# A comparison study between fire-spotting models by a wildfire simulator based on a cellular automata approach

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Fire-spotting.

> Uncentainty source

"Starts new fires (spot-fires).

Accelerate the spread of fire











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Two research approaches in its study.











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Accelerate the spread of fire

Two research approaches in its study.

Good fire-spotting model -> Accurate wildfire simulations.











# Our work: Fire-spotting comparison models

#### Fire-spotting models.

- Trucchia A, Egorova V, Butenko A, Kaur I, Pagnini G (2019) RandomFront 2.3: a physical parameterisation of fire spotting for operational fire spread models – implementation in WRF-SFIRE and response analysis with LSFire+. Geoscientific Model Development 12, 69–87. doi:10.5194/gmd-12-69-2019.
- Alexandridis A, Russo L, Vakalis D, Bafas GV, Siettos CI (2011) Wildland fire spread modelling using cellular automata: evolution in large-scale spatially heterogeneous environments under fire suppression tactics. *International Journal of Wildland Fire* **20**, 633. doi:10.1071/WF09119.
- Perryman HA, Dugaw CJ, Varner JM, Johnson DL (2013) A cellular automata model to link surface fires to firebrand lift-off and dispersal. International Journal of Wildland Fire 22, 428. doi:10.1071/WF11045.











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#### **PROPAGATOR**

Georeferenced map with the burnt probability of each cell.











# The Alexandridis et al. parametrization

- > It was developed in CA framework.
- > Firebrand landing distance:

$$d_p = r_n \cdot P_W = r_n \exp\left(U_{C_2}(\cos \varphi - 1)\right) \tag{1}$$

 $\succ$  r<sub>n</sub> is drawn from a normal distribution.











# The Perryman et al. parametrization

- > It is based in previously research.
- > Firebrand landing distance:

$$d_p = \sqrt{d_{||}^2 + d_{|-}^2} \tag{2}$$

 $\succ$  Paralell to wind direction ( $d_{||}^2$ )-> Lognormal distribution

ightharpoonup Perpendicular-to-wind direction( $d_{|-}^2$ ) -> Normal distribution











# RandomFront parametrization

> Lognormal distribution.

$$p(d) = \frac{1}{(\sqrt{2\pi} \sigma d)} exp\left(\frac{-(\ln(d/\mu)^2)}{(2\sigma^2)}\right)$$
 (3)

 $\triangleright \mu$ ,  $\sigma$  -> Parametrizes the physics of the transport.











# RandomFront parametrization

Mean:

$$\mu = H(H_{ABL}) \left(\frac{3\rho C_d}{2\rho_f}\right)^{1/2} \tag{4}$$

Standard deviation:

$$\sigma = \sigma(\varphi, \omega) = \frac{1}{zp} \ln \left\{ \frac{U \cos \varphi}{\sqrt{gr(1 + Tan^2 \omega)}} + \beta_2 \sqrt{\frac{2\rho_f}{3\rho C_d}} \frac{1.4U \cos \varphi + \sqrt{gh_0(1 + \phi_{wind} + \phi_{slope})^{2/3}} Tan \omega}{\sqrt{gh_0(1 + \phi_{wind} + \phi_{slope})^{2/3}} - 1.4U \cos \varphi Tan \omega} \right\}$$
 (5)











# Study area

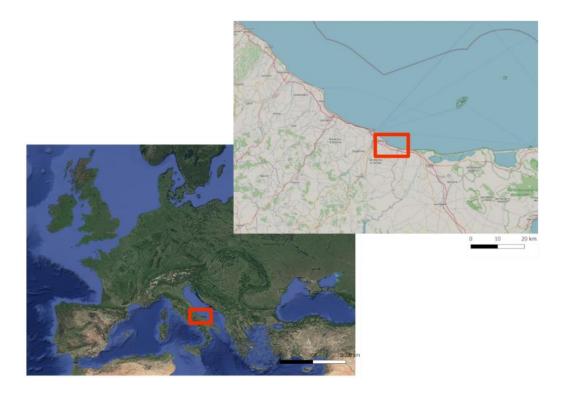


Figure 1











# Study area





(b)

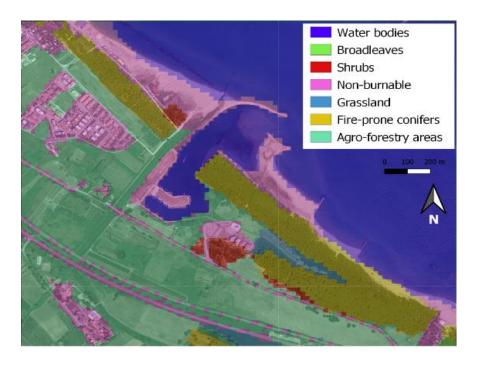


Figure 2











#### Alexandridis Model





Figure 3.











#### > Perryman

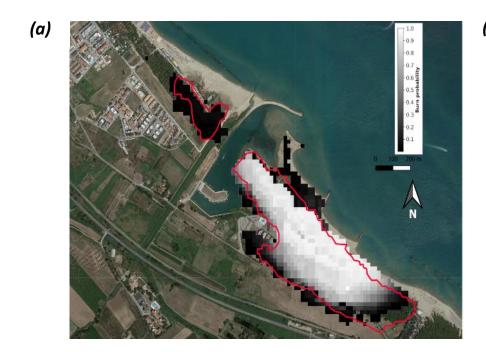




Figure 4











#### RandomFront fomulation





Figure 5











#### Probability generated

Sub-model	East	West
Alexandridis	0.7938	0.0358
Perryman	0.7370	0.1330
RandomFront	0.7303	0.2868

**Table 1** Main probability generated inside the reference Areas.











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- > Richer spread pattern is observed in physical models.
- > Higher probability density with RF.
- > The operational context is conserved.
- > Further work:
  - Analysis of the spread patterns in the ensemble contex.
  - Suitable fire-spotting metrics.
  - Firebrand generation process.
  - Spot ignition probability.











# Thank you for your attendance!!

Trucchia A, Egorova V, Butenko A, Kaur I, Pagnini G (2019) <u>RandomFront 2.3: a physical parameterisation of fire spotting for operational fire spread models – implementation in WRF-SFIRE and response analysis with LSFire+</u>. *Geoscientific Model Development* **12**, 69–87. doi:10.5194/gmd-12-69-2019.

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