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Author/s:

Trouvé, R;Osborne, L;Baker, PJ

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1 The effect of species, size, and fire intensity on tree
2 mortality within a catastrophic bushfire complex

3 Raphaël Trouvé^{1*}, Lisa Osborne², and Patrick J. Baker³

4 ¹CEBRA & School of Ecosystem and Forest Sciences, The University of
5 Melbourne, Parkville, Victoria, 3010, Australia.

6 ²School of Biological Sciences, Monash University, Victoria 3800, Australia.

7 ³School of Ecosystem and Forest Sciences, The University of Melbourne,
8 Parkville, Victoria, 3010, Australia.

9 *Address correspondence to Raphaël Trouvé, The University of Melbourne,
10 Parkville, Victoria, 3010, Australia; raphael.trouve@unimelb.edu.au

11 Running head: Estimating fire-related tree mortality

1 Abstract

2 Infrequent, high-intensity disturbances can have profound impacts on forested landscapes, chang-
3 ing forest structure and altering relative species abundance. However, due to their rarity and the
4 logistical challenges of directly observing such extreme events, both the spatial variability of
5 disturbance intensity and the species-specific responses to this variability are poorly under-
6 stood. We used observed patterns of mortality across a fire severity gradient following the 2009
7 Black Saturday fires in southeastern Australia to simultaneously estimate 1) species- and size-
8 specific susceptibility to fire-induced mortality and 2) fire intensity. We found broad variation in
9 patterns of fire susceptibility among the 10 tree species (five eucalypts and five non-eucalypts)
10 sufficiently abundant for analysis. Among the eucalypts, *Eucalyptus obliqua* was the most resis-
11 tant to fire-induced mortality, with trees of ~ 25 cm DBH having a 50% probability of surviving
12 even the most intense fires. In contrast, *E. regnans* had 100% mortality across all size classes
13 when subjected to high intensity fire. Basal resprouting occurred in six of the study species and,
14 when accounted for, fundamentally changed the mortality profile of these species, highlighting
15 the importance of resprouting as an adaptation to fire in these landscapes. In particular, the two
16 iconic cool temperate rainforest species (*Nothofagus cunninghami* and *Atherosperma moscha-*
17 *tum*) were strong resprouters ($\sim 45\%$ of individuals were able to resprout after being top-killed
18 by fire). We also found evidence for compositional shifts in regeneration above threshold values
19 of fire intensity in cool temperate rainforest and mixed forest sites, both of which have important
20 conservation values within these landscapes. The observed patterns of species- and size-specific
21 susceptibility to fire-induced mortality may be used to anticipate changes in forest structure and
22 composition in the future. In addition, they may also help guide forest management strategies
23 that reduce the length of time individual trees are exposed to potentially lethal fires, thereby
24 increasing the resilience of these forests to future fires.

1 **Keywords:** *Atherosperma moschatum*, Bayesian hierarchical modelling, Black Saturday fire,
2 Disturbance ecology, *Eucalyptus*, Fire severity, Latent variables, Mountain ash, *Nothofagus*
3 *cunninghamii*, Size-related mortality, South-eastern Australia, Wildfire.

4 **Introduction**

5 Fire is the dominant natural disturbance across many regions of the world, impacting native
6 ecosystems from the tropics to the tundra (Bowman *et al.* 2009, McLauchlan *et al.* 2020). While
7 fires are important drivers of many processes in forested ecosystems, such as nutrient dynamics,
8 carbon fluxes, and biodiversity dynamics, they also generate great destructive forces that can
9 dramatically alter the structure, composition, and developmental trajectory of entire forested
10 landscapes (Moritz *et al.* 2014). However, fire is an inherently complex, partially stochastic
11 phenomenon that varies in its impacts across spatial and temporal scales (Turner *et al.* 1989,
12 Agee 1993, McLauchlan *et al.* 2020). In forested ecosystems the energy of a fire per unit
13 time—its *intensity*—is primarily influenced by environmental conditions and fuel availability.
14 The impact of a fire—its *severity*—reflects the interaction between 1) fire intensity and 2) the
15 susceptibility of the trees at a site to fire-induced mortality. Both species- and size-specific
16 differences in fire susceptibility interact to modulate fire severity. Tree species with adaptations
17 such as thick bark or vigorous resprouting are more likely to survive a fire than tree species
18 that lack such adaptations. Within a species, larger trees are more likely to survive fire than
19 smaller trees as they tend to have thicker bark, which delays and reduces heat penetration to the
20 cambium (Gill and Ashton 1968, Agee 1993, Baker and Bunyavejchewin 2006, Keyser *et al.*
21 2006, Ryan *et al.* 1988). As a consequence, two stands subjected to fires of the same intensity
22 may show distinctly different patterns of mortality if they differ in species composition, tree
23 size distribution, or both (Agee 1993). In complex landscapes in which topography, soils, and

1 vegetation structure and composition may vary widely, complex patterns of fire behaviour may
2 result in heterogeneous patterns of tree mortality across the landscape (Lertzman *et al.* 1998).

3 Fire intensity, particularly in extreme fire conditions, is difficult to measure. As a conse-
4 quence, fire severity—typically a measure of the proportion of individual trees killed or canopy
5 area lost—is more commonly measured in the wake of a fire (Turner and Dale 1998, Keeley
6 2009). While disturbance intensity and severity are related, the relationship between the two
7 is rarely a simple, linear one (Lentile *et al.* 2006, Keeley 2009). Disentangling the roles of
8 fire intensity and tree and stand susceptibility presents a challenge because the observed pat-
9 tern of fire severity is influenced by both. For example, an area of forest that had high levels
10 of mortality may have experienced a high-intensity fire or it may have been dominated by tree
11 species or size classes that were particularly susceptible to fire-induced mortality (even from
12 low-intensity fires). Canham *et al.* (2001) proposed a method for simultaneously estimating
13 disturbance intensity and species- and size-specific susceptibility to wind disturbance that takes
14 advantage of variability in wind damage and species-specific responses to wind damage. Their
15 approach is readily generalised to other types of disturbances with spatially variable impacts.
16 Here we adapt it to disentangle various underlying drivers of fire-induced mortality during the
17 2009 Black Saturday fires—one of the most extreme fire events in the history of southeastern
18 Australia prior to the 2019/20 fire season.

19 As the climate in many regions of the world becomes hotter and drier, extreme fire con-
20 ditions are expected to become more frequent, more extensive, and potentially more severe.
21 Unprecedented fire events, like those of the 2017, 2018, and 2020 Californian fires or the 2019-
22 20 Australian fire season, that burn hundreds of thousands to millions of hectares of forest may
23 become more common in the coming decades. If they do, it will be critically important to antici-
24 pate how they will impact and potentially alter forest communities. This requires understanding

1 how different tree species are impacted by fire and whether there are particular combinations of
2 stand structures or composition that are more resistant or resilient to future fires (Lavorel *et al.*
3 2014, Doherty *et al.* 2016, Tepley *et al.* 2017, Koontz *et al.* 2020, Trouvé *et al.* 2020). In this
4 study we addressed two specific questions to explore this issue across a range of species and
5 forest types:

- 6 1. How do species and size influence fire-induced mortality in a large, heterogeneous bush-
7 fire complex?
- 8 2. How does fire intensity vary across a landscape subjected to a catastrophic fire?

9 **Methods**

10 **Study area, fire and forest types**

11 On 7 February 2009, after nearly a decade of drought conditions (Palmer *et al.* 2015), the area
12 around Melbourne, Victoria, had experienced three consecutive days of extreme fire conditions
13 with temperatures $>43^{\circ}\text{C}$, relative humidity $<10\%$, and wind speeds of 65 km h^{-1} , gusting to
14 90 km h^{-1} . The Australian Fire Danger Rating (AFDR) system (Noble *et al.* 1980) classifies
15 days with Forest Fire Danger Index ratings >100 as Catastrophic fire conditions—on 7 February
16 2009 the FFDI was 103. Nearly 600 fires started across the state with a complex of four partic-
17 ularly large fires initiating within 100 km of Melbourne (Teague *et al.* 2010). The Murrindindi
18 fire, which burned through the Central Highlands east of Melbourne, spread rapidly during the
19 day moving nearly 40 km to the small township of Narbethong. In the late afternoon a cold
20 front brought strong southwesterly winds that turned the flanking fires into the main fire line,
21 and increased the size of the fire-affected area dramatically. However, the cooler temperatures
22 and increased relative humidity that accompanied the cold front meant that the fire conditions

1 eased over the course of the night.

2 This led to a complex mosaic of different fire impacts that reflected the interactions of
3 weather, topography, and vegetation on fire behaviour.

4 We conducted our study at six sites that burned in the Murrindindi fire: Acheron Way,
5 Marysville, and Mount Margaret in the Marysville State Forest; the Beeches Rainforest and
6 Cambarville in the Yarra Ranges National Park; and Cathedral Ranges State Park. Study sites
7 were chosen to collectively represent the full range of fire severity (*i.e.*, unburnt, understorey
8 burn, crown scorch, and crown fire) across the four major forest types in the area: dry sclerophyll forest, cool temperate rainforest, tall open forest, and cool temperate mixed forests.
9 Dry sclerophyll forest (DSF) typically occurs on drier, north-facing slopes at elevations <600
10 m.a.s.l. (Ashton and Attiwill 1994). This forest type is typically characterised by an overstorey of
11 *Eucalyptus* species (predominantly *E. obliqua* [messmate], *E. radiata* [narrow-leaf peppermint],
12 and *E. dives* [broad-leaf peppermint]), and a midstorey of *Acacia* species (predominantly *A. deal-*
13 *bata* [silver wattle] and *A. melanoxylon* [blackwood]). Cool temperate rainforest (CTR) is re-
14 stricted to cool, moist gullies and riparian areas typically on southeastern-facing slopes (Busby
15 and Brown 1994). It is characterised by a dense, closed canopy dominated by *Nothofagus cun-*
16 *ninghamii* (myrtle beech) and *Atherosperma moschatum* (southern sassafras) and typically has
17 a sparse and heavily shaded understorey (Busby and Brown 1994). Tall open forest (TOF) occurs
18 at elevations above 400 m in cooler, moister areas than dry sclerophyll forests (Ashton and Atti-
19 will 1994). These forests have a tall overstorey of *Eucalyptus* species (dominated by *E. regnans*
20 [mountain ash] below 900m elevation and *E. delegatensis* [alpine ash] above 900m elevation),
21 a mid-story of *Acacia* (*A. dealbata*, *A. melanoxylon*, and *A. obliquinervia* [mountain hickory
22 wattle]), and a dense understorey of small trees and shrubs (predominantly *Pomaderris aspera*
23 [hazel pomaderris], *Olearia argophylla* [musk daisy-bush], *Lomatia fraseri* [forest lomatia],
24

1 and *Bedfordia arborescens* [blanket-leaf]) (Ashton 2000). In the heavily dissected topography
2 of the Central Highlands, TOF on midslope sites often grades into CTR in the gullies (Ash-
3 ton and Attiwill 1994, Simkin and Baker 2008). The transition from TOF to CTR is readily
4 distinguished due to the presence of eucalypts in the overstorey and rainforest species in the un-
5 derstorey. These ecotones are known as cool temperate mixed forest (MF). Because they have
6 high conservation value within the Central Highlands landscape and are readily identifiable in
7 the field, we included MF as a distinct forest type in this study.

8 A map of the study area and plot location is shown in figure 1.

9 **Data collection**

10 To characterise the impacts of, and species-specific responses to, the 2009 bushfire, we estab-
11 lished 53 circular plots of 20 m radius (0.125 ha) between October 2012 and January 2013.
12 Plots were distributed every 300-600 m along transects in forested areas along minor roads or
13 walking tracks. Transects varied in length from 1.5 to 4 km depending on the length of avail-
14 able roads and tracks. Consequently, the number of plots per transect varied from 2 to 10. The
15 number and location of available sites were constrained by the many roads and forested areas
16 that 3.5 years after the fire were still either inaccessible or unsafe due to the extensive damage
17 caused by the bushfire. Topography and site conditions were highly variable among study loca-
18 tions, from cool moist gullies in Acheron Way and The Beeches Rainforest to steep dry slopes
19 in the Cathedral Ranges. In all, we established 18 plots in DSF, 9 plots in CTR, 16 plots in TOF,
20 and 10 plots in MF.

21 Within each plot we sampled every tree with a diameter at breast height (DBH) >10 cm.
22 Each tree was identified to species, measured for DBH, and tallied as dead or alive. Mortality
23 was determined for all trees rooted within the plot and for recently fallen trees showing evidence

1 of fire damage. We distinguished between trees that had been top-killed but were resprouting
2 from the base and those that were completely killed (*i.e.*, no evidence of basal resprouting;
3 Hoffmann and Solbrig (2003)). The former was recorded as top-kill; the latter as full-kill. Dead
4 trees with all bark removed and showing burnt wood uncharacteristic of the damage to the plot
5 were considered to have died before the 2009 fires and were not recorded. Table 1 shows a
6 breakdown of the experimental design by species. The diameter distributions per species per
7 plot as well as tree status (live and top-kill) in our experimental design are shown in Appendix
8 S1: Figs. S1 and S2. Importantly, there was sufficient overlap of species mixture between the
9 different forest type and plots (Appendix S1: Figs. S1 and S2) to provide some link between
10 the estimates of latent relative fire intensity of each plot in Eq. 1.

11 Within each study plot a central 5-m radius sub-plot (78.5 m²) was established to charac-
12 terise recruitment patterns. For seedlings of large shrubs and trees, each individual was iden-
13 tified to species and the height of the tallest shoot measured. This allowed us to assess how
14 species diversity and seedling density and size correlated with fire intensity, as well as the po-
15 tential filtering effect of fire intensity on post-fire regeneration.

16 **Data analyses**

17 We adapted the approach proposed by Canham *et al.* (2001, 2010) that uses spatial variability
18 in disturbance-induced mortality to simultaneously estimate plot-level disturbance intensity and
19 species- and size-specific susceptibility to disturbance-induced mortality. They used a logistic
20 regression model, in which the probability of mortality is a logistic function of tree species,
21 individual tree size, and plot-level disturbance intensity. We modified their approach by using
22 a hierarchical Bayesian framework to fit the model. The hierarchical approach allows for par-
23 tial pooling of information among species and better estimates of parameter values and their

1 associated uncertainties (Gelman and Hill 2008). We first focus on the top-kill model, which
 2 considers a tree to be have been top-killed in the fire if the aboveground woody stem was dead,
 3 irrespective of the tree's resprouting status. The top-kill model takes the form:

$$y_{ijk} = \text{Bernoulli}(p_{ijk}) \quad (1)$$

$$p_{ijk} = \frac{1}{1 + \exp(-\text{logit}_{ijk})}$$

$$\text{logit}_{ijk} = \log\left(\frac{p_{ijk}}{1 - p_{ijk}}\right) = \alpha_j + \beta_j \times I_i \times D_{ijk}^{\gamma_j}, \text{ where}$$

$$\alpha_j \sim \alpha_0 + N(0, \sigma_\alpha)$$

$$\beta_j \sim \beta_0 + N(0, \sigma_\beta)$$

$$\gamma_j \sim \gamma_0 + N(0, \sigma_\gamma)$$

4 The left-hand term y_{ijk} is the observed mortality status (0 = alive, 1 = top-kill) of the k th indi-
 5 vidual of species j in plot i , which is assumed to be sampled from a Bernoulli distribution with
 6 probability of top-kill p_{ijk} . The probability of top-kill per individual p_{ijk} is modelled on the logit
 7 scale and depends on latent (*i.e.*, unobserved) relative fire intensity, tree diameter, and species.
 8 On the right-hand side of the equation, I_i is the relative fire intensity for plot i (on a scale of 0-1,
 9 with 0 being undisturbed and 1 being the most intensely disturbed plot measured), modelled
 10 as a latent (*i.e.*, unobserved) random variable with a prior uniformly distributed between 0 and
 11 1, D_{ijk} is the DBH (in cm) of the k th individual of species j in plot i , and α_j , β_j , and γ_j are
 12 species-specific parameters to be estimated. α can be interpreted as the baseline top-kill rate
 13 (*i.e.*, when relative fire intensity equals zero), β represents the effect of fire intensity and the
 14 effect of DBH, and γ affects the shape of the top-kill-DBH relationship.

15 To explore the effect of resprouting on our estimates of species- and size-specific suscep-

1 tibility to fire, we refitted Eq. 1 using full-kill as a response variable. We refer to this as the
2 full-kill model where we only considered a tree to be dead if the aboveground parts of the tree
3 were killed *and* there was no evidence of basal resprouting. Those trees that were top-killed,
4 but resprouting from their base, were treated as alive. For the full-kill model, we used the plot-
5 level estimates of I_i from the top-kill model as a predictor instead of re-estimating it as a latent
6 variable.

7 To assess the predictive ability of our top-kill model, we plotted the observed proportion
8 of top-killed trees per plot against the predicted proportion of top-killed trees per plot. Well-
9 calibrated curves should fall along a 1:1 line; that is, the predicted and observed top-killed rates
10 should be roughly the same.

11 **Fire intensity, top-kill patterns, and regeneration**

12 We used univariate linear regression to check whether the estimated values of latent relative fire
13 intensity (I) per plot were associated with landscape attribute details extracted from GIS layers
14 (*i.e.*, elevation, slope, aspect, and topographic wetness index, Gallant and Austin 2012).

15 We also explored the effect of fire intensity on the proportion of top-killed trees per forest
16 type and post-fire forest regeneration patterns. To visualise the relationship between the pro-
17 portion of top-killed trees (both by number of trees and basal area of trees) and relative fire
18 intensity per forest type, we plotted and fitted a separate general additive model (GAM) with a
19 Beta error distribution and a logit link to each forest type. We used the plot-level latent relative
20 fire intensity estimate to determine if regeneration patterns were associated with fire intensity
21 across the plots. To determine if there was any indication of directional changes in species
22 composition, we visually inspected whether there was a change in the dominant forest type of
23 the seedling community vs. the pre-fire adult trees forest type community. All data analyses

1 were conducted in the R statistical package (Version 3.6.0, R core development team, 2019) and
2 the top-kill and full-kill models were fitted using the Stan software (Carpenter *et al.* 2017, Stan
3 Development Team 2020).

4 **Results**

5 We sampled 2863 trees >10cm DBH (including both living and dead trees) across the 53
6 plots. Ten species (*Acacia dealbata*, *Acacia melanoxylon*, *Atherosperma moschatum*, *Eucalyp-*
7 *tus dives*, *Eucalyptus obliqua*, *Eucalyptus radiata*, *Eucalyptus regnans*, *Eucalyptus viminalis*,
8 *Nothofagus cunninghamii*, and *Olearia argophylla*) were sufficiently abundant (>15 individual
9 trees in total) across enough plots and had a sufficiently broad DBH range (> 20 cm) for anal-
10 ysis. These 10 species comprised 97% of all trees recorded in the study plots. Other species
11 present but not included in the analyses due to insufficient numbers or DBH range were *Bedfor-*
12 *dia arborescens* (17), *Eucalyptus cypellocarpa* (6), *Eucalyptus delegatensis* (134), *Hedycarya*
13 *angustifolia* (5), *Kunzea ericoides* (4), *Lomatia fraseri* (24), *Pomaderris aspera* (45), and *Tas-*
14 *mannia lanceolata* (8). It is worth noting that *Eucalyptus delegatensis* was not included in the
15 analyses despite being relatively abundant because every tree found in the study plots had been
16 killed in the fires.

17 The model estimation procedure produced an excellent fit of the model to the dataset. The
18 predictions of our top-kill model were well calibrated as the observed and predicted proportion
19 of top-kill per plot and per species were aligned with the 1:1 line (Figure 2).

20 **Variability in fire intensity**

21 The plot-level estimates of fire intensity (I) were highly variable both within and among study
22 locations, ranging from 0.05 (no fire) to 0.9 (highest observed intensity, resulting in all trees

1 being top-killed) (Figure 3). The distribution of plot-level relative fire intensity scores was not
2 uniform over the range of possible values. While 12 plots had $I < 0.2$ and nine plots had $I > 0.8$,
3 there were clear peaks in relative fire intensity around $I = 0.25$, 0.7 , and 0.9 . The Beeches,
4 Cambarville, and Cathedral Ranges sites experienced the full range of fire intensities—in most
5 cases within several hundred metres. A one-way ANOVA and *post-hoc* pairwise comparisons
6 showed that there were significant differences among the sites in latent relative fire intensity
7 ($F_{(5,47)} = 3.532$, $p = 0.009$). However, this was largely driven by differences between the Mt
8 Margaret plots (mean $I = 0.69$) and the Acheron Way (mean $I = 0.275$) and Cathedral Ranges
9 (mean $I = 0.32$) plots. Latent relative fire intensity was not associated with forest type ($F_{(3,49)}$
10 $= 1.25$, $p = 0.302$). A comparison of plot-level relative fire intensity values to topographic
11 variables (*i.e.*, elevation, aspect, slope, topographic wetness index) showed only weak positive
12 relationship with elevation ($R^2 = 0.12$, $p = 0.007$) and negative relationship with topographic
13 wetness index ($R^2 = 0.14$, $p = 0.007$). None of the other variables was correlated with relative
14 fire intensity.

15 **Species-specific patterns of fire-induced top-kill and full-kill**

16 We found distinct differences among the 10 study species in their susceptibility to fire-induced
17 top-kill and full-kill. Figure 4 shows the size-specific probability of top-kill and full-kill for
18 each tree species across a range of relative fire intensities. The model results show three distinct
19 patterns among the tree species.

20 First, there is a distinct hierarchy of susceptibility to fire-induced top-kill among the eu-
21 calypt species. *Eucalyptus obliqua* is the most fire resistant of the eucalypts. Under the most
22 intense fire conditions (*e.g.*, $I=0.8$, Figure 4), *E. obliqua* has a 50% probability of surviving once
23 it reaches 25 cm DBH. In contrast, *E. regnans*, the least resistant of the eucalypts, has 100%

1 top-kill under similar high-intensity fire conditions regardless of tree size. The three other eu-
2 calypt species are intermediate in their responses, but notably all had probabilities of top-kill
3 that were $<100\%$ under the most extreme fire conditions. They differed only in the probability
4 of top-kill for a given size. The size-specific responses of the five eucalypt species were also
5 strongly influenced by relative fire intensity. At low relative fire intensities all five of the euca-
6 lypt species had high top-kill survival rates. Even the relatively fire-sensitive *E. regnans* had a
7 50% probability of surviving a low intensity fire once individuals reached ~ 25 cm DBH.

8 The second pattern that emerged from our analyses was that fire-induced top-kill of the
9 five non-eucalypt species is primarily driven by relative fire intensity, not tree size. At low fire
10 intensities, the probability of mortality is less than 100% for all non-eucalypt species with the
11 exception of *Olearia argophylla*. However, once $I > 0.4$, the probability of mortality for all
12 five species is 100%, irrespective of DBH. These results suggest a threshold response in fire
13 mortality for the non-eucalypt species in these forests—in low-intensity fires, larger trees may
14 survive (although typically with low to moderate probability), but in high-intensity fires, all of
15 the trees are top-killed.

16 Third, accounting for basal resprouting (full-kill model) dramatically reduces estimated fire
17 susceptibility for those species that are capable of resprouting. The most extreme example of
18 this was *Olearia argophylla*, which experienced 100% top-kill under nearly all fire intensities,
19 irrespective of stem size, but almost complete survival when basal resprouting was included
20 (*i.e.*, full-kill was close to zero). The two cool temperate rainforest canopy dominants, *Atheros-*
21 *perma moschatum* and *Nothofagus cunninghamii*, were also exceptionally resilient to fire when
22 basal resprouting was considered. Under the highest fire intensity *Atherosperma moschatum*
23 and *Nothofagus cunninghamii* had 50% probabilities of full-kill survival for stems that were ~ 6
24 and ~ 20 cm DBH, respectively. All three of the dry sclerophyll forest eucalypt species were

1 capable of basal resprouting, although *E. radiata* was the less vigorous of the three. For *E. obli-*
2 *qua* and *E. dives*, the combination of low probabilities of top-kill (even under high-intensity fire
3 conditions) and the ability to resprout from the base following top-kill makes them extremely
4 resilient to fires.

5 **Patterns of fire-induced mortality among forest types**

6 The amount of top-killed trees per plot, whether calculated as the proportion of individuals
7 killed or the proportion of basal area (BA) killed, showed considerable variability when com-
8 pared to the estimated relative fire intensities. An important component of this variability was
9 differences among the four forest types in their response to fire intensity (Figure 5). For a given
10 fire intensity, DSF sites had a smaller proportion of dead individuals and BA than the three
11 other forest types. The relationships between top-kill and forest type differed for individuals
12 top-killed and BA top-killed. When the number of individuals is considered, the proportion of
13 trees top-killed by the fire was approximately linearly related with latent relative fire intensity
14 in the DSF. However, in the three other forest types, the relationship was non-linear, with an
15 apparent threshold effect. For values of $I > 0.35$, most individuals in CTR, MF, and TOF in
16 our sample plots were top-killed. For stand BA, the proportion top-killed was approximately
17 linearly related to fire intensity for DSF and CTR, although for a given relative fire intensity
18 the proportion of BA top-killed was much lower in DSF than CTR. TOF and MF exhibited a
19 threshold response between relative fire intensity and proportion of BA top-killed with complete
20 loss of BA when $I > 0.6$.

21 **Post-fire regeneration and fire intensity**

22 We compared various measures of regeneration performance (*i.e.*, species richness, maximum
23 height, and density) to fire intensity across the 53 study plots. Neither species richness nor

1 density of post-fire regeneration was correlated with fire intensity. However, maximum height
2 of the regeneration was weakly, but significantly, correlated with fire intensity at the plot level
3 ($Ht_{max} = 1.89 + 1.91I$), d.f. = 49, $p < 0.001$, $R^2 = 0.19$).

4 There was some indication that fire intensity may have had a threshold effect on species
5 composition in CTR and MF plots (Table 3). When $I < 0.5$, regeneration was dominated by
6 rainforest species, but when $I > 0.5$ the regeneration was dominated by TOF species, particularly
7 *Eucalyptus regnans*.

8 **Discussion**

9 Fire is a complex, stochastic phenomenon influenced by a wide range of variables associated
10 with weather, climate, topography, and vegetation (Agee 1993, Bradstock *et al.* 2002, Romme
11 *et al.* 2016) that generates spatially variable patterns of fire damage across multiple scales
12 (Turner *et al.* 1994, Schoennagel *et al.* 2008). Direct, high-resolution measurements of fire
13 intensity are rarely available for active high-intensity fires. This makes it difficult to disentangle
14 the various influences driving fire-induced damage and mortality in forested landscapes. By
15 simultaneously estimating fire intensity and species- and size-specific susceptibility to fire mor-
16 tality, we were able to identify interspecific differences in fire susceptibility and site-specific
17 variability in relative fire intensity. These differences have important implications for under-
18 standing how these forests will respond to fires in the future and suggest forest management
19 strategies to increase the resilience of these native forests to increased fire activity in the future.

20 **Spatial variability in fire intensity**

21 Our latent-variable approach to estimating unobserved relative fire intensity highlighted the
22 broad range of variability in fire intensity in the 2009 Black Saturday fires. Spatially heteroge-

1 neous canopy damage and mortality are a common feature of large disturbances, regardless of
2 forest type (Foster *et al.* 1998, Lertzman *et al.* 1998, Halofsky *et al.* 2011). In extreme fires, such
3 as the 1988 Yellowstone fires (Turner *et al.* 1994), the 2009 Black Saturday fires (Cruz *et al.*
4 2012), or the recent fires in southeastern Australia (NSW Google Earth Engine Burned Area
5 Map (GEEBAM)), large areas within the fire complex experience low- to moderate-severity
6 impacts. In these areas even relatively fire-sensitive trees or tree species may be able to survive,
7 creating a heterogeneous post-fire landscape composed of a mosaic of patches with different
8 forest structures and species. Turner *et al.* (1994) found that the spatial variability in damage
9 after the 1988 Yellowstone fires often occurred over small distances. Our estimates of fire in-
10 tensity for the 2009 Black Saturday fires showed similar patterns, with areas of low-intensity
11 fire ($I < 0.25$) occurring within several hundred metres of high-intensity fires ($I > 0.75$). This
12 spatial heterogeneity has important consequences for forest recovery and potential management
13 because it will generate a diverse set of stand development trajectories across the broader land-
14 scape (Tepley *et al.* 2013, Romme *et al.* 2016). At the broader landscape scale, the spatial
15 arrangement of stands with different developmental trajectories can also have important con-
16 sequences for habitat connectivity and metapopulation dynamics (Levey *et al.* 2008, Brudvig
17 and Damschen 2009), particularly of species associated with specific structural features such
18 as dead trees (Gibbons and Lindenmayer 1996) or specific developmental stages such as old-
19 growth forest (Oliver and Larson 1996).

20 **Variation among species in susceptibility to fire**

21 Our study provides the first empirical estimates of interspecific variation in susceptibility to
22 fire-induced mortality (both top-kill and full-kill) as a function of tree size and fire intensity.
23 We found distinct differences among the species of the dry sclerophyll forest (DSF) and the

1 three other forest types (cool temperate rainforest (CTR), cool temperate mixed forest (MF),
2 and tall open forest (TOF)). Most notably, the eucalypt species that dominated the DSF showed
3 strong size-dependent patterns of top-kill survival even at the highest fire intensities with larger
4 trees having higher probabilities of top-kill survival than smaller trees. In contrast, the tree
5 species from the other forest types exhibited size-independent top-kill under high-intensity fire
6 conditions—that is, all trees were killed aboveground regardless of stem size. Under low- to
7 moderate-intensity fires, top-kill patterns among the non-DSF tree species were more likely to
8 be size-dependent. However, for these species the probability of mortality declined substantially
9 only for the largest trees and for relatively low-intensity fires. Size-dependent patterns of tree
10 survival to fire have long been recognised in the literature (Gill and Ashton 1968, Ryan *et al.*
11 1988, Baker and Bunyavejchewin 2006). As trees increase in stem diameter, bark thickness
12 typically increases, providing greater insulating capacity and protection of the cambium.

13 Among the eucalypt species there was a clear gradient of resistance and resilience to fire-
14 induced mortality. *Eucalyptus obliqua* and *E. dives* were the most resistant to fire—that is, the
15 existing stems were not top-killed at relatively modest sizes even in the hottest fires. This is
16 presumably due to their relatively thick bark and ability to produce epicormic shoots soon after
17 fires (Burrows 2013). However, they were also the most resilient to fire, with small individuals
18 vigorously resprouting from the stem base after high-intensity fires. *Eucalyptus radiata* and *E.*
19 *viminialis* were intermediate in their resistance to fire—that is, for a given relative fire intensity,
20 they required larger trees to have the same probability of top-kill survival as *E. obliqua* or *E.*
21 *dives*. *Eucalyptus regnans* was the least resistant to fire of the modelled species with complete
22 top-kill regardless of tree size in the highest intensity fires, although it did show evidence of
23 size-dependent top-kill survival in low- to moderate-intensity fires. While we were not able
24 to include *E. delegatensis* into our mortality model due to the lack of surviving individuals,

1 all 134 individuals were full-killed (*i.e.*, complete aboveground mortality and no evidence of
2 resprouting), which is consistent with their known sensitivity to fire (Bowman *et al.* 2014).

3 Future fires may act as a filter that shapes the structure and composition of forests across
4 these landscapes in the coming decades (Stevens-Rumann *et al.* 2017, Trouvé *et al.* 2020). For
5 example, if fires become more intense in the DSF, the species-specific differences in suscep-
6 tibility to fire-induced full-kill would lead us to predict a compositional shift towards greater
7 dominance of *E. obliqua* and *E. dives* and gradual loss of *E. radiata* and *E. viminalis*. In addi-
8 tion, size-specific differences in susceptibility within each species would be expected to lead to
9 stands with fewer small trees. However, where young DSF stands dominated by small trees ex-
10 ist, high-intensity fires may lead to widespread top-kill followed by abundant basal resprouting.
11 Anecdotal observations of post-fire recovery after the recent 2019/20 fires suggest that this pat-
12 tern is common across eastern Victoria. It remains unclear whether DSF species would be able
13 to persist through multiple cycles of top-kill and basal resprouting without eventually depleting
14 their stored carbohydrate reserves and ceding the site to a different vegetation type (Fairman
15 *et al.* 2016, 2017).

16 For the TOF dominated by *E. regnans*, high-intensity fires are typically stand replacing.
17 If the burnt trees are of reproductive age (older than ~ 21 years, Fagg *et al.* 2013, von
18 Takach Dukai *et al.* 2018), a new, single cohort of *E. regnans* will establish. If not, the species
19 will become locally extinct and be replaced by communities dominated by non-eucalypts (Fair-
20 man *et al.* 2016). This shift in composition and functional type will persist until a subsequent
21 disturbance provides opportunities for new eucalypt seed dispersing into the site from adjacent
22 stands to germinate and grow. In contrast, low- to moderate-intensity fires in TOF may kill a
23 relatively small proportion of the large trees (Figure 4). This can lead to structurally complex
24 stands with two distinct cohorts made up of surviving trees and new recruits (Ashton 2000). The

1 subsequent dynamics of the stand will depend on the relative abundance of the young and old
2 cohorts. If there are relatively few trees in the older cohort, the new recruits will have access to
3 sufficient resources to grow and survive, allowing the two-cohort structure to persist. However,
4 if the trees from the older cohort are abundant, the younger cohort will be suppressed and ex-
5 perience high mortality rates from competition, leading to the stand reverting to a single-cohort
6 structure (see *e.g.*, Figure 19 in Ashton 2000).

7 Basal resprouting was an important component of resilience, particularly for the CTR species,
8 *Nothofagus cunninghamii* and *Atherosperma moschatum*. CTR has long been considered highly
9 sensitive to fire and there is concern that an increase in fire activity under future climatic con-
10 ditions will lead to contraction and loss of these ecosystems. However, a multi-scale review of
11 CTR dynamics based on evidence from fossil pollen, sub-fossil charcoal, tree rings, and direct
12 observations after the 2009 bushfires showed that cool temperate rainforest had persisted in Vic-
13 toria's Central Highlands despite tens of thousands of years of fire activity (Baker *et al.* 2012).
14 Similarly, Worth *et al.* (2009) used chloroplast DNA to show that *Nothofagus cunninghamii* had
15 persisted *in situ* in Tasmania through periods of global aridity during the Pleistocene. The re-
16 sults from our study demonstrate that the resprouting ability of the dominant canopy species of
17 the cool temperate rainforest, *N. cunninghamii* and *A. moschatum*, is an important component
18 of this resilience. Several previous studies have demonstrated the ability of *Nothofagus cun-*
19 *ninghamii* to resprout after fires in Victoria and Tasmania (Baker *et al.* 2012, Prior *et al.* 2018).
20 When basal resprouting is accounted for, full-kill rates in the highest intensity fire are <50%,
21 even for relatively small individuals (Figure 4). Simkin and Baker (2008) showed that many of
22 the *Nothofagus cunninghamii* and *Atherosperma moschatum* in a cool temperate rainforest near
23 Marysville, Victoria, survived in the extensive 1939 bushfires, even though all of the *Eucalyptus*
24 *regnans* (mountain ash) and *E. delegatensis* (alpine ash) on the adjacent slopes were killed. It

1 is unclear, however, how many cycles of top-kill and basal resprouting CTR species can endure
2 before ceding ground to more resilient species (Fairman *et al.* 2016).

3 Despite the resilience of CTR conferred by basal resprouting, regeneration patterns were
4 linked to fire intensity. Under low to moderate fire intensities ($I < 0.5$), seedling composition
5 was dominated by CTR species. However, when $I > 0.5$, seedling composition was dominated
6 by species from the TOF, particularly *E. regnans*. This interaction between fire intensity and the
7 species composition of the post-fire regeneration has been documented in other forests around
8 the world (*e.g.*, Hollingsworth *et al.* 2013) and raises important questions about the long-term
9 viability of CTR in a warming world.

10 At the landscape scale, spatial variability in fire intensity is a key source of heterogeneity
11 in post-fire forest development (Turner *et al.* 2004, Schoennagel *et al.* 2008). Differences in
12 fire intensity within and among stands and forest types contribute to variation in the density of
13 surviving and dead trees, the predominant mode of regeneration (*e.g.*, seeding, sprouting), and
14 spatial patterns of regeneration. Differences in initial post-fire conditions, both biotic and abi-
15 otic, can lead to differences in post-fire stand development trajectories (Ewel 1980, Christensen
16 *et al.* 1989, Palik and Pregitzer 1993, Ashton 2000, Tepley *et al.* 2013). These, in turn, can drive
17 the future provision of ecosystem services such as carbon sequestration, habitat development,
18 and water uptake.

19 **Implications for managing forests to be more fire resistant**

20 The relationships between top-kill survival and relative fire intensity, tree size, and species also
21 suggest management strategies that may help adapt forested landscapes to fire and reduce the
22 risk of widespread mortality as fire regimes change (*e.g.*, Koontz *et al.* 2020). All but one
23 of the species in this study showed some evidence of size-dependent aboveground mortality

1 (*i.e.*, top-kill) under specific fire intensity conditions. Where size-dependent top-kill occurred,
2 the probability of aboveground mortality was negatively associated with stem size—bigger trees
3 were more likely to survive a fire than smaller trees. This suggests that silvicultural prescriptions
4 that accelerate the growth of individual trees have the potential to reduce the period of time that
5 a tree is exposed to the risk of fire-induced top-kill.

6 Forest management practices, such as thinning, that accelerate the growth of individuals
7 (West and Osier 1995) have the potential to make these stands more resistant to future fires,
8 particularly high-intensity fires, by reducing their exposure to risk of being top-killed. To il-
9 lustrate this point, we provide an example using *E. regnans* for which thinning trial data are
10 abundant. Reducing the initial density through reduced sowing rates or early thinning can lead
11 to substantial increases in growth rates of *E. regnans*. Historical spacing trials established in
12 the 1960s give an indication of the scale of the potential gains (Figure 6). For stands starting at
13 high stem densities, which are typical of current forest management practices (>2500 seedlings
14 ha⁻¹ at four years old), it takes ~35 years to reach 35 cm DBH, the size at which *E. regnans*
15 has a 50% probability of top-kill survival under moderate relative fire intensity ($I = 0.4$). In
16 contrast, at the lowest density in the spacing trials (120 stems ha⁻¹ at four years old), it took
17 only 12 years for individual trees to reach 35 cm DBH. By growing *E. regnans* at such a low
18 density, the length of time individuals are exposed to a high probability of top-kill under a low-
19 to moderate-intensity fire was reduced by more than two decades. Low stem densities, whether
20 achieved through reduced sowing densities or early thinning, may increase the diameter growth
21 of individual trees, but they may also have a variety of drawbacks. For example, at extreme low
22 densities the base of the live crown of individual trees may be closer to the ground and therefore
23 prolong the period that the crown is susceptible to damaging crown scorch. Also, density ef-
24 fects on height growth may be positive, neutral, or negative; although among broad-leaved tree

1 species there is often a positive effect of reduced density on height growth (Trouvé *et al.* 2019).
2 Depending on the fire intensity, individual trees may need to reach a certain height (regardless
3 of stem diameter) to escape being top-killed by fires. Maintaining low density stands may have
4 other negative consequences for fire behaviour such as stimulating shrub growth, increasing
5 fuel loads from thinning debris, and increased insolation and wind movement at the forest floor,
6 which may accelerate fuel drying (see Zylstra *et al.* 2016, although under very dry conditions,
7 surface fuels in open stands have been shown to be not much drier than fuels in denser stands
8 (Bigelow and North 2012)). However, lower density stands also have reduced crown packing
9 within the stand, which lowers the connectivity of crown fuels and reduces fire spread through
10 the forest canopy (Van Wagner 1977, Wilson and Baker 1998).

11 This reduction in the period of time that individual trees are exposed to high risk of fire-
12 induced mortality is of particular importance for landscape-scale forest management planning.
13 While the 2009 Black Saturday fires were amongst the most devastating of the past century,
14 only about one-third of the forest experienced catastrophic (>80% canopy loss), stand-replacing
15 fires (Cruz *et al.* 2012). The remaining two-thirds of the burnt area was evenly divided between
16 forest that experienced partial (20-80%) canopy loss and forest that had little (<20%) canopy
17 loss or only experienced surface fire (Jiang *et al.* in prep.). If such a significant proportion of
18 the landscape experienced low-to-moderate fires—even under some of the most catastrophic
19 fire conditions ever recorded—then managing *E. regnans* to maximise individual growth rates
20 instead of stand-level volume may increase the probability that at least some individuals survive
21 future fires. Because the timing, location, and intensity of future fires is unknown, increasing
22 the proportion of stands with lower densities and higher mean tree sizes for a given age should
23 reduce the total area of stands that suffer high mortality rates across the landscape. While areas
24 with high-intensity fires may have few surviving trees, other areas that only experienced low-

1 to moderate-intensity fires could have significantly lower mortality rates. It is these areas of
2 individual trees or clumps of trees that survive after fires that are critically important sources
3 of shelter and foraging for wildlife in the months and years after the fire has passed (Meddens
4 *et al.* 2018, Coop *et al.* 2019), as well as a critical source of seeds for regeneration after the fire.

5 For the other eucalypt species, which are able to survive moderate- to high-intensity fires,
6 reducing stand density would also limit their exposure to fire-induced mortality. This is partic-
7 ularly important given the context of the 2019-20 bushfire season in southeastern Australian,
8 which has seen nearly 6 million hectares of forest burn in Victoria and New South Wales. A
9 large proportion of these forests are mixed eucalypt forests composed of resprouting eucalypt
10 species similar to those from the DSF in this study. The predicted changes to regional climate
11 are expected to lead to more frequent, more extensive, and potentially more severe fires. This
12 will likely lead to significant changes in species composition and forest structure at the stand,
13 landscape, and regional scale (*e.g.*, Fairman *et al.* 2016). Forest management practices that ac-
14 celerate the growth of individual trees into more fire-safe size classes can reduce the probability
15 of stand- and landscape-wide mortality and increase the chance that at least some forest cover
16 is maintained within a site as fire regimes change. Understanding how different species vary in
17 their susceptibility to fire, conditional on size and fire intensity, may help inform management
18 practices that can mitigate some of this risk and create more resilient landscapes (Schoennagel
19 *et al.* 2017, McWethy *et al.* 2019).

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4 **Open Research statement**

5 Data (Trouvé *et al.* 2021) are available on Github at: [https://doi.org/10.5281/zenodo.](https://doi.org/10.5281/zenodo.4727469)
6 4727469.

7 **References**

- 8 Agee JK, 1993. *Fire Ecology of Pacific Northwest Forests*. Island Press, Washington, DC.
- 9 Ashton DH, 2000. The Big Ash forest, Wallaby Creek, Victoria — changes during one lifetime.
10 *Australian Journal of Botany* 48(1): 1–26.
- 11 Ashton DH, Attiwill PM, 1994. Tall-open forests. In Groves RH (Editor), *Australian Vegetation*.
12 Cambridge University Press, Cambridge, UK.
- 13 Baker PJ, Bunyavejchewin S, 2006. Bark thickness and the influence of forest fire on tree
14 population structures in a seasonal evergreen tropical forest. *Natural History Bulletin of the*
15 *Siam Society* 54: 215–225.
- 16 Baker PJ, Simkin R, Pappas N, Mcleod A, McKenzie GM, 2012. Fire on the mountain: A
17 multi-scale, multi- proxy assessment of the resilience of cool temperate rainforest to fire
18 in Victoria’s Central Highlands. In Haberle SG, David B (Editors), *Peopled landscapes:*
19 *Archaeological and biogeographic approaches to landscapes*, pages 375–391. ANU E-Press.

- 1 Bigelow SW, North MP, 2012. Microclimate effects of fuels-reduction and group-selection
2 silviculture: Implications for fire behavior in sierran mixed-conifer forests. *Forest Ecology*
3 *and Management* 264: 51–59.
- 4 Bowman DMJS, Balch JK, Artaxo P, Bond W, Carlson JM, Cochrane MA, D'antonio CM, De-
5 fries RS, Doyle JC, Harrison SP, Johnston FH, Keeley JE, Krawchuk MA, Kull CA, Marston
6 JB, Moritz MA, Prentice IC, Roos CI, Scott AC, Swetnam TW, Van Der Werf GR, Pyne SJ,
7 2009. Fire in the Earth System. *Science* 324(5926): 481–484.
- 8 Bowman DMJS, Murphy BP, Neyland DLJ, Williamson GJ, Prior LD, 2014. Abrupt fire regime
9 change may cause landscape-wide loss of mature obligate seeder forests. *Global Change*
10 *Biology* 20(3): 1008–1015.
- 11 Bradstock RA, Williams JE, Gill AM (Editors), 2002. *Flammable Australia: Fire Regimes and*
12 *Biodiversity of a Continent*. Cambridge University Press, Cambridge, UK.
- 13 Brown LD, Cai TT, DasGupta A, 2001. Interval estimation for a binomial proportion. *Statistical*
14 *Science* 16(2): 101–133.
- 15 Brudvig LA, Damschen EI, 2009. Landscape connectivity promotes plant biodiversity spillover
16 into non-target habitats. *Proceedings of the National Academy of Sciences* 106(23): 9328–
17 9332.
- 18 Burrows GE, 2013. Buds, bushfires and resprouting in the eucalypts. *Aust J Bot* 61(5): 331–349.
- 19 Busby JR, Brown MJ, 1994. Southern rainforests. In *Australian Vegetation*. Cambridge Uni-
20 versity Press, Cambridge, UK.
- 21 Canham CD, Papaik MJ, Latty EF, 2001. Interspecific variation in susceptibility to windthrow

- 1 as a function of tree size and storm severity for northern temperate tree species. *Canadian*
2 *Journal of Forest Research* 31(1): 1–10.
- 3 Canham CD, Thompson J, Zimmerman JK, Uriarte M, 2010. Variation in susceptibility to
4 hurricane damage as a function of storm intensity in puerto rican tree species. *Biotropica*
5 42(1): 87–94.
- 6 Carpenter B, Gelman A, Hoffman MD, Lee D, Goodrich B, Betancourt M, Brubaker M, Guo J,
7 Li P, Riddell A, 2017. Stan: A Probabilistic Programming Language. *Journal of Statistical*
8 *Software* 76(1): 1–32.
- 9 Christensen NL, Agee JK, Brussard PF, Hughes J, Knight DH, Minshall GW, Peek JM, Pyne
10 SJ, Swanson FJ, Thomas JW, 1989. Interpreting the Yellowstone fires of 1988. *BioScience*
11 39(10): 678–685.
- 12 Coop JD, DeLory TJ, Downing WM, Haire SL, Krawchuk MA, Miller C, Parisien MA, Walker
13 RB, 2019. Contributions of fire refugia to resilient ponderosa pine and dry mixed-conifer
14 forest landscapes. *Ecosphere* 10(7): 39–24.
- 15 Cruz MG, Sullivan AL, Gould JS, Sims NC, Bannister AJ, Hollis JJ, Hurley RJ, 2012. Anatomy
16 of a catastrophic wildfire: The Black Saturday Kilmore East fire in Victoria, Australia. *Forest*
17 *Ecology and Management* 284(C): 269–285.
- 18 Doherty MD, Lavorel S, Colloff MJ, Williams KJ, Williams RJ, 2016. Moving from au-
19 tonomous to planned adaptation in the montane forests of southeastern Australia under chang-
20 ing fire regimes. *Austral Ecology* 42(3): 309–316.
- 21 Ewel J, 1980. Tropical succession: manifold routes to maturity. *Biotropica* 12(2): 2–7.

- 1 Fagg PC, Lutze M, Slijkerman C, Ryan M, Bassett O, 2013. Silvicultural recovery in ash forests
2 following three recent large bushfires in Victoria. *Australian Forestry* 76(3-4): 140–155.
- 3 Fairman TA, Bennett LT, Tupper S, Nitschke CR, 2017. Frequent wildfires erode tree per-
4 sistence and alter stand structure and initial composition of a fire-tolerant sub-alpine forest.
5 *Journal of Vegetation Science* 28(6): 1151–1165.
- 6 Fairman TA, Nitschke CR, Bennett LT, 2016. Too much, too soon? A review of the effects
7 of increasing wildfire frequency on tree mortality and regeneration in temperate eucalypt
8 forests. *International Journal of Wildland Fire* 25: 831–848.
- 9 Foster DR, Knight D, Franklin J, 1998. Landscape patterns and legacies resulting from large,
10 infrequent forest disturbances. *Ecosystems* 1(6): 497–510.
- 11 Gallant J, Austin J, 2012. *Topographic Wetness Index derived from 1" SRTM DEM-H. V2.*
12 CSIRO. Data Collection. <https://doi.org/10.4225/08/57590B59A4A08>.
- 13 Gelman A, Hill J, 2008. *Data Analysis Using Regression and Multilevel Hierarchical Models.*
14 Analytical methods for social research. Cambridge University Press.
- 15 Gibbons P, Lindenmayer DB, 1996. Issues associated with the retention of hollow-bearing trees
16 within eucalypt forests managed for wood production. *Forest Ecology and Management*
17 83(3): 245–279.
- 18 Gill AM, Ashton DH, 1968. The role of bark type in relative tolerance to fire of three central
19 Victorian Eucalypts. *Australian Journal of Botany* 16(3): 491–498.
- 20 Halofsky JE, Donato DC, Hibbs DE, Campbell JL, Cannon MD, Fontaine JB, Thompson JR,
21 Anthony RG, Bormann BT, Kayes LJ, Law BE, Peterson DL, Spies TA, 2011. Mixed-severity

- 1 fire regimes: lessons and hypotheses from the Klamath-Siskiyou Ecoregion. *Ecosphere* 2(4):
2 art40–19.
- 3 Hoffmann WA, Solbrig OT, 2003. The role of topkill in the differential response of savanna
4 woody species to fire. *Forest Ecology and Management* 180(1-3): 273–286.
- 5 Hollingsworth TN, Johnstone JF, Bernhardt EL, Chapin FS, 2013. Fire Severity Filters Regen-
6 eration Traits to Shape Community Assembly in Alaska’s Boreal Forest. *PLoS ONE* 8(2):
7 e56033–11.
- 8 Keeley JE, 2009. Fire intensity, fire severity and burn severity: a brief review and suggested
9 usage. *International Journal of Wildland Fire* 10(3): 393–402.
- 10 Keyser TL, Smith FW, Lentile LB, 2006. Modeling postfire mortality of ponderosa pine fol-
11 lowing a mixed-severity wildfire in the Black Hills: the role of tree morphology and direct
12 fire effects. *Forest Science* 52(5): 530–539.
- 13 Koontz MJ, North MP, Werner CM, Fick SE, Latimer AM, 2020. Local forest structure vari-
14 ability increases resilience to wildfire in dry western U.S. coniferous forests. *Ecology Letters*
15 17: 52–12.
- 16 Lavorel S, Colloff MJ, McIntyre S, Doherty MD, Murphy HT, Metcalfe DJ, Dunlop M,
17 Williams RJ, Wise RM, Williams KJ, 2014. Ecological mechanisms underpinning climate
18 adaptation services. *Global Change Biology* 21(1): 12–31.
- 19 Lentile LB, Smith FW, Shepperd WD, 2006. Influence of topography and forest structure on
20 patterns of mixed severity fire in ponderosa pine forests of the South Dakota Black Hills,
21 USA. *International Journal of Wildland Fire* 15(4): 557–10.

- 1 Lertzman K, Fall J, Dorner B, 1998. Three kinds of heterogeneity in fire regimes: at the cross-
2 roads of fire history and landscape ecology. *Northwest Science* 72: 4–23.
- 3 Levey DJ, Tewksbury JJ, Bolker BM, 2008. Modelling long-distance seed dispersal in hetero-
4 geneous landscapes. *Journal of Ecology* 96(4): 599–608.
- 5 McLauchlan KK, Higuera PE, Miesel JR, Rogers BM, Schweitzer J, Shuman JK, Varner JM,
6 Veblen TT, Adalsteinsson SA, Baker PJ, Balch JK, Batllori Presas E, Bigio E, Brando PM,
7 Cattau ME, Chipman ML, Coen JL, Crandall RM, Daniels LD, Enright NJ, Gross W, Harvey
8 BJ, Hatten JA, Hermann S, Hewitt RE, Kobziar LN, Landesmann JB, Loranty MM, Maezumi
9 SY, Mearns LO, Moritz MA, Myers JA, Pausas JG, Pelligrini AF, Platt WJ, Roozeboom J,
10 Safford HD, Santos F, Scheller RM, Sherriff RL, Smith KG, Smith MD, Watts AC, 2020. Fire
11 as a fundamental ecological process: resaerch advances and frontiers. *Journal of Ecology* In
12 press.
- 13 McWethy DB, Schoennagel T, Higuera PE, Krawchuk M, Harvey BJ, Metcalf EC, Schultz C,
14 Miller C, Metcalf AL, Buma B, Virapongse A, Kulig JC, Stedman RC, Ratajczak Z, Nelson
15 CR, Kolden C, 2019. Rethinking resilience to wildfire. *Nature Sustainability* pages 1–8.
- 16 Meddens AJH, Kolden CA, Lutz JA, Smith AMS, Cansler CA, Abatzoglou JT, Meigs GW,
17 Downing WM, Krawchuk MA, 2018. Fire Refugia: What Are They, and Why Do They
18 Matter for Global Change? *BioScience* 26(7537): 269–11.
- 19 Moritz MA, Batllori E, Bradstock RA, Gill AM, Handmer J, Hessburg PF, Leonard J, McCaf-
20 frey S, Odion DC, Schoennagel T, SYPHARD AD, 2014. Learning to coexist with wildfire.
21 *Nature* 515(7525): 58–66.
- 22 Noble IR, Bary GAV, Gill AM, 1980. McArthur’s fire-danger meters expressed as equations.
23 *Australian Journal of Ecology* 5(2): 201–203.

- 1 Oliver CD, Larson BC, 1996. *Forest stand dynamics*. John Wiley and Sons, New York.
- 2 Palik B, Pregitzer K, 1993. The repeatability of stem exclusion during even-aged development
3 of bigtooth aspen dominated forests. *Canadian Journal of Forest Research* 23(6): 1156–
4 1168.
- 5 Prior LD, French BJ, Bowman DMJS, 2018. Effect of experimental fire on seedlings of aus-
6 tralian and gondwanan trees species from a tasmanian montane vegetation mosaic. *Australian*
7 *Journal of Botany* 66(7): 511–517.
- 8 Romme WH, Whitby TG, Tinker DB, Turner MG, 2016. Deterministic and stochastic processes
9 lead to divergence in plant communities 25 years after the 1988 Yellowstone fires. *Ecological*
10 *Monographs* 86(3): 327–351.
- 11 Ryan KC, Peterson DL, Reinhardt ED, 1988. Modeling long-term fire-caused mortality of
12 Douglas fir. *Forest Science* 34(1): 190–199.
- 13 Schoennagel T, Balch JK, Brenkert-Smith H, Dennison PE, Harvey BJ, Krawchuk MA, Mi-
14 etkiewicz N, Morgan P, Moritz MA, Rasker R, Turner MG, Whitlock C, 2017. Adapt to
15 more wildfire in western North American forests as climate changes. *Proceedings of the*
16 *National Academy of Sciences* 114(18): 4582–4590.
- 17 Schoennagel T, Smithwick EAH, Turner MG, 2008. Landscape heterogeneity following large
18 fires: insights from Yellowstone National Park, USA. *International Journal of Wildland Fire*
19 17(6): 742–12.
- 20 Simkin R, Baker PJ, 2008. Disturbance history and stand dynamics in tall open forest and
21 riparian rainforest in the Central Highlands of Victoria. *Austral Ecology* 33(6): 747–760.
- 22 Stan Development Team, 2020. RStan: the R interface to Stan. R package version 2.21.2.

- 1 Stevens-Rumann CS, Kemp KB, Higuera PE, Harvey BJ, Rother MT, Donato DC, Morgan P,
2 Veblen TT, 2017. Evidence for declining forest resilience to wildfires under climate change.
3 *Ecology Letters* 21(2): 243–252.
- 4 Teague B, McLeod R, Pascoe S, 2010. Final report of the Teague Commission on the 2009
5 Black Saturday bushfires. Technical report, 2009 Victorian Bushfires Royal Commission.
- 6 Tepley AJ, Swanson FJ, Spies TA, 2013. Fire-Mediated Pathways of Stand Development in
7 Douglas-Fir/Western Hemlock Forests of the Pacific Northwest, USA. *Ecology* 94: 1729–
8 1743.
- 9 Tepley AJ, Thompson JR, Epstein HE, Anderson-Teixeira KJ, 2017. Vulnerability to forest loss
10 through altered postfire recovery dynamics in a warming climate in the Klamath Mountains.
11 *Global Change Biology* 23(10): 4117–4132.
- 12 Trouvé R, Bontemps JD, Collet C, Seynave I, Lebourgeois F, 2019. When do dendrometric
13 rules fail? insights from 20 years of experimental thinnings on sessile oak in the gis coop
14 network. *Forest Ecology and Management* 433: 276–286.
- 15 Trouvé R, Osborne L, Baker PJ, 2021. Data accompanying the manuscript ‘the effect of species,
16 size, and fire intensity on tree mortality within a catastrophic bushfire complex’. ”<https://doi.org/10.5281/zenodo.4727469>”.
- 18 Trouvé R, Bunyavejchewin S, Baker PJ, 2020. Disentangling fire intensity and species’ suscep-
19 tibility to fire in a species-rich seasonal tropical forest. *Journal of Ecology* 102(3): 9–13.
- 20 Turner M, Dale VH, 1998. Comparing large, infrequent disturbances: What have we learned?
21 *Ecosystems* 1: 493–496.

- 1 Turner M, Tinker D, Romme W, Kashian D, Litton C, 2004. Landscape patterns of sapling
2 density, leaf area, and aboveground net primary production in postfire lodgepole pine forests,
3 Yellowstone National Park (USA). *Ecosystems* 7(7): 751–775.
- 4 Turner MG, Gardner RH, Dale VH, O'Neill RV, 1989. Predicting the spread of disturbance
5 across heterogeneous landscapes. *Oikos* 55: 121–129.
- 6 Turner MG, Hargrove WW, Gardner RH, Romme WH, 1994. Effects of fire on landscape
7 heterogeneity in Yellowstone National Park, Wyoming. *Journal of Vegetation Science* 5(5):
8 731–742.
- 9 Van Wagner CE, 1977. Conditions for the start and spread of crown fire. *Canadian Journal of*
10 *Forest Research* 7: 23–34.
- 11 von Takach Dukai B, Lindenmayer DB, Banks SC, 2018. Environmental influences on growth
12 and reproductive maturation of a keystone forest tree: Implications for obligate seeder sus-
13 ceptibility to frequent fire. *Forest Ecology and Management* 411: 108–119.
- 14 West PW, Osier G, 1995. Growth response to thinning and its relation to site resources in
15 *Eucalyptus regnans*. *Canadian Journal of Forest Research* 25(1): 69–80.
- 16 Wilson JS, Baker PJ, 1998. Mitigating fire risk to late-successional forest reserves on the east
17 slope of the Washington Cascade Range, USA. *Forest Ecology and Management* 110: 59–75.
- 18 Worth JRP, Jordan GJ, Mckinnon GE, Vaillancourt RE, 2009. The major Australian cool tem-
19 perate rainforest tree *Nothofagus cunninghamii* withstood Pleistocene glacial aridity within
20 multiple regions: evidence from the chloroplast. *New Phytologist* 182(2): 519–532.
- 21 Zylstra P, Bradstock RA, Bedward M, Penman TD, Doherty MD, Weber RO, Gill AM, Cary GJ,
22 2016. Biophysical Mechanistic Modelling Quantifies the Effects of Plant Traits on Fire Sever-

- 1 ity: Species, Not Surface Fuel Loads, Determine Flame Dimensions in Eucalypt Forests.
- 2 *PLoS ONE* 11(8): e0160715–24.

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Table 1: Summary statistics of the data. The resprout column corresponds to the percentage of basal resprouting trees among the trees that were top-killed (*i.e.*, $resprout = \frac{topkilled - fullkilled}{topkilled}$).

	Number of trees	DBH (cm)		Top-killed %	Full-killed %	Resprout %
		Mean	Range			
<i>Acacia dealbata</i>	405	24	(10, 70)	79	79	0
<i>Acacia melanoxylon</i>	166	18	(10, 48)	64	57	11
<i>Atherosperma moschatum</i>	61	18	(10, 77)	84	48	43
<i>Nothofagus cunninghami</i>	807	23	(10, 164)	79	44	45
<i>Olearia argophylla</i>	56	14	(10, 31)	100	4	96
<i>Eucalyptus obliqua</i>	295	35	(10, 112)	35	23	33
<i>Eucalyptus dives</i>	127	32	(10, 80)	47	24	48
<i>Eucalyptus radiata</i>	348	36	(10, 162)	46	36	20
<i>Eucalyptus viminalis</i>	18	56	(19, 100)	22	22	0
<i>Eucalyptus regnans</i>	318	68	(12, 376)	42	42	0

Table 2: Mean and 95% credible interval of parameter values per species for the top-kill and full-kill models

Model	Species	α			β			γ		
		Mean	95% CI		Mean	95% CI		Mean	95% CI	
			Min	Max		Min	Max		Min	Max
Top-kill	<i>Acacia dealbata</i>	-3.87	-6.77	-1.93	196.0	69.2	470.9	-0.53	-0.84	-0.26
	<i>Acacia melanoxylon</i>	-3.66	-5.83	-2.15	189.5	60.9	497.6	-0.66	-1.05	-0.33
	<i>Atherosperma moschatum</i>	-3.61	-7.73	-0.99	68.9	19.3	197.7	-0.28	-0.85	-0.02
	<i>Eucalyptus dives</i>	-3.00	-4.73	-1.68	65.9	21.4	171.4	-0.74	-1.18	-0.43
	<i>Eucalyptus obliqua</i>	-2.77	-3.75	-2.02	479.4	88.5	1919.5	-1.42	-2.05	-0.94
	<i>Eucalyptus radiata</i>	-2.59	-3.50	-1.86	151.4	57.4	344.7	-0.84	-1.15	-0.58
	<i>Eucalyptus regnans</i>	-6.50	-9.41	-4.51	182.1	79.2	381.9	-0.53	-0.71	-0.36
	<i>Eucalyptus viminalis</i>	-3.10	-6.19	-0.87	158.5	12.4	801.5	-1.16	-4.80	-0.15
	<i>Nothofagus cunninghami</i>	-4.23	-7.15	-2.47	34.5	21.5	54.5	-0.07	-0.25	-0.01
	<i>Olearia argophylla</i>	-3.34	-7.26	0.29	306.6	40.7	1472.5	-0.29	-0.84	-0.01
Full-kill	<i>Atherosperma moschatum</i>	-10.79	-17.36	-5.30	86.1	4.3	397.4	-2.67	-8.50	-0.15
	<i>Eucalyptus dives</i>	-11.02	-18.25	-5.85	79.8	2.3	327.8	-2.19	-11.64	-0.23
	<i>Eucalyptus obliqua</i>	-11.13	-18.72	-5.94	115.7	2.1	512.4	-2.25	-8.03	-0.22
	<i>Eucalyptus radiata</i>	-11.12	-17.06	-6.45	93.7	20.7	343.2	-0.39	-0.86	-0.07
	<i>Nothofagus cunninghami</i>	-10.57	-16.68	-6.21	167.3	43.4	706.0	-0.76	-1.35	-0.38
	<i>Olearia argophylla</i>	-10.99	-18.83	-6.00	105.6	2.4	307.2	-2.72	-12.96	-0.18

Table 3: Mean estimates (and range) of latent relative fire intensity (I) and proportion of trees killed by the fire for each study location. Forest type describes the pre-fire forest classification based on tree species composition (DSF–dry sclerophyll forest; TOF–tall open forest; CTR–cool temperate rainforest; MF–mixed forest). Seedling composition reflects the species composition of post-fire regeneration observed in each forest type. Where the species composition of the post-fire regeneration differed from pre-fire tree species composition, the forest type association of the regeneration is highlighted in bold. At the Beeches CTR sites, post-fire seedling composition was conditional on fire intensity. So, we present the post-fire species composition for sites that experienced fire intensities above and below the observed threshold intensity value. Estimates of proportion dead are calculated as the basal area (BA) of trees within each plot that were killed in the fire divided by the total BA of the plot.

Location	Forest type	N	Relative fire intensity (I)	Proportion of top-kill	Seedling composition
Acheron Way	CTR	1	0.05	0.05	CTR
	DSF	2	0.55 (0.48-0.63)	0.21 (0.14-0.29)	DSF
	MF	1	0.09	0.09	MF
	TOF	6	0.25 (0.20-0.32)	0.18 (0.14-0.39)	TOF
The Beeches	CTR	4	0.28 (0.13-0.29)	0.63 (0.31-0.83)	$I < 0.5 =$ CTR
	MF	3	0.79 (0.73-0.83)	1.00	$I > 0.5 =$ TOF
Cambarville	CTR	3	0.53 (0.35-0.88)	0.55 (0.26-1.00)	TOF
	MF	3	0.51 (0.08-0.83)	0.58 (0.04-1.00)	TOF
	TOF	5	0.33 (0.18-0.83)	0.33 (0.17-1.00)	TOF
Cathedral Ranges	DSF	11	0.32 (0.06-0.89)	0.28 (0.07-0.84)	DSF
Marysville	DSF	5	0.54 (0.38-0.71)	0.47 (0.37-0.76)	DSF
Mt Margaret	CTR	1	0.68	1.00	MF
	MF	2	0.67	1.00	MF
	TOF	5	0.7 (0.60-0.90)	1.00	TOF

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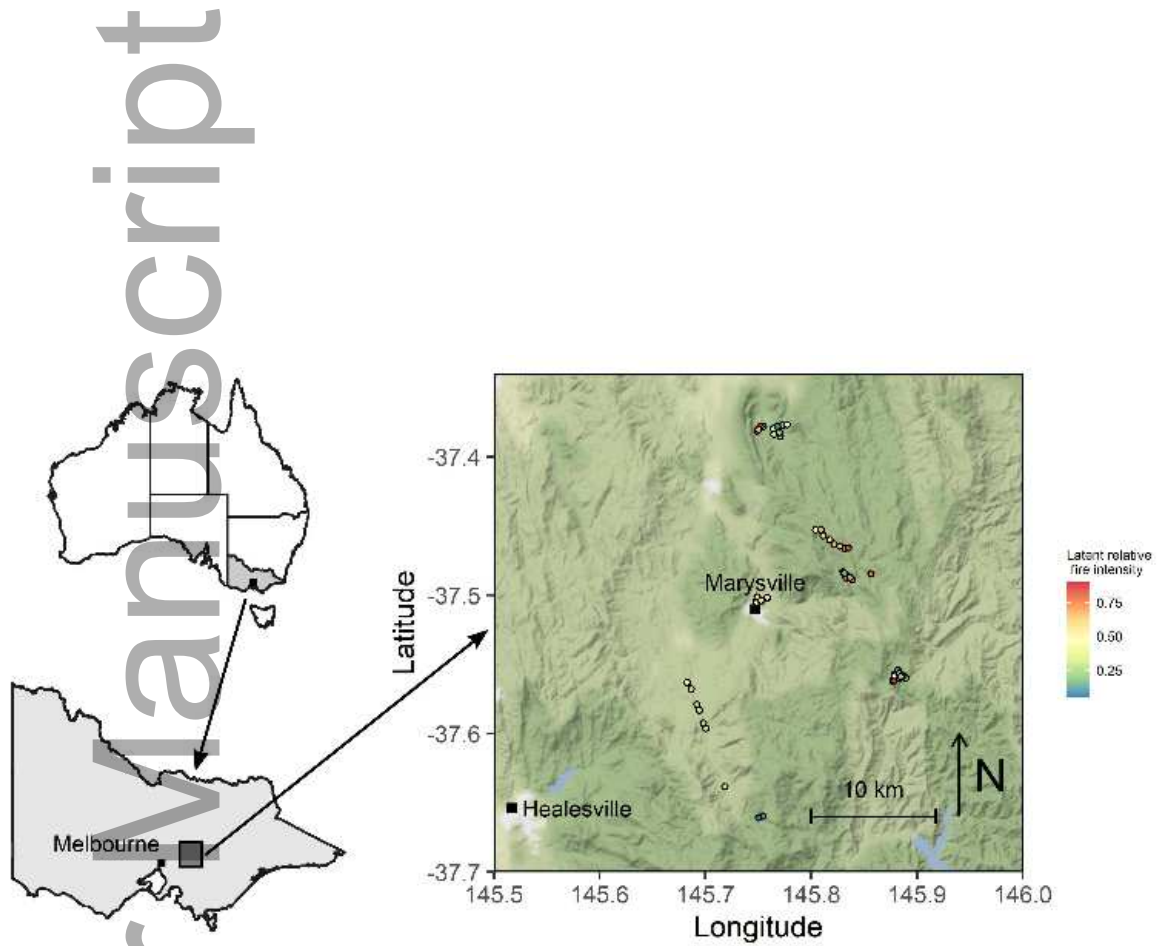


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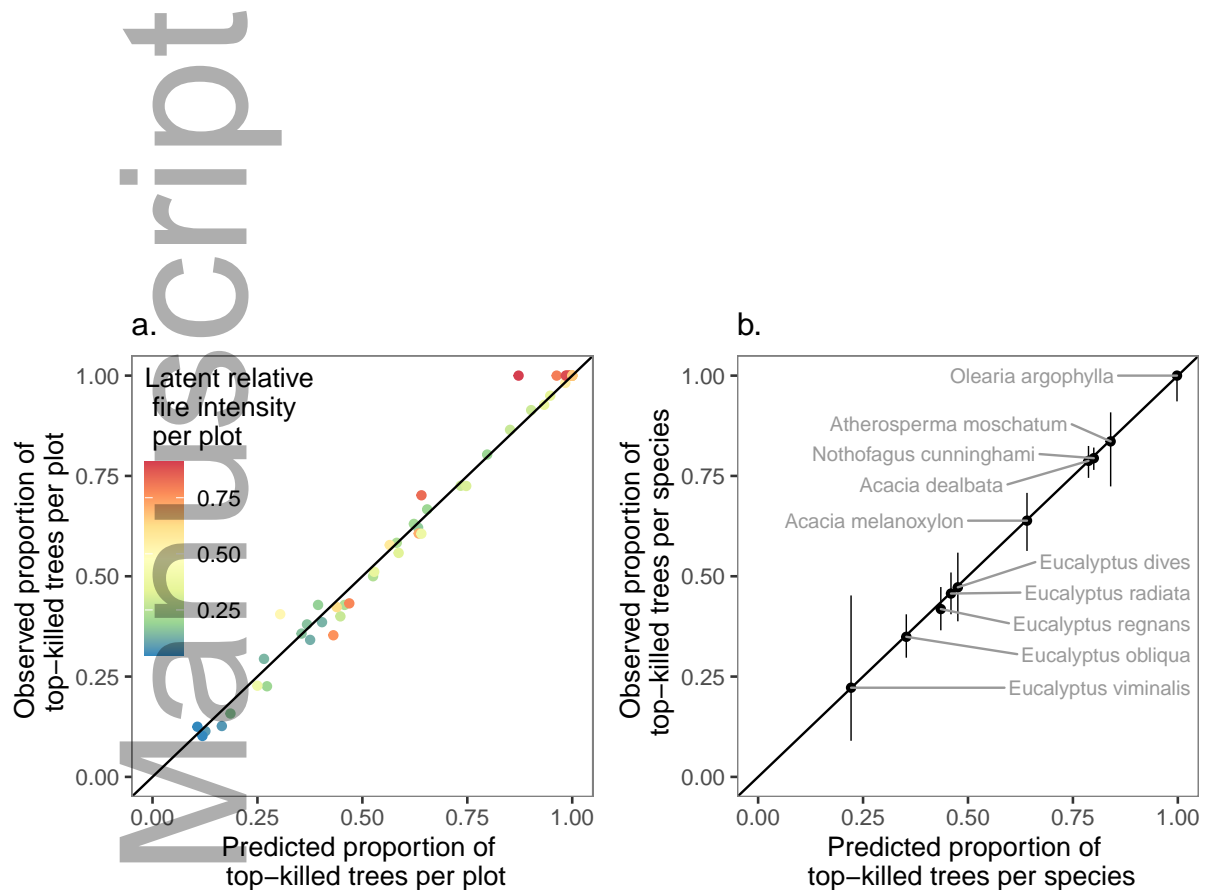


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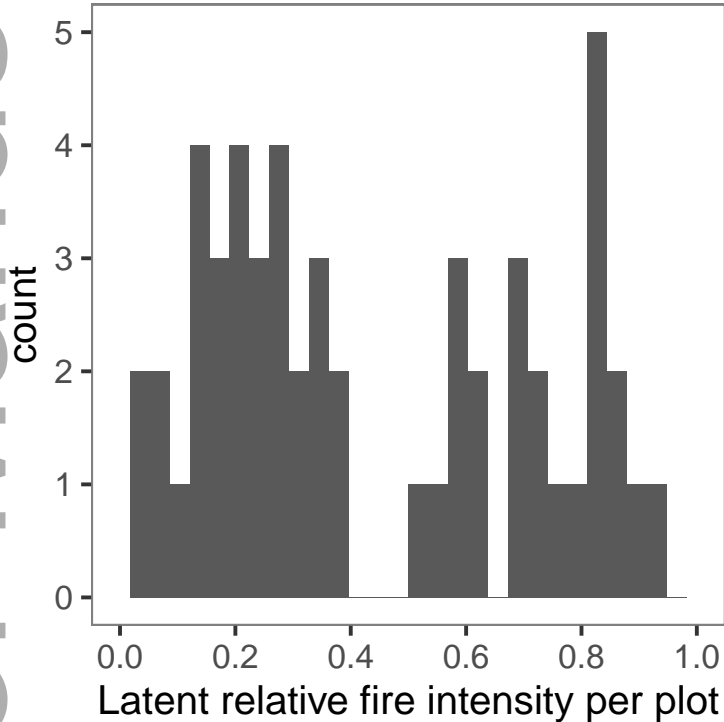


Figure 3: Distribution of latent relative fire intensity (I) scores for all plots ($n=53$) in the study.

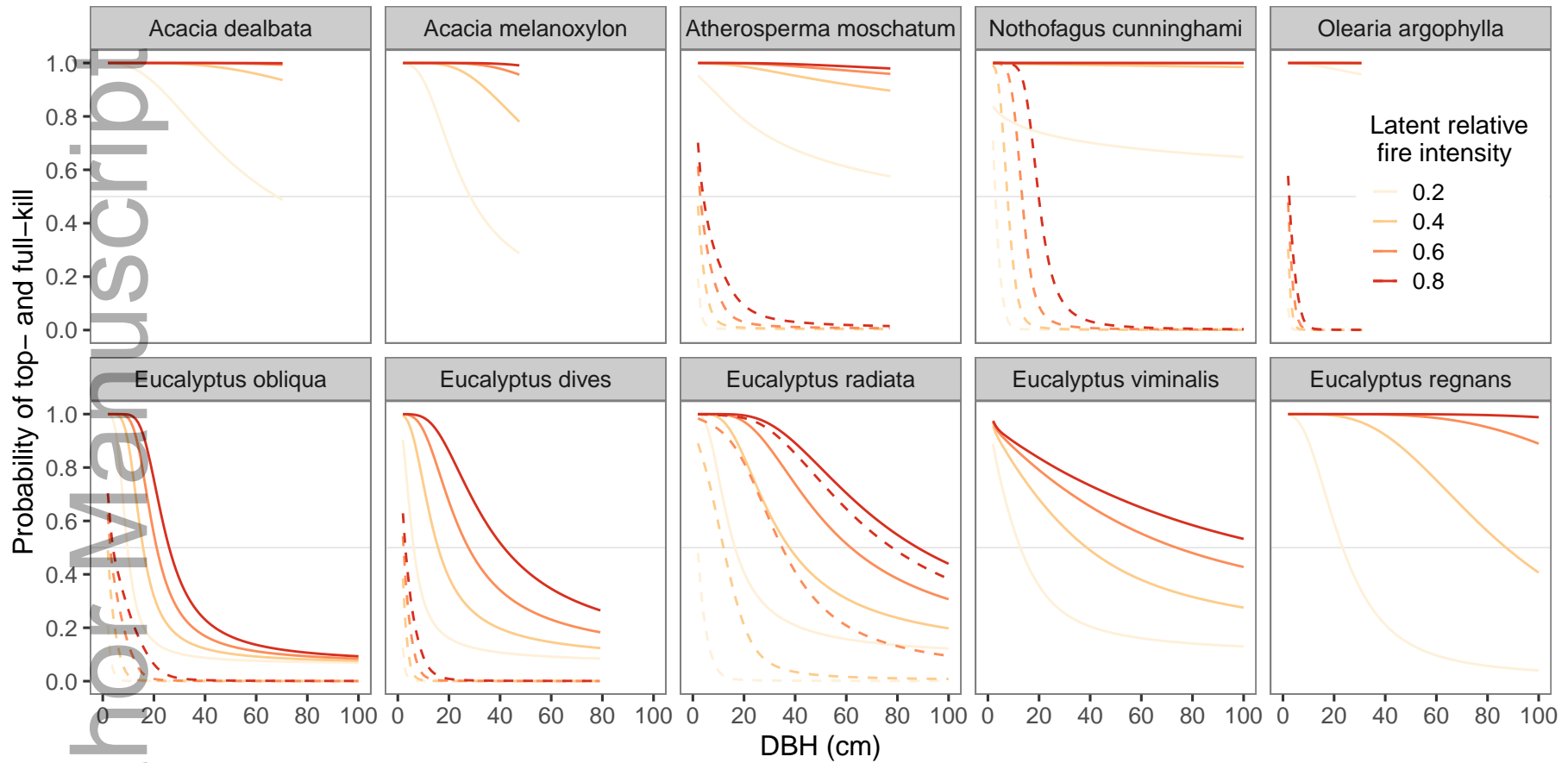


Figure 4: Susceptibility to fire-induced top- and full-kill as a function of species, tree size, and latent relative fire intensity for the 10 abundant tree species in the study. For those species that had >10 individuals resprouting, we calculated mortality with and without resprouts separately (see Methods for details). The solid lines represent results for the top-kill model; the dashed lines represent results for the full-kill model (top-killed individuals with basal resprouts are counted as surviving). The four relative fire intensities were selected to represent the range of estimated relative fire intensities.

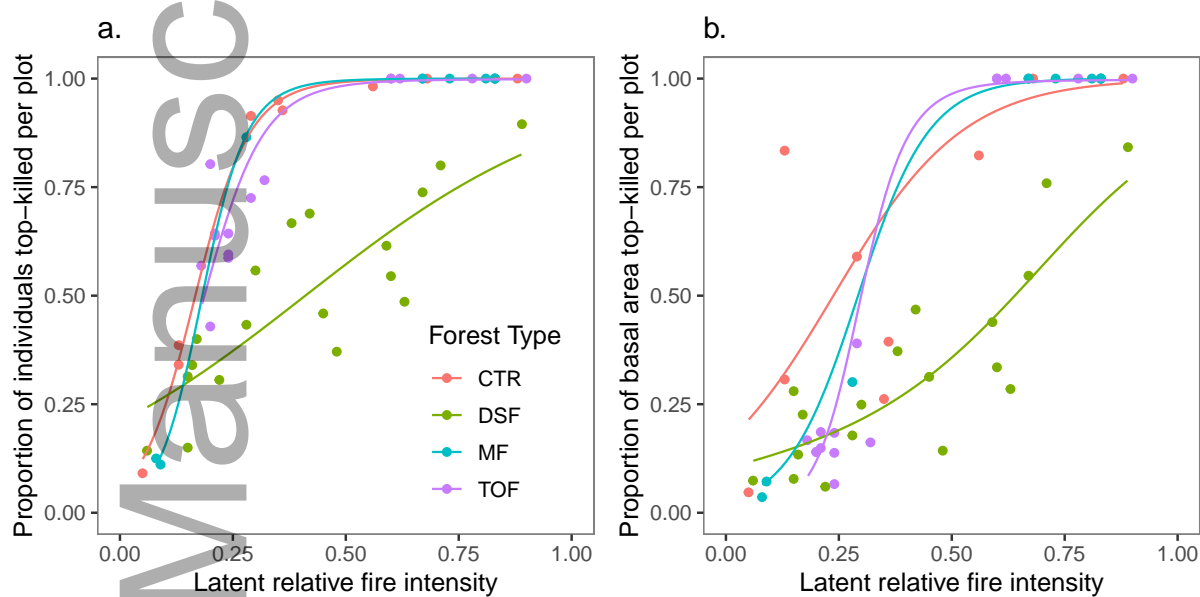


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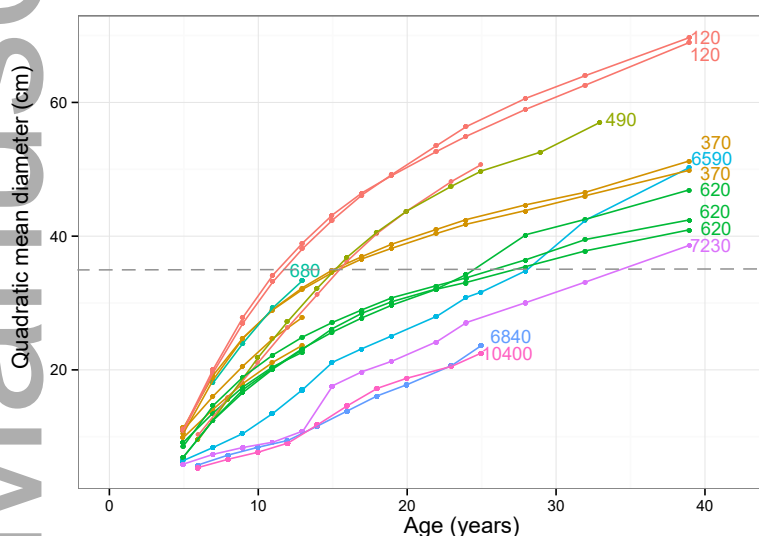
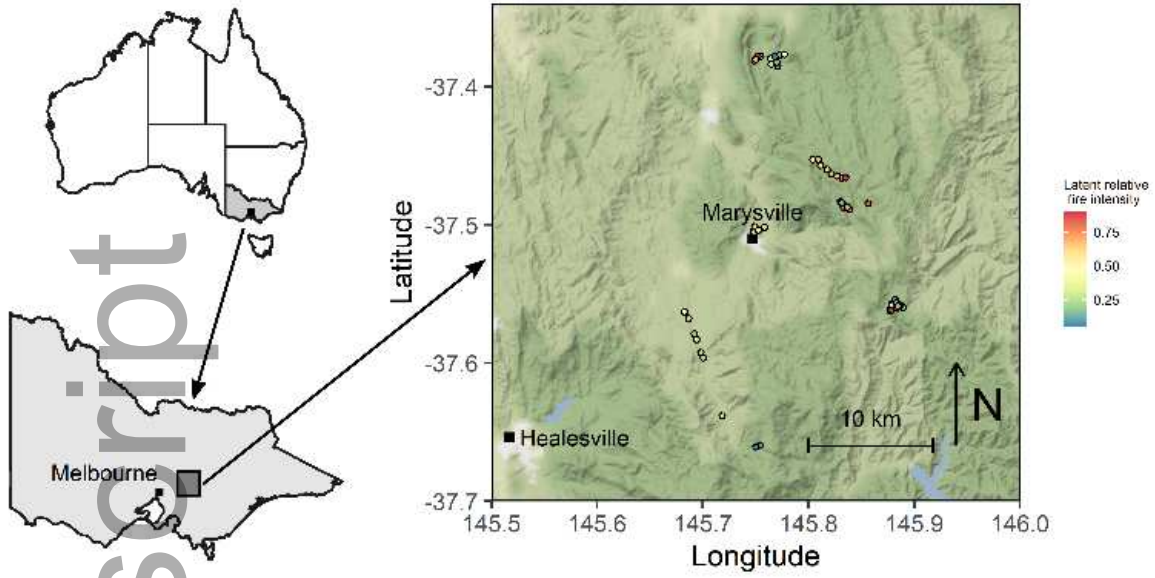
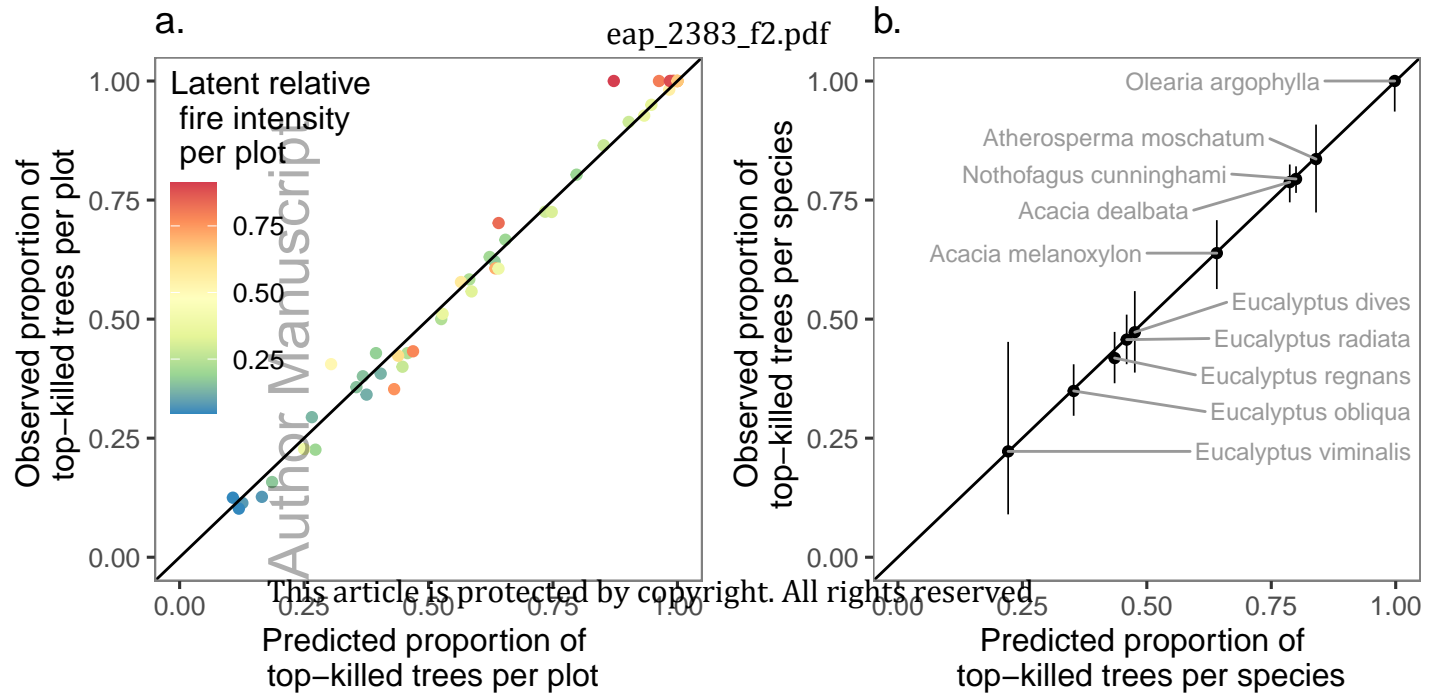


Figure 6: The effect of early stand density on diameter growth of *Eucalyptus regnans*. The stands were thinned at Year 4 to a range of residual densities (120-10400 trees ha⁻¹), shown to the right of each line. The y-axis values are plot-level quadratic mean diameters. The gray dashed line is the diameter (35 cm) at which mountain ash has a 50% probability of top-kill survival when $I=0.2$. The data are from long-term thinning trials originally initiated by the Victorian Forest Commission in the 1950s and 60s and subsequently maintained and managed by the Victorian Department of Environment, Land, Water, and Planning and VicForests.



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