r]	8.0[o], 11.0[r]	Leuning 1990
Rainfall interception coefficient, mm LAI ⁻¹ day ⁻¹	0.30	0.30	Dunin et al. 1988, Hatton et al. 1992
Aerodynamic resistance, s m ^{-1}	15[o], 20[r]	30	Leuning et al. 1991, van de Griend and van Boxel 1989
Maximum assimilation rate, μ mols m ⁻² s ⁻¹	24.0	24.0	Leuning 1990, van den Driessche et al. 1971
Optimum temperature for assimilation, °C	20.0	20.0	Hatton et al. 1992, Küppers et al. 1987
Half saturation temperature, °C	10.0	10.0	Küppers et al. 1987
Light saturation value, μ mols m ⁻² s ⁻¹	1000.0	800.0	Küppers et al. 1987
Leaf respiration coefficient, kg kg ⁻¹ °C ⁻¹ day ⁻¹	0.0140	0.0008	Calibrated parameter
Stem respiration coefficient, kg kg ^{-1} °C ^{-1} day ^{-1}	0.0010	0.0012	Calibrated parameter
Root respiration coefficient, kg kg ⁻¹ °C ⁻¹ day ⁻¹	0.0010	0.0010	Calibrated parameter
Leaf growth respiration, kg kg ⁻¹	0.65	0.65	Calibrated parameter
Stem growth respiration, kg kg ⁻¹	0.35	0.35	Calibrated parameter
Root growth respiration, kg kg ⁻¹	0.50	0.50	Calibrated parameter
Initial leaf carbon value, kg m ⁻²	0.16[o], 0.03[r]	0.16[o], 0.006[r]	Field observations
Initial stem carbon value, kg m ⁻²	2.6[o], 0.03[r]	1.6[o], 0.006[r]	Field observations
Initial root carbon value, kg m^{-2}	0.164[o], 0.012[r]	0.102[o], 0.002[r]	Field observations

Forest growth

Eight observations of live stem carbon content were made between Years 9 and 20. Figure 4 compares observed mean overstory stem carbon with model predictions for two locations in the catchment possessing extremes in radiation loading. The model slightly overestimated stem growth between Years 9 and 12, but at Year 20, the observed catchment mean stem carbon value of 2.4 kg m⁻² was within the spatial range of predicted values. The observed trend of overstory stem carbon was generally steeper than the model predictions, but stem carbon increments decreased after Year 15. The understory stem carbon reached a peak of 0.42 kg m⁻² at Year 9, and then declined slightly to 0.39 kg m⁻² at Year 20. At the end of the simulation, overstory root carbon was 1.05 kg m⁻².

Simulated mean catchment LAI for the mountain ash regrowth following logging increased monotonically to about 4.0 at Year 20 (Figure 5); however, because of spatial differences in radiation across the catchment, LAI at particular



Figure 4. Observed and predicted live stem carbon values for the mountain ash overstory in the Picaninny catchment for 1969-1991. The range in predicted values over the catchment is indicated, along with the mean of observed values. Spatial variation in stem carbon was due to the pattern of solar radiation across the catchment.



Figure 5. Predicted regrowth LAI values for the mountain ash overstory and hardwood understory in the Picaninny catchment for 1969– 1991. Observed overstory LAI data point taken from Orr et al. (1986); this value is a catchment mean of 45 growth plots.

locations varied between 3.1 and 4.7. Understory LAI increased to a maximum of 0.9 at Year 10, and then declined slowly to 0.8 by Year 20.

The availability of soil water was near its maximum value throughout the simulation period. With nutrition assumed to be optimal, the IRM scalar for assimilation of CO_2 covaried almost solely with the temporal pattern of solar radiation; temperature effects on photosynthesis were minimal.

Throughfall and evapotranspiration

Throughfall measurements commenced 2.5 years after forest clearance and totaled 176.55 cm by Year 20 (86.5% of gross rainfall). The model estimate of throughfall for the same period was 178.15 cm. Figure 6 compares cumulative observed and predicted throughfall. The two curves diverge slightly at the beginning, but by Year 20 there is only a 1% difference between the cumulative totals.

Simulated evaporation (spatially averaged over the catchment) from the overstory, understory and soil appear in Figure 7 as annual totals. There was little variation in annual evaporation rates before forest clearance, despite variation in



Figure 6. Observed and predicted cumulative daily throughfall in the Picaninny catchment for 1974–1991. Observed throughfall includes estimated stemflow, taken as 4% of gross rainfall.



Figure 7. Simulated annual evaporation from overstory, understory and soil in the Picaninny catchment for 1969–1991. These estimates do not include interception losses. Arrow shows time of forest clearance.

rainfall over this period. In the first few years following logging, there was an immediate increase in soil evaporation, and transpiration was mostly limited to the unlogged buffer vegetation. Regrowth of the mountain ash led to increases in transpiration over the rest of the simulation and exceeded pretreatment rates after only 5 years. Transpiration from the understory recovered more slowly and approximated soil evaporation after 7 years of regrowth. Simulated soil water potentials did not significantly limit transpiration in any year.

Before logging, maximum summer daily evaporation rates for the overstory, understory and soil were 2.0, 1.1 and 0.4 mm day⁻¹, respectively, and the corresponding values at Year 20 were 4.1, 0.5 and 0.5 mm day⁻¹. In midwinter, typical daily evaporation rates for the overstory, understory and soil were 0.14, 0.07 and 0.06 mm day⁻¹, respectively, before logging, and the corresponding values at Year 20 were 0.2, 0.02 and 0.05 mm day⁻¹.

Soil water

Mean winter time soil water storage within the catchment rose from about 4500 mm in the pretreatment period to 5000 mm about 2.5 years after forest clearance (Figure 8). Soil water storage returned to pretreatment values around Year 6, then declined to a stable value of about 4100 mm in Year 11. Despite falling water levels, the seasonal amplitude of soil water change tended to increase as regrowth vegetation became established. Mean summer soil water depletion increased from 14 mm m⁻¹ in the pretreatment phase to 23 mm m⁻¹ between Years 10 and 20 (Figure 8).

There was good correspondence between observed and predicted soil water storage in the upper 1.5 m of soil between Years 4.5 and 8.5 (January 1977 to January 1981) (Figure 9). The predicted curve shows mean daily soil water storage for all catchment elements outside the buffer region (Figure 2A). The observations are mean values from 12 bore holes scattered throughout the catchment (Figure 3). Both the seasonal amplitude and the timing of changes in soil water storage are well represented by the model, and predictions are within the observed 95% confidence limits.



Figure 8. Predicted mean monthly catchment soil water storage in the Picaninny catchment for 1969–1991. Soil depth is 16 m.

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Figure 9. Observed and predicted mean catchment soil water storage, 0–1.5 m depth, in the Picaninny catchment: (A) January 1977 to January 1979, and (B) January 1979 to January 1981. Dots are the observed values with 95% confidence intervals.

Daily and monthly runoff

Model predictions of daily catchment runoff generally agreed with field observations (Figures 10A and 10B). Although the model tended to underpredict base flow rates and overpredict peak flows, these errors compensated one another and became insignificant when runoff was aggregated on a monthly basis as shown in Figure 11, which shows good agreement between observed and predicted monthly runoff ($r^2 = 0.76$, n = 286).

Annual runoff

Although annual catchment rainfall fluctuated during the 23year simulation, there was a clear temporal trend in annual catchment runoff with an initial increase after forest clearance, followed by a sustained decrease in runoff relative to pretreatment flows (Figure 12). Observed annual catchment runoff from Picaninny rose from around 300 mm in the pretreatment phase to 700 mm in Year 2 (1974) and then declined to about 200 mm in Year 7 (1979). This trend is reinforced when annual runoff from Picaninny is compared to annual runoff from the adjacent Slip Creek control catchment (Figure 13). In the

4 3 2 DAILY RUNOFF (mm) 1 0 740 780 660 700 580 620 4 В OBSERVED 3 PREDICTED 2 1 0 6880 6840 692O 6760 6800 6720 DAYS SINCE 1/1/69

Figure 10. Observed and predicted daily runoff in the Picaninny catchment: (A) 200-day sequence in the pretreatment phase, and (B) 200-day sequence in the regrowth phase.

pretreatment phase, Picaninny yielded about 80% of the runoff volume of the wetter Slip Creek catchment; however, within 2 years of forest clearance, Picaninny yielded twice as much runoff as Slip Creek and then gradually returned to pretreatment flow levels by Year 6 after logging. From Year 6 onward,



OBSERVED MONTHLY RUNOFF (mm)

Figure 11. Observed versus predicted monthly runoff from the Picaninny catchment for 1969–1991. The diagonal line indicates a 1/1 relationship.



Figure 12. Annual rainfall, and observed and predicted annual runoff for the Picaninny catchment for 1969–1991. Reduced discharge following forest regrowth is largely independent of the rainfall pattern. Arrow shows time of forest clearance.



Figure 13. Ratio of observed annual runoff values from Picaninny (logged) and Slip Creek (unlogged control) catchments. Arrow shows time of forest clearance.

annual runoff from Picaninny declined below pretreatment ratios until 1983 when it stabilized at about 45% of the Slip Creek runoff.

Predicted annual runoff generally agreed with observed values throughout the 23-year simulation (Figure 12), though there was a systematic overprediction in the years 1975 to 1983 (Years 3 to 11 after forest clearance). Total observed and predicted runoff values for the entire simulation period were 5947 and 6317 mm, amounting to a 6.2% overprediction in runoff.

Scenario modeling

Table 3 shows the results from the two applications to two hypothetical catchments with varying aspect and elevation. Rotating the Picaninny catchment to a northerly aspect resulted in a 9% increase in LAI at Year 20, with a 6% increase in total annual evaporation (E), and thus reduced runoff efficiency. Application of a high elevation climate resulted in only a 1% increase in LAI, but a significant decrease (21%) in evaporation. This and the enhanced rainfall resulted in a trebling of runoff efficiency.

Discussion

Throughfall and evapotranspiration

There was close agreement between observed and predicted throughfall (Figure 6) despite the simplicity of the interception algorithm employed in Topog-IRM, which makes no allow-ance for differences in interception caused by variations in rainfall intensity. Although it is likely that discrete daily throughfall predictions would often be in error, it appears that these errors are compensatively smoothed over time. We note that the throughfall results are based on an *a priori* estimate for the leaf storage coefficient of 0.3 mm LAI⁻¹ as reported for *E. maculata* Hook. forests by Dunin et al. (1988) and Hatton et al. (1992).

The seasonal range in simulated evaporation was greater than expected, with winter rates being low. By scaling leaf conductance measurements, Legge and Connor (1985) estimated peak summer evaporation rates from a mature mountain ash overstory to be 4.1 mm day⁻¹, which is slightly more than the peak rates for total evapotranspiration in our pretreatment simulations. Based on sap flow measurements at a higher elevation site than Picaninny, Dunn and Connor (1993) re-

Table 3. Results of Topog-IRM simulations for the Picaninny catchment during the regrowth phase (1972–1991), indicating effects of changing aspect and elevation (climate). Bracketed numbers indicate percent change relative to baseline simulation. All LAI values shown are catchment averages.

Simulation	Total E (mm year ⁻¹)	Runoff efficiency (runoff/rainfall)	LAI
Regular Picaninny	798	0.22	3.7
Rotated Picaninny	848 (+6%)	0.18 (-18%)	4.1 (+9%)
Elevated Picaninny	632 (-21%)	0.66 (+200%)	3.8 (+1%)

ported overstory summer transpiration for 50- and 230-yearold mountain ash stands of 2.04 and 0.77 mm day⁻¹, respectively. Their annual estimates for 50- and 230-year-old stands were 679 and 296 mm, respectively. Given that the simulated LAI at Year 20 is apparently approaching a maximum, these values agree well with the model estimates at that age (669 mm) and for the old growth pretreatment condition (350 mm). Further, Dunn and Connor (1993) estimated understory contributions to total plant water use in the old growth forest to be 27%; in the pretreatment simulations, we predicted a value of 34%.

Simulated annual evaporation rates also agree with estimates by Langford et al. (1982) for the Picaninny catchment based on the difference between streamflow and estimated interception losses; these authors report pretreatment total evaporation losses of 620 mm in both 1969 and 1975 compared with values of 633 and 610 mm, respectively, in our simulations. Feller (1981) used soil water, rainfall and runoff data to estimate evaporative losses for the Year 1978, exclusive of interception, for a nearby, higher elevation mountain ash catchment receiving 1631 mm of rainfall and reported a value of 620 mm. The model estimate for Picaninny in 1978 was 710 mm, which is expected because there are more annual sunshine hours and warmer temperatures in this catchment.

Most of the annual variation in evaporation from canopies or soil was accounted for by changes in LAI, even in the presence of large variations in annual rainfall. Our soil water balance calculations indicate that soil matric potentials remain favorable for transpiration (and soil evaporation) throughout the year. We conclude that evaporation in mountain ash forests is almost entirely energy limited, with annual patterns of net radiation and temperature being conservative from year to year. This may simplify inferences about the effect of forest management on runoff (at least in this forest and edaphic environment), particularly when combined with observed (imposed) estimates of LAI development. If LAI data can be made available from remote sensing technology (*sensu* Running et al. 1989), the need for a complex growth and evaporation model may be eliminated.

Forest growth

Apart from stem carbon measurements, there are few observations with which to compare the simulated growth responses following logging in the Picaninny catchment. The simulated value of average catchment LAI for the mountain ash regrowth at Year 15 was 3.6, compared to a measured average value of 3.4 (range of 0.97 to 6.62) (Orr et al. 1986). Using an LAI-2000 leaf area meter (Li-Cor Inc., Lincoln, NE), O'Sullivan (pers. comm.) determined that total LAI was 4.7 at stand age 22 years at Picaninny; the sum of simulated overstory and understory LAI was 4.8 at age 20 years. The monotonicity of the LAI response in our simulations is somewhat at variance with observations for a regrowth plot at a nearby catchment with mountain ash in which LAI peaked at 5.1 at age 9 years (Vertessy et al. 1995) and declined to 4.4 at age 12 years (O'Sullivan, pers. comm.). Ashton (1976) argued that radiation interception increases in these forests to about age 40 years,

though this includes understory effects. Our simulations indicated that LAI would continue to increase past age 20 years.

Feller (1980) estimated a root carbon equivalent of 3.1 kg m⁻² in the top 100 cm of soil of a stand at age 40 years, which is almost three times the amount simulated at 20 years in this study, although Feller's estimate included nonliving root material. Incoll (1979) found 99.7% of fine roots (< 5 mm in diameter) in the upper 105 cm of soil, but with some roots extending to 2.5 m. Ashton (1975) indicated that trees of age 23 years have rooting distributions to over 4 m, compared to simulated root carbon at age 20 years which grew to 5 m depth in this study. We conclude that although our simulated rooting distributions develop appropriately, the predicted root biomass may be underestimated.

Our simulations suggest that after 10 years of regrowth, understory begins to thin. Vertessy et al. (1995) estimated understory live stem carbon at stand age 15 years to be 0.97 kg m⁻² which is twice the simulated value for Picaninny. However, their estimate was for a stand where the overstory had been repeatedly subject to severe insect attack that aided understory development.

There are strong dependencies of throughfall and evaporation on LAI within Topog-IRM. Although there is good overall agreement between observed and predicted throughfall, the cumulative estimates (Figure 6) depart in the early half of the simulation and are roughly parallel thereafter, indicating that the model did not develop the canopy quickly enough. Similar inferences may be drawn from the runoff results (Figure 12). Although the predicted annual runoff values for the pretreatment phase and the final 8 years of the simulation are good, there is a period between Years 6 and 11 that indicates too little evaporation from the system. Given the previous observation that too much stem carbon was accrued during this phase, this may indicate that the carbon allocation algorithm favored stem too much over leaves during this phase of the simulation.

Extrapolation of the overstory LAI results indicates a longterm maximum LAI at about 30--40 years after forest clearance, which is in broad agreement with observations by Ashton (1976). After this age, stand thinning and the apparent lack of gap-replacement recruitment of mountain ash leads to a longterm decline in overstory LAI and subsequent increase in understory LAI to roughly equivalent values (2.0). Topog-IRM is a "big leaf" model and hence does not represent trees as individuals, nor does it represent the stochastic processes leading to forest thinning. Consequently, the model would need to be modified to simulate forest growth beyond about 40 years in this forest type.

Although the agreement between observed and predicted live stem carbon was largely dependent on calibration of the respiration coefficients, it should be noted that Topog-IRM does not simulate leaf, root and stem growth independently. Thus the resulting balance among these components is neither predetermined nor subject to independent calibration. In this sense, the agreement with observations of these components is a reasonable test of the assimilation and allocation logic of the model, at least over the time scales involved in the reported simulations.

Soil water and runoff

The observed soil water data support our model predictions of soil water drainage and plant water uptake (Figure 9). The comparison is restricted to the upper 1.5 m of the soil profile because deeper readings were taken at only a few of the 12 bores in the catchment. We conclude that 12 bores was a marginal number for estimating mean water content across the catchment; however, the accuracy of the cumulative runoff predictions over the 23-year simulation period is probably within the accuracy of the measured climatic inputs and stream gauging.

The model simulated base flow behavior well but performed less well during and immediately after large storm events. The model had a tendency to overpredict peak flows (suggesting exaggeration of the contributing runoff area) and predicted flow recessions which were too steep (indicating exaggeration of the drainage rate) (Figure 10A). During heavy rain events, the model produced daily flow predictions that were worse than those shown in Figure 10A. It may be possible to rectify this problem with better parameterization of the catchment soils.

The model was able to simulate monthly and annual runoff from Picaninny during a period of variable land cover (Figures 11 and 12), although there are some systematic errors, particularly in the annual flow predictions. One of the discrepancies between observed and predicted annual flows is the overprediction of runoff between 1978 and 1983 (Years 6–11). This result suggests that we underestimated LAI during this period, though this conflicts with our finding that throughfall was modeled accurately over this period. It is possible that stomatal conductances are higher than we assumed in the first few years of regrowth and begin to decline from age 11 years onward.

Scenario modeling

In catchments where land cover is static and streamflow is gauged, empirical relationships between rainfall and runoff are robust predictors of catchment runoff (e.g., Post et al. 1993). However, when land cover is dynamic, as in vigorously growing forest, such relationships cannot be used to predict runoff because the water balance is determined by the course of leaf area development. Similarly, empirical runoff relationships are problematic to transfer to ungauged sites where catchment properties differ (Moran 1988). Physically based, distributed parameter models such as Topog-IRM provide a sounder basis for predicting the hydrologic impacts of forest disturbance in systems characterized by changing land cover and spatially variable catchment properties.

The scenarios imposed for two hypothetical catchments illustrate the advantages and pitfalls of using process-based models for scenario modeling. By changing only aspect, the enhanced radiation led to increased evapotranspiration and canopy development. However, we know that at this low elevation, mountain ash forests are largely restricted to south-facing slopes and are likely to be replaced by a more xerophytic mixed eucalypt forest on northerly aspects (Ashton 1976). The failure of the model to predict a water-stressed forest in this

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instance can be attributed to the use of inappropriate temperature and humidity data and soil profiles associated with southerly aspects. For instance, we know that soils at this elevation and aspect tend to be thinner or even skeletal; had we used these less favorable soils in this simulation, the forest canopy would not have performed so well. The high elevation scenario attempted to covary all climate parameters with elevation, but retained the same soil properties and topography. This resulted in elevated catchment runoff efficiencies of about 66% of rainfall, a value quite similar to the runoff for a real catchment (Myrtle II) in years with rainfall approaching this amount (Vertessy et al. 1993).

If models such as Topog-IRM are to be used in scenario modeling across and between regions, it is essential to represent the spatial covariance of landscape properties accurately. For this reason, continuing effort is needed in the acquisition and handling of regional geographical data pertaining to topography, soils, land cover and climate. Further, the scenario modeling is only valid if the baseline calibration has resulted in a parameterization that minimizes compensating errors in processes (Sorooshian et al. 1983, Beven 1993).

We note that the calibration procedure adopted in these simulations was based on reproducing observed patterns of pretreatment runoff and stand development during the regrowth phase. The resulting predictions of evaporation, interception and soil water distribution generally agreed well with observations based on *a priori* parameter values obtained from the literature and direct measurement. The finding that not only the resulting runoff estimations following logging were accurate but that the partitioning of the rest of the water balance was correct suggests that our model parameterization is sound and appropriate for extrapolation to other sites. Further, we note that the calibrated vegetation parameters related to respiration and allocation (Table 2) are in principle species specific (i.e., genetically determined) and thus transferable to other mountain ash sites.

Conclusions

After clear-felling in 1972, the Picaninny catchment underwent profound water balance changes that persisted for at least 20 years. Catchment runoff increased initially then decreased to well below pretreatment values as vigorous regrowth forest was established. This response is qualitatively consistent with the empirical model of Kuczera (1985), although the magnitude and timing of the response differed because of the specific climatic conditions of the Picaninny site. The Kuczera model is based on a regionally averaged response to disturbance but is inadequate for planning at small catchment scales where site differences are important. We argue that to model this response over the range of environments in the Melbourne Water supply area, it is essential to model the forest recovery and its interactions with the water and energy balances given local site characteristics.

Model estimates of stem carbon and leaf area development during the regrowth phase in the Picaninny catchment agreed with field observations. This agreement, in combination with accurate representation of the surface energy balance, rainfall interception and catchment drainage meant that we were able to simulate catchment runoff accurately. The strong dependence of mountain ash forest growth and evaporation on available energy suggests a potential for simple representations of these processes and demonstrates the utility of physically based ecohydrological models in the development of practical management tools.

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