

# ForeFire: A Modular, Scriptable C++ Simulation Engine and Library for Wildland-Fire Spread

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## Software

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## Summary

Wildfire forecasting is both an active research area and an important need for decision support systems. **ForeFire** is a modular, high-performance, scriptable, discrete-event-driven simulation engine (Filippi et al., 2009) focusing computational effort on the active region of a fire front defined as a dynamic mesh (or multipolygons) of fire markers. It is designed to model the spread of wildfire perimeters over large landscapes at meter scale resolution in seconds, serving both as a research platform and a tool for operational forecasting. The core C++ library has Fortran and Python bindings and is accompanied by a lightweight scriptable interpreter (a custom FF language), can load, save and export data in NetCDF, GeoJson, KML, PNG and JPG, and includes a local HTTP service with customizable graphical user interface. ForeFire can also account for fire-atmosphere interaction by two-way coupling with the MesoNH (Lac et al., 2018) atmospheric model (Filippi et al., 2013).

## Statement of need

Wildfire modeling tools have historically been split between **complex combustion research models** and **streamlined operational tools**, each with distinct limitations. Computational combustion and fluid dynamics (CFD) based models (e.g., FIRETEC (Linn & Cunningham, 2005) or WFDS (Mell et al., 2007)) are highly computationally intensive and yet unable to provide large wildfire forecasts faster than real time. Atmospheric coupled codes, such as WRF/SFire (Mandel et al., 2011) must be run within an atmospheric model and require a large amount of processing power and data. Operational wildfire simulators, such as widely used Farsite (Mark A. Finney, 1998) (now Flammap (Mark A. Finney et al., 2023)), or Canadian Prometheus (Garcia et al., 2008), are able to simulate fire fronts spanning tens of kilometers in a matter of seconds, but have definite built-in modeling assumptions and are distributed as compiled software with graphical interfaces with limited scriptability. Other open source libraries are ElmFire (Lautenberger, 2013) or Cell2Fire (Pais et al., 2021) that are tied to a single spread models and do not include scripting language, or deep learning based (Xia & Cheng, 2025).

ForeFire was developed as a community tool to fill the gap between highly complex customizable models and more rigid operational tools: a **unified** wildfire simulator that is both **adaptable** (highly scriptable with multiple bindings) and **high-performing** (discrete-event-driven simulation with dynamic mesh allows to concentrate computation at meter scale resolution only on the active part of the front to perform speed over 100Ha per second on a single CPU). It is intended to serve both as a research platform and a tool for operational forecasting.

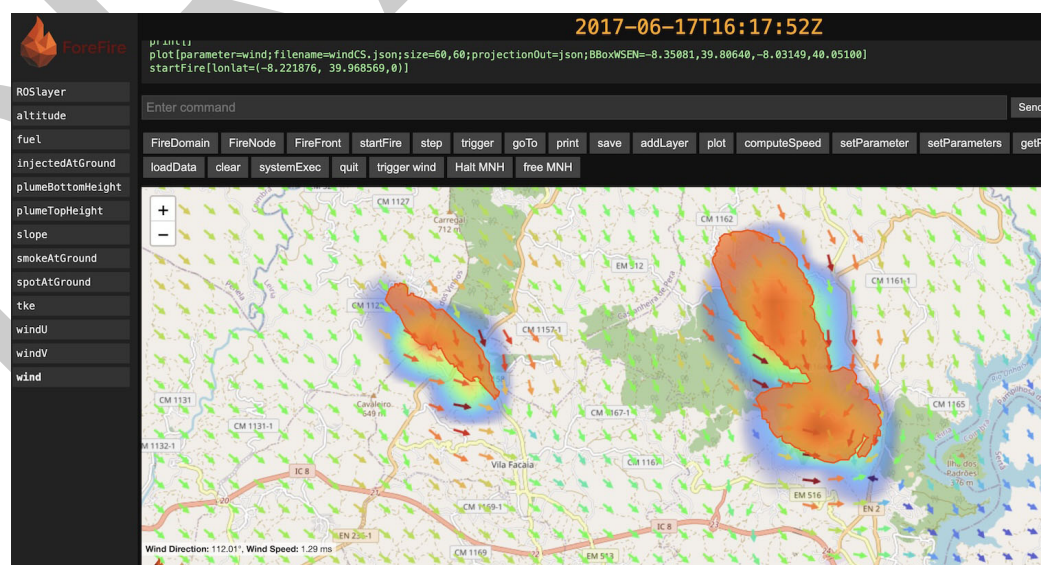
## Typical Use Cases

### Rapid prototyping of new models

ForeFire implements several standard fire flux and spread rate models, such as Rothermel (Andrews, 2018) or Balbi (Balbi et al., 2009), but also makes it trivial to switch, extend or add to this base with a single .cpp using any existing model file as a template. Internally data is handled as *layers* that can come from a NumPy array, read from NetCDF or generated on the fly by ForeFire (e.g. slope derived from the elevation layer, fuel loaded as index map with tabulated fuel (with part of (Scott & Burgan, 2005) fuel table already available)). Developing a Rate Of Spread wildfire model was the original purpose of this simulation code and helped to iterate versions of the Balbi Rate Of Spread formulation on case studies in (Balbi et al., 2009) and (Santoni et al., 2011). It also served to implement various heat and chemical species flux models used for volcanic eruption in (Filippi et al., 2021), plume chemistry (Strada et al., 2012) or industrial fires in (Baggio et al., 2022). In addition, the code includes a generic ANNPPropagationModel, which implements a feedforward artificial neural network (ANN) that expects a pre-trained graph file.

### Batch simulations with the ForeFire scripting

Custom FF language allows users to easily generate multiple scenarios, including fire-fighting strategies, model evaluation (Filippi et al., 2014), ensemble forecasts (Allaire et al., 2020) or generate a deep learning database (Allaire et al., 2021). A FF script is a set of scheduled instructions that are interpreted real-time, advancing the simulation clock with a step[dt=] or a goTo[t=] command. Each of these commands (such as goTo[t=42], include[state.ff], startFire[lonlat=(-8.1, 39.9,0)]@t=42, setParameter[propagationModel=Rothermel] or plot[parameter=speed;filename=ROS.png]) can also be called from HTTP, C++, Fortran or Python, and constitutes the core logic of the library. Help and autocompletion are directly available in the interactive shell interpreter that also includes a batch mode. The graphical user interface is web-based through an embedded HTTP service (command listenHTTP[host:port]) with user-defined or default pages as shown in Figure 1.



**Figure 1:** Default web interface with data layers on the left pane, commands displayed as buttons and displaying an atmospheric coupled simulation of a wildfire in Portugal.

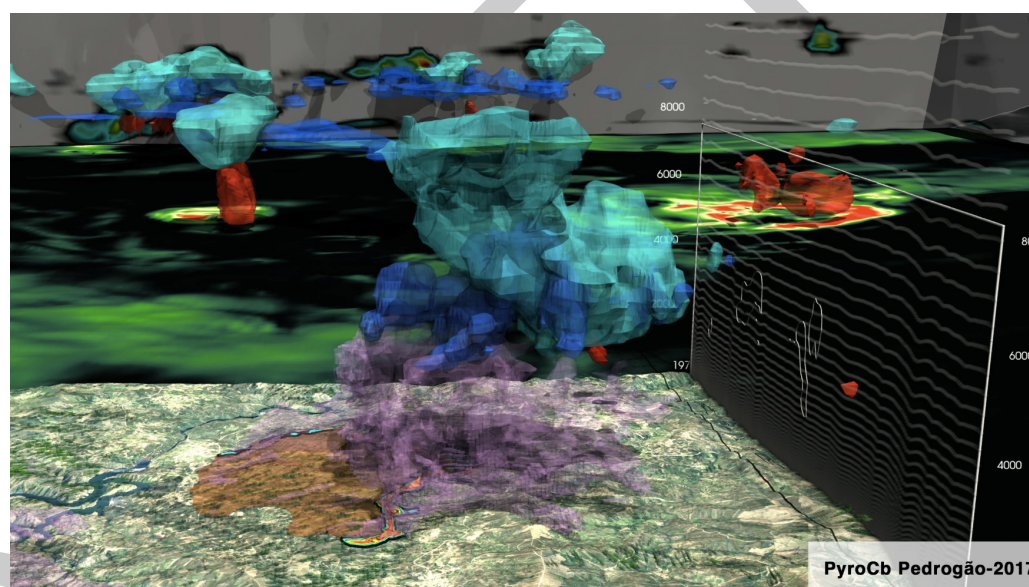
By utilizing pre-compiled datasets over extensive regions, this approach supports continent-

70 wide operational forecasting services. It has been deployed to identify optimal escape routes  
71 ([Kamilaris et al., 2023](#)), integrated into the French National WildFire Decision Support System  
72 [OPEN DFCI](#), showcased on the [FireCaster demonstration platform](#), and also currently used  
73 in commercial simulation services [AriaFire Firecaster](#), [UmGraueMeio Pantera](#) and [Ororatech](#)  
74 [FireSpread](#).

## 75 Two-way coupling with the MesoNH atmospheric model

76 The same scripts can be executed in coupled mode with the Open-Source atmospheric model  
77 [MesoNH](#) ([Lac et al., 2018](#)) with fire propagating using surface fields (wind) from MesoNH and  
78 forcing heat and other flux fields into the atmosphere. An idealized coupled simulation can be  
79 run on a laptop at field scale ([Filippi et al., 2013](#)), but also on a supercomputer to forecast  
80 fire-induced winds of large wildfires ([Filippi et al., 2018](#)), fire-induced convection ([Couto et al.,](#)  
81 [2024](#)), ([Campos et al., 2023](#)) or even to estimate wildfire spotting ([Alonso-Pinar et al., 2025](#)).

82 Coupled simulations generate gigabytes of 3D data that can be converted to VTK/VTU files  
83 using Python helper scripts to visualize in the open-source tool ParaView as shown in [Figure 2](#).



**Figure 2:** Coupled simulation of the Pedrogao Grande wildfire ([Couto et al., 2024](#)) (Paraview). On the ground, the burned area is in orange, while among atmospheric variables, downbursts are highlighted in red and pyro-cumulonimbus clouds in blue.

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## 89 References

- 90 Allaire, F., Filippi, J.-B., & Mallet, V. (2020). Generation and evaluation of an ensemble  
91 of wildland fire simulations. *International Journal of Wildland Fire*, 29(2), 160. <https://doi.org/10.1071/wf19073>  
92



- 93 Allaire, F., Mallet, V., & Filippi, J.-B. (2021). Emulation of wildland fire spread simulation  
94 using deep learning. *Neural Networks*, 141, 184–198. [https://doi.org/10.1016/j.neunet.](https://doi.org/10.1016/j.neunet.2021.04.006)  
95 [2021.04.006](https://doi.org/10.1016/j.neunet.2021.04.006)
- 96 Alonso-Pinar, A., Filippi, J.-B., & Filkov, A. (2025). Modelling aerodynamics and combustion  
97 of firebrands in long-range spotting. *Fire Safety Journal*, 152, 104348. [https://doi.org/10.](https://doi.org/10.1016/j.firesaf.2025.104348)  
98 [1016/j.firesaf.2025.104348](https://doi.org/10.1016/j.firesaf.2025.104348)
- 99 Andrews, P. L. (2018). *The rothermel surface fire spread model and associated developments:*  
100 *A comprehensive explanation*. U.S. Department of Agriculture, Forest Service, Rocky  
101 Mountain Research Station. <https://doi.org/10.2737/rmrs-gtr-371>
- 102 Baggio, R., Filippi, J.-B., Truchot, B., & Couto, F. T. (2022). Local to continental scale  
103 coupled fire-atmosphere simulation of large industrial fire plume. *Fire Safety Journal*, 134,  
104 103699. <https://doi.org/10.1016/j.firesaf.2022.103699>
- 105 Balbi, J. H., Morandini, F., Silvani, X., Filippi, J. B., & Rinieri, F. (2009). A Physical Model  
106 for Wildland Fires. *Combustion and Flame*, 156(12), 2217–2230. [https://doi.org/10.1016/](https://doi.org/10.1016/j.combustflame.2009.07.010)  
107 [j.combustflame.2009.07.010](https://doi.org/10.1016/j.combustflame.2009.07.010)
- 108 Campos, C., Couto, F. T., Filippi, J.-B., Baggio, R., & Salgado, R. (2023). Modelling  
109 pyro-convection phenomenon during a mega-fire event in portugal. *Atmospheric Research*,  
110 290, 106776. <https://doi.org/10.1016/j.atmosres.2023.106776>
- 111 Couto, F. T., Filippi, J.-B., Baggio, R., Campos, C., & Salgado, R. (2024). Numerical  
112 investigation of the pedrógão grande pyrocumulonimbus using a fire to atmosphere coupled  
113 model. *Atmospheric Research*, 299, 107223. [https://doi.org/10.1016/j.atmosres.2024.](https://doi.org/10.1016/j.atmosres.2024.107223)  
114 [107223](https://doi.org/10.1016/j.atmosres.2024.107223)
- 115 Filippi, J.-B., Bosseur, F., Mari, C., & Lac, C. (2018). Simulation of a large wildfire in a coupled  
116 fire-atmosphere model. *Atmosphere*, 9(6), 218. <https://doi.org/10.3390/atmos9060218>
- 117 Filippi, J.-B., Durand, J., Tulet, P., & Bielli, S. (2021). Multiscale modeling of convection  
118 and pollutant transport associated with volcanic eruption and lava flow: Application to  
119 the april 2007 eruption of the piton de la fournaise (reunion island). *Atmosphere*, 12(4).  
120 <https://doi.org/10.3390/atmos12040507>
- 121 Filippi, J.-B., Mallet, V., & Nader, B. (2014). Evaluation of forest fire models on a large  
122 observation database. *Natural Hazards and Earth System Sciences*, 14(11), 3077–3091.  
123 <https://doi.org/10.5194/nhess-14-3077-2014>
- 124 Filippi, J.-B., Morandini, F., Balbi, J. H., & Hill, D. R. (2009). Discrete event front-  
125 tracking simulation of a physical fire-spread model. *SIMULATION*, 86(10), 629–646.  
126 <https://doi.org/10.1177/0037549709343117>
- 127 Filippi, J.-B., Pialat, X., & Clements, C. (2013). Assessment of FOREFIRE/MESONH for  
128 wildland fire/atmosphere coupled simulation of the FireFlux experiment. *PROCEEDINGS*  
129 *OF THE COMBUSTION INSTITUTE*, 34(2), 2633–2640. [https://doi.org/10.1016/j.proci.](https://doi.org/10.1016/j.proci.2012.07.022)  
130 [2012.07.022](https://doi.org/10.1016/j.proci.2012.07.022)
- 131 Finney, Mark A. (1998). *FARSITE: Fire Area Simulator-model development and evaluation*  
132 (RMRS-RP-4). U.S Department of Agriculture, Forest Service; Res. Pap. RMRS-RP-4,  
133 Revised 2004, Ogden, UT: U.S. Department of Agriculture, Forest Service, Rocky Mountain  
134 Research Station. 47 p. <https://doi.org/10.2737/rmrs-rp-4>
- 135 Finney, Mark A., Brittain, S., Seli, R. C., McHugh, C. W., & Gangi, L. (2023). *FlamMap: Fire*  
136 *Mapping and Analysis System (Version 6.2) [Software]*. [https://www.firelab.org/project/](https://www.firelab.org/project/flammap)  
137 [flammap](https://www.firelab.org/project/flammap).
- 138 Garcia, T., Braun, J., Bryce, R., & Tymstra, C. (2008). Smoothing and bootstrapping the  
139 PROMETHEUS fire growth model. *Environmetrics*, 19(8), 836–848. [https://doi.org/10.](https://doi.org/10.1002/env.907)  
140 [1002/env.907](https://doi.org/10.1002/env.907)

- 141 Kamilaris, A., Filippi, J. B., Padubidri, C., Koole, R., & Karatsiolis, S. (2023). Examining the  
142 potential of mobile applications to assist people to escape wildfires in real-time. *Fire Safety*  
143 *Journal*, 136, 103747. <https://doi.org/10.1016/j.firesaf.2023.103747>
- 144 Lac, C., Chaboureaud, J.-P., Masson, V., Pinty, J.-P., Tulet, P., Escobar, J., Leriche, M., Barthe,  
145 C., Aouizerats, B., Augros, C., Aumond, P., Auguste, F., Bechtold, P., Berthet, S., Bielli,  
146 S., Bosseur, F., Caumont, O., Cohard, J.-M., Colin, J., ... Wautelet, P. (2018). Overview  
147 of the meso-NH model version 5.4 and its applications. *Geoscientific Model Development*,  
148 11(5), 1929–1969. <https://doi.org/10.5194/gmd-11-1929-2018>
- 149 Lautenberger, C. (2013). Wildland fire modeling with an eulerian level set method and  
150 automated calibration. *Fire Safety Journal*, 62, 289–298. <https://doi.org/10.1016/j.firesaf.2013.08.014>
- 152 Linn, R. R., & Cunningham, P. (2005). Numerical simulations of grass fires using a coupled  
153 atmosphere–fire model: Basic fire behavior and dependence on wind speed. *Journal of*  
154 *Geophysical Research: Atmospheres*, 110(D13). <https://doi.org/10.1029/2004jd005597>
- 155 Mandel, J., Beezley, J. D., & Kochanski, A. K. (2011). Coupled atmosphere-wildland fire  
156 modeling with WRF 3.3 and SFIRE 2011. *Geoscientific Model Development*, 4(3), 591–610.  
157 <https://doi.org/10.5194/gmd-4-591-2011>
- 158 Mell, W., Jenkins, M. A., Gould, J., & Cheney, P. (2007). A physics-based approach  
159 to modelling grassland fires. *International Journal of Wildland Fire*, 16(1), 1. <https://doi.org/10.1071/wf06002>
- 161 Pais, C., Carrasco, J., Martell, D. L., Weintraub, A., & Woodruff, D. L. (2021). Cell2Fire: A  
162 cell-based forest fire growth model to support strategic landscape management planning.  
163 *Frontiers in Forests and Global Change*, 4. <https://doi.org/10.3389/ffgc.2021.692706>
- 164 Santoni, P.-A., Filippi, J.-B., Balbi, J.-H., & Bosseur, F. (2011). Wildland fire behaviour case  
165 studies and fuel models for landscape-scale fire modeling. *Journal of Combustion*, 2011(1).  
166 <https://doi.org/10.1155/2011/613424>
- 167 Scott, J. H., & Burgan, R. E. (2005). *Standard fire behavior fuel models: A comprehensive*  
168 *set for use with rothermel's surface fire spread model*. U.S. Department of Agriculture,  
169 Forest Service, Rocky Mountain Research Station. <https://doi.org/10.2737/rmrs-gtr-153>
- 170 Strada, S., Mari, C., Filippi, J.-Baptiste., & Bosseur, F. (2012). Wildfire and the atmosphere:  
171 Modelling the chemical and dynamic interactions at the regional scale. *Atmospheric*  
172 *Environment*, 51, 234–249. <https://doi.org/10.1016/j.atmosenv.2012.01.023>
- 173 Xia, Z., & Cheng, S. (2025). PyTorchFire: A GPU-accelerated wildfire simulator with  
174 differentiable cellular automata. *Environmental Modelling & Software*, 188, 106401.  
175 <https://doi.org/10.1016/j.envsoft.2025.106401>