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# Size of wildfires in the Euro-Mediterranean region: observations and theoretical analysis

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## Abstract

MODIS satellite observations of fire size and ERA-Interim meteorological reanalysis are used to derive a relationship between burnt area and wind speed over the Mediterranean region and Eastern Europe. As intuitively expected, the burnt area associated

- to the largest wildfires is an increasing function of wind speed in most situations. It is always the case in Eastern Europe. It is also the case in the Mediterranean for moderate temperature anomaly. In situations of severe heatwaves and droughts, the relationship between burnt area and wind speed displays bimodal shape. Burnt areas are large for low 10 m wind speed (lower than 2 m s<sup>-1</sup>), decrease for moderate wind speed values (lower than 5 m s<sup>-1</sup> and larger than 2 m s<sup>-1</sup>) and increase again for large wind speed (larger than 5 m s<sup>-1</sup>). To explain such behavior fire propagation is investigated using a probabilistic cellular automaton model. The observed relationship between burnt area and wind speed can be interpreted in terms of percolation threshold which mainly depends on local terrain slope and vegetation state (type, density, fuel moisture). In east-
- ern Europe, the percolation threshold is never exceeded for observed wind speeds. In the Mediterranean Basin we see two behaviors. During moderately hot weather, the percolation threshold is passed when the wind grows strong. On the other hand, in situations of severe Mediterranean heatwaves and droughts, moderate wind speed values impair the propagation of the wildfire against the wind and do not sufficiently accelerate the forward propagation to allow a growth of wildfire size.
- <sup>20</sup> the forward propagation to allow a growth of wildfire size.

#### 1 Introduction

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Large wildfires can have a significant impacts on natural, social and economic systems. If most fires are extinguished in the initial stages and remain small, the largest forest fires events represent most of the area burned and can have catastrophic consequences. In the boreal areas, large wildfires have a significant impact on carbon cycling (Amiro et al., 2001; Zhuang et al., 2002; Balshi et al., 2007), vegetation dynamics



(Thonicke et al., 2001) and the climate system (Randerson et al., 2006). The Mediterranean basin is also the major European region where wildfires occur, with about 90 % of the European burnt area (Chuvieco, 2009). Between 1980 and 2003, 57% of the total burnt area and 38 % of all ignited fires in Europe burned in Portugal (EFFIS, 2003).

- The year 2003 was one of the most severe years in Europe (Hernandez et al., 2015). Nearly 800 000 ha were burnt across southern Europe and in Portugal 4218 km<sup>2</sup> were burnt. A total of 20864 wildfires were ignited that year for this country. Such wildfires have a low probability of occurrence and are high consequence events because they behave differently than the other smaller fires (Rothermel, 1991).Large wildfire occurrence is also affected by special weather conditions which do not affect most of other 10

fires (Alvarado et al., 1998). The propagation of these wildfires is impacted by various factors such as fuel mois-

ture and load, human activities and short to long term weather (Flannigan et al., 2009). Wind speed is one of the main factors regarding the weather dependency, as it drives the rate of spread and direction of propagation (Rothermel, 1972). Fuel moisture is

15 also an important factor for the propagation of wildfires (Sharples, 2008). It depends on short (i.e. hour time scale) to medium (i.e. seasonal time scale) variability of temperature, precipitations and relative humidity.

The relationships between weather, vegetation and wildfire burnt area have been investigated in several works. Flannigan et al. (2001) link seasonal burnt area with 20 weather at various time and spatial scales (several years to a few days; global to local) and show that warm, dry and windy days are correlated to higher burnt area, and that certain synoptic conditions such as the breakdown of an upper blocking ridge leads to particularly favorable conditions for wildfires. Flannigan and Harrington (1988) show

the weather dependency of monthly burnt area with long sequences of dry and low 25 precipitations days. These relationships have been extrapolated for forecasting future wildfire severity in a changing climate. The review by Flannigan et al. (2009) indicates with a high confidence that climate change should lead a positive trend of burnt area. This is already measurable in the western US, as suggested by Westerling et al. (2006)



and Running (2006). Increase in the fire season duration, higher summer temperature, earlier melting of snowpacks and expansion of the areas susceptible to wildfires to elevated forests combine their effects, leading in this region to higher seasonal burnt area since the mid 1980s.

- In order to further these studies, our aim is to analyze the role of short and mid-term weather on the burnt area in the Euro-Mediterranean region, characterized by a variety of land-uses (natural forests, chaparral, agricultural fields), high exposure to heat waves (Stéfanon et al., 2012) and a high wildfire activity. Wildfires in the Mediterranean region tend to be generally caused by accidents (Silva et al., 2010) and in eastern Europa they are meatly linked to agricultural prestines (Turguety et al., 2014).
- rope they are mostly linked to agricultural practices (Turquety et al., 2014). Here we will not study the causes of wildfire activity but the dependency between fire size and weather, in particular wind speed and temperature anomaly. To complete our goal, we need a comprehensive database of recent wildfires that occurred in this region and the accompanying weather. The availability of remote sensing observations since the
- 15 1990s allows the analysis of large, consistent datasets on wildfire activity. In this study we use the MODIS-retrieved burnt area with the ERA-Interim meteorology to analyze the dependency of fire size with wind speed and temperature anomaly. In order to better understand the variations seen in the observations we need to use a model of fire spread. In this study we choose to focus on probabilistic cellular automata (PCA). Our
- study relies on the following data-driven observation: under certain circumstances wildfires tend to decrease in size with small wind speed. In order to investigate this assessment we will perform a theoretical study of percolation threshold in PCA with varying wind speed using the modelling framework shown on Alexandridis et al. (2008). The robustness of the study with respect to the model parameters will be tested.
- <sup>25</sup> After the introduction in Sect. 1, Sect. 2 details the data source, the methodology of data processing and analyzes the wildfire burnt area with respect to wind speed and temperature. Section 3 describes the wildfire model used for the theoretical analysis of the dependence of burnt area with respect to wind speed. Section 4 provides an



in-depth discussion of the processes explaining the relationship between the wildfire burnt area and the weather. Section 5 concludes the study.

# 2 Wildfire burnt area observations: methodology and analysis

# 2.1 Data source

- <sup>5</sup> Meteorological data are the 10 m wind speed and 2 m air temperature from the ERA-Interim reanalysis of the European Center for Mean-Range Weather Forecast (ECMWF), which provides data from 1979 onwards, and continues in real time (Simmons et al., 2006; Dee et al., 2011). The data are available every 6 h at a 0.75° × 0.75° longitude/latitude resolution. In the present study, the 12:00 UTC data are used only.
- Wildfire burnt areas result from the fire detection based on the observations from the MODIS instrument (Moderate Resolution Imaging Spectroradiometer) on board of Aqua and Terra polar heliosynchronous orbiting satellites. Daily burnt areas are identified when land cover, and thus surface reflectance, changes drastically and rapidly due to the wildfires (Roy et al., 2008). In the present study, the MCD45 burnt area product is used over a period of 10 years (2003–2012) and provides the daily burnt area on a 500 m horizontal grid. Only the fraction of each grid cell covered by vegetation is allowed to burn, following Turguety et al. (2014).

# 2.2 Data processing

A regridded version at 10 km resolution of the MCD45 product is used. A 3-D (latitude, longitude and time) connected component algorithm is applied to extract distinct fire events. Two fire detections on 26-adjacent 3-D grid meshes are considered to be part of the same event. The principal strength of this method is that it allows the detection of wildfires extending over more than 10 000 ha, since adjacent burning cells are connected. This is of particular importance since only large wildfires are investigated in the present study. Its main weakness is that it does not take into account cloud cover



impairment of remote sensing. Another problem is that two independent fire event occurring close to one another (less than 20 km of distance and less than a day between the end of the first event and the beginning of the second) are considered the same with this method.

#### 5 2.3 Data analysis

Figure 1 shows the cumulated wildfire burnt areas as retrieved by MODIS over the 10 year period (2003–2012). If a grid point has burnt entirely during this 10 year period, the cumulated burnt area is 10 000 ha. We see that the most wildfire-prone regions are the Mediterranean Basin and central eastern Europe. Wildfires in the Mediterranean area are mainly caused by accidents or criminal behavior (Silva et al., 2010), whereas in eastern Europe wildfires are generally linked to agricultural practices. In the following, we differentiate the two regions and the Mediterranean Basin will be referred as MED, whereas central eastern Europe will be referred as EAST.

- In the following, large wildfires are considered as extreme fires and are identified <sup>15</sup> when the value of the burnt area exceeds the upper 95th quantile of the local probability density function. Figure 2 shows the dependence of the 95th quantile of burnt area to the 10 m wind speed for an anomaly of the 2 m air temperature with respect to the climatology (1979–2012) ( $\Delta T_2$ ) smaller than 3 °C and exceeding 3 °C (heatwave). Large temperature anomalies are generally associated with an anticyclonic anomaly,
- <sup>20</sup> persisting heat waves (Black et al., 2004; Cassou et al., 2005; Barriopedro et al., 2011; Stéfanon et al., 2012) and larger burnt area (Pereira et al., 2005). In the EAST domain, wildfire size is nearly independent of the wind speed for  $\Delta T_2 < 3$ K and slightly increase with wind speed for  $\Delta T_2 > 3$ K. In the MED domain, a totally different behavior is found. For  $\Delta T_2 < 3$ K, wildfire burnt areas increase with wind speed, similarly to what is found
- <sup>25</sup> in the EAST domain for larger temperature anomalies. For  $\Delta T_2 > 3$  K, a bimodal dependence is found. A large number of wildfires occur for low wind speed (i.e. weaker than  $2 \text{ m s}^{-1}$ ) and for high wind speed (i.e. stronger than  $4 \text{ m s}^{-1}$ ).



For robustness assessment, the dependence of the wildfire burnt areas to the wind speed has also been checked for the MED domain using the European Forest Fire Information System (EFFIS) dataset over the 2006–2012 period (San-Miguel-Ayanz et al., 2012; the EAST domain is not covered by the EFFIS dataset). A similar behavior <sup>5</sup> has been found, consistent with the MODIS products, for both low and high temperature anomalies (not shown). The bimodal dependence of wildfire burnt areas to wind speed in the MED region for situations associated with heatwaves is counter-intuitive since the rate of spread is a growing function of wind speed (Rothermel, 1972). Indeed, the main reason is that those models give the rate of spread as a function of wind and other parameters (such as slope and vegetation state) but do not derive total burnt areas. To understand these observations, a probabilistic cellular automaton model is adapted in Sect. 3 to include the impact of wind speed on fire propagation and derive the burnt

#### 3 Wildfire burnt area simulations: sensitivity to wind speed

#### 15 3.1 Probabilistic cellular automaton

area.

Several types of fire models exist. One of them is the probabilistic cellular automaton (PCA), which relies on a stochastic fire propagation set by a probability of propagation which depends on wind speed and direction, terrain slope and vegetation state. Despite its simplicity, such model presents enough similarities with fire propagation dynamics to ensure accurate verification against observations and to allow in-depth analysis of fire propagation dynamics for a moderate computational cost (Berjak and Hearne, 2002; Trunfio, 2004; Hernández-Encinas, 2007a, b; Alexandridis et al., 2008). Unlike coupled fire-atmosphere models such as FIRETEC (Linn et al., 2002), the PCA model does not rely on physical processes that govern the fire behavior. However its simple implementation allows easier theoretical analysis.



The PCA implements a grid of cells that can be in a finite number of states and a local rule that determines the probability of transition between states at each time step. In our case we use the most elementary form of PCA for fire simulation, taking into account only the effect of wind speed on fire spread. The grid is a regular square lattice of cells and we choose 3 different states for the cells.

- State 0 corresponds to an unburnt cell;
- State 1 corresponds to the situation of a burning cell. The cells in this state can propagate the fire;
- State 2 corresponds to the situation of an already burnt cell.
- <sup>10</sup> The probability that a cell in state 1 can spread the fire to its neighbors is defined by Eq. (1):

 $\rho = \min\left(\gamma \rho_0 f\left(V, \theta\right), 1\right)$ 

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with *V* the surface wind speed,  $\theta$  the angle between the direction of propagation and the wind and  $\gamma$  a corrective factor for diagonal spread. The quantity  $p_0$  is a constant and is the spreading probability in absence of wind. The function *f* of wind is defined in Alexandridis et al. (2008) as:

 $f(V,\theta) = \exp[V(c_1 + c_2(\cos\theta - 1))].$ 

The quantities  $c_1$  and  $c_2$  are empirical factors set respectively to  $0.045 \text{ sm}^{-1}$  and  $0.131 \text{ sm}^{-1}$  (Alexandridis et al., 2008). The shape of  $f(V, \theta)$  is shown in Fig. 3 for various values of wind speed. Function *f* is isotropic with zero wind and gradually expands in the direction of the wind. The corrective factor  $\gamma$  is set at  $\frac{1}{\sqrt{2}}$  such as the fire propagation is isotropic in absence of wind speed. Without such correcting factor the fire would take a square form in the absence of wind.



(1)

(2)

Figure 4 shows the fraction of burnt cells  $y_2^{\text{final}}$  – with  $y_i$  the fraction of cells in state *i* in the PCA grid – as a function of  $p_0$  when the fire is extinct in a grid constituted of 101 × 101 cells. The quantity  $y_2^{\text{final}}$  is computed from a Monte Carlo simulation. Figure 4 displays a transition at  $p_0 \sim 0.3$  between a mode where very few cells are burnt ( $y_2^{\text{final}} \sim 0$ ) and another one where the fire spreads to nearly all the grid ( $y_2^{\text{final}} \sim 0.9$ ).

## 3.2 Percolation threshold analysis

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Above a certain value of the propagation probability p the fire can thus propagate indefinitely in the PCA. The critical value  $p_0^{\text{crit}}$  of the  $p_0$  constant is shown in Fig. 4 for the zero wind case. This phenomenon is called percolation and can be analyzed theoretically. Pak and Hayakawa (2011) studied the percolation threshold for an elementary form of PCA, with no diagonal propagation or wind speed effect. They state with the help of a mean-field approximation that the fraction of cells in states 1 and 2 are governed by Eqs. (3)–(5).

$$\frac{dy_1}{dt} = 4\rho y_1 y_0 - a_{12} y_1 \tag{3}$$

$$\frac{\mathrm{d}y_2}{\mathrm{d}t} = a_{12}y_1\tag{4}$$

$$y_0 + y_1 + y_2 = 1 \tag{5}$$

with  $y_i$  the fraction of cells in state *i* in the PCA grid, *p* the probability of fire spread and  $a_{12}$  the probability that a cell in state 1 at time *t* goes to state 2 at time *t* + 1. In our case, the quantity  $a_{12}$  is set to 1. In the study, we adapt the PCA model to account for wind speed effect and diagonal propagation. Equation (3) is thus modified as follows:

$$\frac{\mathrm{d}y_1}{\mathrm{d}t} = C(V)p_0y_1y_0 - y_1$$

(6)

with C(V) a weighting coefficient which allows to account for fire spreading in various directions as a function of the wind speed. It writes:

$$C(V) = \exp(Vc_1) + 2\exp[V(c_1 - c_2)] + \exp[V(c_1 - 2c_2)]$$
$$+ 2\gamma \exp\left\{V\left[c_1 + c_2\left(\frac{\sqrt{2}}{2} - 1\right)\right]\right\}$$
$$+ 2\gamma \exp\left\{V\left[c_1 - c_2\left(\frac{\sqrt{2}}{2} + 1\right)\right]\right\}.$$

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The first term of the sum on the right-hand-side of Eq. (7) corresponds to a propagation along the wind direction, the second term to a propagation across the wind direction and the third to a propagation against the wind. The other terms correspond to diagonal propagation ( $\sqrt{2}/2$  corresponds to  $\cos(\pi/4)$ ). The details of the derivation of Eq. (7) is provided in the Appendix. Equation (7) can be simplified:

$$C(V) = \exp(Vc_1) \times \left\{ [1 + \exp(-Vc_2)]^2 + \sqrt{2} \left( \exp\left[Vc_2\left(\frac{\sqrt{2}}{2} - 1\right)\right] + \exp\left[-Vc_2\left(\frac{\sqrt{2}}{2} + 1\right)\right] \right) \right\}.$$

Following the methodology of Pak and Hayakawa (2011), we derive a relationship between  $p_0^{\text{crit}}$  and the wind speed-dependent coefficient C(V):

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$$p_0^{\text{crit}} = \frac{1.51552}{C(V)}$$
. (9)

The value of the numerator comes from site percolation theory. For further insight, Pak and Hayakawa (2011) give an introduction to bond and site percolation theory. This



(7)

(8)

equation is plotted in Fig. 5, along with the numerically derived  $p_0^{\text{crit}}$ . To compute this quantity numerically we consider that percolation occurs when a fire initiated in the middle of our 101 × 101 PCA grid, reaches the edges of the domain. The  $p_0$  value increases from 0.2 with 0.001 increments. The critical value  $p_0 = p_0^{\text{crit}}$  is set when more than 90% of a 300 simulation ensemble percolate. We can see that both the theoretical and computed values  $p_0^{\text{crit}}$  increase with wind speed for low wind values, reach a maximum and then decrease with wind speed. The difference between the theoretical and computed curves is methodology of derivation of  $p_0^{\text{crit}}$ . The shape of this curve will be essential in the analysis of our observations in Sect. 2.

In the following, we refer to  $p_{0,0}^{\text{crit}}$  as the value in the absence of wind and  $p_{0,\max}^{\text{crit}}$  as the maximum value reached by  $p_0^{\text{crit}}$ .

#### 4 Discussion

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The PCA gives us a theoretical framework to understand the results of Fig. 2. The mean fire size of the biggest wildfires (above the 95th quantile) is a growing function of <sup>15</sup> wind speed in most cases, excepted in the Mediterranean Basin for large temperature anomalies. The dependence of wildfire burnt area to the temperature anomaly and soil moisture anomaly (not shown) can be inferred theoretically by the fact that is a function of vegetation and fuel moisture. It is also dependent of the slope of the local topography but this has no consequence on the observed behavior of the burnt area. A proposed <sup>20</sup> expression of  $p_0$  is described in Alexandridis et al. (2008):

$$p_0 = p_h(1 + p_{den})(1 + p_{ved})p_s(\theta_s)$$
(10)

where  $p_h$  is a constant used for homogeneous fuel beds,  $p_{den}$  and  $p_{veg}$  are factors depending on vegetation density and type, respectively. The quantity  $\theta_s$  is the local slope in degrees and the expression of  $p_s(\theta_s)$  is given by:

<sup>25</sup>  $p_s(\theta_s) = \exp(a\theta_s).$ 



(11)

This formulation is very simple and other factors such as fuel moisture could be taken into account for more realistic simulations. We do not consider the effect of slope in this study. It may be a differentiating factor between EAST and MED, which could lead to quantitative differences between fire sizes in the two regions but for a given region the slope value is the same whatever the wind speed, vegetation state or temperature and soil moisture anomaly. The main cause for such behavioral differences is the value of  $p_0$  relative to  $p_0^{\text{crit}}$ . The final burnt area can vary drastically because of percolation as seen in Fig. 4. Figure 6 displays the fraction of burnt cells (normalized to its value after 50 time steps in the absence of wind speed and for p = 1, see Fig. 6) as a function of wind speed for 3 different values of  $p_0$ . Three  $p_0$  values are chosen in the 3 different domains:  $p_{0,1}$  set at 0.26, below the 0 wind percolation threshold,  $p_{0,2} = 0.3$  between  $p_{0,0}^{\text{crit}}$  and  $p_{0,3}^{\text{crit}} = 0.35$  above the maximum of the percolation threshold (Figs. 5 and 6).

The saturation for strong winds visible is due to the fact that p = 1 in the direction <sup>15</sup> of the wind so the fire reaches the edge of the simulation domain and the fraction of burnt cells saturates. For  $p_0$  values smaller than  $p_{0,0}^{crit}$ , percolation is only possible for high winds so the burnt area increases with wind speed similarly to what is observed over the EAST domain and over the MED domain for low temperature anomaly ( $\Delta T_2 <$ 3°C). For intermediate  $p_0$  values ranging between  $p_{0,0}^{crit}$  and  $p_{0,max}^{crit}$ , large burnt area are simulated for low wind values (smaller than about 5 m s<sup>-1</sup>) and large wind speed values

<sup>20</sup> Simulated for low wind values (smaller than about 5 ms<sup>-1</sup>) and large wind speed values (larger than about 15 ms<sup>-1</sup>) (Fig. 6). Conversely, smaller burnt area are simulated for intermediate wind ranging between 5 and  $15 \text{ ms}^{-1}$  (Fig. 6). For  $p_0$  values larger than  $p_{0,\text{max}}^{\text{crit}}$ , similar qualitative results are obtained with however larger burnt area.

We can therefore analyze Fig. 2 as if to each subfigure corresponds a value of  $p_0$ . The value of  $p_0$  corresponding to the EAST domain and MED domain for low temperature anomaly should be lower than  $p_{0,0}^{crit}$ , so that the burnt area increases with wind speed. The rate of burnt area increase with respect to wind speed is a function of the departure of  $p_0$  from  $p_{0,0}^{crit}$ . Over the EAST domain, the value of  $p_0$  is expected to be



farther below  $p_{0,0}^{crit}$  than over the MED domain. It may also explain why a change of propagation regime is observed over the MED domain when shifting from low to high temperature anomalies. Indeed, for large temperature anomalies corresponding to severe droughts associated with heatwaves ( $\Delta T_2 > 3^{\circ}$ C) over the MED domain, a value

- of  $p_0$  exceeding  $p_{0,0}^{\text{crit}}$  is expected. A simple way to understand such behavior is to consider the probability of extinction of a fire occupying one cell. Let p be the spread probability. The cell has 8 neighbors on a square grid. In the absence of wind, the extinction probability is  $(1 - p)^8$ . When wind blows, the fire spreads in the direction of the wind and diagonally. The probability of extinction becomes  $(1 - p)^3$ . Let us consider
- <sup>10</sup> very weak wind speed values for which isotropic spreading can be assumed and *p* increases with wind speed. Then the extinction probability  $(1 p)^8$  drops very quickly. When isotropic spreading does not stand anymore, then the probability of extinction transitions to  $(1 p)^3$  and increases as 3 is a lower exponent than for the isotropic spreading. With increasing wind speed, the probability of extinction decreases again.
- <sup>15</sup> So instead of enhancing wildfire propagation, moderate winds produce smaller wildfires by cutting off the propagation against the wind. Such analysis explains the 2 modes observed in the MODIS and EFFIS data over the MED domain.

The function of wind  $f(V, \theta)$  in the calculation of p is critical in the explanation of the observed behavior (see Eq. 9). The choice of this function by Alexandridis et al. (2008) was done empirically to match with the behavior of the rate of spread of real wildfires

- 20 was done empirically to match with the behavior of the rate of spread of real wildfires as a function of the wind and the angle between the wind direction and the direction of propagation. Its shape is coherent with real wildfire propagation but concerns can be raised (Fig. 3). The backpropagation is very small and the overall shape is nonelliptical. This last point could be due to the indirect impact of factors that alter the normally elliptical shape of small wildfires, such as the small variations in wind direction that make the wildfire slightly conical. Despite such concerns, this function has been
- that make the wildfire slightly conical. Despite such concerns, this function has been efficient in the simulation with PCA of the 1990 burnt scar in the Spetses island and thus does fairly well for burnt scar modeling purposes.



Another concern is that the propagation speed in the direction of the wind tends to grow very quickly with the wind speed. With the expression of  $f(V, \theta)$  used in this study, the propagation speed only becomes very high when the wind speed exceeds  $15 \text{ ms}^{-1}$  which is much too high compared to observed wildfires. Tests have been performed by using other expressions of  $f(V, \theta)$ , such as the elliptical form and varying the  $c_1$  and  $c_2$  coefficients of Eq. (2), and no qualitative difference in the behavior of  $p_0^{\text{crit}}$ has been evidenced (not shown). In detail, the choice of  $c_1$  and  $c_2$  coefficients can be questioned, as it was only validated on one case study. A key issue is to know if  $p_{0,\text{max}}^{\text{crit}}$  is reached for a positive value of the wind speed, which conditions the existence of the "2 modes" regime. Figure 7 shows the value of the ratio  $\left(p_{0,\text{max}}^{\text{crit}} - p_{0,0}^{\text{crit}}\right) / p_{0,0}^{\text{crit}}$  as a function of  $c_1$  and  $c_2$  calculated from Eq. (9). For a value of  $c_2$  larger than about  $1.25c_1$  this property is verified. This condition can be interpreted as the backpropagation of the wildfire being impaired in a sufficiently large angle domain by the wind.

A possible reason for the burnt area to decrease with wind speed for moderate wind speed over the MED domain is that in order to reach a  $p_0$  value larger than  $p_{0,0}^{crit}$ , higher flammability of the fuel bed or a more fire-prone environment are needed. This could be caused by the severe heatwaves and droughts regularly occurring in the MED region, which would lead to lower fuel moisture. This could also be due to different vegetation type between the EAST and MED regions.

#### 20 5 Conclusions

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The behavior of the burnt area with temperature anomaly and wind speed was analyzed with the help of MODIS observations and ERA-Interim reanalysis. In the EAST region and in MED for low temperature anomalies, the burnt area is a growing function of wind, which is rather intuitive. For large temperature anomalies in the MED region, corresponding to severe heatwaves and droughts, the burnt area behaves differently. It displays a "2 modes" shape with a minimum value of burnt area for medium values



of the wind speed. The possible cause of such counter-intuitive behavior has been investigated theoretically by means of probabilistic cellular automata (PCA). It was found that when the wind speed is moderate, the backpropagation is impaired when the forward propagation is not sufficiently fast to compensate the loss in burnt area. Therefore

- if the wind is not strong enough to make the propagation along the wind direction "sure enough", percolation happens in the PCA resulting in smaller wildfires. The percolation threshold is never reached in the EAST region. It is only exceeded for moderate wind speed in the MED region. Such shift between 2 regimes of propagation may be caused by the favorable occurrence of severe heatwaves and droughts in the MED region.
- <sup>10</sup> A natural follow-up of this study would be to model the impact of fuel moisture on the propagation probability by adapting the known variations of the rate of spread of the fire with this quantity (Sharples, 2008). It could be complemented by a thorough analysis of the behavior of dead and live fuel moisture in Mediterranean ecosystems with respect to air temperature anomaly and the incorporation of such relation in the spression of the probability of fire propagation. However, fuel moisture is a quantity
- which is hard to assess. Live fuel moisture depends on the plant species and dead fuel moisture on the size of the fuel particle. Finally, for a more holistic approach, a more realistic fire propagation modelling framework such as FIRETEC would be needed to better identify and quantify the processes driving fire propagation.

#### 20 Appendix A

In this Appendix, the calculation of C(V) is detailed. The quantity C(V) corresponds to the sum of the probabilities that the fire propagates from a neighboring cell to the center cell (Fig. 8).

For all 8 neighbors we can write:

25  $p_i = k_i p_0$ 



(A1)

with  $k_i$  being a coefficient depending on the direction of propagation. One can write

$$k_i = \gamma \exp\left\{V\left[c_1 + c_2(\cos\theta_i - 1)\right]\right\}$$
(A2)

with  $\theta_i$  being the angle between the direction of propagation and the direction of the wind for the *i* component. One can write:

$$_{5} \quad k_{1} = k_{7} = \gamma \exp\left\{V\left[c_{1} - c_{2}\left(1 - \sqrt{2}/2\right)\right]\right\}$$
(A3)

$$k_{2} = k_{6} = \exp[V(c_{1} - c_{2})]$$
(A4)  

$$k_{3} = k_{5} = \gamma \exp\left\{V\left[c_{1} - c_{2}\left(1 + \sqrt{2}/2\right)\right]\right\}$$
(A5)

$$k_{4} = \exp[V(c_{1} - 2c_{2})]$$
(A6)  

$$k_{8} = \exp(Vc_{1}).$$
(A7)

<sup>10</sup> The quantity C(V) is the sum of the  $k_i$  coefficients and thus writes:

$$C(V) = \exp(Vc_1) + 2\exp[V(c_1 - c_2)] + \exp[V(c_1 - 2c_2)]$$
  
+  $2\gamma \exp\left\{V\left[c_1 + c_2\left(\frac{\sqrt{2}}{2} - 1\right)\right]\right\}$   
+  $2\gamma \exp\left\{V\left[c_1 - c_2\left(\frac{\sqrt{2}}{2} + 1\right)\right]\right\}$ 

which can be simplified into:

<sup>15</sup> 
$$C(V) = \exp(Vc_1) \times \left\{ [1 + \exp(-Vc_2)]^2 + \sqrt{2} \left( \exp\left[Vc_2\left(\frac{\sqrt{2}}{2} - 1\right)\right] + \exp\left[-Vc_2\left(\frac{\sqrt{2}}{2} + 1\right)\right] \right) \right\}.$$



(A8)

(A9)

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**Figure 1.** 10 years cumulated burnt area (in ha) for each regridded fire product grid cell over the Euro-Mediterranean region (in colors). The Mediterranean domain (MED) and central eastern Europe domain (EAST) are identified by black boxes.





**Figure 2.** Wildfire burnt areas (BA) (i.e. 95th quantile of the probability density function) as a function of 10 m wind speed for the EAST (top row panels) and MED domains (bottom row panels) for  $\Delta T_2 < 3$  °C (left column panels) and  $\Delta T_2 > 3$  °C (right column panels). The black line shows the 95th quantile of the burnt area probability density function and the red lines show the lower and upper 1 –  $\sigma$  limits.





**Figure 3.** Function  $f(V, \theta)$  for values of wind speed V equal to 0 (blue), 5 (green), 10 (red), 20 (cyan) and  $30 \text{ m s}^{-1}$  (purple).





**Figure 4.** Evolution of the fraction of burnt cells  $y_2^{\text{final}}$  – with  $y_i$  the fraction of cells in state *i* in the PCA grid – as a function of  $p_0$  when the fire is extinct in a grid constituted of 101 × 101 cells. The simulation is performed in the absence of wind.





**Figure 5.**  $p_0^{\text{crit}}$  as a function of wind. Dashed curve is the theoretical value found using meanfield approximation and the full curve is the numerically derived value. Dotted lines ( $p_{0,1}$ ), ( $p_{0,2}$ ) and ( $p_{0,3}$ ) correspond to the curves plotted in Fig. 6.





**Figure 6.** Fraction of burnt cells  $y_2/y_2^{50}$  (normalized to its value after 50 time steps in the absence of wind speed and for p = 1) as a function of wind speed for  $p_0$  values set at  $p_{0,1} = 0.26$  (solid line),  $p_{0,2} = 0.3$  (dashed line) and  $p_{0,3} = 0.35$  (dotted line) in Fig. 5.





**Figure 7.** Ratio  $\left(p_{0,\text{max}}^{\text{crit}} - p_{0,0}^{\text{crit}}\right) / p_{0,0}^{\text{crit}}$  for different values of the  $c_1$  and  $c_2$  parameters. The black diamond sign shows the  $(c_1, c_2)$  couple chosen by Alexandridis et al. (2008).





**Figure 8.** Schematic of the calculation of C(V).

