



THE UNIVERSITY *of* EDINBURGH

Edinburgh Research Explorer

Clarifying the meaning of mantras in wildland fire behaviour modelling: reply to Cruz et al. (2017)

Citation for published version:

Mell, W, Simeoni, A, Morvan, D, Hiers, JK, Skowronski, N & Hadden, RM 2018, 'Clarifying the meaning of mantras in wildland fire behaviour modelling: reply to Cruz et al. (2017)' *International Journal of Wildland Fire*, vol. 27, no. 11, pp. 770-775. DOI: 10.1071/WF18106

Digital Object Identifier (DOI):

[10.1071/WF18106](https://doi.org/10.1071/WF18106)

Link:

[Link to publication record in Edinburgh Research Explorer](#)

Document Version:

Peer reviewed version

Published In:

International Journal of Wildland Fire

General rights

Copyright for the publications made accessible via the Edinburgh Research Explorer is retained by the author(s) and / or other copyright owners and it is a condition of accessing these publications that users recognise and abide by the legal requirements associated with these rights.

Take down policy

The University of Edinburgh has made every reasonable effort to ensure that Edinburgh Research Explorer content complies with UK legislation. If you believe that the public display of this file breaches copyright please contact openaccess@ed.ac.uk providing details, and we will remove access to the work immediately and investigate your claim.



1 Clarifying the meaning of mantras in wildland fire behavior modeling: reply to Cruz et al.

2 William Mell^{a,g}, Albert Simeoni^b, Dominique Morvan^c, J. Kevin Hiers^d, Nicholas Skowronski^e, Rory Hadden^f

3 ^aPacific Wildland Fire Sciences Laboratory, U.S. Forest Service, Seattle, WA 98074

4 ^bFire Protection Engineering Department, Worcester Polytechnic Institute, Worcester, MA 01609

5 ^cAix-Marseille Univ, CNRS, Centrale Marseille, M2P2, Marseille, France

6 ^dTall Timbers Research Station, 13093 Henry Beadel Dr., Tallahassee, FL 32312

7 ^eNorthern Research Station, U.S. Forest Service, Morgantown, WV 26505

8 ^fSchool of Engineering, University of Edinburgh, Edinburgh, United Kindgdom

9 ^gCorresponding author. Email: wemell@fs.fed.us; 206-430-2072

10

11 **Abstract.** In a recent communication, Cruz et al., 2017 called attention to a number of recurring
12 statements (mantras) in the wildland fire literature regarding empirical and physical fire behavior
13 models. Motivated by concern that these mantras have not been fully vetted and are repeated blindly,
14 Cruz et al. seek to verify five mantras they identify. This is a worthy goal and we seek here to extend the
15 discussion and provide clarification to a number of confusing aspects of the Cruz et al. communication.
16 In particular, their treatment of what they call physical models is inconsistent, neglects reference to
17 current research activity focused on combined experimentation and model development, and misses an
18 opportunity to discuss the potential use of physical models to fire behavior outside the scope of
19 empirical approaches.

20 **Brief summary:** The validity of a number commonly held beliefs regarding fire behavior models is
21 discussed with an emphasis on physical (or physics-based) models.

22 **Additional keywords:** empirical models, physics-based models, CFD

23

24 **Introduction**

25 In a recent commentary on fire behavior models, Cruz et al. (2017) identify five statements, or mantras,
26 they believe have gained “currency as facts — or truths” regarding empirical and physical (sometimes
27 called physics-based or process-based) wildland fire models. Cruz et al. are concerned that an
28 unquestioning acceptance of the mantras will lead to poorly informed use of the models in question.
29 They seek, therefore, “to discuss the validity” of these mantras. We agree that model users should be
30 aware of the strengths and weaknesses of a given model. However, inconsistencies between how the
31 mantras are represented by Cruz et al, and how they appear in the literature add confusion, rather than
32 clarity, to a broader discussion. In some cases, the authors discussion of the mantras is not even
33 consistent within their own framework. Regarding physical models, the largely negative critique is
34 confused by inconsistent definitions, inaccuracies, and falls short of understanding how model
35 advancement in engineering science is coupled to appropriate measurements. The authors appear to
36 favor empirical models for prediction while not recognizing the capabilities of physical models,
37 especially those based on computational fluid dynamics, for improving our understanding of the
38 mechanisms and their role in driving fire behavior.

39 While we appreciate the motivation and goal of Cruz et al., our intent in this response is to provide a
40 constructive critique of Cruz et al. by clarifying the particularly confusing elements and providing
41 viewpoints from the engineering and management perspectives. In Cruz et al., empirical (as opposed to
42 semi-empirical) models are the subject of the first two mantras and what they call “physical” models are
43 considered in the last three mantras. These mantras are:

44 Mantra 1 (M1): Empirical models work well over the range of their original data.

45 Mantra 2 (M2): Empirical models are not appropriate for and should not be applied to
46 conditions outside the range of the original data.

47 Mantra 3 (M3): Physical models provide insight into the mechanisms that drive wildland fire
48 spread and other aspects of fire behavior.

49 Mantra 4 (M4): Physical models give a better understanding of how fuel treatments modify fire
50 behavior.

51 Mantra 5 (M5): Physical models can be used to derive simplified models to predict fire behavior
52 operationally.

53

54

55 **The discussion regarding physical models is flawed**

56 The discussion related to the mantras for the physical models displays a limited understanding of
57 modeling approaches that attempt to include (explicitly or implicitly) physical processes driving wildland
58 fire. In the first paragraph of Cruz et al., the authors define a physical modeling approach as one that
59 “employs a mathematical description of fundamental physical and chemical processes underpinning
60 combustion, fluid flow and heat transfer”. We take this to mean that the processes driving fire behavior
61 are explicitly accounted for in “physical models”. Cruz et al. then use the term “physical model” for both
62 simpler models that, for example, neglect the process of convective heat transfer (in M3 and M5) and
63 for more comprehensive physical models based on computational fluid dynamics (CFD) that explicitly
64 account for all the recognized driving processes (in M3 and M4), including convective heat transfer.

65 A consequence of this inconsistent use of the term physical model is confusion and lack of
66 completeness. For clarity, here we place physical models into two groups: one group uses CFD, and the
67 other does not. Both have model equations that are the result of approximations based on physically
68 motivated assumptions. To be more precise, we use CFD based physical models to denote

69 comprehensive approaches that explicitly model the recognized processes driving fire behavior. This is
70 consistent with references cited for CFD based models and statements made by Cruz et al.

71 In Cruz et al., nearly all the cited non-CFD physical models do not explicitly model convective heat
72 transfer. Cruz et al. appear to mistakenly assume that convective heat transfer was neglected because
73 the model developers assumed it is not relevant to fire spread, which is clearly not the case. If one reads
74 the cited literature, it is clear that the model developers are fully aware that convective heat transfer, in
75 some environmental conditions, will be relevant; but these are not the environmental conditions for
76 which they derive their model. The assumption of radiation dominance in these models is not,
77 therefore, an “example of our ignorance of the fundamental processes governing wildland fire behavior”
78 as stated in the third paragraph of the M3 discussion.

79 Adding to the confusion, Cruz et al. incorrectly interpret findings in the cited literature (Anderson et al.,
80 2010; Butler, 2010) when they write (end of second paragraph of M3 discussion) “recent experimental
81 evidence suggests it is convective heat transfer ... that is the dominant heat transfer mechanism
82 determining wildland fire propagation”. Anderson et al. (2010) don’t measure radiation and, therefore,
83 do not compare radiative and convective heat fluxes. Butler (2010) finds that convective and radiative
84 heat flux can be comparable in magnitude at certain times, and do not state that convective heat
85 transfer dominates. Finney et al. (2015) do state that “repetitive convective heating thus appears to be
86 the critical heat transfer mechanism causing ignition and spread of these fires”. In addition, Morandini
87 and Silvani (2010) (this study was not referenced in Cruz et al.) conducted five field experiments and
88 found that, depending on the fire experiment, radiative heat transfer either dominated convective heat
89 transfer, or they were of similar magnitude. Morandini and Silvani (2010) considered shrub fires. Butler
90 (2010) considered full-scale crown fires. Finney et al. (2015) considered laboratory-scale surface fire in
91 highly uniform fuel beds. Clearly, more work is needed to determine why the findings of these
92 experiments differ. This point is missed by Cruz et al.

93 The latter part of the discussion of M3 and most of the M4 discussion is focused on the challenges facing
94 CFD based physical models, including the need for some empiricism and more model validation. While
95 space limits a comprehensive response, some statements are notably incorrect and demonstrate a
96 limited understanding of CFD modeling. For example, it is not possible to model buoyant flow driven by
97 combustion while assuming (as stated by Cruz et al., in the M3 section) constant density, incompressible
98 flow.

99 Significantly, what Cruz et al. do not convey is that the reason they can list challenges to CFD based
100 modeling is precisely because these models are well characterized, both in their modeling approach and
101 in areas needing improvement. CFD based fire behavior models are constructed from coupled numerical
102 models, for the governing processes, that vary in their degree of maturity and proven physical fidelity.
103 For example, the models for fluid flow (including buoyancy induced flow) and radiation are significantly
104 more advanced and validated than models for the processes of thermal degradation and momentum
105 drag in vegetation. Cruz et al. give an incomplete picture of the advances made and the state of activity
106 (including new experiments) in pursuit of improvements to these models (e.g., Mueller 2017a, Anand et
107 al. 2017, Lamorlette et al. 2018).

108 In the last sentence of the M3 section Cruz et al. summarize their view of CFD based physical modeling:
109 “Until a complete and robust understanding of the processes ... we question how much is to be gained
110 from pure modeling exercises” This statement is problematic for a number of reasons. Physical
111 models have approximations and will not be “complete”, but they can be useful and their failings can be
112 characterized and addressed, making this a specious criticism. In addition, the suggestion that the
113 developers of CFD based physical models are in some way focused on “pure modeling exercises”
114 displays a lack of familiarity with fire engineering science. It is fundamental to the scientific method and
115 well established in the fire engineering community that the development of physical models requires
116 comparison with observations and experiments (see Mell et al. 2007, Tihay et al. 2008, Mell et al. 2009,

117 Morvan et al. 2009, Tihay et al. 2009, Hoffman et al. 2016, El Houssami et al. 2018). The necessity to
118 have detailed comparisons between numerical results and experimental data (i.e., not just rate of
119 spread observations) often push experimentalists to use more and more sophisticated experimental
120 diagnostic methods in the laboratory (Marcelli et al 2004, Morandini et al 2005, Zhou et al 2007, Lozano
121 et al 2010) and in the field (Frankman et al 2013, Mueller et al. 2017b). This list of experimental studies,
122 using advanced diagnostics, is only a sampling, many more exist.

123 **Mantra 2 is not representative of statements in literature**

124 There is no acknowledgment or discussion of how the particular wording of any given mantra, which
125 impacts the mantra's meaning, required choices by the authors. For example, consider mantra 2 which
126 is stated to be "likely the most commonly used fire behaviour modelling mantra". In the literature cited
127 in Table 1 of Cruz et al. for M2, the following text can be found (note, Cruz et al. do not provide these
128 excerpts):

129 Catchpole and de Mestre (1986): "While such models may be very successful over fuel and
130 environmental conditions similar to those occurring in the test fires, their lack of a physical basis means
131 that the use of such models outside of these conditions must be treated with caution."

132 Morvan and Larini (2001): "The predicted values for the ROS [rate of spread] remain valid for conditions
133 close to the experimental conditions which were used to gauge the parameters of the model. ...
134 Unfortunately the results obtained with this type of approaches are not easily applicable for more
135 general fire conditions."

136 Balbi et al. (2009): "... but the model is only valid in the range of experiments for which it was validated.
137 Peculiarly, the change from laboratory to field scale experiments is not supported, but involves a new
138 calibration of the parameters."

139 Mell et al. (2010) "... strictly speaking, their application to environmental conditions outside of those for
140 which they were derived is not justified."

141 Pastor et al. (2003): "These are only applicable to systems in which conditions are identical to those
142 used in formulating and testing the models." Later in the paper it is stated, regarding McArthur meters
143 for dry Eucalypt forest, that: "Nevertheless, the use of this model in landscapes with vegetation different
144 from that of dry Eucalypt forest in Australia should be done with caution."

145 At first glance, these quoted statements seem to be well represented by M2 of Cruz et al. However,
146 most of the statements allow for the possibility of applying an empirical model outside its original data
147 set, but with appropriate caution. Thus, the wording of the Cruz et al. version of this mantra is *stricter*
148 than that of the authors cited because Cruz et al. make no allowance for the possibility that an empirical
149 model may work outside the original environmental conditions. This sets the stage for easily invalidating
150 M2 by finding any case where an empirical model works sufficiently well outside its originating
151 environmental conditions. This is what Cruz et al. do in their discussion of M2.

152 Cruz et al. go further and state that "empirical models are likely to be valid for far drier and windier
153 conditions than those involved in the model development". But this statement *required* sufficient
154 measurements in the new environment to show that the original model actually worked outside its
155 dataset. Also, there are contrary examples. The work by Fernandes (2014) had the opposite finding: an
156 empirical model could not be successfully extended to environmental conditions outside its original
157 dataset unless it was recalibrated using the new data.

158 While many scientists would allow that an empirical model may work for environmental conditions
159 outside its originating dataset, they would also agree that without measurements confirming it, there is
160 no justification for asserting that the empirical model will do so with quantifiable confidence. Caution is
161 inherent to the process of using empirically fit models beyond their domain of inference and is taught in

162 basic statistics (Sokal and Rohlf 1995). In essence, Cruz et al. agree with this when they state, at the end
163 of M2, “evaluation should always precede the use of models within operational contexts”.

164

165 **Are mantras 3 through 5 valid?**

166 We agree that the wording of M3 is representative of the literature and believe it to be valid. As an
167 example, we provide a simple demonstration of how of CFD based models can provide insight into the
168 roles of convective and radiative heat transfer. Figure 1 shows results from a three-dimensional, time
169 dependent, simulation (using the wildland-urban interface fire dynamics simulator (WFDS); Mell et al.
170 2009; Perez-Ramirez et al. 2017 have model details), of a surface fire spreading, with no ambient wind,
171 through a 10 cm deep, 80 cm wide, 1.8 m long excelsior fuel bed. Figure 1 shows the time histories of
172 the gas and vegetation temperatures and the contribution of the convective ($\nabla \cdot q_{\text{CONV}}$) and radiative
173 ($\nabla \cdot q_{\text{RAD}}$) heat fluxes to the rate of change of the vegetation’s temperature. These quantities are
174 plotted at two vertical locations (both at a distance of 1 m from the ignition region): $z=35$ cm above the
175 fuel bed (i.e., a location subjected to the combustion generated buoyant plume and intermittent flame)
176 and at $z = 0$ cm (i.e., top of fuel bed and subjected to a relatively slower and less variable flow and
177 radiation from a continuous fire front). Consistent with the findings of Finney et al. (2015) (see their Fig.
178 5A), the vegetation temperature at $z = 35$ cm follows a “stair-stepped” rise that is controlled by a varying
179 convective heat flux (Figs. 1a and 1b). At the $z = 0$ cm on top of the fuel bed (Figs. 1c, 1d), radiation
180 dominates until near ignition ($T_{\text{veg}} \approx 350$ C, time = 36 s), at which point radiation and convection are
181 comparable, at no point does convection exhibit the large oscillations seen at $z = 35$ cm. The
182 experimental configuration of Finney et al. (2015) is a surface fire and their measurement location is
183 similar to Figs. 1c and 1d (i.e., at the top of the fuel bed). Their results are similar to Figs. 1a and 1b

184 because their imposed wind increases the unsteady behavior of the flame. Simulations with WFDS give
185 similar results with an imposed wind (not shown).

186 Regarding M4, we believe that Cruz et al. chose a wording that is stricter than in the literature. This
187 mantra should read: “Physical models *have the potential to* give a better understanding of how fuel
188 treatments modify fire behavior”, which we believe is valid. It is not clear why Cruz et al. did not write
189 M4 this way, especially since their opening sentence introducing M4 does. CFD based models have been
190 used to simulate the influence of the spatial heterogeneity of vegetation on fire behavior (e.g., in
191 addition to the references in Cruz et al.: Pimont et al. 2011, Hoffman et al. 2015, Ziegler et al. 2017). The
192 challenge is to evaluate how well these simulations represent reality, which requires well designed
193 experiments. This is well recognized by physical modelers and the community would be better served if
194 Cruz et al. discussed the need for well-designed experiments to support model development and current
195 activity. Instead, Cruz et al. present an obstructive discussion on model approximations and the lack of
196 model validation. Also, with their emphasis that the physical models are not ready for operational use,
197 the discussion deviates from M4. The wording of M4 does not explicitly state that it refers to either CFD
198 based physical models (which is the only type of physical model cited) or operational objectives.

199 We agree with Cruz et al.’s statement in M5 that models applied to operational objectives need to be
200 properly used and their limitations known. However, their M5 discussion suffers from another
201 inconsistent use of the term “physical model”. In this section, they write: “the physical model is an
202 acceptable representation of the fire processes and that the only limitations for model implementation
203 are extraneous to the modelling of the fire processes, such as numerical implementation issues and
204 computational time demands”. This is followed by their declaration that Albin’s model (Albin 1996,
205 2000) is a physical model of crown fire spread. But Albin’s model does not meet the characteristics of a
206 physical model as described above by Cruz et al. Instead, Albin’s model is a simpler approach and Butler
207 et al. (2004) combine four existing simpler models for different components of the problem (see bottom

208 right of page 1590 in Butler et al. 2004). Thus, the Cruz et al. use of Butler et al. (2004) has no relevance
209 to M5.

210 While we do not find compelling evidence that M5 appears in the references cited, we agree with the
211 mantra in the sense that it is possible to use ROS predictions from CFD based models to develop
212 “empirical” formulas. For example, the study of Mell et al. (2007) found good agreement of the head fire
213 ROS determined from numerical predictions and an empirical model based on field observations. This
214 included predictions of fireline acceleration dependent on the head fire width. Thus, these simulations
215 could have been the basis of an empirical model. But model developers, as a matter of course, are
216 reluctant to provide such empirical models without sufficient characterization of model performance,
217 which requires a range of appropriate experiments. Examples of analysis leading to a reduced model
218 from a more comprehensive physical model include the works of Simeoni et al. (2001), who use the
219 approach of model reduction, and Margerit and Sero-Guillaume (2002) who use asymptotic analysis.

220 **Management implications**

221 From the perspective of a land manager, the changing landscapes in which wildfires and prescribed fires
222 are managed, demand a more robust toolset for understanding the processes at play. Operational tools
223 for predicting fire behavior lag far behind the science of fire-atmospheric interactions, and a continued
224 reliance on empirical models becomes less “predictive” as managers face increasingly novel
225 combinations of fuels (from non-native species), weather, climate, and heterogeneity across landscapes
226 (Kraaij et al. 2018). Furthermore, by definition, empirical models cannot capture, with well-
227 characterized confidence, the limits/extremes of observed fires (see discussion of M2). This limitation
228 creates the need for caution, which is often not adequately relayed to the management community,
229 when employing empirical models beyond their domain of origin. Also, managing fire in conditions for
230 which measurements are incomplete creates an important operational decision space for the use of CFD

231 based approaches for understanding the potential physical mechanisms in increasingly complex
232 contexts. Empirical modeling focuses almost exclusively on the ROS. The use of ROS as a gold standard
233 for validation further misses a critical management need to understand complex fire-atmospheric
234 feedbacks, multiple fireline development, and canopy induced flows on planned ignitions. There are
235 simply too many management tactics and decisions that involve critical fire behavior phenomena
236 outside the domain of empirical inference. Because managers are themselves empirical modelers, tools
237 that operate at conditions and fire behavior at the edge of their experience are the most critical for
238 enhancing decision making in operational contexts.

239 Using CFD or other physical modeling tools is needed for the evaluation, either retrospectively or
240 proactively, of processes and mechanisms that generate unexpected fire behaviors. Such lessons
241 learned for fire reconstructions has proven useful in understanding rare events (e.g., Cunningham and
242 Reeder 2009). It is equally important for managers to understand when CFD or other physics-based
243 modeling tools approach the limits of their applicability. If skepticism of CFD and trust in empirical
244 models is the ultimate point of Cruz et al., then they sadly miss the opportunities that each approach
245 provides as managers tackle a range of operational contexts.

246 **Conclusions**

247 We believe that there is a need for all types of models for research and for operational purposes. We
248 also firmly reject the assertion that because all the physical processes and their interaction driving fire
249 behavior are not fully understood, physical modeling should be discouraged or held suspect. History and
250 the scientific method have shown that progress in physical modeling is made with initial simplifying
251 approximations to be tested against well designed experiments. The idea that the two approaches
252 (experimental and theoretical/numerical) are complementary is widely shared in the scientific
253 community (as, notably, stated in Cruz et al 2011).

254 Recurrent in Cruz et al. is the recognized need for well-designed experiments for the development and
255 evaluation of both empirical and physical models. We heartily agree and emphasize that for physical
256 models, especially in the field, these measurements are challenging (e.g., Mueller et al. 2017b; Mueller
257 et al. 2018) and require careful consideration of model needs in order to adequately provide
258 information on vegetation, wind, and fire behavior.

259 Uncertainty is and will always be part of a fire manager's risk calculations, and most managers clearly
260 understand that models are tools. Nearly all managers are also anxiously awaiting tools that provide
261 insight into fire behaviors not already self-evident through their own observations. The critical targeting
262 of new approaches based on physical modeling, especially CFD based, by Cruz et al. runs the risk of
263 undermining innovation and opportunities for managers to learn from this branch of fire research.

264 **Acknowledgements**

265 The authors would like to thank the Chad Hoffman, Russ Parsons, and Morgan Varner for their
266 suggestions and comments during the development of this manuscript.

267 **Conflicts of interest**

268 The authors declare no conflicts of interest.

269 **References**

270 Alexander ME, Cruz MG (2013) Are the applications of wildland fire behaviour models getting ahead of
271 their evaluation again? *Environmental Modelling & Software* 41, 65-71.

272 Anand C, Shotorban B, Mahalingham S, McAllister S, Weise DR (2017) Physics-Based Modeling of Live
273 Wildland Fuel Ignition Experiments in the Forced Ignition and Flame Spread Test Apparatus. *Combustion
274 Science and Technology*, 89(9) 1551-1571.

275 Cruz MG, Alexander ME, Sullivan A.L. (2017) Mantras of wildland fire behaviour modelling: facts or
276 fallacies? *International Journal of Wildland Fire* 26, 973-981.

277 Cunningham P, Reeder MJ (2009). Severe convective storms initiated by intense wildfires: Numerical
278 simulations of pyro-convection and pyro-tornadogenesis. *Geophysical Research Letters*, 36(12).

279 El Houssami M, Lamorlette A, Morvan D, Hadden RM (2018) Framework for submodel improvement in
280 wildfire modeling. *Combust. Flame*, 190, 12–24.

281 Fernandes PM (2014) Upscaling the estimation of surface-fire rate of spread in maritime pine (*Pinus*
282 *pinaster* Ait.) forest. *iForest* 7, 123-125.

283 Frankman D, Webb BW, Butler BW, Jimenez D, Forthofer JM, Sopko P, Shannon KS, Hiers JK, Ottmar RD
284 (2013) Measurements of convective and radiative heating in wildland fires. *International Journal of*
285 *Wildland Fire* 22 (2), 157-167.

286 Hoffman CM, Linn R, Parsons R, Sieg C, Winterkamp J (2015). Modeling spatial and temporal dynamics of
287 wind flow and potential fire behavior following a mountain pine beetle outbreak in a lodgepole pine
288 forest. *Agricultural and Forest Meteorology*, 204, 79-93.

289 Hoffman CM, Ziegler J, Canfield J, Linn RR, Mell W, Sieg CH, Pimont F (2016) Evaluating crown fire rate of
290 spread predictions from physics-based models. *Fire Technology* 52, 221-237.

291 Kraaij T, Baard JA, Arndt J, Vhengani L, van Wilgen BW (2018). An assessment of climate, weather and
292 fuel factors influencing a large, destructive wildfire in the Knysna region, South Africa. *Fire Ecology* 14:
293 (In Press)

294 Lamorlette A, Houssami ME, Morvan D (2018) An improved non-equilibrium model for the ignition of
295 living fuel. *International J. Wildland Fire* 27(1), 29-41.

296 Lozano J, Tachajapong W, Weise DR, Mahalingam S, Princevac M (2010) Fluid dynamic structures in a
297 fire environment observed in laboratory-scale experiments. *Combustion Science and Technology* 182 (7),
298 858-878.

299 Marcelli T, Santoni PA, Simeoni A, Leoni E, Porterie B (2004) Fire spread across pine needle fuel beds:
300 characterisation of temperature and velocity distributions within the fire plume. *International Journal of*
301 *Wildland Fire*, 13: 37-48.

302 Margerit J, Sero-Guillaume O (2002) Modelling forest fires. Part II: reduction to two-dimensional models
303 and simulation of propagation. *International Journal of Heat and Mass Transfer*, 45 1723-1737.

304 Mell W, Jenkins MA, Gould J, Cheney P (2007) A physics-based approach to modelling grassland fires.
305 *International J. Wildland Fire*, 16, 1-22.

306 Mell W, Maranghides A, McDermott R, Manzello SL (2009) Numerical simulation and experiments of
307 burning Douglas fir trees. *Combust. Flame*, 156, 2023–2041.

308 Mell WE, McDermott RJ, Forney GP (2010) Wildland fire behavior modeling: perspectives, new
309 approaches and applications. In “Proceedings of the Third Fire Behavior and Fuels Conference”, 25-29
310 October 2010, Spokane, WA, USA (EDs DD Wade, M Robinson) (CD-ROM) International Association of
311 Wildland Fire, Birmingham, AL, USA.

312 Morandini F, Simeoni A, Santoni PA, Balbi JH (2005) A model for the spread of fire across a fuel bed
313 incorporating the effects of wind and slope. *Combustion Science and Technology*, 177: 1381-1418.

314 Morandini F, Silvani X (2010) Experimental investigation of the physical mechanisms governing the
315 spread of wildfires. *International J. Wildland Fire* 19(50), 570-582.

316 Morvan D, Méradji S, Accary G (2009) Physical modelling of fire spread in grasslands. *Fire Safety Journal*,
317 44, 50-61.

318 Mueller EV (2017a) Examination of the underlying physics in a detailed wildland fire behavior model
319 through field-scale experimentation. PhD dissertation University of Edinburgh.
320 <http://hdl.handle.net/1842/22039>

321 Mueller EV., Skowronski N, Clark K, Gallagher M, Kremens R, Thomas JC, El Houssami M, Filkov A,
322 Hadden RM, Mell W, Simeoni A (2017b) Utilization of remote sensing techniques for the quantification
323 of fire behavior in two pine stands, *Fire Safety Journal*, 91, 845-854.

324 Mueller EV, Skowronski N, Thomas JC, Hadden RM, Mell W, Simeoni A. (2018) Local measurements of
325 wildland fire dynamics in a field-scale experiment. *Combustion and Flame*, to appear.

326 Perez-Ramirez Y, Santoni PA, Tramoni JB, Bosseur F, Mell WE (2017) Examination of WFDS in modeling
327 spreading fire in a furniture calorimeter. *Fire Technology* 53, 1795-1832.

328 Pimont F, Dupuy JL, Linn RR, Dupont S (2011). Impacts of tree canopy structure on wind flows and fire
329 propagation simulated with FIRETEC. *Annals of Forest Science*, 68(3), 523.

330 Simeoni A, Santoni PA, Larini M, Balbi JH (2001) Proposal for theoretical contribution for improvement
331 of semi-physical forest fire spread models thanks to a multiphase approach: application to a fire spread
332 model across a fuel bed. *Combustion Science and Technology*, 162, 59-84.

333 Sokal RR, Rohlf FJ (1995) Biometry the principle and practices of statistics in biological research. 3rd
334 Edition, W.H. Freeman and Company, New York, NY

335 Tihay V, Simeoni A, Santoni PA, Rossi L, Bertin V, Bonneau L, Garo JP, Vantelon JP (2008) On the interest
336 of studying degradation gases for forest fuel combustion modeling. *Combustion Science and Technology*,
337 180(9): 1637-1658.

338 Tihay V, Santoni PA, Simeoni A, Garo JP, Vantelon JP (2009) Skeletal and global mechanisms for the
339 combustion of gases released by crushed forest fuels. *Combustion and flame*, 156(8): 1565-1575.

340 Ziegler JP, Hoffman C, Battaglia M, Mell W (2017). Spatially explicit measurements of forest structure
341 and fire behavior following restoration treatments in dry forests. *Forest Ecology and Management*, 386,
342 1-12.

343 Zhou X, Mahalingam S, Weise D (2007) Experimental study and large eddy simulation of effect of terrain
344 slope on marginal burning in shrub fuel beds. *Proceedings of the Combustion Institute* 31 (2), 2547-2555.

345

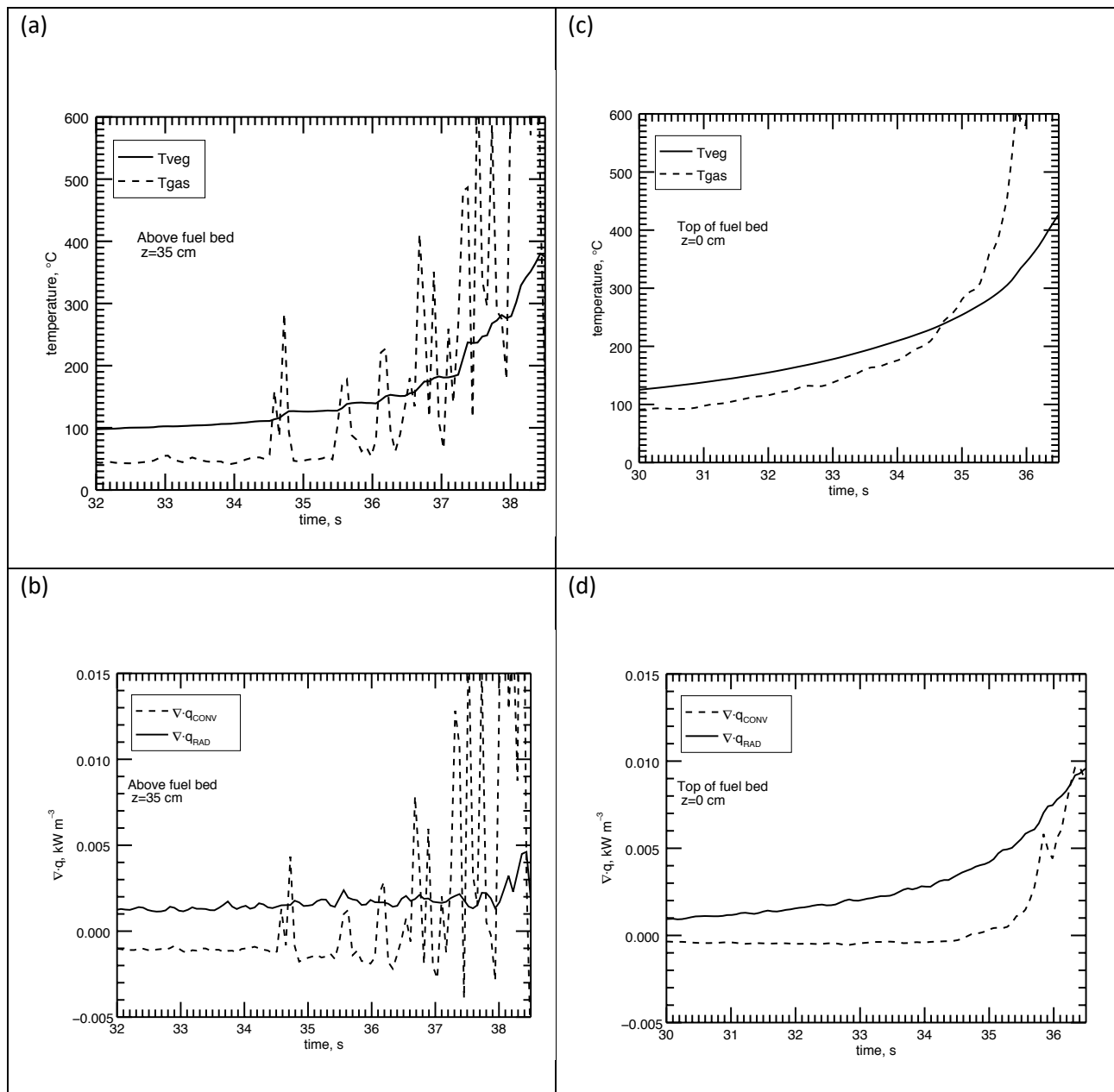


Figure 1: Results from a CFD based physical model simulation of a fire spreading through an excelsior fuel bed in the absence of an ambient wind. The gas temperature, the vegetation temperature, and measures of the convective ($\nabla \cdot q_{\text{CONV}}$) and radiative ($\nabla \cdot q_{\text{RAD}}$) flux into a 2 cm^3 volume of excelsior are plotted versus time. The left-side column (figures (a) and (b)) show these quantities at a location $z = 35 \text{ cm}$ above the fuel bed. The right-side column (figures (c) and (d)) are for a location at the top of the fuel bed, $z = 0 \text{ cm}$.

