Exploring how groundwater buffers the influence of heatwaves on vegetation function during multi-year droughts

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Abstract. The co-occurrence of droughts and heatwaves can have significant impacts on many socioeconomic and environmental systems. Groundwater has the potential to moderate the impact of droughts and heatwaves by moistening the soil and enabling vegetation to maintain higher evaporation, thereby cooling the canopy. We use the Community Atmosphere Biosphere Land Exchange (CABLE) land surface model, coupled to a groundwater scheme, to examine how groundwater influences ecosystems under conditions of co-occurring droughts and heatwaves. We focus specifically on South East Australia for the period 2000–2019 when two significant droughts and multiple extreme heatwave events occurred. We found groundwater plays an important role in helping vegetation maintain transpiration, particularly in the first 1–2 years of a multi-year drought. Groundwater impedes gravity-driven drainage and moistens the root zone via capillary rise. These mechanisms reduced forest canopy temperatures by up to 5°C during individual heatwaves, particularly where the water table depth is shallow. The role of groundwater diminishes as the drought lengthens beyond 2 years and soil water reserves are depleted. Further, the lack of deep roots or stomatal closure caused by high vapour pressure deficit or high temperatures can reduce the additional transpiration induced by groundwater. The capacity of groundwater to moderate both water and heat stress on ecosystems during simultaneous droughts and heatwaves is not represented in most global climate models, suggesting model projections may overestimate the risk of these events in the future.

1 Introduction

Droughts and heatwaves are important socio-economic and environmental phenomena, impacting regional food production (Kim et al., 2019; Lesk et al., 2016), water resources (Leblanc et al., 2009; Orth and Destouni, 2018) and the resilience of ecosystems (Ibáñez et al., 2019; Ruehr et al., 2019; Sandi et al., 2020). When droughts and heatwaves co-occur (a “compound event”) the consequences can be particularly severe, reducing the terrestrial carbon sink (Ciais et al., 2005), potentially accelerating tree die-off (Allen et al., 2010, 2015; Birami et al., 2018) and setting conditions conducive for wildfires (Jyoteshkumar reddy et al., 2021). One region experiencing severe coincident heatwaves and drought is Australia (Mitchell et al., 2014). Drought in Australia is associated with large-scale modes of variability, including the El Niño-Southern Oscillation and the Indian Ocean Dipole (van Dijk et al., 2013), and periods of below average rainfall can extend for multiple years (Verdon-Kidd and Kiem, 2009). Heatwaves are commonly synoptically driven, associated with blocking events that can be sustained over many days (Perkins-Kirkpatrick et al., 2016; Perkins, 2015). Modes of variability and synoptic situations are important in setting up conditions conducive to drought and heatwave. However, once a heatwave or drought has become established, land-atmosphere interactions can intensify and prolong both heatwaves and droughts (Miralles et al., 2019), affect their intensity and influence the risk of their co-occurrence (Mukherjee et al., 2020). The role of the land surface in amplifying or dampening heatwaves and droughts is associated with the partitioning of available energy between sensible and latent heat (Fischer et al., 2007; Hirsch et al., 2019) and is regulated by sub-surface water availability (Teuling et al., 2013; Zhou et al., 2019). As soil
moisture becomes more limiting, more of the available energy is converted into sensible heat, reducing evaporative cooling via latent heat. Changes in the surface turbulent energy fluxes influence the humidity in the boundary layer, the formation of clouds, incoming solar radiation and the generation of rainfall (D’Odorico and Porporato, 2004; Seneviratne et al., 2010; Zhou et al., 2019). The sensible heat fluxes warm the boundary layer, leading to heat that can accumulate over several days and exacerbate heat extremes (Miralles et al., 2014), which can in turn increase the atmospheric demand for water and intensify drought (Miralles et al., 2019; Schumacher et al., 2019).

Vegetation access to groundwater has the potential to alter these land-atmosphere feedbacks by maintaining vegetation function during extended dry periods, supporting transpiration and moderating the impact of droughts and heatwaves (Marchionni et al., 2020; Miller et al., 2010). Where the water table is relatively shallow, capillarity may bring water from the groundwater towards the surface root zone, increasing plant water availability. Where the water table is deeper, phreatophytic vegetation with tap roots can directly access groundwater (Zencich et al., 2002). The presence of groundwater, and the access to groundwater by vegetation is therefore likely to buffer vegetation drought and heatwave stress. For example, groundwater may help vegetation sustain transpiration and consequently cool plant canopies via evaporation. This is particularly critical during compound events where cessation of transpiration would increase the risk of impaired physiological function and the likelihood that plants would exceed thermal limits and risk mortality (Geange et al., 2021; O’sullivan et al., 2017; Sandi et al., 2020).

Quantifying the influence of groundwater on vegetation function has remained challenging as concurrent observations of groundwater dynamics, soil moisture, and energy and water fluxes are generally lacking over most of Australia and indeed many parts of the world. Land surface models (LSMs) provide an alternative tool for studying the interactions between groundwater, vegetation, and surface fluxes in the context of heatwaves and droughts (Gilbert et al., 2017; Martinez et al., 2016a; Maxwell et al., 2011; Shrestha et al., 2014). However, there has been very little work focused on the influence of groundwater on droughts and heatwaves occurring at the same time (Keune et al., 2016; Zipper et al., 2019). Our key goal in this paper is therefore to examine the timescales and extent to which vegetation utilises groundwater during drought and heatwaves, and determine the degree to which groundwater can mitigate the impacts of compound extremes. We focus on droughts and heatwaves occurring over south-eastern (S.E.) Australia between during 2000–2019 using the Community Atmosphere Biosphere Land Exchange (CABLE) LSM. S.E. Australia is an ideal case study since its forest and woodland ecosystems are known to be dependent on groundwater (Eamus and Froend, 2006; Kuginis et al., 2016; Zencich et al., 2002) and it has experienced two multi-year droughts and record-breaking heatwaves over the last two decades. By examining the role of groundwater in influencing droughts and heatwaves, and by understanding how well CABLE can capture the relevant processes, we aim to build confidence in the simulations of land-atmosphere interactions for future droughts and heatwaves.

2 Methods

2.1 Study area

The climate over S.E. Australia varies from humid temperate near the coast to semi-arid in the interior. In the last 20 years, S.E. Australia experienced the 9-year Millennium drought during 2001–2009 (van Dijk et al., 2013) where rainfall dropped from a climatological average (1970–1999) of 542 mm yr$^{-1}$ to 449 mm yr$^{-1}$, and a 3-year intense recent drought during 2017-2019 where rainfall dropped to 354 mm yr$^{-1}$ (Figure S1). It has also suffered record-breaking summer heatwaves in 2009, 2013, 2017, and 2019 (Bureau of Meteorology, 2013, 2017, 2019; National Climate Centre, 2009). Here we investigate groundwater interactions during the period 2000–2019, focusing on the Millennium drought (MD, 2001–2009) and the recent drought (RD, 2017–2019).
2.2 Overview of CABLE

CABLE is a process-based LSM that simulates the interactions between climate, plant physiology and hydrology (Wang et al., 2011). Above ground, CABLE simulates the exchange of carbon, energy and water fluxes, using a single layer, two-leaf (sunlit/shaded) canopy model (Wang and Leuning, 1998), with a treatment of within-canopy turbulence (Raupach, 1994; Raupach et al., 1997). CABLE includes a 6-layer soil model (down to 4.6 m) with soil hydraulic and thermal characteristics dependent on the soil type and soil moisture content. CABLE has been extensively evaluated (e.g., Abramowitz et al., 2008; Wang et al., 2011; Zhang et al., 2013) and benchmarked (Abramowitz 2012; Best et al. 2015) at global and regional scales. Here we adopt a version of CABLE (Decker, 2015; Decker et al., 2017) which includes a dynamic groundwater component with aquifer water storage. This version, CABLE-GW, has been previously evaluated by Decker (2015), Ukkola et al. (2016b) and Mu et al. (2021) and shown to perform well for simulating water fluxes. CABLE code is freely available upon registration (https://trac.nci.org.au/trac/cable/wiki); here we use CABLE SVN revision 7765.

2.3 Hydrology in CABLE-GW

The hydrology scheme in CABLE-GW solves the vertical redistribution of soil water via a modified Richards equation (Zeng and Decker, 2009):

\[
\frac{\partial \theta}{\partial t} = - \frac{\partial}{\partial z} K \frac{\partial}{\partial z} (\Psi - \Psi_E) - F_{\text{soil}}
\]

where \( \theta \) is the volumetric water content of the soil (mm\(^3\) mm\(^{-3}\)), \( K \) is the hydraulic conductivity (mm s\(^{-1}\)), \( z \) is the soil depth (mm), \( \Psi \) and \( \Psi_E \) are the soil matric potential (mm) and the equilibrium soil matric potential (mm), and \( F_{\text{soil}} \) is the sum of a sink term related to subsurface runoff and transpiration (mm s\(^{-1}\)) (Decker, 2015; Zeng and Decker, 2009; Decker, 2015). To simulate groundwater dynamics, an unconfined aquifer is added to the bottom of the soil column with a simple water balance model:

\[
\frac{dW_{\text{aq}}}{dt} = q_{\text{re}} - q_{\text{aq,sub}}
\]

where \( W_{\text{aq}} \) is the mass of water in the aquifer (mm), \( q_{\text{aq,sub}} \) is the subsurface runoff in the aquifer (mm s\(^{-1}\)), and \( q_{\text{re}} \) is the water flux between the aquifer and the bottom soil layer (mm s\(^{-1}\)) computed by the modified Darcy’s law:

\[
q_{\text{re}} = K_{\text{aq}} \left( \frac{\Psi_{\text{aq}} - \Psi_n - (\Psi_{E, \text{aq}} - \Psi_{E,n})}{z_{\text{wt}} - z_n} \right)
\]

where \( K_{\text{aq}} \) and \( K_{\text{bot}} \) are the hydraulic conductivity within the aquifer and in the bottom soil layer (mm s\(^{-1}\)), \( \Psi_{\text{aq}} \) and \( \Psi_{E, \text{aq}} \) are the soil matric potentials for the aquifer (mm), and \( \Psi_n \) and \( \Psi_{E,n} \) are the soil matric potentials for the bottom soil layer (mm). \( z_{\text{wt}} \) and \( z_n \) are the depth of the water table (mm) and the lowest soil layer (mm), respectively. The positive \( q_{\text{re}} \) refers to the downward water flow from soil column to aquifer (i.e. vertical drainage, Dr), and the negative \( q_{\text{re}} \) is the upward water movement from aquifer to soil column (i.e., recharge, Qrec). CABLE-GW assumes the groundwater aquifer sits above impermeable bedrock, giving a bottom boundary condition of:

\[
q_{\text{out}} = 0
\]

CABLE-GW computes the subsurface runoff (\( q_{\text{sub}}, \) mm s\(^{-1}\)) using:
where \( \bar{x} \) is the mean subgrid-scale slope, \( \dot{q}_{\text{sub}} \) is the maximum rate of subsurface drainage (mm s\(^{-1}\)) and \( f_p \) is a tunable parameter. \( q_{\text{sub}} \) is generated from the aquifer and the saturated deep soil layers (below the third soil layer).

### 2.4 Experiment design

To explore how groundwater influences droughts and heatwaves, we designed two experiments, with and without groundwater dynamics, driven by the same 3-hour time-evolving meteorology forcing and the same land surface properties (see section 2.5 for datasets) for the period 1970-2019, at a 0.05° spatial resolution with a 3-hour time step. To correct a tendency for high soil evaporation, we implemented a parameterisation of soil evaporation resistance that has previously been shown to improve the model (Decker et al., 2017; Mu et al., 2021).

#### 2.4.1 Groundwater experiment (GW)

This simulation uses the default CABLE-GW model, which includes the unconfined aquifer to hold the groundwater storage and simulates the water flux between the bottom soil layer and the aquifer. We first ran the default CABLE-GW with fixed CO\(_2\) concentrations at 1969 levels for 90 years by looping the time-evolving meteorology forcing over 1970–1999. At the end of the 90-year spin-up, moisture in both the soil column and the groundwater aquifer reached an effective equilibrium when averaged over the study area. We then ran the model from 1970 to 2019 with time varying CO\(_2\). We omit the first 30 years of this period and analyse the period 2000–2019 to allow for further equilibrium with the time-evolving CO\(_2\).

#### 2.4.2 Free drainage experiment (FD)

Many LSMs, including those used in the Coupled Model Intercomparison Project 5 (CMIP5), still use a free drainage assumption and neglect the parameterisation of the unconfined aquifer. To test the impact of this assumption we decoupled the aquifer from the bottom soil layer and thus removed the influence of groundwater dynamics (experiment FD). In FD, at the interface between the bottom soil layer and the aquifer, soil water can only move downwards as vertical drainage at the rate defined by the aquifer bottom soil layer’s hydraulic conductivity:

\[
q_{re} = K_{a} K_{bot}
\]

This vertical drainage is added to the subsurface runoff flux:

\[
q_{\text{sub}} = q_{\text{sub}} + q_{re}
\]

The simulated water table depth (WTD, i.e., \( z_{WTD} \)) in CABLE-GW affects the water potential gradient between the soil layers via \( \Psi_{\theta} \) (Zeng and Decker, 2009) and impacts \( q_{\text{sub}} \) (Equation 5). However, in FD, decoupling the soil column from the aquifer and adding vertical drainage directly to subsurface runoff causes an artificial and unrealistic decline in WTD. To solve this problem, we assume a fixed WTD in the FD simulations at 10 m in order to remove this artefact from the simulation of \( \Psi_{\theta} \) and \( q_{\text{sub}} \).
The FD simulations are initialised from the near-equilibrated state at the end of the 90-year spin-up used in GW. The period 1970–2019 is then simulated using varying CO₂ and the last 20 years are used for analysis.

2.4.3 Deep root experiment (DR)

The parameterisation of roots, including the prescription of root parameters in LSMs, is very uncertain (Arora and Boer, 2003; Drewniak, 2019) and LSMs commonly employ root distributions that are too shallow (Wang and Dickinson, 2012). The vertical distribution of roots influences the degree to which plants can utilise groundwater, and potentially the role groundwater plays in influencing droughts and heatwaves. To explore the uncertainty associated with root distribution, we added a “deep root” (DR) experiment by increasing the effective rooting depth in CABLE for tree areas. In common with many LSMs, CABLE-GW defines the root distribution following Gale and Grigal (1987):

\[ f_{\text{root}} = 1 - \beta_{\text{root}}^z \] (8)

where \( f_{\text{root}} \) is the cumulative root fraction (between 0 and 1) from the soil surface to depth \( z \) (\( \text{cm} \)), and \( \beta_{\text{root}} \) is a fitted parameter specified for each plant functional type (PFT) (Jackson et al., 1996). In CABLE, the tree areas in our study region are simulated as evergreen broadleaf PFT (Figure S2a) with a \( \beta_{\text{root}} = 0.962 \), implying that only 8% of the simulated roots are located below the soil depth of 0.64 cm (Figure S2b). However, field observations (Canadell et al., 1996; Eberbach and Burrows, 2006; Fan et al., 2017; Griffith et al., 2008) suggest that the local trees tend to have a far deeper root system, possibly to help cope with the high climate variability. We therefore increased \( \beta_{\text{root}} \) for the evergreen broadleaf PFT to 0.99, which assumes 56% of roots are located in depths below 6.4 cm and 21% of roots below 1.7 m (Figure S2b). This enables the roots to extract larger quantities of deep soil water moisture, which is more strongly influenced by groundwater.

This is given we lack the detailed observations to set root distributions across S.E. Australia, we undertake the DR experiment as a simple sensitivity study, and therefore, we only run this experiment during January 2019, when the record-breaking heatwaves compound with the severe recent drought. The DR experiment uses identical meteorology forcing and land surface properties as GW and FD, and is initialised by the state of the land surface on the 31st December 2018 from the GW experiment.

2.5 Datasets

Our simulations are driven by the atmospheric forcing from the Australian Water Availability Project (AWAP), which provides daily gridded data covering Australia at 0.05° spatial resolution (Jones et al., 2009). This dataset has been widely used to force LSMs for analysing the water and carbon balances in Australia (Haverd et al., 2013; De Kauwe et al., 2020; Raupach et al., 2013; Trudinger et al., 2016). The AWAP forcing data include observed fields of precipitation, solar radiation, minimum and maximum daily temperatures and vapour pressure at 9 am and 3 pm. Since AWAP forcing does not include wind and air pressure we adopted the near-surface wind speed data from McVicar et al. (2008) and assume a fixed air pressure of 1000 hPa. Due to missing observations before 1990, the solar radiation input for 1970–1989 was built from the 1990–1999 daily climatology. Similarly, wind speeds for 1970–1974 are built from the 30-year climatology from 1975 to 2004. We translated the daily data into 3-hourly resolution using a weather generator (Haverd et al., 2013).

The land surface properties for our simulations are prescribed based on observational datasets. Land cover type is derived from the National Dynamic Land Cover Data of Australia (DLCD) (https://www.ga.gov.au/scientific-topics/earth-obs/accessing-satellite-imagery/landcover). We classify DLCD’s land cover types to five CABLE PFTs: crop (irrigated/rainfed crop, pasture and sugar DLCD classes), broadleaf evergreen forest (closed/open/scattered/sparse tree), shrub (closed/open/scattered/sparse shrubs and open/scattered/sparse chenopod shrubs), grassland (open/sparse herbaceous) and barren land (bare areas). We then...
resample the DLCD dataset from the 250 m resolution to the 0.05° resolution. The leaf area index (LAI) in CABLE is prescribed using a monthly climatology derived from the Copernicus Global Land Service product (https://land.copernicus.eu/global/products/lai). The climatology was constructed by first creating a monthly time series by taking the maximum of the 10-daily timesteps each month and then calculating a climatology from the monthly data over the period 1999–2017. The LAI data was resampled from the original 1 km resolution to the 0.05° resolution following De Kauwe et al. (2020). Soil parameters are derived from the soil texture information (sand/clay/silt fraction) from SoilGrids (Hengl et al., 2017) via the pedotransfer functions in Cosby et al. (1984) and resampled from 250 m to 0.05° resolution.

To evaluate the model simulations, we use monthly total water storage anomaly (TWSA) at 0.5° spatial resolution from the Gravity Recovery and Climate Experiment (GRACE) and GRACE Follow On products (Landerer et al., 2020; Watkins et al., 2015; Wiese et al., 2016, 2018). The RLM06M release is used for February 2002 — June 2017 and for June 2018 – December 2019. We also use the total land evaporation–evapotranspiration from the 2000–2018 monthly Derived Optimal Linear Combination Evapotranspiration (DOLCE version 2, Hobeichi et al., 2021) at 0.25° resolution, as well as the 2000–2019 daily Global Land Evaporation Amsterdam Model (GLEAM version 3.5, https://www.gleam.eu/; Martens et al., 2017; Miralles et al., 2011) at 0.5° spatial resolution. For daytime land surface temperature (LST) we use the Moderate Resolution Imaging Spectroradiometer (MODIS) datasets from Terra and Aqua satellites (products MOD11A1 and MYD11A1, Wan and Li, 1997; Wan 2015a, 2015b) at 1 km spatial resolution. We only consider pixels and time steps identified as good quality (QC flags 0). Only the day-time LST values are used due to the lack of good quality night-time LST data. The Terra overpass occurs at 10 am and Aqua at 2 pm local time. To analyse the compound events in January 2019, we linearly interpolate the 3-hourly model outputs to 2 pm to match the overpass time of the Aqua LST. The GRACE, GLEAM and MODIS datasets were resampled to the AWAP resolution using bilinear interpolation.

To evaluate model performance during heatwaves, we identify heatwave events using the excess heat factor index (EHF, Nairn and Fawcett, 2014). EHF is calculated using the daily AWAP maximum temperature, as the product of the difference of the previous 3 day mean to the 90th percentile of the 1970–1999 climatology and the difference of the previous 3 day mean to the preceding 30 day mean. A heatwave occurs when the EHF index is greater than 0 for at least three consecutive days. We only focus on summer heatwaves occurring between December and February of the following year.

3 Results

3.1 Simulations for the Millennium Drought and the recent drought

Previous studies have shown that simulations by LSMS diverge as the soil dries (Ukkola et al., 2016a), associated with systematic biases in evaporative fluxes and soil moisture states in the models (Mu et al., 2021; Swenson and Lawrence, 2014; Trugman et al., 2018). We therefore first evaluate how well CABLE-GW captures the evolution of terrestrial water variability during two recent major droughts.

Figure 1a shows the total water storage anomaly during 2000–2019 observed by GRACE and simulated in GW and FD. Both GW and FD accurately capture the interannual variability in total water storage for S.E. Australia (r = 0.96 in GW, and 0.90 in FD). Both model configurations simulate a decline in TWSA through the first drought period (up to 2009, see Figure S1), the rapid increase in TWSA from 2010 associated with higher rainfall, a decline from around 2012 due to the re-emergence of drought conditions, and the rapid decline during the recent drought after conditions had eased in 2016 (Figure S1). FD underestimates the magnitude of monthly TWSA variance (standard deviation, SD = 37.18 mm) compared to GRACE (47.74
mm) or GW (47.67 mm). This underestimation, particularly during the wetter periods (2000, 2011-2016) and the first ~2 years of the droughts (2001-2, 2017-8) (Figure 1a). This underestimation in FD compared to GW is linked with the lack of aquifer water storage in the FD simulations which provides a reservoir of water that changes slowly and has a memory of previous wet/dry climate conditions (Figure 1a).

Figure 1b shows the accumulated precipitation (P) minus evapotranspiration (E) over the two drought periods. GW increases the evapotranspiration relative to FD such that the accumulated P–E decreases from about 786 mm to 455 mm during the Millennium drought, which is much closer to within the range of DOLCE (460 mm) and GLEAM estimates (97 mm). A similar result, although over a much shorter period, is also apparent for the recent drought (Figure 1b). The lower P–E in GW suggests that the presence of groundwater storage can alleviate the vegetation water stress during droughts, and reduces the reliance of E on P, indicated by a small reduction in the correlation (r) between E and P from 0.28 in FD to 0.24 in GW for MD, and a reduction from 0.42 to 0.37 for RD (Figure 1b).

Although the evapotranspiration products display some differences, the GW simulations are also closer overall to both the DOLCE and the GLEAM, observational-constrained estimates which suggests, The better match of GW than FD to the two evapotranspiration products implies that adding groundwater improves the simulations during droughts. The difference in E is also demonstrated spatially in Figure 2a-c. During the Millennium drought, the GW simulations show a clear improvement over FD in two aspects. GW shows smaller biases in E along the coast where FD underestimates E strongly (Figure 2b). The areas where E is underestimated are also smaller in extent in GW, suggesting that GW overall reduces the dry bias. The magnitude of the bias in GW reaches around 300 mm over small areas of S.E. Australia while in the FD simulations biases are larger, reaching 400 mm over a larger area. Plant photosynthesis assimilation rates are associated with transpiration via stomata conductance, Figure 4 presents the spatial maps of gross primary productivity (GPP) during the two droughts. GW simulations increase carbon uptake by 50–300 g C yr⁻¹, particularly along the coasts (Figure 4c). However, since CABLE uses a prescribed LAI and does not simulate any feedback between water availability and plant growth (e.g., defoliation) and its impact on GPP, we only focus on how GW influences evapotranspiration and the surface energy balance in the subsequent sections.

Overall, Figure 1 and Figure 2 show that representing groundwater in the model improves the simulation of the inter-annual variability in the terrestrial water cycle and storage, particularly during droughts.

### 3.2 The role of groundwater in sustaining evapotranspiration during droughts

We next explore the mechanisms by which including groundwater modifies the simulation of evapotranspiration. Figure 2 displays the overall influence of groundwater on water fluxes during the recent drought. GW simulates 50–200 mm yr⁻¹ more E over coastal regions where there is high tree cover (Figure 2a; see Figure 2S2a for land cover). Adding groundwater also increases E in most other regions, although the impact is negligible in many inland and non-forested regions (i.e., west of 145°E). We identified a clear connection between E (Figure 2a) and the simulated WTD in the GW simulations (Figure 2S5). GW simulates 110 mm yr⁻¹ more E when the WTD is shallower than 5 m deep, 22 mm yr⁻¹ when the WTD is 5–10 m deep, but only 3 mm yr⁻¹ more when the WTD is below 10 m. Higher transpiration (ΔEt) in GW explains 78% of the total evaporative evapotranspiration difference between GW and FD where WTD is shallower than 5 m (Figure 2b). This is confirmed by the change in the soil evaporation (ΔEs) (Figure 2c) where adding groundwater increases Es by negligible amounts over most of S.E. Australia, but by up to 25 mm y⁻¹ in regions underlain by shallow groundwater (Figure 2S5), which is consistent with field observations that indicate that Es can be substantial under conditions of a very shallow water table (Thorburn et al., 1992). In the very shallow WTD areas, the excess Es in GW results from the capillary rise of moisture from the shallow...
A significant factor in explaining how groundwater influences E is through changes in vertical drainage and recharge from the aquifer to the soil column. Figure 2d shows that the vertical drainage (Dr) both increases and decreases depending on the location. The addition of groundwater reduces vertical drainage by 74 mm yr\(^{-1}\) where WTD is shallower than 5 m. In some regions, the drainage increases with the inclusion of groundwater by up to 100 mm yr\(^{-1}\), especially in the areas where WTD is ~5 m. This is associated with the WTD being slightly below the bottom of the soil column (4.6 m). When the groundwater aquifer is nearly full and in GW, the wetter soil in the bottom soil layer is relatively wet, the calculated leads to a much higher hydraulic conductivity (\(K_{\text{sat}}\)) in GW is much larger than in FD where the bottom soil layer is drier due to a lack of groundwater contribution. This leads to higher vertical drainage in GW and a positive ΔDr. Inland, where the WTD tends to be much deeper there is no significant difference in Dr between GW and FD.

Figure 2e shows the difference in recharge into the upper soil column (ΔQrec) between GW and FD. The recharge from the aquifer into the bottom soil layer provides 17 mm yr\(^{-1}\) extra moisture in the GW simulations in regions where has a WTD between 5–10 m and 10 mm yr\(^{-1}\) where the WTD is deeper than 10 m, helping to explain the changes in E and Et in areas with deep WTD. However, there is no significant ΔQrec in regions with a shallow WTD (~5 mm yr\(^{-1}\)), suggesting the influence of groundwater is mainly via reduced drainage in these locations. Recharge from the aquifer to the soil column can only occur when WTD is below the soil column (bottom boundary at 4.6 m depth). If WTD is shallow and within the soil column, the interface is saturated and no recharge from the aquifer to the soil column can occur and water only moves downwards by gravity.

The combined impact of reduced drainage in GW (Figure 2d) and recharge from the aquifer into the root-zone (Figure 2e) is an increased water potential gradient between the drier top soil layers and the wetter deep soil layers, encouraging overnight capillary rise. Taking the hot and dry January 2019 as an example, when the compound events occurred, Figure 2f shows the maximum water stress factor difference (Δβ) overnight (between 9 pm and 3 am, i.e. predawn when soil is relatively moist following capillary lift overnight). We only consider rainless nights to exclude the impact of drainage induced by precipitation. The water stress factor (β) is based on the root distribution and moisture availability in each soil layer and represents the soil water stress on transpiration as water becomes limiting. Figure 2f implies that while the redistribution of moisture is small overall, in some locations it can reduce moisture stress by up to 4–6%.

### 3.3 The impact of groundwater during heatwaves

We next explore whether the higher available moisture due to the inclusion of groundwater enables the canopy to cool itself via evaporation during heatwaves by examining the temperature difference between the simulated canopy temperature (\(T_{\text{canopy}}\), °C) and the forced air temperature (\(T_{\text{air}}\), °C). We focus on the forested regions (Figure S3S2a) as the role of groundwater in enhancing plant water availability was shown to be largest in these regions (Figure 2).

Figure 3a shows the average \(T_{\text{canopy}} - T_{\text{air}}\) (ΔT, °C) over the forested regions for summer heatwaves from the GW and FD simulations, with the grey line indicating the median ΔT difference. During heatwaves, the inclusion of groundwater moistens the soil and supports higher transpiration, cooling the canopy and reducing ΔT relative to FD by up to 0.76°C (e.g. January summer heatwaves in 2013). As the drought lengthens in time (Figure 1a), the depletion of moisture gradually reduces this effect, from an average reduction of 0.52°C of the first 3 years to 0.16°C of the last 3 years in Millennium Drought (Figure 3a).

The impact of groundwater is clear in the evaporative fraction (Figure 3b) where in periods of higher rainfall (e.g. 2010–2011; Figure S1), and at the beginning of a drought (2001, 2017), the EF is higher (0.03 to 0.18). This implies more of the available energy is exchanged with the atmosphere in the form of latent, rather than sensible heat. However, the strength of the cooling...
effect decreases as the droughts extend. and the transpiration difference ($\Delta E_t$, mm d$^{-1}$) diminishes quickly (Figure 3c) because the vegetation becomes increasingly water-stressed (Figure 3d) which consequently limits transpiration (Figure 3e). For all variables ($AT$, $EF$, $Et$ and $\beta$), the difference between GW and FD is greatest during the wetter periods (e.g., 2013) and in the first 1–2 years of the multi-year drought (2001–2002 for the Millennium Drought or 2017–2018 for the recent drought). After the drought becomes well established, the FD and GW simulations converge as depleting soil moisture reservoirs reduce the impact of groundwater on canopy cooling and evaporative fluxes.

Figure 4a shows the spatial map of $\Delta T$ simulated in GW during heatwaves in the 2017–2019 drought. It indicates both land cover type (Figure S4S2a) and WTD (Figure S4S5) contribute to the $\Delta T$ pattern. The evaporative cooling via transpiration is stronger over the forested areas compared to crop-rorpland or grassland, and stronger in the regions with a wetter soil associated with a shallower WTD. However, $EF$ is mainly determined by WTD (compare Figure 4b and Figure S4S5). Inland, where the WTD is deeper and the soil is drier, most of the net radiation absorbed by the land surface is partitioned into sensible rather than latent heat (Figure 4b). However, in the coastal regions with a shallow WTD, the wetter soil reduces the water stress (Figure 4c), enables a higher $EF$ (Figure 4b), and alleviates heat stress on the leaves (Figure 4a). Along the coast where WTD is shallow, GW simulates a cooler canopy temperature due to the higher evaporative cooling (Figure 4e) which is the consequence of a lower soil water stress (Figure 4f) linked to the influence of groundwater (Figure S4S5).

Figure 5 shows the density scatter plot of $\Delta T$ versus WTD in S.E. Australia forested areas during heatwaves in 2000–2019. A shallow WTD moderates the temperature difference between the canopy and the ambient air during heatwaves leading to a smaller temperature difference. Meanwhile, as the WTD increases, due to the limited rooting depth in the model, the ability of the groundwater to support transpiration and offset the impact of high air temperatures is reduced. Figure 5 shows a large amount of variations, but nonetheless implies a threshold of $\sim$6 m whereafter there is a decoupling and little influence from groundwater during heatwaves. However, the absolute value of the threshold is likely CABLE-specific and associated with the assumption of a 4.6 m soil depth, which also sets the maximum rooting depth (roots can only extend to the bottom of the soil and cannot directly access the groundwater aquifer in CABLE). The CABLE soil depth comes from observational evidence of most roots being situated within the top 4.6 m (Canadell et al. 1996). Since the model assumes no roots exist in the groundwater aquifer, when the water table is below this depth, the water fluxes become largely uncoupled between the soil column and the groundwater aquifer, leading to a negligible impact of GW below $\sim$6 m depth.

### 3.4 The impact of groundwater during the drought and heatwave compound events

To examine the influence of groundwater on heatwaves occurring simultaneously with drought, we focus on a case study of the record-breaking heatwaves in January 2019, which is the hottest month on record for the study region (Bureau of Meteorology, 2019). The unprecedented prolonged heatwave period started in early December 2018 and continued through January 2019 with three peaks. We select two days (15th and 25th January 2019), when heatwaves spread across the study region, from the second and third heatwave phases (Figure S5S6).

We evaluate CABLE $T_{canopy}$ against MODIS LST observations, concentrating on forested areas where MODIS LST should more closely reflect vegetation canopy temperatures, but note that this comparison is not direct as the satellite estimate will contain contributions from the understorey and soil. Figures 6a-b show the good quality MODIS LST minus $T_{air}$ at 2 pm ($\Delta T_{MOD, 2pm}$) over forested regions on the 15th and 25th January 2019, and Figures 6c-d display the matching GW-simulated $\Delta T$ at 2 pm ($\Delta T_{GW, 2pm}$). Overall, $\Delta T_{GW, 2pm}$ increases from the coast to the interior in both heatwaves, consistent with the $\Delta T_{MOD, 2pm}$ pattern in both heatwaves, albeit that $\Delta T_{GW, 2pm}$ appears to be biased high relative to $\Delta T_{MOD, 2pm}$ along the coastal forests (Figure S6aS7a-b).
Figure 6e-f shows the $\Delta T_{2pm}$ difference between GW and FD. Access to groundwater can reduce canopy temperature by up to 5°C, in particular where the WTD is shallow. While reductions of 5°C are clearly limited in spatial extent, the overall pattern of cooling associated with groundwater access is quite widespread, and coincident with the groundwater-induced Et increase (Figure S8a-b), implying a reduction in heat stress experienced by the woody vegetation along coastal regions with a shallow WTD during heatwaves. Generally, GW matches MODIS LST better than FD despite the bias in both simulations (compare Figure S6a-S7a-b and Figure S6a-S7c-d). Nevertheless, the temperature reduction between GW and FD is still modest (< 1°C) for most of the forested regions. This may be related to the shallow root distribution assumed in most LSMs, which prevents roots from directly accessing the moisture stored in the deeper soil (note, CABLE assumes 92% of forest’s roots are in the top 0.64 m, Figure S2b). To examine this possibility, we performed the deep root (DR) sensitivity experiment which prescribed more roots in the deeper soil for forests (56% below 0.64 m depth). Figure 6e-f illustrates the difference between $\Delta T_{2pm}$ in DR and $\Delta T_{OWZpm}$. By enabling access to moisture in the deeper soil, the LSM simulates further cooling by 0.25–5°C across the forests, associated with an Et increase of 25–250 W m$^{-2}$ (Figure S8c-d). The prescribed deeper roots also lead to an overall better simulation of $\Delta T$ at 2 pm relative to the MODIS LST (compare Figure S6g-h with S7e-f vs Figure S6a-S7a-b).

Figure 6g-h shows the diurnal cycles of $\Delta T$ for the two selected regions (red boxes in Figure 6e-f) compared with the MODIS LST estimates. The region highlighted for the 15th January (Figure 6g7a) has a WTD of 4–7 m, while the region highlighted for the 25th January (Figure 6h7b) has a WTD < 4 m (Figure S4S5). In both regions, the simulated $\Delta T$ is highest in FD, lower in GW and lowest in DR. Where the WTD is 4–7m (Figure 6g7a), the three simulated $\Delta T$ are slightly lower than $\Delta T$ calculated by MODIS LST (red squares). However, in the shallower WTD region (Figure 6h7b), the simulated $\Delta T$ between experiments is more dispersed across experiments and exceeds the MODIS $\Delta T$ at both time points, implying that neglecting groundwater dynamics and deep roots is more likely to cause an overestimation of heat stress in the shallower WTD region. The shallower WTD region (Figure 7b) tends to have a high LAI coverage, implying that the MODIS LST represents a good approximation of the canopy temperature over this region. Consequently, the lower MODIS $\Delta T$ implies that CABLE is likely underestimating transpiration, leading to an overestimation of $\Delta T$ in all three simulations.

### 3.5 Constraints on groundwater mediation during the compound events

We finally probe the reasons for the apparent contradiction between the large impact of groundwater on E during drought (Figure 2a) but a smaller impact on $\Delta T$ during the compound events (Figure 6e-f). Figure 7g shows three factors ($\beta$, vapour pressure deficit (D) and $T_{air}$) that constrain the impact of groundwater on $\Delta T$ in CABLE during heatwaves in January 2019. Figure 7g8a shows the difference in $\Delta T$ between GW and FD as a function of $\Delta \beta$, suggesting that the inclusion of groundwater has a large impact on $\Delta T$ when there is a coincidental and large difference in $\beta$ between the GW and FD simulations. Figure 7g8b indicates a clear threshold at D = 3 kPa where GW and FD converge, while Figure 7g8c shows a convergence threshold when the $T_{air}$ exceeds 32°C. Above these two thresholds, access to groundwater seemingly becomes less important in mitigating plant heat stress. There are two mechanisms in CABLE that explain this behaviour. First, as D increases, CABLE predicts that stomata begin to close following a square root dependence (De Kauwe et al., 2015; Medlyn et al., 2011). Second, as $T_{air}$ increases, photosynthesis becomes inhibited as the temperature exceeds the optimum for photosynthesis. In both instances, evaporative cooling is reduced, regardless of the root zone moisture state dictated by groundwater access. That is to say, access to groundwater has limited capacity to directly mediate the heat stress on plants during a compound event when the air is very dry, or very hot.
4 Discussion

In the absence of direct measurements, we used the CABLE-GW LSM, constrained by satellite observations to investigate how groundwater influences ecosystems under conditions of co-occurring droughts and heatwaves. We found that representing the influence of groundwater was most important during the onset of drier wetter periods and the first ~ two years of a multi-year drought (~2001–2002 and 2017–2018; Figure 1 and 3). This primarily occurred via impeding gravity-driven drainage (Figure 2d) but also via capillary rise from the groundwater aquifer (Figure 2e). This moistening enabled the vegetation to sustain higher E for at least a year (Figure S2–S9). As the droughts progressed into multi-year events, the impact of groundwater diminished due to a depletion of soil moisture stores regardless of whether groundwater dynamics were simulated.

When a heatwave occurs during a drought, and in particular early in a drought, the extra transpiration enabled by representing groundwater dynamics helps reduce the heat stress on vegetation (e.g. the reduction of 0.64°C of ΔT over the forests in 2002, Figure 3a). This effect is particularly pronounced in regions with a shallower WTD (e.g. where the groundwater was within the first 5±5 m, there was a 1°C–0.5°C mean reduction in ΔT in the recent drought, Figure 4d). Importantly, the role played by groundwater diminishes as the drought lengthens beyond two years. (Figure 3). Additionally, either the lack of deep roots or stomatal closure caused by high D/Tair can reduce the additional transpiration induced by groundwater. The latter plant physiology feedback dominates during heatwaves co-occurring with drought, even if the groundwater’s influence has increased root-zone water availability.

Our results highlight the impact of groundwater on both land surface states (e.g. soil moisture) and on surface fluxes and how this impact varies with the length and intensity of droughts and heatwaves. The results imply that the dominant mechanism by which groundwater buffered transpiration was through impeding gravity-driven drainage. We found a limited role for upward water movement from aquifer due to simulated shallow WTD (which was broadly consistent with the observations in Fan et al., 2013). Further work will be necessary to understand how groundwater interacts with droughts and heatwaves and what these interactions mean for terrestrial ecosystems and the occurrence of the compound extreme events, particularly under the projection of intensifying droughts (Ukkola et al., 2020) and heatwaves (Cowan et al., 2014).

4.1 Changes in the role of groundwater in multi-year droughts

Groundwater is the slowest part of the terrestrial water cycle to change (Condon et al., 2020) and can have a memory of multi-year variations in rainfall (Martínez-de la Torre and Miguez-Macho, 2019; Martinez et al., 2016a). Our results show that the effect of groundwater on the partitioning of available energy between latent and sensible heat fluxes is influenced by the length of drought. As the drought extends in time, the extra E sustained by groundwater decreases (e.g. during the Millennium drought, Figure S2-S9). The role of a drying landscape in modifying the partitioning of available energy between latent and sensible heat fluxes is well known and has been extensively studied (Fan, 2015; Miralles et al., 2019; Seneviratne et al., 2010). Our results add to the knowledge by quantifying the extent of the groundwater control, and eliciting the timescales of influence and the mechanisms at play. The importance of vegetation-groundwater interactions on multi-year timescales has been identified previously. Humphrey et al. (2018) hypothesised that climate models may underestimate the amplitude of global net ecosystem exchange because of a lack of deep-water access. Our regional based results support this hypothesis and in particular highlight the importance of groundwater for explaining the amplitude of fluxes in wet regions (Figure 1), periods, as well as sustaining evapotranspiration during drought. (Figure 1).

4.2 Implications for land-atmosphere feedbacks during compound events

Our results show that during drought-heatwave compound events, the existence of groundwater eases the heat stress on the forest canopy and reduces the sensible heat flux to atmosphere. This has the potential to reduce heat accumulating in the
boundary layer and help ameliorate the intensity of a heatwave (Keune et al., 2016; Zipper et al., 2019). The presence of groundwater helps dampen a positive feedback loop whereby during drought-heatwave compound events, the high exchange of sensible and low exchange of latent heat can heat the atmosphere and increase the atmospheric demand for water (De Boeck et al., 2010; Massmann et al., 2019), intensifying drying (Miralles et al., 2014). The lack of groundwater in many LSMs suggests a lack of this moderating process and consequently a risk of overestimating the positive feedback on the boundary layer in coupled climate simulations. Our results show that neglecting groundwater leads to an average overestimate in canopy temperature by 0.2–1°C where the WTD is shallow (Figure 4d), but as much as 5°C in single heatwave events (Figure 6e-f), leading to an increase in the sensible heat flux (Figure 4e).

The capacity of groundwater to moderate this positive land-atmosphere feedback is via modifying soil water availability. Firstly, soil water availability influenced by WTD affects how much water is available for E. In the shallow WTD regions, the higher soil water is likely to suppress the mutual enhancement of droughts and heatwaves (Keune et al., 2016; Zipper et al., 2019), particularly early in a drought. However, this suppression becomes weaker as the WTD deepens, in particular at depths beneath the root zone (e.g. 4.6 m in CABLE-GW) or as a drought lengthens. Our results imply the land-amplification of heatwaves is likely stronger in the inland regions (Hirsch et al., 2019) where the WTD is lower than 5m5 m and the influence of groundwater diminishes (Figure S455), and once a drought has intensified significantly.

On a dry and hot heatwave afternoon, plant physiology feedbacks to high D and high T air dominate transpiration and reduce the influence of groundwater in moderating heatwaves. In CABLE, stomatal closure occurs either directly due to high D (>3 kPa) (De Kauwe et al. 2015) or indirectly due to biochemical feedbacks on photosynthesis as high T air (>32°C) (Kowalczyk et al., 2006); both processes lead to transpiration to near zero, eliminating the buffering effect of groundwater on canopy temperatures. While the timing of the onset of these physiology feedbacks varies across LSMs due to different parameterised sensitivities of stomatal conductance to atmospheric demand (Ball et al., 1987; Leuning et al., 1995) and different temperature dependence parameterisations (Badger and Collatz, 1977; Bernacchi et al., 2001; Crous et al., 2013), importantly, stomatal closure during heat extremes would be model invariant.

4.3 Uncertainties and future directions

Our study uses a single LSM and consequently the parameterisations included in CABLE-GW influence the quantification of the role of groundwater on droughts and heatwaves. We note CABLE-GW has been extensively evaluated for water cycle processes (Decker, 2015; Decker et al., 2017; Mu et al., 2021; Ukkola et al., 2016b), but evaluation for groundwater interactions remains limited due to the lack of suitable observations (e.g. regional WTD monitoring or detailed knowledge of the distribution of root depths). Figure 1 gives us confidence that CABLE-GW is performing well, based on GRACE data, and other the evaluation of CABLE highlights against the GRACE, DOLCE and GLEAM products, as well as previous work that showed the capacity of CABLE-GW to simulate E well (Decker, 2015; Decker et al., 2017). However, we also note here that key model parameterisations that may influence the role of groundwater are particularly uncertain.

We need to be cautious about the “small” groundwater impact on the canopy temperature and associated turbulent energy fluxes during high D or high T air (Figure 3, 4, 6 and 7). The thresholds of D and T air currently assumed by LSMs are in fact likely to be species specific. Australian trees in particular have involved a series of physiological adaptations to reduce the negative impact of heat extremes. It is important to note that most LSMs parameterise their stomatal response to VPD for moderate ranges (< 2 kPa), which leads to significant biases at high D (Yang et al., 2019), a feature common in Australia and during heatwaves in general. New theory is needed to ensure that models adequately capture the full range of stomatal response to variability in D (low and high ranges). Similarly, while there is strong evidence to suggest that the optimum temperature for
photosynthesis does not vary predictably with the climate of species origin (Kumarathunge et al., 2019) (implying model parameterisations do not need to vary with species), findings from studies do vary (Cunningham and Reed, 2002; Reich et al., 2015). Moreover, evidence that plants acclimate their photosynthetic temperature response is strong (Kattge and Knorr, 2007; Kumarathunge et al., 2019; Mercado et al., 2018; Smith et al., 2016; Smith and Dukes, 2013). As a result, it is likely that LSMs currently underestimate groundwater influence during heatwaves due to the interaction with plant physiology feedbacks. This is a key area requiring further investigation. For example, Drake et al. (2018) demonstrated that during a 4-day heatwave >43°C, Australian Eucalyptus parramattensis trees did not reduce transpiration to zero as models would commonly predict, allowing the trees to persist unharmed in a whole-tree chamber experiment. Although De Kauwe et al. (2019) did not find strong support for this phenomenon across eddy covariance sites, if this physiological response is common across Australian woodlands, it would change our view on the importance of soil water availability (therefore groundwater) on the evolution of heatwave or even compound events. Coupled model sensitivity experiments may be important to determine the magnitude that such a physiological feedback would present and could guide the direction of future field/manipulation experiments.

Root distribution and root function and thereby how roots utilise groundwater are uncertain in models (Arora and Boer, 2003; Drewniak, 2019; Wang et al., 2018; Warren et al., 2015) and indeed in observations (Fan et al., 2017; Jackson et al., 1996; Schenk and Jackson, 2002). Models often ignore how roots forage for water and respond to moisture heterogeneity, limiting the model’s ability to accurately reflect the plant usage of groundwater (Warren et al., 2015). In LSMs, roots are typically parameterised using a fixed distribution and normally ignore water uptake from deep roots. This assumption neglects any climatological impact of root distribution and the differentiation in root morphology and function (fine roots vs tap roots), leading to a potential underestimation of groundwater utilization in LSMs (see our deep root experiment, Figure 6g-h). This assumption may be particularly problematic in Australia where vegetation has developed significant adaptation strategies to cope with both extreme heat and drought, including deeply rooted vegetation that can access groundwater (Bartle et al., 1980; Dawson and Pate, 1996; Eamus et al., 2015; Eberbach and Burrows, 2006; Fan et al., 2017). We also note that CABLE does not directly consider hydraulic redistribution, defined as the passive water movement via plant roots from moister to drier soil layers (Burgess et al., 1998; Richards and Caldwell, 1987). Neglecting hydraulic redistribution has the potential to underestimate the groundwater transported upwards and underestimate the importance of groundwater on ecosystems.

On the atmosphere side, the existence of groundwater increases the water flux from the land to atmosphere, particularly in regions of shallow WTD, during the first 1–2 years of a drought. This has the potential to moisten the lower atmosphere and may encourage precipitation (Anyah et al., 2008; Jiang et al., 2009; Martinez et al., 2016b; Maxwell et al., 2011). However, our experiments are uncoupled from the atmosphere so while there is the potential for the higher E to affect the boundary layer moisture (Bonetti et al., 2015; Gilbert et al., 2017; Maxwell et al., 2007), clouds and precipitation, we cannot conclude that it would until we undertake future coupled simulations.

Finally, we note we have focused on the role of groundwater in a natural environment. Humans extract large quantities of groundwater in many regions (Döll et al., 2014; Wada, 2016). Adding human management of groundwater into LSMs enables an examination of how this affects the vulnerability of ecosystems to heatwaves and drought, and may ultimately identify those vulnerable ecosystems close to tipping points that are priorities for protection.

5 Conclusions

In conclusion, we used the CABLE LSM, constrained by satellite observations, to explore the timescales and extent to which groundwater influences vegetation function and turbulent energy fluxes during multi-year droughts. We showed that
groundwater moistened the soil during the first ~two years of a multi-year drought which enabled the vegetation to sustain higher evaporation (50–200 mm yr\(^{-1}\) over the coastal forest regions) during drought onset. This cooled the forest canopy on average by 0.03–0.76\(^\circ\)C in heatwaves during 2001–2019 and as much as 5–6\(^\circ\)C in regions of shallow water table depths in the heatwave in January 2019, helping to moderate the heat stress on vegetation during heatwaves. However, the ability of groundwater to buffer evapotranspiration varied with the length and intensity of droughts and heatwaves, with its influence decreasing with prolonged drought conditions. Importantly, we also demonstrated that the capacity of the groundwater to buffer evaporative fluxes during heatwaves is dependent on constrained by plant physiology feedbacks which regulate stomatal control, irrespective of soil water status. Given increased risk of regional heatwaves and droughts in the future, the role of groundwater on land-atmosphere feedbacks and on terrestrial ecosystems needs to be better understood in order to constrain future projections.

**Code and data availability**

The CABLE code is available at [https://trac.nci.org.au/trac/cable/wiki](https://trac.nci.org.au/trac/cable/wiki) (last access: 30 April 2021) (NCI, 2021) after registration. Here, we use CABLE revision r7765. Scripts for plotting and processing model outputs are available at [https://github.com/bibivking/Heatwave/tree/master/GW_DH](https://github.com/bibivking/Heatwave/tree/master/GW_DH). DOLCE version 2 dataset is available from the NCI data catalogue at [http://dx.doi.org/10.25914/5f1664837ef06](http://dx.doi.org/10.25914/5f1664837ef06) (Hobeichi, 2020). GRACE land is available at [http://grace.jpl.nasa.gov](http://grace.jpl.nasa.gov), supported by the NASA MEaSUREs Program. GLEAM dataset is available at [https://www.gleam.eu/](https://www.gleam.eu/). MODIS/Terra Land Surface Temperature and the Emissivity Daily L3 Global 1km and MYD11A1 MODIS/Aqua Land Surface Temperature and the Emissivity Daily L3 Global 1km datasets were acquired from the NASA Land Processed Distributed Active Archive Center (LP DAAC), located in the USGS Earth Resources Observation and Science (EROS) Center in Sioux Falls, South Dakota, USA ([https://lpdaacsvc.cr.usgs.gov/appeears/](https://lpdaacsvc.cr.usgs.gov/appeears/)).

**Author contributions**

MM, MGDK, AJP and AMU conceived the study, designed the model experiments, investigated the simulations and drafted the manuscript. SH and PRB provided the evaluation and the meteorology forcing datasets. All authors participated in the discussion and revision of the manuscript.

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Figure 1 (a) Total water storage anomaly (TWSA) during 2000–2019 and (b) accumulated P–E for the two droughts over S.E. Australia. In panel (a), observations from GRACE are shown in black, the GW simulation in green, FD in orange and the aquifer water storage anomaly in GW in blue. The shading in panel (a) highlights the two drought periods. The left top corner of panel (a) displays the correlation ($r$) between GRACE and GW/FD, as well as the standard deviation (SD, mm) of GRACE, GW and FD over the periods when GRACE and the simulations coincide. Panel (b) shows the accumulated P–E for two periods; the dark lines show the 2001–2009 Millennium drought (MD) and the light lines show the 2017-2019 recent drought (RD). The correlation ($r$) between the P and E is shown in the legend of panel (b).
Figure 2. The overall influence of groundwater during the recent drought. (a)-(e) are the difference (GW-FD) in total evapotranspiration (ΔE), transpiration (ΔEt), soil evaporation (ΔEs), vertical drainage (ΔDr) and recharge from the aquifer to soil column (ΔQrec), respectively. In the bottom right of panels (a)-(e), the average of each variable over selected water table depths (WTD) is provided. (f) is the maximum night-time water stress factor difference (Δβ) between 3 am (i.e. predawn when the soil is relatively moist following capillary lift overnight) and 9 pm the previous day. We only include rainless nights in January 2019 to calculate Δβ to remove any influence of overnight rainfall.
Figure 3. Groundwater-induced differences in (a) $T_{canopy} - T_{air}$ ($\Delta T$), (b) evaporative fraction (EF), (c) transpiration (Et), and (d) water stress factor ($\beta$) during 2000-2019 summer heatwaves over forested areas (the green region in Figure S2a). The left y-axis is the scale for boxes. The blue boxes refer to the GW experiment and the red boxes to FD. For each box, the middle line is the median, the upper border is the 75th percentile, and the lower border is 25th percentile. The right y-axis is the scale for the grey lines which display the difference in the medians (GW-FD). The shadings highlight the two drought periods.
Figure 4. Land response to heatwaves during the recent drought. Panels (a)-(c) are the mean $T_{\text{canopy}}-T_{\text{air}}$ ($\Delta T$), evaporative fraction (EF), and soil water stress factor ($\beta$) in GW, respectively, during 2017-2019 summer heatwaves. Panel (d)-(f) are the difference (GW-FD) of $T_{\text{canopy}}$, EF and $\beta$. In the bottom right of each plot, the average of each variable over selected water table depths (WTD) is provided. Note that the colour bar is switched between (d) and (e)-(f).

Figure 5. A density scatter plot of $T_{\text{canopy}}-T_{\text{air}}$ ($\Delta T$) versus water table depth (WTD) in GW simulations over forested areas in all heatwaves during 2000–2019. Every tree pixel on each heatwave day accounts for one record and the darker colours show higher recorded densities.
Figure 6. The simulation of two heatwaves on 15th (left column) and 25th January 2019 (right column). The first row shows the difference between MODIS land surface temperature (LST) and $T_{\text{air}}$ at 2 pm ($\Delta T_{\text{MOD, 2pm}}$) (only forested areas with good LST quality data are displayed). The second row is the GW simulation of $\Delta T$ at 2 pm ($\Delta T_{\text{GW, 2pm}}$). The third row is the difference of $\Delta T$ at 2 pm between GW and FD simulations ($\Delta T_{\text{GW, 2pm}} - \Delta T_{\text{FD, 2pm}}$). The last row is the diurnal same as the third row but for the difference between the DR and GW simulations ($\Delta T_{\text{DR, 2pm}} - \Delta T_{\text{GW, 2pm}}$). Note that the comparison between GW/FD/DR and MODIS LST is shown in Figure S7.
Figure 7. Diurnal cycle of \( \Delta T \) on 15\textsuperscript{th} (left column) and 25\textsuperscript{th} January 2019 (right column) over the selected regions shown by the red boxes in panels (e) and (f). In panel (e) and (f), the shadings show the uncertainty in every simulation defined as one standard deviation (SD) among the selected pixels. The red dots are MODIS LST minus \( T_{air} \) with the uncertainty shown by the red error bars. For both regions, only pixels available in MODIS are shown.

Figure 7. Density scatter plots showing the three factors that influence the difference in \( T_{canopy} - T_{air} \) between GW and FD (\( \Delta T \), expressed as GW-FD difference). (a) is \( \Delta T \) difference against the \( \beta \) difference (GW-FD) (\( \Delta \beta \)), (b) is \( \Delta T \) difference against vapour pressure deficit (D), and (c) is \( \Delta T \) difference against \( T_{air} \). Each point corresponds to a tree pixel on a heatwave day in January 2019. The darker colours illustrate where the records are more dense. The correlation (r) between the x- and y-axis is shown in the bottom left of each panel.