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A model inter-comparison study to examine limiting factors in modelling Australian tropical savannas

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Abstract

Savanna ecosystems are one of the most dominant and complex terrestrial biomes that derives from a distinct vegetative surface comprised of co-dominant tree and grass populations. While these two vegetation types co-exist functionally, demographically they are not static, but are dynamically changing in response to environmental forces such as annual fire events and rainfall variability. Modelling savanna environments with the current generation of terrestrial biosphere models (TBMs) has presented many problems, particularly describing fire frequency and intensity, phenology, leaf biochemistry of C_3 and C_4 photosynthesis vegetation, and root water uptake. In order to better understand why TBMs perform so poorly in savannas, we conducted a model inter-comparison of 6 TBMs and assessed their performance at simulating latent energy (LE) and gross primary productivity (GPP) for five savanna sites along a rainfall gradient in northern Australia. Performance in predicting LE and GPP was measured using an empirical benchmarking system, which ranks models by their ability to utilise meteorological driving information to predict the fluxes. On average, the TBMs performed as well as a multi-linear regression of the fluxes against solar radiation, temperature and vapour pressure deficit, but were outperformed by a more complicated nonlinear response model that also included the leaf area index (LAI). This identified that the TBMs are not fully utilising their input information effectively in determining savanna LE and GPP, and highlights that savanna dynamics cannot be calibrated into models and that there are problems in underlying model processes. We identified key weaknesses in a model's ability to simulate savanna fluxes and their seasonal variation, related to the representation of vegetation by the models and root water uptake. We underline these weaknesses in terms of three critical areas for development. First, prescribed tree-rooting depths must be deep enough, enabling the extraction of deep soil water stores to maintain photosynthesis and transpiration during the dry season. Second, models must treat grasses as a co-dominant interface for water and carbon exchange, rather than a secondary one to trees. Third, models need a dynamic representation of

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LAI that encompasses the dynamic phenology of savanna vegetation and its response to rainfall interannual variability. We believe this study is the first to assess how well TBMs simulate savanna ecosystems, and that these results will be used to improve the representation of savannas ecosystems in future global climate model studies.

1 Introduction

Savanna ecosystems are a diverse and important biome that play a significant role in global land-surface processes (van der Werf et al., 2008). Globally, they occupy regions around the wet-dry tropical to sub-tropical equatorial zone, covering approximately 15 to 20 % of the terrestrial surface and contribute ~ 30 % to global net primary production (Grace et al., 2006; Lehmann et al., 2014). Savannas are water-limited ecosystems where rainfall is often seasonal or monsoonal, and have a spatial extent that can cover an area with annual rainfall in the range of 500 to 2000 mm (Bond, 2008; Kanniah et al., 2010; Sankaran et al., 2005). The variability in the amount and timing of annual rainfall, coupled with local topo-edaphic properties, and the frequency and intensity of seasonal fires strongly influences the structure and function of savanna vegetation (Beringer et al., 2007; Kanniah et al., 2010; Ma et al., 2013; Sankaran et al., 2005). Savannas are characterised by a multi-layer stratum of vegetation, where an open and discontinuous canopy overstorey is seasonally dominated by understorey grasses (Scholes and Archer, 1997). These tree and grass layers are distinctly and functionally different, fixing carbon using different photosynthetic pathways, C_3 and C_4 photosynthesis respectively (Bond, 2008; Scholes and Archer, 1997; Williams et al., 1996b). The canopy overstorey can be either evergreen or deciduous (depending on the evolutionary history), while the grass understorey is annual: active only in the wet season and senescing at the end of this period (Williams et al., 1996b). Consequently, water, carbon and nutrient cycling in savannas is largely determined from the balance and co-existence of these two life forms (Lehmann et al., 2009; Sankaran et al., 2005).

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Given the complex nature of savannas, modelling the land surface exchange and vegetation dynamics for this biome is challenging, for terrestrial biosphere models (TBMs). Here we define TBMs to broadly encompass stand, land-surface, and dynamic global vegetation models (Pitman, 2003). Most land surface schemes that feed into larger earth system models use simplistic representations of vegetation, and these will have difficulty describing the complex structure of savannas ecosystems. Such issues may simplistic assumptions in relation to rooting depth and inadequate responses to drought (De Kauwe et al., 2015; Li et al., 2012); ignoring the multilayered nature of savannas and the differing structural (including radiation), functional (including different plant functional types) and phenological differences (Whitley et al., 2011); and in some cases neglecting the C_4 photosynthetic pathway entirely (Parton et al., 1983; Schymanski et al., 2007) It is therefore critical that TBMs meet the challenges that savanna dynamics present if water and carbon exchange are to be correctly simulated in response to global change.

Despite these issues, there have been significant advances in modelling savanna dynamics in recent years, and this has been focused on integrating important features specific to savanna ecosystems, namely frequent fire and tree-grass competitive interactions, processes that shape savanna structure and function (Haverd et al., 2015; Higgins and Scheiter, 2012; Scheiter and Higgins, 2007; Scheiter et al., 2014; Simioni et al., 2003). Nevertheless, little work has been undertaken to critically evaluate the performance and processes of TBMs when used to capture water and carbon cycling in savannas, with notably: in west Africa (Simioni et al., 2000) and Australia (Schymanski et al., 2007, 2008, 2009; Whitley et al., 2011). Many global ecosystem models more-over use broad plant functional types (PFTs) with single parameter values to describe whole biomes (Pitman, 2003), making them unable to represent changing vegetation structure (tree:grass ratio) in the continuum of grassland to woodland savanna. Approaches have been developed that can account for savanna dynamics, such as using mixed tiles, whereby trees and grasses are simulated as separate surfaces that are

then aggregated together (Kowalczyk et al., 2006). However, this approach fails to capture the competition between trees and grasses for light, water and nutrient resources.

In this study, we take 6 TBMs of distinctly different conceptual frameworks, and assess their ability at simulating savanna water and carbon exchange along the North Australian Tropical Transect (NATT) that is defined by a strong rainfall gradient. Australian tropical savannas can be considered largely intact compared to South American and African savannas, and provide a “living laboratory” to understand the links between vegetation structure and function and how it responds to environmental change (Hutley et al., 2011). We challenge the models by evaluating them along the rainfall gradient, which extends over a broad biogeographical extent and strong interannual variability in climate (Koch et al., 1995). The aim of this study is to highlight critical processes that may be missing in current TBMs and are required to adequately simulate savanna ecosystems. Specifically, we examine whether a TBM’s structural framework, such as the representation of the understorey grasses (C_4 photosynthesis), tree rooting depth, and description of phenology (prescribed vs. dynamic) can adequately replicate observed carbon and water fluxes. To achieve this we measure the performance of each TBM by comparing its predictions to a set of empirical benchmarks that describe a priori expected levels of model performance. We identify regions of low performance among sites and seasons, to diagnose under what climate conditions reduced model performance occurs. We then infer what processes (present or missing) may be the cause for reduced performance when applied to savannas ecosystems. Our intention is that these results can be used to flag high priorities for future development by the terrestrial biosphere modelling community.

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2 Methodology

2.1 Observational data

The North Australian Tropical Transect (NATT) is a sub-continental rainfall gradient in the wet-dry tropical climate zone of Northern Australia, which encompasses a distance of approximately 1000 km over a latitudinal range of -12 to -23° S and a decline in mean annual precipitation (MAP) from 1700 to 300 mm (Hutley et al., 2011). It is one of three savanna transects established in the mid 1990's, forming part of the International Geosphere Biosphere Program (IGBP) along with the Savannas in the Long Term (SALT) transect in West Africa and the Kalahari Transect (KT) in Southern Africa (Koch et al., 1995). Soils range from sand dominated red Kandosols to black, cracking clay soils that are more extensive in the southern end of the NATT that are limiting to woody plant growth (Hutley et al., 2011; Williams et al., 1996b). Kandosols are ancient and weathered, such that they have been leached of nutrients by the large monsoonal rainfall (McKenzie et al., 2004). Close to the northern coastline, vegetation is comprised primarily of evergreen *Eucalyptus* and *Corymbia* tree species that overly an understorey of C_4 *Sorghum* and *Heteropogon* spp. grasses. Inland, tree biomass, leaf area index (LAI) and cover tends to decline and by 18° S savanna vegetation transitions to less dense *Acacia* woodlands, shrublands and grasslands dominated by *Astrebla* grass species (Hutley et al., 2011). Fires occur regularly in these environments, increasing in frequency with higher rainfall (MAP > 1000 mm), fuelled by the accumulation of understorey C_4 grasses that cure in the dry season (Beringer et al., 2014; Russell-Smith and Edwards, 2006).

The five flux tower sites along the NATT used in this study are outlined in Table 1, which describes stand soil and vegetation characteristics, as well as a summary of local meteorology (Hutley et al., 2011). These sites represent a sampling of savanna environments covering a wide range of MAP and a much smaller range of mean annual temperature (MAT) (Fig. 1). At each site, an eddy covariance system was used to measure the ecosystem-atmosphere exchange of radiation, heat, water and CO_2 .

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during the SPECIAL program (Beringer et al., 2011). Parameters relating to soil sand and clay content were taken from the Australian Soil Classification (Isbell, 2002), while root profile information was sourced from Chen et al. (2003) and Eamus et al. (2002). TBMs that require LAI as an input (all models with the exception of LPJ-GUESS) were provided with MODIS derived approximations that are well matched to ground-based estimation of LAI at the SPECIAL sites (Sea et al., 2011). Additionally, in the case of MAESPA and SPA, the models allowed for time-varying tree and grass fractions to be assigned, and these time-varying fractions were determined using the method of Donohue et al. (Donohue et al., 2009). Model simulations were driven using observations of solar radiation, air temperature, relative humidity (or vapour pressure deficit; VPD), rainfall, atmospheric CO₂ concentration and LAI (if prescribed), and included a spin-up period of 5 years to allow internal states, such as the soil water balance and soil temperature to reach equilibrium. The exception to the above was the BIOS2 model, which was run using gridded meteorological inputs and had its model parameter optimised through a model-data fusion process (see Haverd et al., 2013).

Simulations for each savanna site covered a period of 2 to 10 years depending on the availability of data from each flux site (Table 1) and results were standardised to the ALMA (Assistance for Land-surface Modelling Activities) convention. Model predictions of LE and GPP were then evaluated against local observations at each site from the eddy covariance datasets and benchmarked following the methodology proposed by the PALS Land Surface Model Benchmarking Evaluation Project (PLUMBER) (Abramowitz, 2012; Best et al., 2015) as described below.

2.4 Empirical benchmarking

The paradigm for model assessment set out in the Protocol for the Analysis of Land-surface models (PALS) (Abramowitz, 2012) suggests that model assessment is more meaningful when a priori expectations of performance in any given metric can be defined. Such benchmarks can be created using simple empirical models, built on statistical relationships between the fluxes and drivers, and establish the degree to which

models utilise the information available in their driving data about the fluxes they aim to predict. Additionally, these empirical models are simple in the sense that they are purely instantaneous response to time-varying meteorological forcing and contain no internal states or expression of ecophysiological processes. This is in comparison to TBMs that are complex, having some 20+ soil and vegetation parameters, internal states, partitioning of light, as well as soil and vegetation, carbon and nitrogen pools (Abramowitz et al., 2008).

We created a set of 3 empirical models of increasing complexity following the procedure of Abramowitz (2012), which we compare with the TBMs. The first benchmark (emp1) is simply a linear relationship between a turbulent flux (LE or GPP) and downward short-wave radiation (R_s). The second benchmark (emp2) is slightly more complex, and is a multi-linear regression between a flux and R_s , air temperature (T_a), and vapour pressure deficit (VPD). Finally, the third benchmark (emp3) is the most complex and is a nonlinear regression of the fluxes against R_s , T_a , VPD and LAI, determined from an ANN. This benchmark is constructed using a self-organising linear output map that clusters the four covariates into 10^2 distinct nodes and performs a multi-linear regression between the fluxes and the 4 covariates at each node, resulting in a nonlinear (piece-wise linear) response to the meteorological forcing data (Abramowitz et al., 2008; Hsu, 2002). In a departure from Abramowitz (2012), we include LAI as an additional covariate, as the seasonal variance of savanna water and carbon exchange is strongly coupled to the phenology of the grasses and to the deciduous and semi-deciduous woody species (Moore et al., 2015). Empirical benchmarks are created for each of the five flux sites using non gap-filled data, and are parameterised *out-of-sample*, such that they use data from all sites except the one in question. For example, the Howard Springs empirical benchmark models would use information from Adelaide River, Daly Uncleared, Dry River and Sturt Plains to establish their parameter values, but would exclude Howard Springs itself. Constructing the benchmarks out-of-sample results in what is effectively a generalised response to an independent dataset. Once

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the empirical models were calibrated for each site, benchmarks were then created for each flux using the same meteorological forcing used to run the TBMs.

Finally, we assess ecosystem model performance in terms of a ranking system, following the PLUMBER methodology of Best et al. (2015). The performance of each individual ecosystem model in predicting both LE and GPP at each site was determined using four statistical metrics that describe the mean and variability of a model compared to the observations. These metrics included the correlation coefficient (r), standard deviation (SD), normalised mean error (NME), and mean bias error (MBE) (see Supplement Table S1). Similarly, the same metrics were determined for each of 3 benchmarks at each savanna site. Each ecosystem model was then ranked against the benchmarks (independently of the other models) for each of the metrics listed above. The ranking is between 1 and 4 (1 model + 3 benchmarks), where the best performing model for a given metric is ranked as 1. An average ranking is then determined across all metrics for each ecosystem model and benchmark to give a final ranking of performance at the savanna site in question. The ranks denote the number of metrics being met by the models and are not a measure of the smallest absolute error. To determine the average ranks, the metrics were evaluated at the daily time scale, as this was the lowest temporal resolution common amongst the models. Additionally, days where either driver or flux had been gap-filled were removed in the determination of these metrics. Herewith we use the term *performance* to relate to how well the TBMs compare to the benchmarks as expressed by the ranks.

3 Results

3.1 Model predictions

Figure 2 shows the daily time-course of LE and GPP from the flux tower, models, and benchmarks at each of the five savanna sites. Models, benchmarks and observations are represented as a smoothed time-series (7 day running mean) and have been ag-

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gregated into an ensemble year to express the typical seasonality of savanna water and carbon exchange. Visually, the TBMs showed varying levels of performance across the rainfall gradient. None of the models showed a clear consistency in simulating either flux and responded differently to the meteorological drivers across sites. Additionally, some of the models, such as CABLE and LPJ-GUESS, showed difficulty in simulating the seasonality of the fluxes across the transect, particularly GPP. Differences among model simulated LE and GPP were larger in the wet season than the dry season. However, modelled LE and GPP appeared to co-vary quite strongly; overall both fluxes were underestimated across sites by most models. Simulations by SPA and MAESPA were the exception to this, broadly capturing tower GPP, despite consistently underestimating LE across sites.

Figure 3 shows the probability density functions (PDFs) for the wet (November–April) and dry season (May–October) fluxes at each savanna site. Tower and model PDFs were determined by binning each flux into the respective seasons and using kernel density estimation (Bashtannyk and Hyndman, 2001) to determine smoothed distributions. The shape and spread of the distributions highlight possible biases in the models (over- or underestimating the tower fluxes), as well as their ability to capture the spread of values. The PDFs for the tower fluxes tended to shift to low values and became narrower as annual rainfall declined, and this was most prominent in the dry season. By contrast, the PDFs of the model simulations did not replicate this trend, being mostly stationary (i.e. density of values do not shift) across sites, especially for wet season GPP. For example, the LPJGUESS and MAESPA models showed distributions with higher densities at low LE ($20\text{--}40\text{ W m}^{-2}$) across all sites and seasons, whereas during the wet season almost all models except MAESPA and SPA had higher densities of GPP at low values ($3\text{--}4\text{ }\mu\text{mol m}^{-2}\text{ s}^{-1}$). In models that did replicate the observed spread and shape of the tower fluxes (CABLE, LPJ-GUESS), PDFs were biased towards lower values particularly for LE, where there was little overlap. A consistent pattern emerged among the models, suggesting that drier conditions resulted in PDFs with higher densities at low values for both fluxes.

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The benchmarks showed that there was enough information in R_s (emp1) to predict the magnitude and daily time-course of the tower fluxes (data not shown). However, additional information contained in T_a and VPD (emp2) was needed to capture the water and carbon flux seasonality. Importantly additional phenological information (provided as LAI in emp3) was required in order to fully capture the seasonality of the savannas fluxes and provide predictions with the lowest absolute error. This indicates that in order for models to achieve the best possible performance simulating savanna ecosystems they will require correct information on phenology.

3.2 Residual analysis

Figure 4 shows how model structure (internal processes) is affecting predictions of savanna fluxes across the rainfall gradient. To do this we examine the standardised model residuals from each TBM where the scale is expressed in terms of standard deviations. Figure 4 shows the LE and GPP residuals at each savanna site represented in two ways: (i) against time, and (ii) against the flux prediction itself. Plotting the residuals against time provides an effective way of examining how a model responds to progressive changes in the environment, while plotting the residuals against the model prediction gives a measure of a model's bias (Medlyn et al., 2005). A linear regression has been applied to the residuals vs. flux prediction scatter plots in order to better visualise the degree of bias.

Model residuals for both LE and GPP illustrated a larger tendency towards underestimation in the wet than in the dry season. The LE residuals showed a larger between-day variance than the GPP residuals in the wet season, as LE. A possible explanation for this is that LE is the sum of multiple evaporation components: vegetation transpiration, soil evaporation, wet canopy, i.e. 3 sources for potential error. However, the error and variance of the LE residuals reduced into the dry season, and as the site itself became drier. It is likely that this reduced error and variance in LE was a combination of the increasing contribution of woody transpiration to the bulk land-surface LE as the grass transpiration and soil evaporation declined (via senescence and low surface soil

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models was at best equivalent to that of a multi-linear regression against R_s , T_a and VPD (emp2) and in some cases no better than a linear regression against R_s (emp1). Given that this subset of TBMs are sophisticated process-based models that represent our best understanding of land-surface, atmospheric exchange processes, we would expect them to perform as well as a neural network prediction (emp3). Consequently there is an evident underutilisation of the driving information (i.e. a failure to describe the underlying relationships in the data) impeding the performance of these models when predicting savanna fluxes. However, there were instances where some of the TBMs were able to reach similar levels of performance with the emp3 benchmark, and strongly suggests that each of these models is capable of replicating savanna dynamics under certain conditions (e.g. during the dry season).

Our results suggest that errors among models are likely to be systematic, rather than related to calibration of existing parameters. For example, BIOS2 had previously optimised model parameters for Australian vegetation (see Haverd et al.2013), but was still unable to out-perform the emp3 benchmark in most cases, although performed better than an un-calibrated CABLE in most cases, to which it is functionally similar. Similarly, MAESPA and SPA, which used considerable site characteristic information to parameterise their simulations, did not significantly outperform un-calibrated models (e.g. CABLE). Additionally, despite these models using the same leaf, root and soil parameterisations, both SPA and MAESPA displayed markedly different performances in predicting LE. Consequently, improving how models represent key processes that drive savanna dynamics is critical to improving model performance across this ecosystem.

There is certainly enough information in the time-varying model inputs to be able to adequately simulate wet and dry season dynamics, as is evidenced by the benchmarks. We therefore consider the implications of our results, and present possible reasons below for why this group of TBMs is failing to capture water and carbon exchange along the NATT, and make suggestions as to how this could be improved.

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4.1 Water access and tree rooting depth

During the late dry season surface soil moisture in the sandy soils declines to less than 3% volumetric water content, with an equivalent matric potential of 3 to 4 MPa (Prior et al., 1997). During this seasonal phase, the grass understorey becomes inactive and LE can be considered as equivalent to tree transpiration, such that it is the only active component during this period (O'Grady et al., 1999). Using this equivalence, one can infer the relative effect that rooting depth has on LE during this period. Previous studies have shown that for these savanna sites along the NATT, tree transpiration is maintained throughout the dry season by deep root systems that access deep soil-water stores, which in turn are recharged over the wet season (Eamus et al., 2000; Hutley et al., 2001; Kelley et al., 2007; O'Grady et al., 1999). In order for models to perform well they will need to set adequate rooting depths and distributions, along with root water uptake process, to enable a model response to such seasonal variation. Examining performance across the models, we can infer this to be a key deficiency. As expected, TBMs that prescribed shallow rooting depths (e.g. LPJ-GUESS) did not simulate this process well, and underestimated dry season LE at 3 of the 5 savanna sites by up to 30 to 40%. The two sites at Adelaide River and Sturt Plains were an exception to this with the TBMs displaying a low residual error, which is likely to be a consequence of heavier textured soils and trees at these sites having shallow root profiles. At Adelaide River shallow root profiles are a consequence of shallow, heavier textured soils, however dry season transpiration is sustained due to the presence of saturated yellow hydrosol soils. Sturt Plains is a grassland (end member of the savanna continuum) where C_4 grasses dominate and no trees are present such that transpiration is close to zero in the dry season. The few small shrubs that have established have shallow root profiles that have adapted to isolated rainfall events driven by convective storms (Eamus et al., 2001; Hutley et al., 2001, 2011). Consequently, the TBMs would be expected perform better at these sites, as water and carbon exchange will be modulated by the soil-water status of the sub-surface soil layers. For the other

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the sum of woody and herbaceous transpiration (E_{veg}) as well as soil and wet-surface evaporation (E_{soil}); in contrast dry season LE is predominantly woody transpiration as described previously. During the wet season, up to 75 % of total LE arises from understorey herbaceous transpiration and soil evaporation (Eamus et al., 2001; Hutley et al., 2000; Moore et al., 2015) and of this fraction the C_4 understorey contributes a significant daily amount (Hutley et al., 2000). In the absence of observations of grass LE it can be difficult to determine whether grass transpiration is being simulated correctly (lacking observed grass LE). However, separating out the components of wet season LE into soil and vegetation can help identify which of these components are causes for error.

Separating the outputs of simulated E_{veg} and E_{soil} from each TBM (excluding BESS which did not determine these as outputs during the study) shows that simulated wet season E_{veg} was particularly low for a lot of the models, despite high LAI and non-limiting soil-water conditions (Fig. 6). A previous study at Howard Springs by Hutley et al. (2000) observed that during the wet season, the grass understorey could transpire $\sim 2.8 \text{ mm d}^{-1}$, while the tree canopy transpired only 0.9 mm d^{-1} ($E_{veg} = 3.7 \text{ mm d}^{-1}$). Of the 6 TBMs at Howard Springs, only CABLE and SPA were able to predict an E_{veg} close to this level, while the other models predicted values closer to tree transpiration (i.e. an under-estimate). This pattern is similar for other NATT sites, where predicted wet season E_{veg} remained low and was dominated by E_{soil} at the southern end of the NATT. An underestimation of wet season LE could be due to underestimated E_{soil} in some of the models. Conversely, CABLE and BIOS2 predicted a higher E_{soil} than the other models, and this could be a reason for their higher LE performance during the wet season. Although E_{soil} has been reported to reach as high as 2.8 mm d^{-1} at Howard Springs (Hutley et al., 2000), predicted E_{soil} by these models may still be overestimated, given that vegetation cover during this period is at a seasonal peak (limiting energy available at the soil surface) and transpiration is only limited by available energy not water (Hutley et al., 2000; Ma et al., 2013; Schymanski et al., 2009; Whitley et al., 2011). Given the limited data for E_{soil} along the NATT, it is difficult to determine how large E_{soil} should

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tude of the fluxes to achieve good performance. LAI integrates the observed structural changes of the savanna as annual rainfall declines with reduced woody stem density; driving water and carbon exchange as a result (Kanniah et al., 2010; Ma et al., 2013; Sea et al., 2011). If LAI is prescribed in a model, it is important that leaf area is partitioned correctly between the trees and grass layers to describe their respective phenology. This partitioning is important, as the C₄ grass understorey explains most of the seasonal variation in LAI, and is a consequence of an annual phenology that exhibits rapid growth at the onset of the wet season and senescence at the onset of the dry (Williams et al., 1996b). By contrast the evergreen eucalypt canopy shows modest reductions in canopy leaf area during the dry season, especially as mean annual rainfall declines (Bowman and Prior, 2005; Kelley et al., 2007). The strong seasonal dynamics of the grasses result in large changes in LAI, with levels varying between 0.7 and 2.5 at high rainfall sites (Sea et al., 2011). The phenological strategy of the C₄ grasses also changes with rainfall interannual variability, with the onset of the greening period becoming progressively delayed as sites become drier, to become eventually rain-pulse driven as the monsoonal influence weakens (Ma et al., 2013).

With the exception of LPJ-GUESS, all models prescribed LAI as an input driver. Prescribing LAI can be problematic depending on the time-scale and how it is partitioned between trees and grass layers. At large time-steps (months) it will fail to capture the rapidly changing dynamics of vegetation during the transition periods, and this is particularly true for the onset of the wet season (September–November) especially at drier sites that are subject to larger inter-annual rainfall variability (Hutley et al., 2011). Additionally, as the sites become drier the tree:grass ratio will become smaller and this dynamic can be difficult to predict, although methods do exist (see Donohue et al., 2009). From the results, we infer that TBMs that prescribe LAI and allow for a dynamic representation of tree and grass ratios are better able to capture the changing dynamics of the savanna system. This is a possible explanation for the better performance of MAESPA and SPA models in simulating GPP as both models dynamically partition leaf area between trees and grasses at the sub-monthly time-scale, rather than use a bulk

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value. However, there are limitations to using prescribed LAI, predominantly in that it describes a stable system, of which savannas are typically not; having a large sensitivity to changes in climate; particularly rainfall variability and disturbance (Sankaran et al., 2005). DGVMs that consider dynamic vegetation and use a prognostic LAI can simulate the feedback between the climate and the relative and absolute abundance or cover of trees and grasses, which shapes the savanna continuum. This feedback allows the simulated savanna structure to potentially shift to alternate states (e.g. grassland or forest) in response to changes in annual rainfall and fire severity (Scheiter and Higgins, 2007, 2009). While LPJ-GUESS was the only TBM to use a prognostic LAI in our study, it achieved only moderate performance, and this may be due to how carbon is allocated from the pool on an annual time step, such that it is not as dynamic as it could be. However, its capability to simulate the feedback between climate and LAI is critical for simulating how savanna dynamics may change from year to year. There may also be issues with how phenology is simulated, particularly as it is determined from empirical formulations, which are: (i) not specifically developed for savanna environments and (ii) calculated before the growing season begins. Such formulations are therefore not mechanistic, responding to actual season dynamics (e.g. limiting soil water), but are empirically determined (Richardson et al., 2013).

5 Conclusions

This study set out to assess how well a set of functionally different, state-of-the-art TBMs perform at predicting the bulk exchanges of carbon and water over savanna land surfaces. Our model inter-comparison has identified key weaknesses in the assumptions of biosphere–atmosphere processes, which do not hold for savanna environments. Our benchmarking has identified low model performance by TBMs is likely a result of incorrect assumptions related to: (i) deep soil water access, (ii) a systematic under-estimation of the contribution of the grass understorey in the wet season, and (iii) the use of static phenology to represent dynamic vegetation. Our results showed

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that these assumptions, as they currently exist in TBMs, are not wholly supported by “observations” of savanna water and carbon exchange and need to be addressed if more reliable projections are to be made on how savannas respond to environmental change. Despite this, our benchmarking has shown that all TBMs could potentially operate well for savanna ecosystems, provided that the above issues are developed on. We suggest that further work investigates how particular processes in the models may be affecting overall predicted water and carbon fluxes, and may include testing variable rooting depths, alternate root-water uptake schemes and how these might affect leaf-level outputs (e.g. stomatal conductance, leaf water potential) among TBMs, and different phenology schemes. The issues highlighted here also have scope beyond savanna environments, and are relevant to other water-limited ecosystems. The results from this study provide a foundation for improving how savanna ecosystem dynamics are simulated.

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Table 1. Summarised dataset information for each of the five savanna sites used in this study. This includes site descriptions pertaining to local meteorology, vegetation and below ground soil characteristics. Where data were not available, the abbreviation NA is used. Definitions for the species genus mentioned in the table are as follows: *Eucalyptus* (*Eu.*), *Erythrophleum* (*Er.*), *Terminalia* (*Te.*), *Corymbia* (*Co.*), *Planchonia* (*Pl.*), *Buchanania* (*Bu.*), *Themda* (*Th.*), *Heteropogon* (*He.*), and *Chrysopogon* (*Ch.*). Eddy covariance datasets relating to each of the 5 sites here can be download from www.ozflux.org.au and hdl references are given by order of column (Jason Beringer, 2013).

	Howard Springs ^a	Adelaide River ^b	Daly Uncleared ^c	Dry River ^d	Sturt Plains ^e
Years (inclusive)	2001–2012	2007–2009	2008–2012	2008–2012	2008–2012
Co-ordinates	12°29′39.12″ S 131°09′09″ E	13°04′36.84″ S 131°07′04.08″ E	14°09′33.12″ S 131°23′17.16″ E	15°15′31.62″ S 132°22′14.04″ E	17°09′02.76″ S 133°21′01.14″ E
Elevation (m)	64	90	110	175	250
^f Meteorology					
Annual Rainfall (mm)	1714	1460	1170	850	535
Min/Max Daily Temperature (°C)	22.0/33.0	21.8/35.3	20.8/35.0	20.0/34.8	19.0/34.2
Min/Max Absolute Humidity (gm ⁻³)	11.0/18.5	8.9/17.7	8.6/15.1	7.8/12.3	6.1/9.0
Min/Max Soil Moisture (m ³ m ⁻³)	0.06/0.1	0.09/0.14	0.03/0.06	0.03/0.05	0.04/0.1
Soil Temperature (°C)	32.7	35.7	32.8	NA	30.2
Solar Radiation (W m ⁻²)	256.5	258.1	270.6	266.5	269.7
Bowen Ratio	1.7	3.1	3.2	4.6	15.8
^f Vegetation					
Overstorey species	<i>Eu. Miniata</i> <i>Eu. tetradonta</i>	<i>Eu. tectifera</i> <i>Pl. careya</i>	<i>Te. grandiflora</i> <i>Eu. tetradonta</i>	<i>Eu. tetradonta</i> <i>Co. terminalis</i>	NA
Understorey species	<i>Er. chlorostachys</i> <i>Sorghum</i> spp. <i>He. triticeus</i>	<i>Co. latifolia</i> <i>Sorghum</i> spp. <i>Ch. fallax</i>	<i>Co. latifolia</i> <i>Sorghum</i> spp. <i>He. triticeus</i>	<i>Eu. dichromophloia</i> <i>Sorghum intrans</i> <i>Th. Tiandra</i> <i>Ch. fallax</i>	<i>Astrabla</i> spp.
Basal Area (m ² ha ⁻¹)	9.7	5.1	8.3	5.4	NA
Canopy Height (m)	18.9	12.5	16.4	12.3	0.2
LAI (m ² m ⁻²)	1.04 ± 0.07	0.68 ± 0.07	0.80 ± 0.12	0.58 ± 0.11	0.39 ± 0.11
Total Leaf Nitrogen (gm ⁻³)	1.42 ± 0.20	1.27 ± 0.18	1.35 ± 0.19	1.97 ± 0.15	2.37 ± 0.17

^a hdl: 102.100.100/14228, ^b hdl: 102.100.100/14239, ^c hdl: 102.100.100/14229, ^d hdl: 102.100.100/14234, ^e hdl: 102.100.100/14230. Site meteorology is given as 30 year averages with values taken from ^f Hutley et al. (2011). Soil descriptions are taken from the Digital Atlas of Australian Soils (www.asris.csiro.au), ^g Isbell (2002).

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Table 1. Continued.

	Howard Springs ^a	Adelaide River ^b	Daly Uncleared ^c	Dry River ^d	Sturt Plains ^e
^g Soil					
Type	Red kandosol	Yellow hydrosol	Red kandosol	Red kandosol	Grey vertosol
A Horizon					
Texture	Sandy loam	Sandy loam	Loam	Clay	loam
Clay PSD (%)	15	20	20	50	20
Sand PSD (%)	60	50	40	25	40
Thickness (m)	0.30	0.30	0.20	0.15	0.20
Bulk Density (Mg m ⁻³)	1.29	1.60	1.39	1.20	1.39
Hydraulic Conductivity (mm h ⁻¹)	9	7	9	3	9
Field Capacity (mm m ⁻¹)	156	132	147	140	147
B Horizon					
Texture	Clay loam	Clay	Clay loam	Clay	Clay loam
Clay PSD (%)	40	55	35	55	35
Sand PSD (%)	30	20	30	20	30
Thickness (m)	1.20	0.60	0.69	1.29	0.69
Bulk Density (Mg m ⁻³)	1.39	1.70	1.39	1.39	1.39
Hydraulic Conductivity (mm h ⁻¹)	8	5	7	2	7
Field Capacity (mm m ⁻¹)	146	31	146	107	146

^a hdl: 102.100.100/14228, ^b hdl: 102.100.100/14239, ^c hdl: 102.100.100/14229, ^d hdl: 102.100.100/14234, ^e hdl: 102.100.100/14230. Site meteorology is given as 30 year averages with values taken from ^f Hutley et al. (2011). Soil descriptions are taken from the Digital Atlas of Australian Soils (www.asris.csiro.au), ^g Isbell (2002).

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Table 2. Summary table of the ecosystem models used in the experiment; highlighting differences and similarities in model structure and shared processes. Information is broken down into how each model describes aboveground canopy and belowground soil processes.

Model Name	SPA	MAESPA	CABLE	BIOS2	BESS	LPJ-GUESS
Model definition	Soil-Plant-Atmosphere Model	MAESTRA-SPA	Community Atmosphere Biosphere Land-surface Exchange Model	Modified CABLE (CABLE + SLI + CASA-CNP)	Breathing Earth System Simulator	Lund-Potsdam-Jena General Ecosystem Simulator
Version Reference	1.0 Williams et al. (1996a)	1.0 Duursma and Medlyn (2012)	2.0 Kowalyzck et al. (2006), Wang et al. (2011)	2.0 Haverd et al. (2013)	1.0 Ryu et al. (2011, 2012)	2.1 Smith et al. (2001)
Temporal resolution	30 min	30 min	30 min	Daily (30 min time-steps are generated from daily time-series)	Snap shot with MODIS overpass, then up-scaled to a daily and 8 day time series	Daily
Spatial resolution	Point	Point	0.05° (5 km)	0.05° (5 km)	0.05° (5 km)	Patch (c. 0.1 ha)
Functional class	Stand model	Individual Plant or Stand Model	Land-Surface Model	Land-Surface Model	Remote Sensing Model	Dynamic Global Vegetation Model
Canopy Description						
C ₃ Assimilation	Farquhar et al. (1980)	Farquhar et al. (1980)	Farquhar et al. (1980)	Farquhar et al. (1980)	Farquhar et al. (1980)	Collatz et al. (1991)
C ₄ Assimilation	Collatz et al. (1992)	Collatz et al. (1992)	Collatz et al. (1992)	Collatz et al. (1992)	Collatz et al. (1992)	Collatz et al. (1992)
Stomatal conductance	Williams et al. (1996a)	Medlyn et al. (2011)	Leuning (1995)	Leuning (1995)	Ball et al. (1987)	Haxeltine and Prentice (1996)
Transpiration	Penman–Monteith calculated at leaf-scale accounting for g_b and limitation of soil-water supply via Ψ_f	Penman–Monteith calculated at the leaf scale	Penman–Monteith	Penman–Monteith	Penman–Monteith	Haxeltine and Prentice (1996)
Boundary layer resistance	f (wind speed, leaf width, air temperature)	f (wind speed, leaf width, air temperature and atmospheric pressure)	f (wind speed, leaf width, air temperature)	f (wind speed, leaf width, air temperature)	<i>Not Modelled</i>	Huntingford and Monteith (1998)
Aerodynamic resistance	f (wind speed, canopy height)	Not calculated unless transpiration is calculated at the canopy scale, in which case g_b above isn't calculated.	f (wind speed, canopy height)	f (wind speed, canopy height)	f (wind speed, canopy height)	Huntingford and Monteith (1998)
Leaf area index	Prescribed (MODIS)	Prescribed (MODIS)	Prescribed (MODIS)	Prescribed (MODIS)	Prescribed (MODIS)	Prognostic (C allocation)
Canopy structure	Canopy + understorey divided into 10 layers	Individual plant crowns, spatially explicit locations and uniform understorey	2 (tree/grass) big leaf (sunlit/shaded)	2 (tree/grass) big leaf (sunlit/shaded)	2 (tree/grass) big leaf (sunlit/shaded)	5 year age/size cohorts for trees, single-layer grass understorey
C ₃ : C ₄ fraction	Dynamic ratio variable with time. Compete for water and light.	Dynamic ratio variable with time. Compete for water and light.	Simulated as independent layers	Dynamic ratio variable with time. Compete for water not light.	Still et al. (2003) Ratio changes 70 : 30 to 10 : 90 down transect	Prognostic, determined as the outcome of the competition with trees
Canopy interception	YES	YES	YES	YES	NO	YES
Simulates growth	NO	NO	NO	NO	NO	YES

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Table 2. Continued.

Model Name	SPA	MAESPA	CABLE	BIOS2	BESS	LPJ-GUESS
Soil Profile Description						
Soil profile structure	Profile divided into N layers (prescribed – 20 in this case.)	Profile divided into N layers (prescribed – 20 in this case.)	Profile divided into 6 layers	Profile divided into 12 layers (adjustable)	<i>Not Modelled</i>	2 layers (0–0.5, 0.5–2 m) with 10 cm evaporation sub-layer
Soil hydraulic properties	Function of sand and clay particle size distributions	Function of sand and clay particle size distributions	Prescribed	Australian Soils Resource Information System (ASRIS)	<i>Not Modelled</i>	Sitch et al. (2003)
Soil depth	6.5 m	5.0 m	4.5 m	10.0 m	<i>Not Modelled</i>	2 m
Root depth	6.5 m	5.0 m	4.5 m	0.5 m (grasses), 5.0 m (trees)	<i>Not Modelled</i>	2 m
Root distribution	Prescribed; exponential decay as a function of surface biomass and the total root biomass of the column	Prescribed; exponential decay as a function of surface biomass and the total root biomass of the column	Prescribed; exponential decay	Prescribed; exponential decay	<i>Not Modelled</i>	PFT-specific, trees have deeper roots on average
Soil-water stress modifier	E_s via g_s is increased to meet atmospheric demand while Ψ_l remains above a critical threshold	Maximum transpiration rate calculated from hydraulic conductance (soil-to-leaf) sets limit on actual transpiration, OR uses the Tuzet et al. (2003) model of stomatal conductance	Supply/Demand	g_s scaled by a soil moisture limitation function related to extractable water accessible by roots	Assumes LAI and seasonal variation of V_{cmax} reflect soil water stress	Supply/Demand
Hydraulic pathway resistance	$R_{soil} + R_{plant}$	$R_{soil} + R_{plant}$	<i>Not Modelled</i>	<i>Not Modelled</i>	<i>Not Modelled</i>	Not explicit, min(supply, demand) determines sapflow

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Table A1. Definition of common metrics used to determine ranks against the empirical benchmarks. The terms M and O stand for model and observations respectively, while n denotes the length of the data, and i is the datum.

Statistical Metric	Definition
Correlation coefficient (r)	$\frac{n \sum_{i=1}^n (O_i M_i) - \sum_{i=1}^n O_i \sum_{i=1}^n M_i}{\sqrt{\left(n \sum_{i=1}^n O_i^2 - \left(\sum_{i=1}^n O_i \right)^2 \right) \left(n \sum_{i=1}^n M_i^2 - \left(\sum_{i=1}^n M_i \right)^2 \right)}}$
Standard Deviation (SD)	$\left 1 - \frac{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (M_i - \bar{M})^2}}{\sqrt{\frac{1}{n-1} \sum_{i=1}^n (O_i - \bar{O})^2}} \right $
Normalised mean error (NME)	$\frac{\sum_{i=1}^n M_i - O_i }{\sum_{i=1}^n \bar{O} - O_i }$
Normalised mean bias (MBE)	$\frac{1}{n} \sum_{i=1}^n (M_i - O_i)$

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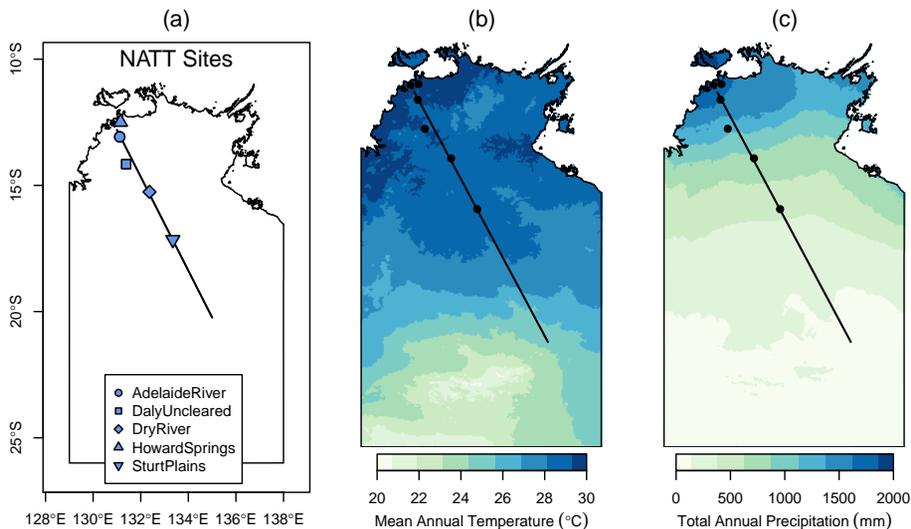


Figure 1. The Northern Territory of Australia and the North Australian Tropical Transect (NATT) showing **(a)** the flux site locations with an accompanying 30 year (1970 to 2000) expression of the average meteorological conditions for **(b)** mean annual temperature, and **(c)** total annual precipitation derived from ANUCLIM v6.1 climate surfaces (Hutchinson and Xu, 2010).

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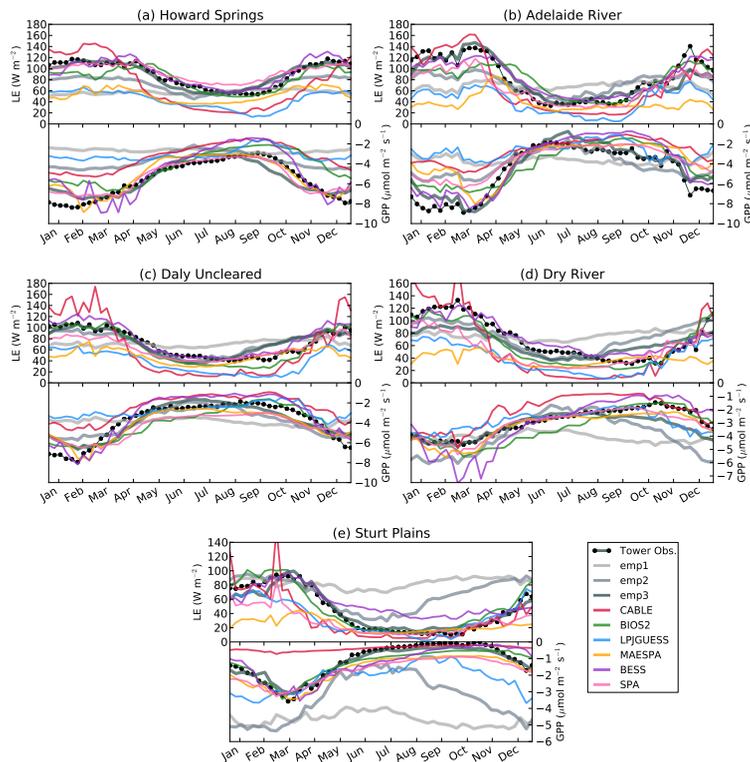


Figure 2. Time-series of daily mean latent heat (LE) flux and gross primary productivity (GPP) depicting an average year for each of the 6 savanna sites using a smoothed, 7 day moving average. The sites are ordered from wettest to driest; **(a)** Howard Springs, **(b)** Adelaide River, **(c)** Daly River, **(d)** Dry River and **(e)** Sturt Plains. The joined, black dots are the tower flux time-series, while the grey lines are the performance benchmarks (emp1, emp2, emp3). Predictions of LE and GPP for each of the six terrestrial biosphere models are given by a spectrum of colours described in the legend.

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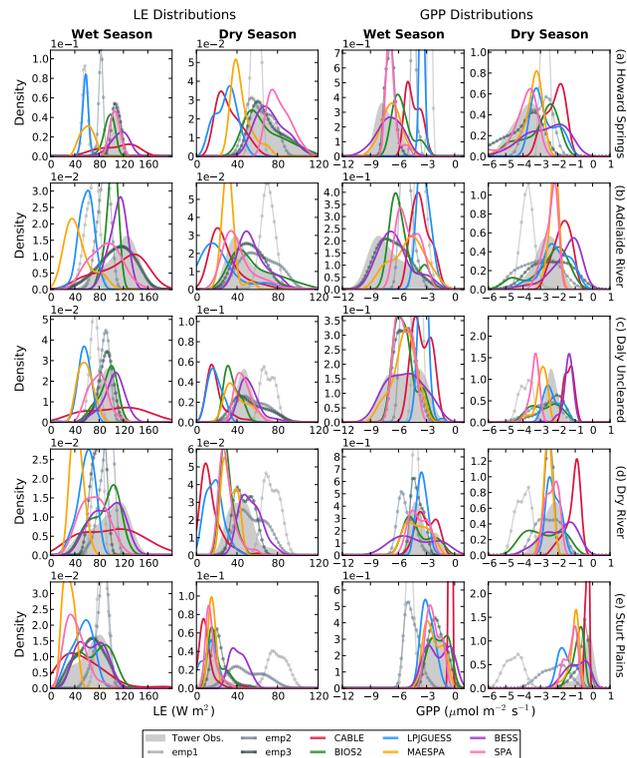


Figure 3. Probability densities (expressed in scientific notation) of daily mean latent heat (LE) flux and gross primary productivity (GPP) at each of the 6 savanna sites, where the distributions for each flux are partitioned into wet and dry seasons. The order of the sites are from wettest to driest; **(a)** Howard Springs, **(b)** Adelaide River, **(c)** Daly River, **(d)** Dry River and **(e)** Sturt Plains. The grey region is the tower flux, while the dotted lines are the empirical benchmarks. Predicted LE and GPP probability densities from each of the six process-based models are given by a spectrum of colours described in the legend.

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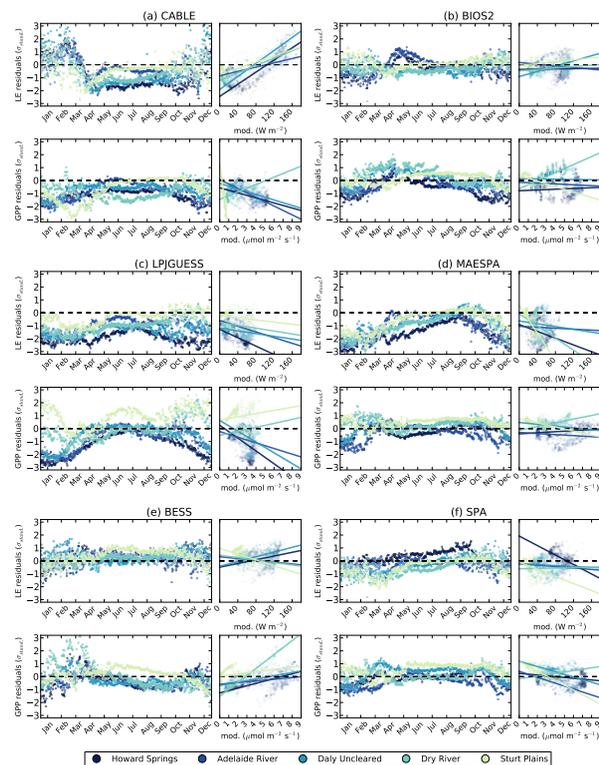


Figure 4. Standardised model residuals for latent energy (LE) and gross primary productivity (GPP) expressed in units of standard deviations (SD) [(modelled flux – observed flux)/SD(observed flux)]. Residuals are presented for each model: **(a)** CABLE, **(b)** BIOS2, **(c)** LPJGUESS, **(d)** MAESPA, **(d)** BESS and **(e)** SPA, where each flux site is represented by a blue-green-yellow gradient. For both fluxes, the residuals are plotted against time (ensemble average year) and against the flux prediction (bias).

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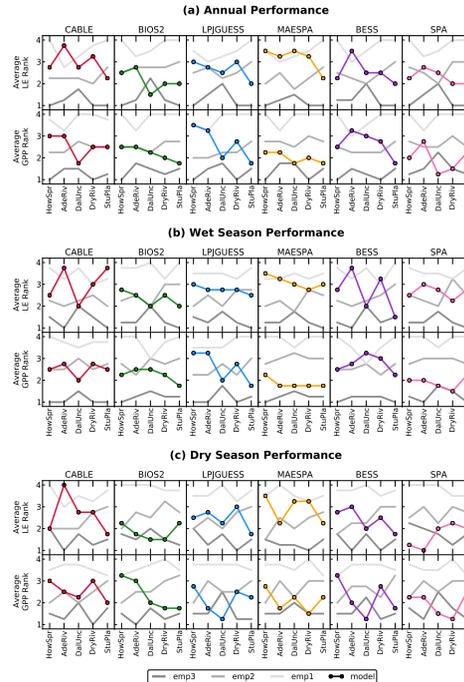


Figure 5. Average rank plot showing the performance of the terrestrial biosphere models for all sites across the North Australian Tropical Transect (NATT) ordered in terms of annual rainfall as follows: Howard Springs (HowSpr), Adelaide River (AdrRiv), Daly Uncleared (DalUnc), Dry River (DryRiv), and Sturt Plains (StuPla). Models are individually ranked against the benchmarks in order of 1 to 4 (1 model + 3 benchmarks) and express the amount of metrics the models are meeting listed in Table S1 in the Supplement. The rankings are determined individually for latent energy (LE) and gross primary productivity (GPP). The coloured lines represent each of the 6 models in the study, while the grey lines represent the empirical benchmarks. The average ranking for each model was determined for **(a)** a complete year, **(b)** the wet season and **(c)** the dry season.

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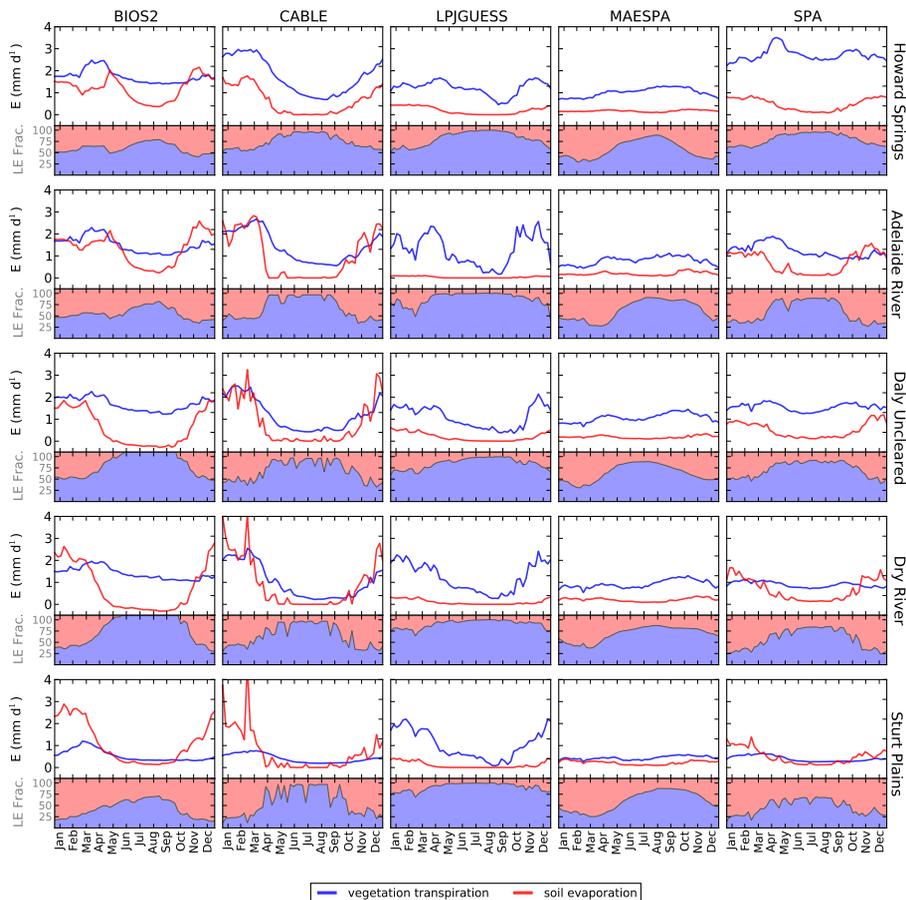


Figure 6. Average year outputs of vegetation transpiration (grass + trees) and soil evaporation, as well as their percentage contributions to total latent energy (LE) for each of the 6 terrestrial biosphere models at each of the 5 savanna sites.

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