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Flammability dynamics in the Australian Alps

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Abstract

Forests of the Australian Alps (SE Australia) are considered some of the most vulnerable to climate change in the country, with ecosystem collapse considered likely for some due to frequent fire. It is not yet known, however, whether increasing fire frequency may stabilize due to reductions in flammability related to reduced time for fuel accumulation, show no trend, or increase due to positive feedbacks related to vegetation changes. To determine what these trends have been historically, dynamics were measured for 58 years of mapped fire history. The 1.4 million ha forested area was divided into broad formations based on structure and dominant canopy trees, and dynamics were measured for each using flammability ratio, a modification of probability of ignition at a point. Crown fire likelihood was measured for each formation, based on satellite-derived measurements of the 2003 fire effects across a large part of the area. Contrary to popular perception but consistent with mechanistic expectations, all forests exhibited pronounced positive feedbacks. The strongest response was observed in tall, wet forests dominated by Ash-type eucalypts, where, despite a short period of low flammability following fire, post-disturbance stands have been more than eight times as likely to burn than have mature stands. The weakest feedbacks occurred in open forest, although post-disturbance forests were still 1.5 times as likely to burn as mature forests. Apart from low, dry open woodland where there was insufficient data to detect a trend, all forests were most likely to experience crown fire during their period of regeneration. The implications of this are significant for the Alps, as increasing fire frequency has the potential to accelerate by producing an increasingly flammable landscape. These effects may be semi-permanent in tall, wet forest, where frequent fire promotes ecosystem collapse into either the more flammable open forest formation, or to heathland.

Disciplines

Medicine and Health Sciences | Social and Behavioral Sciences

Publication Details

Zylstra, P. (2018). Flammability dynamics in the Australian Alps. *Austral Ecology: a journal of ecology in the Southern Hemisphere*, 43 (5), 578-591.

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3 **Abstract**

4 Forests of the Australian Alps (SE Australia) are considered some of the most vulnerable to
5 climate change in the country, with ecosystem collapse considered likely for some due to
6 frequent fire. It is not yet known, however, whether increasing fire frequency may stabilise
7 due to reductions in flammability related to fuel accumulation, show no trend, or increase due
8 to positive feedbacks related to vegetation changes. To determine what these trends have
9 been historically, dynamics were measured for 58 years of mapped fire history. The 1.4
10 million ha forested area was divided into five broad formations based on structure and
11 dominant canopy trees, and dynamics were measured for each using Probability of Ignition at
12 a Point, modified to minimise noise from small outliers. Crown fire likelihood was measured
13 for each formation, based on satellite-derived measurements of the 2003 fire effects across
14 the NSW and ACT Alps. Contrary to popular perception but consistent with mechanistic
15 expectations, all forests exhibited pronounced positive feedbacks. The strongest response was
16 observed in tall, wet forests (TWF) dominated by Ash-type eucalypts, where despite a short
17 period of low flammability following fire, regenerating stands have been more than eight
18 times as likely to burn than have mature stands. The weakest feedbacks occurred in open
19 forest (OF), although regenerating forests were still 1.5 times as likely to burn as mature
20 forests. Aside from low, dry open woodland where there was insufficient data to detect a
21 trend, all forests were most likely to experience crown fire during their period of
22 regeneration. The implications of this are significant for the Alps, as increasing fire frequency
23 has the potential to accelerate by producing an increasingly flammable landscape. These
24 effects may be semi-permanent in TWF, where frequent fire promotes ecosystem collapse
25 into either the more flammable OF formation, or to heathland.

26 **Key-words** Australian Alps, climate change impacts, ecosystem collapse, flammability

27 INTRODUCTION

28 The fire season is both lengthening and becoming more severe in Australia, in line with
29 global trends (Jolly *et al.* 2015; Flannigan *et al.* 2013; Clarke *et al.* 2013). Coupled with the
30 increase in lightning activity (Reeve and Toumi 1999), this has the potential to greatly
31 increase the impact of fire on forest and other ecosystems (Krause *et al.* 2014) through
32 ‘interval squeeze’ (Enright *et al.* 2015), with consequent feedbacks into the carbon cycle (e.g.
33 (Fisher *et al.* 2016)). Although much of the focus of this impact has gone to climate; the
34 flammability dynamics of ecosystems may play a comparable role. These have the capacity to
35 either reduce or amplify the effect of external drivers, yet they are poorly understood, and our
36 lack of knowledge in this area represents a major barrier to understanding the earth system
37 (Harris *et al.* 2016).

38

39 If regenerating forests are on average less flammable than mature forest, the system remains
40 stable when external drivers increase fire frequency, because those pressures are balanced by
41 reduced flammability (negative feedback). Conversely, where the feedback is positive and
42 regenerating forest is on average more flammable than mature forest, the frequency and/or
43 severity of fire is amplified. This makes the ecosystem more vulnerable to collapse (Keith *et*
44 *al.* 2013; Lindenmayer *et al.* 2011), while increasing the threat to human values. Positive
45 feedbacks therefore have the potential to cause tipping points in both the likelihood and
46 consequence components of wildfire risk.

47

48 *The Australian Alps*

49 The Australian Alps require consideration in this regard, as they are vulnerable to both
50 bioclimatic and abiotic changes, and as an ‘elevationally restricted mountain ecosystem’ they
51 fall into the category of first concern for Australian ecosystems at risk from climate change

52 (Laurance *et al.* 2011). Using the criteria of this panel, the susceptibility of the Australian
53 Alps is increased by the dependence of some montane and subalpine forests on cloud-
54 stripping for moisture inputs (Costin 1961), the declining seasonal snow cover of the
55 mountains (Sánchez-Bayo and Green 2013; Davis 2013), and their value as habitat to high
56 numbers of restricted endemic species (Pickering *et al.* 2004). The alpine area is of particular
57 concern in this regard (Hughes 2010).

58

59 As described earlier, fire can operate to amplify change through feedbacks between changing
60 fire regimes and landscape flammability. In the montane zone, (Bowman *et al.* 2014) have
61 measured localised ecosystem collapses produced by fire in tall, obligate-seeding Alpine Ash
62 (*Eucalyptus delegatensis* subsp. *delegatensis*, Myrtaceae) forests. Although the likelihood of
63 total ecosystem collapse has not yet been measured for these forests, there is significant risk
64 of total collapse for some alpine communities such as snow patch herbfield or feldmark
65 (Green and Pickering 2009; Williams *et al.* 2015). The risk of this occurring due to shrub
66 encroachment was considered greater under changed fire regimes, and increases in fire
67 frequency are likely not only because of the changing climate, but are amplified by the
68 positive feedbacks in the surrounding alpine (Camac *et al.* 2017) and lower sub-alpine
69 communities (Zylstra 2013) through which fire generally enters the alpine area. Fire can also
70 act as a catalyst for climate-driven change along ecotones between formations, as observed in
71 Yellowstone National Park (Donato *et al.* 2016). The strength and direction of feedbacks in
72 the wider forested area of the Alps may then be critical in either mitigating this risk by
73 providing a buffer that loses flammability when fire becomes more frequent, or compounds it
74 by increasing landscape connectivity and fire contagion as they are burnt more often.

75

76

77 *Defining flammability*

78 Such concerns however may for some be counterintuitive in Australian eucalypt forests,
79 which are frequently described as ‘fire adapted’, with arguments for deliberate management
80 increases in fire frequency coming from both political and some academic sources (Attiwill
81 and Adams 2013; Baker and Catterall 2016; Teague *et al.* 2010). While debate continues over
82 the ecological veracity of this characterisation (Bradshaw *et al.* 2011), care is needed that
83 ecological requirements do not become conflated with plant or stand flammability trends.

84

85 The concept of flammability is frequently misunderstood or poorly applied, and this is largely
86 due to the lack of clear definition and objective units of measurement. At the heart of this is
87 the reality that the three components of ignitability, combustibility and sustainability operate
88 either independently or as counter-measures to each other (Gill and Zylstra 2005). A fuel that
89 burns quickly for example is highly combustible, but has low sustainability. As illustrated in
90 (Zylstra *et al.* 2016), if other factors are equal, a thin leaf is more readily ignitable than a
91 thick one, but has lower sustainability. If a plant is composed from thin leaves however, it is
92 more likely that ignition will spread through it rapidly, which may result in a more
93 combustible plant; or in a less combustible one if some burning leaves expire before the rest
94 of the plant ignites. One level of flammability therefore emerges from its components, so that
95 measures are not readily scalable and it is not valid to say that a plant or an ecosystem is
96 more or less flammable due to measured components, unless that complexity is modelled.
97 Some of this ambiguity in plant flammability was succinctly captured in the “non-flammable,
98 the fast-flammable and the hot-flammable” categories described by (Pausas *et al.* 2017).

99

100 Reference to flammability at the scale of a stand, an ecosystem or a landscape then has the
101 potential to become meaningless. Tall forests can at times burn with extraordinary intensity,

102 so that the word “flammable” appears an apt description. But if this has a natural return
103 interval of centuries, can it objectively be considered more or less flammable than the
104 grassland that burns every few years? Such questions emphasise the need to reference
105 flammability as both an emergent, and relative property of the ecosystem. Emergent because
106 of the complexity producing the behaviour (Zylstra 2011), and relative because its definition
107 depends on the way that fire relates to the components of the ecosystem. The difference
108 between two different fire behaviours is only of importance to a life if one of those
109 behaviours will not harm it; otherwise the distinction loses significance. This is expressed in
110 the consideration of flammability as a biological concept (Pausas and Moreira 2012), where it
111 is measured relative to the influence that resultant flames exert in niche construction,
112 population dynamics, and other evolutionary and ecological mechanisms. This can be
113 quantified for a point using measures of severity, and for a landscape, using Probability of
114 Ignition at a Point (PIP, e.g. (Gill *et al.* 2000; de Ligt 2005)). In risk terms, these express
115 consequence and likelihood.

116

117 *Aims*

118 In order to better define the risk posed to the Australian Alps through fire, the aim of this
119 paper is to quantify the fire-flammability feedbacks across all broad forest formations of the
120 Australian Alps.

121

122 **MATERIALS AND METHODS**

123 *Location*

124 The study was conducted in the forested part of the Australian Alps bioregion contained
125 within the Australian Alps National Parks. This covers 1.4m ha across the states of New
126 South Wales and Victoria along with the Australian Capital Territory.

127

128 The environment is highly diverse, largely due to topography. Elevation ranges from close to
129 sea level in the south east, to the highest forested areas in Australia at 1960m.a.s.l., and the
130 area spans two temperate climatic divisions with distinctly dry summers ranging from mild to
131 hot (Stern *et al.* 2000). Mean annual rainfall ranges from 500mm to over 2300mm, and snow
132 can persist for up to six months in areas above 1400m.a.s.l. (NSW Office of Environment and
133 Heritage 2016; Clayton-Greene and Ashton 1990). Pre-European fire return-intervals ranged
134 from multiple decades to centuries (Zylstra 2006).

135

136 Forests were broadly grouped in formations divided by structure (Specht 1970) and dominant
137 tree species, combining units from Victorian Ecological Vegetation Groups (Department of
138 Sustainability and Environment 2010) and NSW/ACT Formations (Gellie 2005). Mapping is
139 not consistent between the two systems, but the formations shared the same dominant canopy
140 species, and mapped patches matched reasonably well where they bordered. These formed
141 one tall wet forest (TWF), one open forest (OF), one subalpine open forest-woodland (SFW),
142 one dry open forest (DOF), and one low, dry open woodland (LDOW, Table 1, Fig. 1). Of
143 these, previous analysis has demonstrated positive feedbacks in SFW (Zylstra 2013, 2016)
144 and the *Eucalyptus regnans* (Myrtaceae) component of TWF (Taylor *et al.* 2014). Consistent
145 with increases in local fire weather severity (Clarke *et al.* 2013), fire frequency in the study
146 area has increased sharply in the 21st century (Fairman *et al.* 2016). As a result, much of this
147 formation is expected to be lost over the remainder of this century (Bowman *et al.* 2014;
148 Ferguson 2010). No feedbacks have been quantified for the remaining area.

149

150 INSERT FIG. 1 AROUND HERE

151

152 *Fire History*

153 The primary data used for all fires mapped for the 58 years from 1957 to 2015 were records
154 sourced from State Government agencies (New South Wales Office of Environment and
155 Heritage, Australian Capital Territory Emergency Services Agency, and the Victorian
156 Department of Environment and Primary Industries, unpublished data). These were checked
157 for duplicate polygons, and fires that had not been entered from the archived paper records
158 were added to the dataset from that used by (Zylstra 2013). Some fires had been mapped
159 prior to this period, however the quality and frequency of this was inconsistent, rendering the
160 data too poor for analysis.

161

162 *Analysis 1: Flammability feedbacks*

163 Flammability feedbacks were determined by relating time since fire to two measures of
164 Probability of Ignition at a Point (PIP) – probability of a cell burning, and probability of
165 crown fire within a burning cell. The overall feedback was measured from the first of these,
166 whereas the second was used to provide further insight.

167

168 The required inputs for each analysis were collected in ArcGIS (ESRI 2015) from 1ha grid
169 cells, and analysis performed in the R Statistical Environment (R Core Team 2016).

170

171 To find the probability of a cell burning, PIP was calculated using Flammability Ratio FR
172 (Zylstra 2013), as this includes a scaling factor to account for noise produced by small age-
173 class samples. Simple PIP is defined as the proportion of a given stand age that is burnt. But
174 if a certain age is only represented by a small pocket, a large fire may easily consume all of it
175 and give a probability of one, regardless of its characteristic flammability. To account for
176 this, FR scales the influence of each fire year on the final statistic by producing the Area

177 Factor (AF , Equation 1), which is the standard PIP (left hand term) multiplied by the ratio of
178 average age class area $\bar{x}A$ to total burnt area Σ_b (right hand term). Years in which the area of
179 fire is large compared to the average age class area are scaled down in influence, and those
180 years where fires are smaller are given greater influence.

181

$$AF = \frac{A_b}{A} \cdot \frac{\bar{x}A}{\Sigma_b}$$

182

Equation 1.

183

184 Where A_b is the area of the age class burnt, and A is the total area of that age class.

185

186 FR for a given age and year is equal to the AF divided by the average of all AF values across
187 all ages and years, and the FR for that age across all years is the average of all FR values
188 collected for that age.

189

190 In keeping with (McCarthy *et al.* 2001) and the introductory comments on flammability, FR
191 incorporates the term in preference to ‘hazard’ or ‘probability’, as the probability of ignition
192 at a point represents both an emergent and a relative outcome. In a landscape such as the
193 Alps, where fire is actively suppressed, there is greater probability of a point burning if fires
194 not only spread faster, but are more difficult to suppress due to factors such as larger flames
195 and / or more spot fires over greater distances. These behaviours are all aspects of
196 flammability at a landscape level (Gill and Zylstra 2005), and FR combines them into the
197 single emergent feature of measured burnt area.

198

199 Spatial autocorrelation is likely within a given year, as fire burns discrete patches rather than
200 random locations. This can however be disregarded when multiple years are examined, as
201 new patches are formed in each year.

202

203 *Analysis 2: Likelihood of crown fire ignition*

204 To find the likelihood of crown fire L_{cb} in any burning cell, crown fire measurements were
205 made from difference Normalised Burn Ratio dNBR (e.g. (Keeley 2009)), mapped for a
206 632,448ha subset of the 2003 bushfires in NSW and the ACT (Barrett 2006), using only the
207 highest class of severity to denote crown fire.

208

209 As PIP was only measured for a single year, data could not be corrected via the FR, which
210 weights values by comparison between years. The smaller dataset was therefore statistically
211 weaker than that used in the first analysis, due to the possibility of small, highly influential
212 outliers burnt in this very large fire, and to spatial autocorrelation.

213

214 To account for noise from small outliers, these were identified using Grubbs test (Grubbs
215 1950) via the ‘outliers’ package in R (Komsta 2011), and the largest outlier for each
216 formation was removed if $p < 0.05$ for Grubbs’ statistic. To limit the loss to data, a maximum
217 of one outlier constituting less than 0.5% of the area of each formation was removed.

218

219 To account for spatial autocorrelation, severity measurements were made using a random grid
220 of cells that covered 75% of each formation.

221

222 *Fitting functions*

223 In order to find trends in the dynamics, a series of functions were fit to the data using the
 224 NLS package for R (Bates and Chambers 1992). Where more than one contender was
 225 identified, the best was chosen by comparing the Akaike Information Criterion (AIC (Akaike
 226 1974)) using Akaike weight (Symonds and Moussalli 2011) to give the probability that this
 227 was the best approximating model. For the smaller datasets used in the second analysis, AIC_c
 228 was used (Symonds and Moussalli 2011).

229

230 All dynamics were assumed to commence with a period of low, increasing flammability,
 231 where ground fuels are absent immediately following fire. Five functions were used for this
 232 purpose (Fig. 2) - the Olsen function ((Olsen 1963), Equation 3) and a logistic function
 233 (Equation 4) describe negative feedbacks, however the logistic function allows for a longer
 234 period of initial reduced flammability compared to Olsen. The other functions ((Burr 1942),
 235 Equation 5), the standard binomial distribution (Equation 6) and the Moisture function
 236 (Equation 7, (McCarthy *et al.* 2001)) allow for positive feedbacks; with Burr capable of
 237 representing both.

238

$$FR = a(1 - e^{-b.T})$$

239

Equation 3

$$FR = \frac{K}{1 + e^{(a-r.T)}}$$

240

Equation 4

$$FR = ab \frac{0.1T^{a-1}}{(1 + 0.1T^a)^{b+1}}$$

241

Equation 5

$$FR = \frac{sc}{s\sqrt{2\pi}} \cdot e^{-\frac{(-T-\bar{x})^2}{2s^2}}$$

242

Equation 6

$$FR = a(1 - e^{-bT})(c + e^{-dT})$$

243

Equation 7

244

245 Where T is years since the last fire, a to d are constants, r is the biotic rate of increase, K is an
246 asymptote, sc is a scaling factor, s is the standard deviation, and \bar{x} is the mean.

247

248 INSERT FIG. 2 AROUND HERE

249

250 Models were excluded if they could not be fit using NLS, or if when fit, all constants did not
251 have $p < 0.01$. When this occurred in the FR analysis, outliers were successively identified
252 using Grubbs test (Grubbs 1950; Komsta 2011) and subsequently removed before re-
253 examination. As this dataset was much larger than that used in the second analysis; rather
254 than set an arbitrary limit to the size of outliers, the percentage burnt area that was removed
255 was reported as qualifying data. As already described, the largest outlier in the dataset used
256 for the second analysis was routinely removed, if it constituted less than 0.5% of the
257 formation area.

258

259 A model was chosen if it achieved significance, and where more than one did so, if Akaike
260 weight was greater than 0.5. If no model satisfied these criteria, the null hypothesis was not
261 rejected.

262

263 In order to best represent the data, any FR model chosen that did not achieve the highest level
264 of significance ($p < 0.001$) was again examined for outliers. If these outliers constituted only
265 a very small part of the burnt study area and did not change support from one alternate

266 hypothesis to another, they were removed until maximum significance was achieved, if that
267 was possible. Again, the area of forest removed as outliers was reported to provide
268 qualification to these models.

269

270 *Measuring feedbacks*

271 To determine feedbacks in the first analysis, the null hypothesis H_0 of no feedback (i.e.,
272 flammability remains constant) was tested against two alternate hypotheses of negative
273 feedback H_1 and positive feedback H_2 . Feedbacks were defined from FR, as this provided the
274 most comprehensive measure of flammability. These were termed positive if the average
275 flammability of mature forests was less than the average taken across the whole period of
276 regeneration, negative if mature forests were more flammable, or no feedback if no model
277 could be fit, or if the null model provided the best fit.

278

279 Flammability dynamics in the Olsen and logistic models were divided into a young period
280 (Y), in which modelled $FR < 1$, and a mature period (M) of all ages beyond this. For the other
281 functions where FR dropped below unity at older age classes, Y was calculated, followed by
282 a regrowth period (R) where modelled $FR \geq 1$, then a mature period of all ages where FR was
283 once again < 1 . The mean FR for each of these periods was measured from the raw data, after
284 the removal of any outliers.

285

286 The strength of each feedback was quantified as feedback strength (FS), which is the mean
287 FR for non-mature forest (Y, or Y+R depending on the function) divided by the mean FR for
288 M. Where the feedback was positive, young equivalent (YE) was calculated as the number of
289 years for which the mean modelled flammability of regrowing forests was less than or equal
290 to the mean flammability of mature forests.

291

292 For both analyses, the strength of the best model for each community was indicated by
293 residual standard error for the full datasets, and by R^2 . For FR, R^2 was measured against
294 mean FR values for each age in preference to raw values, to minimise the noise arising from
295 inter-annual stochastic differences and thereby providing a better representation of the
296 feedback.

297

298 *Final models*

299 To find the annual likelihood of fire L_f in each age and formation, the relevant FR models
300 were multiplied by the mean likelihood of fire L across the formation being studied (Equation
301 8).

302

$$L = \frac{\Sigma_{bY}}{\Sigma_{AY}}$$

303

Equation 8.

304 Where Σ_{AY} is the sum of all areas of known age for all years, and Σ_{bY} is the sum of all burnt
305 areas within those known ages, for all years.

306

307 To find the annual likelihood of crown fire L_c in each age and formation, the relevant models
308 for likelihood of crown fire in a burning stand L_{cb} were multiplied by L_f .

309

310 **RESULTS**

311

312 *Flammability feedbacks*

313 Flammability Ratio could be measured for the 52 years following 1964 in all formations
314 except LDOW, where the first fire to burn forest of a known age was in 1965 (Appendix S1).

315

316 The only significant models in all formations were Burr and the binomial distribution (Table
317 2, Fig. 3), with Burr providing the strongest fit of the two in all cases. DOF, SFW, and TWF
318 (Figs. 3a, d and e) all had the highest category of significance for the study ($p = 0.0001$) when
319 modelled from the full dataset using Burr. OF was significant with the full dataset ($p = 0.01$),
320 but removal of two outliers constituting 0.007% of the studied burnt area increased this to $p =$
321 0.0001 (Fig. 3c). LDOW was only significant if the largest outlier was removed, but removal
322 of another two outliers constituting 0.183% of the burnt area produced a model with $p =$
323 0.0001.

324

325 INSERT FIG. 3 AROUND HERE

326

327 Further outliers were apparent in some formations such as OF, but were not removed if a
328 function had already been fit with the highest category of significance.

329

330 Feedback strength was greater than one for all formations, indicating that flammability of
331 regenerating forests was higher than that of mature forests.

332

333 *Likelihood of crown fire ignition*

334 The number of age classes burnt in 2003 and for which data was available varied between 20
335 (LDOW) and 40 (TWF), allowing functions to be fit to all formations except for LDOW
336 (Table 3, Appendix S2). The trend was binomial in three cases, and followed a Burr curve for
337 DOF due to a weak trend of more crown fire in the first two decades of post-fire recovery
338 (Fig. 4). Although this model was weakly significant, it had little explanatory power ($R^2 =$
339 0.01).

340

341 INSERT FIG. 4 AROUND HERE

342

343 *Final models*

344 The mean likelihood of fire per annum L varied only slightly between formations when all
345 ages were combined (Table 4), however the formations OF and SFW differed notably from
346 the others when age effects were taken into account by finding L_f (Fig. 5a.). This pattern
347 varied again when likelihood of crown fire at a burning site L_{cb} was considered (Fig 5b), and
348 as a result, the combined function L_c produced large differences in the annual likelihood of
349 crown fire at a point (Fig 5c).

350

351 INSERT FIG. 5 AROUND HERE

352

353 **DISCUSSION**

354 Contrary to widely held perceptions of eucalypt forests, feedbacks in the study area have
355 been pronounced and positive in all formations over the past half century. Forest stands have
356 burnt 1.5 (OF) to 8.3 (TWF) times more often in regenerating forest than in mature forest,
357 and crown fires appear to have been mostly confined to regenerating stands.

358

359 The degree to which stand age drove flammability varied between formations. The highest R^2
360 was measured for DOF, where stand age has contributed more than one third of the
361 variability in fire size, regardless of weather and other effects. The lowest R^2 was measured
362 for OF and LDOW, reflecting the influence of a small number of remaining outliers rather
363 than a lack of general trend (Figs. 3b & 3c.), however these were not removed as the model
364 had already been successfully fit at the highest level of significance.

365

366 *Driving mechanisms*

367 These findings are consistent with broad mechanistic expectations that have been proposed
368 for angiosperm forests (Bond and Midgley 2012), and challenge the expectation of a negative
369 feedback due to fuel accumulation. The “angiosperm revolutions” model of Bond and
370 Midgley (2012) points to the characteristic formation of a gap between ground and canopy
371 fuels in mature forests, which acts to prevent crown fire ignition and leaves an intact canopy
372 to slow wind speeds and resulting fire spread at ground level.

373

374 The theory that these aspects of plant structure and spatial relationships drive flammability
375 was compared to standard expectations from fuel load, using the biophysical, mechanistic
376 Forest Flammability Model (FFM, (Zylstra 2011; Zylstra *et al.* 2016)). By validating both
377 approaches against flame heights measured from burn patterns across a large area of the
378 Australian Alps burnt in 2003, the model found that fuel load explained only 11% of the
379 observed variability in flame height, whereas the addition of plant structure and species’
380 composition increased the R^2 to 0.80 (Zylstra *et al.* 2016). When the FFM was used to model
381 SFW flammability dynamics from plant species and structure, the positive feedback
382 measured from FR was reproduced. In contrast, modelling using the fuel load-based approach
383 of McArthur (Noble *et al.* 1980; McArthur 1967) falsely predicted a negative feedback
384 (Zylstra 2013). This model has long been the primary industry-standard approach used for
385 bushfire in eastern Australia (Gould and Cruz 2012).

386

387 Fire-induced factors that increased flammability in these studies included the loss of canopy
388 protection for low vegetation, the germination and/or fertilisation of understorey plants, and
389 the loss of a gap between ground and regenerating canopy fuels. Such patterns of regrowth

390 are common in the forests of the Alps, whether due to resprouting or the regrowth of trees
391 from seed (Fig. 6a., b.). Feedback strength is likely to derive from the degree to which these
392 factors change between regenerating and mature forests, so that the strongest feedbacks are
393 evident in forests that change from dense regrowth close to the ground, to mature stands with
394 large separations between dense tree crowns and the lower fuels.

395

396 INSERT FIG. 6 AROUND HERE

397

398 An implication of this is that less severe fire impacts may produce weaker feedbacks. This
399 may be the case where fires do not kill the crown either through scorch or cambium heating,
400 do not promote growth close to the ground, or do not by some other mechanism such as
401 promotion of epicormic growth remove the gap between ground and canopy fuels. It cannot
402 yet be determined whether such fires can change the direction of feedback by creating a
403 regeneration period that is less flammable than the mature period, but they may produce
404 weaker positive feedbacks. The strength of this effect will depend on the formation in
405 question, as structure and species' sensitivity to fire could have overriding effects (Fig. 6c. to
406 6f.).

407

408 It is notable that the peak in crown fire likelihood occurred later than the FR peak in the three
409 formations that had strong responses (c.f. Figs. 5a and 5b.). This pattern is similar to that
410 modelled for SFW, which predicted a later peak in the effects of flame height than in rate of
411 spread (Zylstra 2011, 2013). If this pattern holds for all communities examined here, it
412 suggests that FR and hence annual burnt area is driven more by spread rate than by difficulty
413 of suppression due to flame height.

414

415 In addition to the probable effect of fire severity, the analyses in this study are limited in that
416 they describe historical trends only. If changing atmospheric and climatic influences act to
417 vary species' dominance, structure or leaf traits such as thickness or chemistry, the
418 flammability of component species will also vary, with unforeseen effects on stand
419 flammability. For example, (Prior and Bowman 2014) found that large eucalypt species are
420 likely to grow more slowly in a warmer environment, and as the onset of the mature phase is
421 related to the formation of a sufficient gap between canopy and lower plants, this may act to
422 extend the flammable period of regeneration.

423

424 *Implications*

425 The findings of this study have significant implications for the Australian Alps, as they
426 demonstrate that positive flammability feedbacks are not only present in the alpine area, but
427 are pronounced and prevalent across the entire forested bioregion. Increases in fire frequency
428 can be expected to result in increased landscape flammability, with greater consequent
429 impacts on resident species, ecosystem processes and human values. As frequency increases,
430 connectivity will increase between areas of flammable landscape, accelerating the effect of
431 exogenous drivers and facilitating fire spread into previous refugia. In some formations, this
432 may push forests toward tipping points leading to ecosystem collapse and alternative stable
433 states.

434

435 Such concern has already been raised in regard to the Ash-type forests in the TWF formation
436 (Burns *et al.* 2015; Bowman *et al.* 2014), which are experiencing significant decline due to
437 frequent fire, and to logging in areas adjacent to Park. Two forms of transformation in these
438 forests fit the definition of ecosystem collapse (Keith *et al.* 2013). Firstly, where the canopy
439 is heavily dominated by the obligate-seeding Ash, successive crown deaths within the

440 primary juvenile period result in a distinct tipping point, and consequent transition to
441 shrubland. This transition has been observed in several locations since 2003 (Bowman *et al.*
442 2014; Wright and Robertson 2014). Secondly, where a resprouting sub-dominant canopy tree
443 such as *E. dalrympleana* is present in sufficient numbers, frequent fire leads to a shift in
444 formation from TWF to OF. This second form of collapse is potentially well advanced
445 already, although differences in survey technique introduce uncertainty in that determination.
446 Early surveys found that *E. delegatensis* almost entirely dominated Ash forest canopies in the
447 NSW Alps (Byles 1932), yet data (G. Wright and G. Robertson, unpublished data, 2014)
448 from recent surveys (Wright and Robertson 2014) showed dominance in only 68% of sites in
449 mapped Ash forest, with other species typical to the OF formation dominating the remaining
450 fraction.

451

452 Transition from TWF to OF introduces changes to landscape flammability in the Alps that
453 may be considered permanent on a scale of human management, as mature OF is more
454 flammable than mature TWF (Table 2). In terms of L_f and L_c (Fig. 5a and c) averaged for
455 these mature periods, this represents a doubling in the likelihood of fire at any point, and four
456 times the likelihood that any point will experience crown fire in a given year. This has
457 potential to increase fire contagion and transform the flammability of the entire bioregion.

458

459 *Concluding remarks*

460 This study constitutes the first examination of fire-flammability feedbacks for these forests,
461 and thereby provides a basis by which fire risk assessment and mitigation can transition from
462 assumed feedbacks to an evidential basis. Across the Australian Alps, recently burnt forests
463 have been on average more flammable than mature forests, consistent with the mechanistic
464 understanding based on plant growth and species' change. Increases in fire frequency are

465 therefore likely to create a more flammable landscape, with implications for both natural and
466 built assets. Drivers of post-fire succession such as fire severity or expected climatic and
467 atmospheric changes may vary the pattern or strength of these feedbacks, but this has yet to
468 be shown through either empirical measurement or mechanistic modelling.

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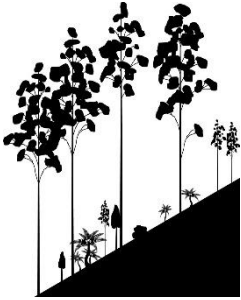



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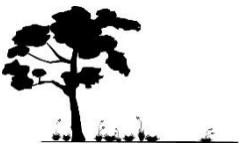
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618

619 **Table 1.** Forest formations within the study area, showing details of the component formations[†] in the New
 620 South Wales and Australian Capital Territory forests, followed by the Ecological Vegetation Community (EVC)
 621 groups[‡] in Victoria.

Mapped area (ha)	Elevation range (m.a.s.l.)	Canopy height (m)	Typical canopy species	Formations EVC Groups	Indicative structure
Tall Wet Forest (TWF)					
228888	<1400	20-50	<i>Eucalyptus delegatensis</i> , <i>E. regnans</i> , <i>E. dalrympleana</i> , <i>E. pauciflora</i> , <i>E. fastigata</i> (Myrtaceae)	Ash eucalypt forests, Rainforests Wet or damp forests, Rainforests	
Open Forest (OF)					
486891	700- 1300	20-35	<i>Eucalyptus dalrympleana</i> , <i>E. robertsonii</i> subsp. <i>robertsonii</i> , <i>E. macrorhyncha</i> , <i>E. bridgesiana</i> , <i>E. pauciflora</i> , <i>E. viminalis</i> , <i>E. rubida</i> subsp. <i>rubida</i> , <i>E. aggregata</i> , <i>E. stellulata</i> (Myrtaceae)	Moist eucalypt forests, Montane tableland forests, Swamp forests/sedgeland Montane shrublands, grasslands, or woodlands	
Subalpine Forest and Woodland (SFW)					
189598	1000- 1960	5-15	<i>Eucalyptus debeuzevillei</i> , <i>Eucalyptus niphophila</i> , <i>Eucalyptus pauciflora</i> (Myrtaceae)	Subalpine low forests Subalpine shrublands, grasslands or woodlands	
Dry Open Forest (DOF)					
					

			<i>E. macrorhyncha, E. rossii, E. dives, E. mannifera</i> (Myrtaceae)	Grass/shrub forests	
417005	300-1100	15-20		Dry forests	
Low, Dry Open Woodland (LDOW)					
			<i>Eucalyptus blakelyi, E. melliodora, E. bridgesiana, E. albens, E. polyanthemos subsp. polyanthemos</i> (Myrtaceae), <i>Callitris glaucophylla</i> (Cupressaceae)	Grassy woodlands/grasslands Lower slopes or hills woodlands	
94102	< 550	15-30			

TOTAL

1416586

622 †(Gellie 2005), ‡(Department of Sustainability and Environment 2010)

623

624 **Table 2.** Statistics for testing hypotheses and characterising feedbacks for the models that could be fit to the
 625 data.

Area burnt (ha)	Model	Model strength			Constants			Dynamics, Max age, \bar{x} FR			Feedback statistics	
		RSE	R ²	Adj. R ²	a / sc	b / sd	\bar{x}	Y	R	M	YE (years)	FS
Tall Wet Forest (TWF)												
141,255	Burr	10.01	0.16		2.340***	1.602***		3,	21,			
	Olsen				1.046***	0.575	0.98	1.54	0.17	0	8.3	
Open Forest (OF)												
338,432	Burr	15.59	0.07		2.324***	5.510*						
	Binomial				129.84	35.27	-12.50					
	Olsen				1.033*	0.846						
	Burr_O				7.39	0.01	2.221***	1.065***	6,	28,	0.73	4
							0.48	1.29				
Subalpine Forest & Woodland (SFW)												
85,693	Burr	8.50	0.09	0.63	2.230***	1.163***		6,	25,			
	Binomial				43.319**	11.982*	17.143***	0.81	1.31	0.50	3	2.3
	Olsen				1.066***	0.382						
Dry Open Forest (DOF)												
331,946	Burr	5.85	0.36	0.80	2.306***	1.671***		3,	19,			
	Binomial				31.240***	6.496***	11.415***	0.62	1.49	0.50	2	2.6
	Olsen				1.048***	0.631						
Low, Dry Open Woodland (LDOW)												
84,254	Burr	8.45	0.07		2.197***	0.521						
	Olsen				1.109**	0.403						
	Burr_O				2.303***	2.285***	2,	14,	0.67	2	2.0	
							0.67	1.50				

626

627 Models that could be fit using NLS are shown, regardless of their goodness of fit. The constants a and b fit
628 either Burr or Olsen as indicated, and sc , sd and \bar{x} fit the binomial model, showing significance for each
629 constant (*0.01, **0.001, ***0.0001). Where all constants were significant for more than one model, Akaike
630 Weight is shown for significant models to indicate the likelihood of the best approximating model. Where this
631 was >0.5 , the model was further refined for some formations with the removal of outliers, in which case this
632 adjusted model is shown with the suffix $_O$. Strength of the best model for each formation is given by Residual
633 Standard Error RSE, and R^2 .

634

635 The Dynamics columns divide the forest into the three age classes young Y (flammability ratio $FR < 1$),
636 regrowth R ($FR \geq 1$), and mature M (second period where $FR < 1$). The classes Y and R are described by
637 maximum age, and all classes show mean flammability in italics.

638

639 Two feedback statistics are given; these are young equivalent YE, which is the number of years after fire for
640 which the $FR \leq$ the mean FR for M, and feedback strength FS, which is the mean FR for regenerating forests (Y
641 and R) divided by the mean FR for M. In all cases, FS is greater than one, indicating that mean flammability of
642 regrowing forest is greater than the mean flammability of mature forests.

643

644

645 **Table 3.** Statistics for the models that could be fit to crown fire data.

Formation	Model	Model strength			Constants		
		Samples	RSE	R ²	a / sc	b / s	\bar{x}
TWF	Binomial	40	0.057	0.17	1.494 ^{***}	6.655 ^{***}	23.704 ^{***}
OF	Binomial	30	0.078	0.12	4.184 ^{***}	11.184 ^{***}	23.206 ^{***}
DOF	Mean	30	0.167	0.01	1.455 ^{***}	0.992 [*]	
SFW	Binomial	26	0.132	0.11	4.010 ^{***}	6.63 ^{**}	20.356 ^{***}
LDOW	Mean	20					0.107

646

647 Model strength is given by Residual Standard Error RSE and R². The constants *sc*, *s* and
 648 \bar{x} describe the binomial model (Equation 6), and asterisks indicate their significance (*0.01, **0.001, ***0.0001).

649 Where no model could be fit, the mean of the dataset is given as the Null value.

650

651

652 **Table 4.** Mean annual likelihood *L* of fire at a point in a formation.

Formation	<i>L</i>	Standard error
TWF	0.026	0.013
OF	0.025	0.014
DOF	0.032	0.012
SFW	0.023	0.019
LDOW	0.032	0.016

653

654

655 **Figure 1.** Location and forest formations composing the study area in south eastern Australia.

656

657 **Figure 2.** Contender functions used to describe flammability dynamics, as per equations 3 - 7

658

659 **Figure 3.** Modelled flammability ratio FR (curved lines), and mean FR values grouped into 5-year clusters. The
660 horizontal broken line shows $FR = 1$, so that values above the line represent ages that are more flammable than
661 average for that community, and those below the line are less flammable than average. The formations are DOF
662 (a), LDOW with outliers removed (b), OF with outliers removed (c), SFW (d), and TWF (e). Box plots show
663 standard quartile divisions for annual mean data, with outliers up to the default of 1.5 times the interquartile
664 range from the box.

665

666 **Figure 4.** Frequency of crown fire (the highest dNBR class) in decadal groupings of each formation. As per
667 (Hintze and Nelson 1998) violin plots are composed of box plots indicating data range, quartiles, and median,
668 and the shaped area shows the density trace.

669

670 **Figure 5.** Flammability trends for each formation, where the x -axis gives years since the last fire, and the y -axis
671 gives likelihood for (a) fire burning a point (L_p), (b) crown fire occurring if that point is burning (L_{cb}); and (c)
672 crown fire occurring at any point (L_c).

673

674 **Fig. 6.** Post-fire recovery of (a) DOF with a canopy both resprouting and reseeding, and (b) TWF regenerating
675 from seed alone. Both formations have produced abundant low fuel with little canopy protection. Severity
676 effects vary, depending on the formation. so that in OF, low severity fire can leave the canopy intact, but
677 stimulate ground growth (c), whereas high severity can change both factors (d). In SFW, both low (e) and high
678 (f) severity tend to kill the canopy and promote ground growth, although to differing extents.

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