



August 12, 2005



ISMM 07

*Rio de Janeiro
October 2007*

*Random spread
and
Forest Fires*

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Classical maps

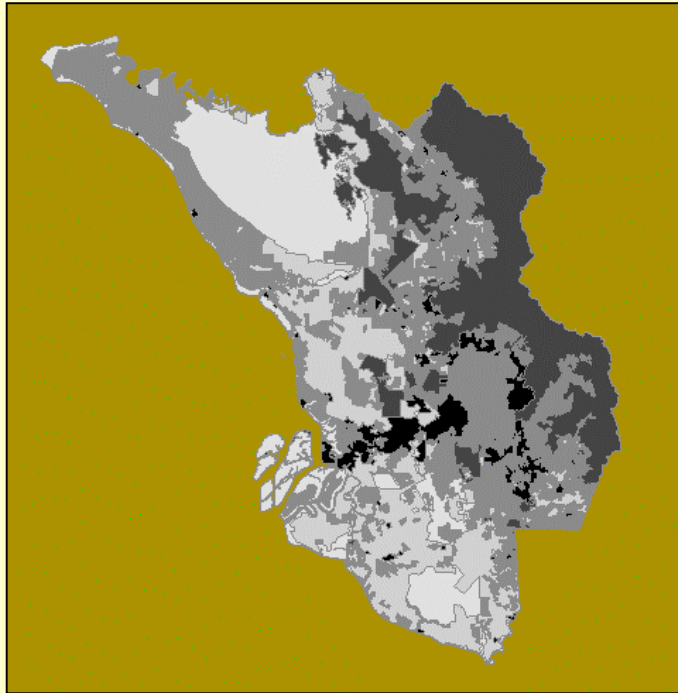


- Usual maps symbolize objects that *do exist* in the physical world,
- e.g. the Malaysian peninsula is *prior* to any geographer, and independent of him.

Risks maps

Consider now the two maps of forest fires parameters

Fuel amount : θ



Relative Scale :
the lighter, the more fuel

Fire spread : ρ



*Scale : 0 meter/mn (black)
to 3 meters/mn (light)*

Selangor State, Malaysia

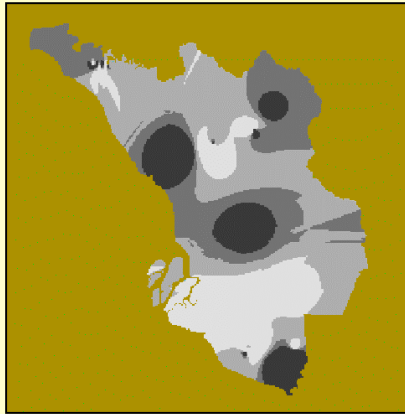
A missing link

- When we draw the map of a risk, e.g. a spread fire, we describe a *scientific assumption*.
- There is *no actual object* in the physical world that the map symbolizes: it represents potentialities only.
- If we want to go from potentialities to the actual events, an *additional element* turns out to be necessary.

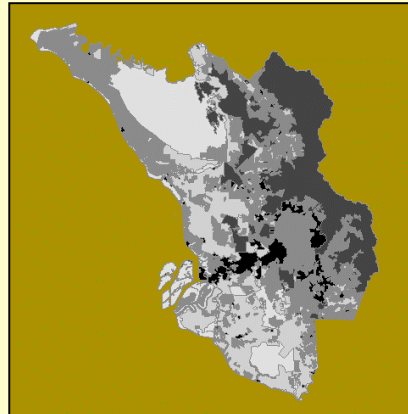
.....But how to handle it ?

The missing link

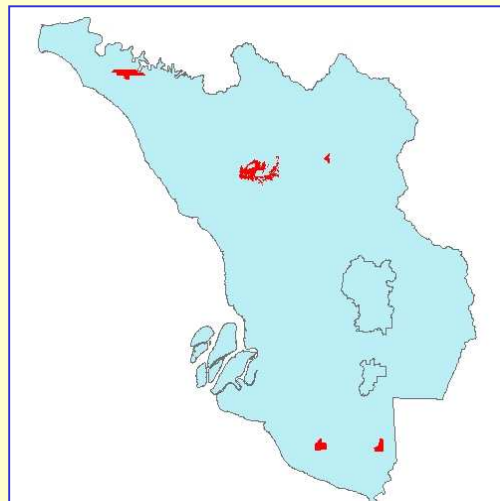
Fire spread : ρ



Fuel amount : θ



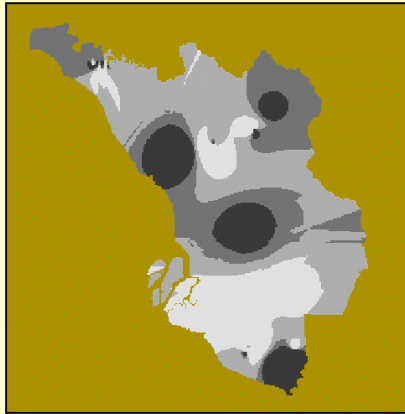
- How to go from these two maps to the burnt regions?
- Can we derive from them the duration of a fire ?
- and the size distribution of the burnt regions ?
- **Without a model, surely not !**



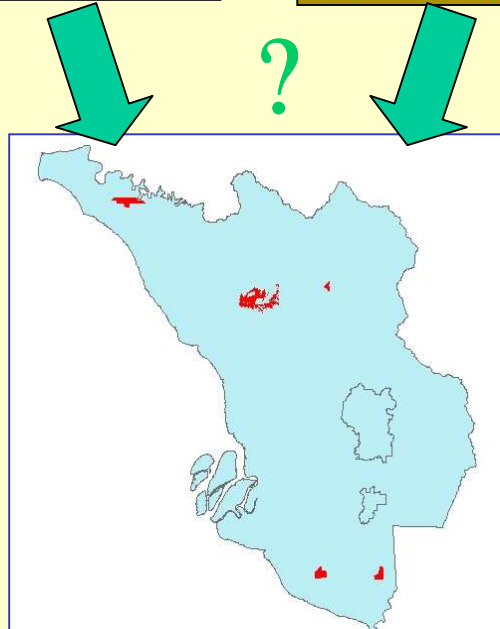
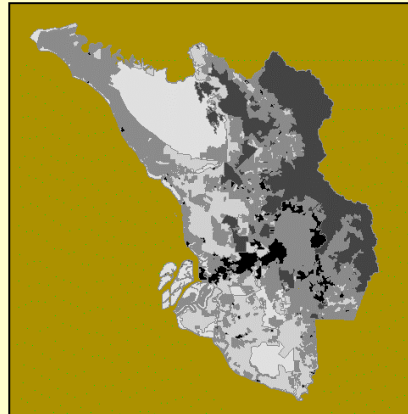
Scars from 2000 to 2004

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A Pde ?

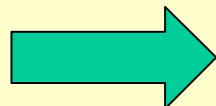
- Classically, each Pde summarizes a conflict of elementary variations,
- but here, we are facing a space-time process whose all parameters act in the sense of the space invasion.

How to introduce an element that balance the invasion?

A Pde ?

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How to introduce an element that balance the invasion?

 the *hot spots* provide a third piece of information.

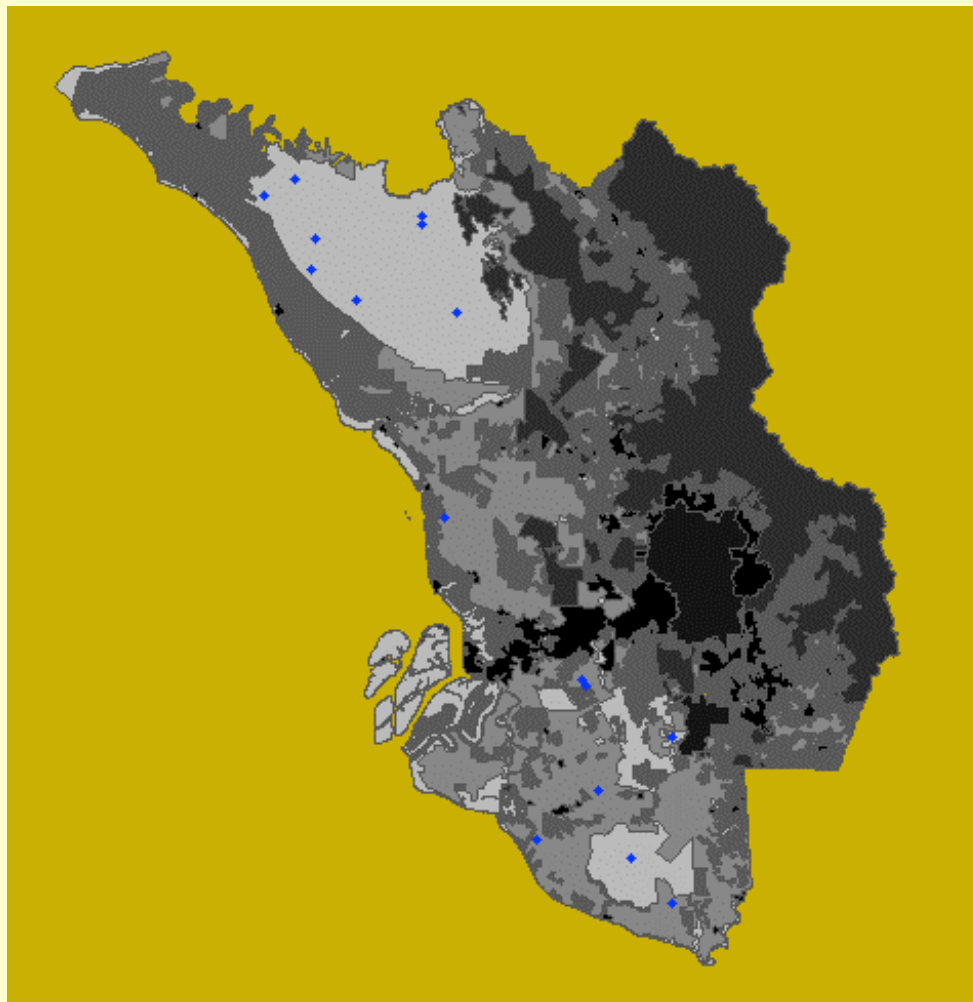


Hot Spots

In red, the hot spots detected in Selangor, on August 12, 2005

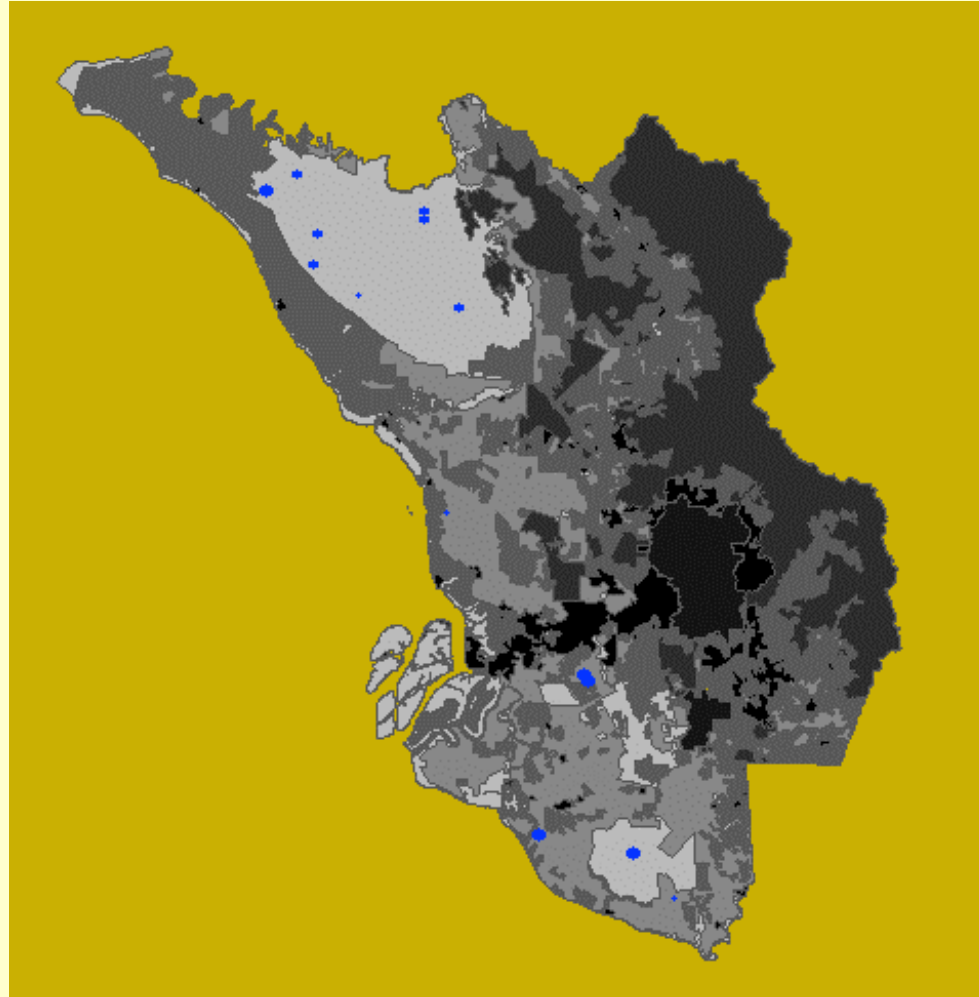
The spread map based Pde

Initial hot spots

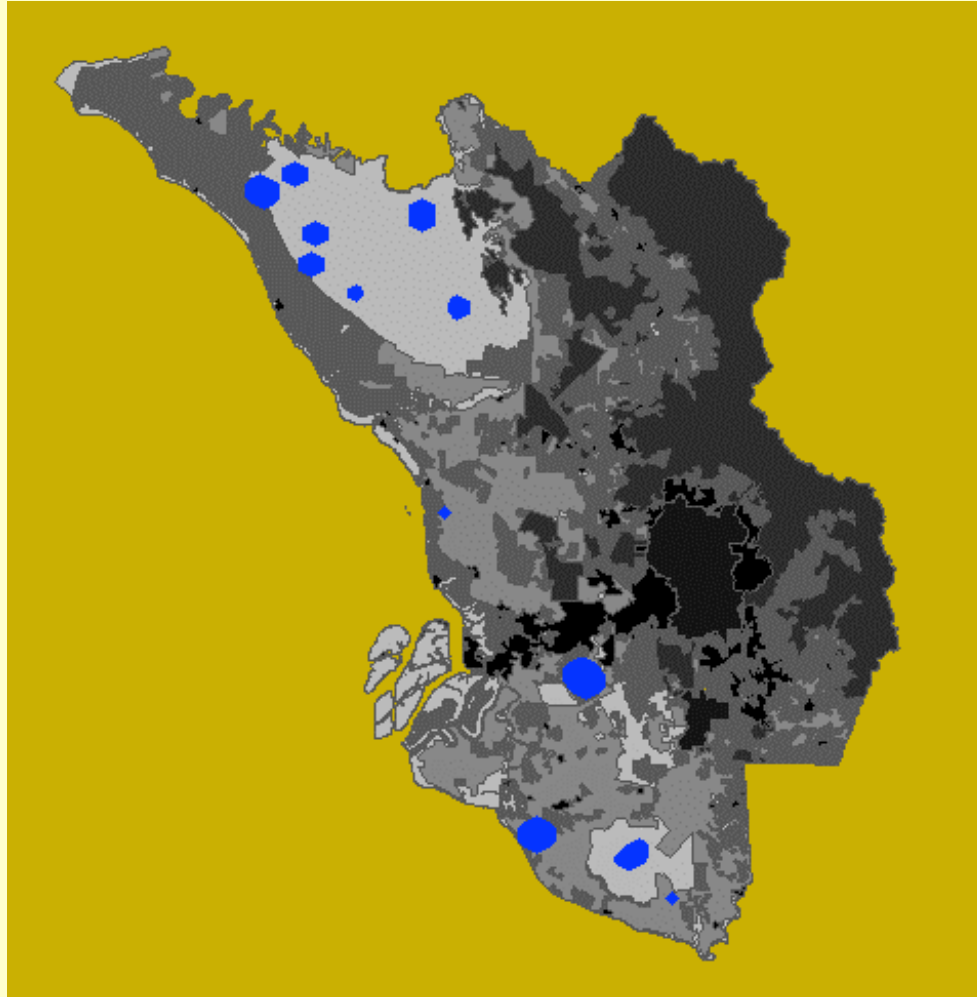


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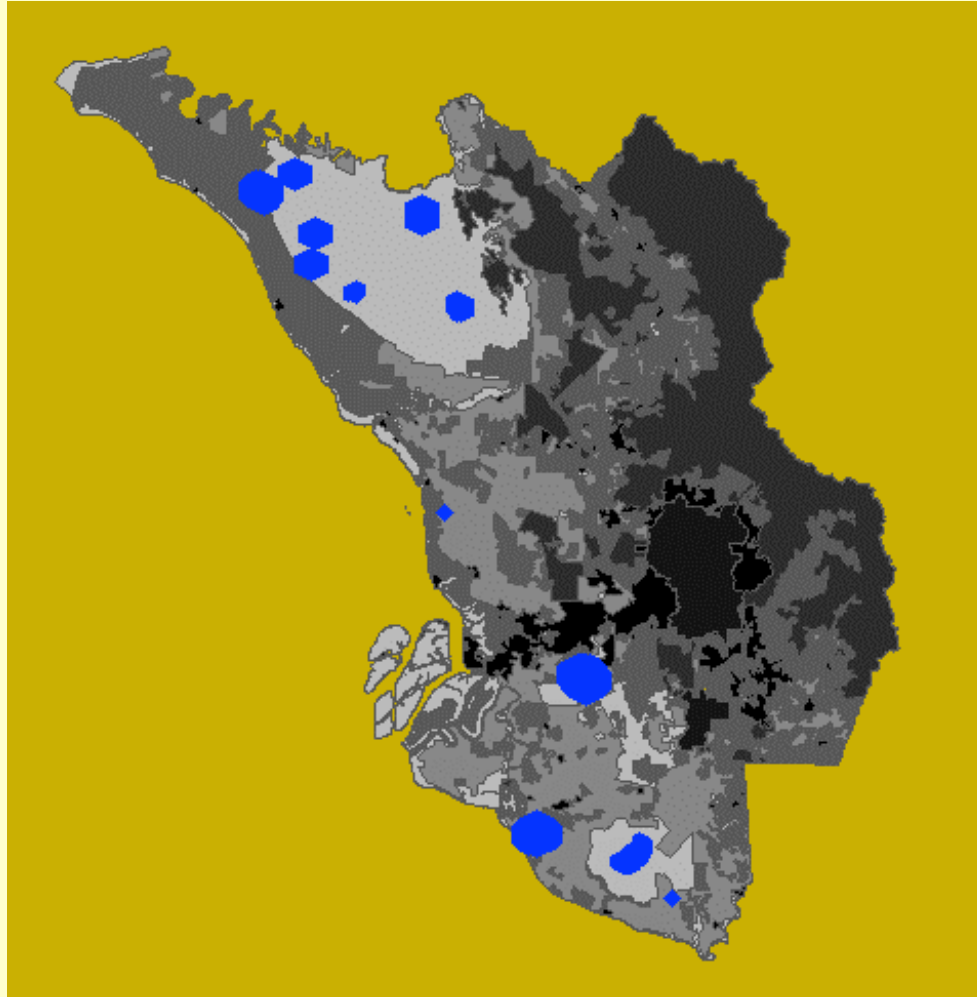
The speed of expansion is given by the spread rate map



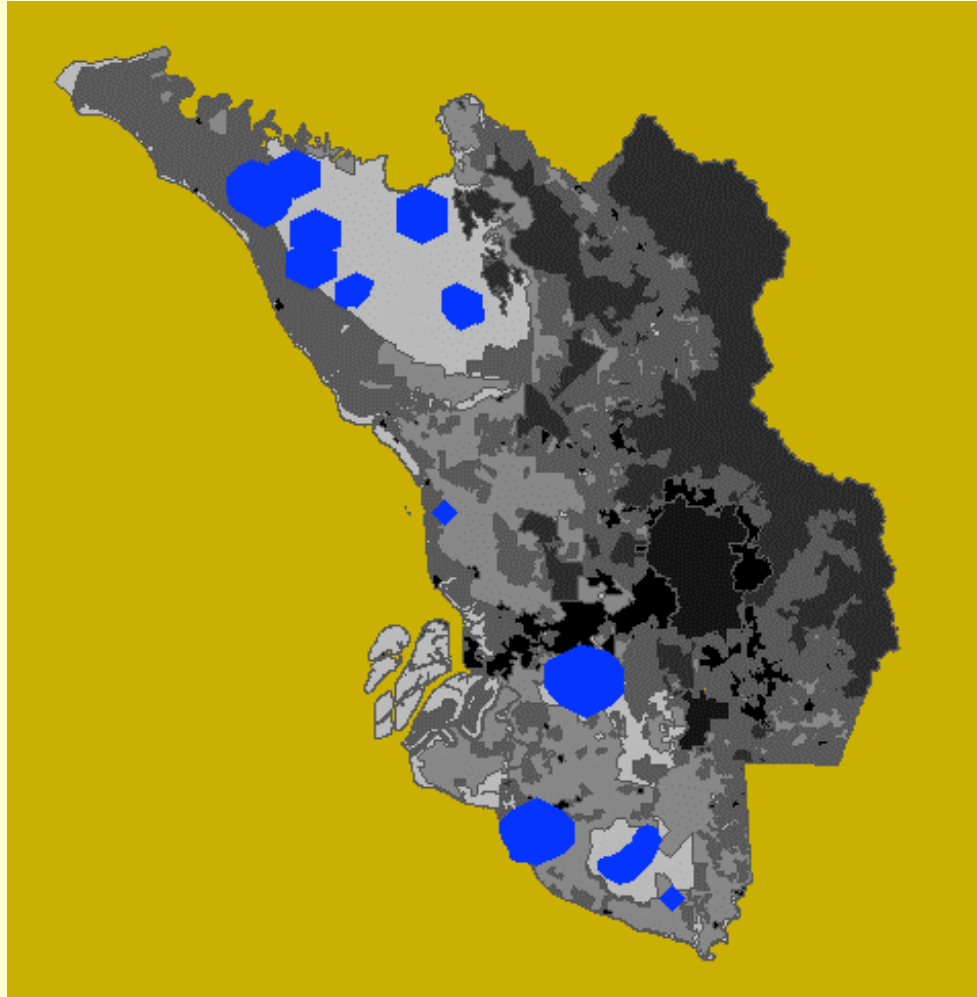
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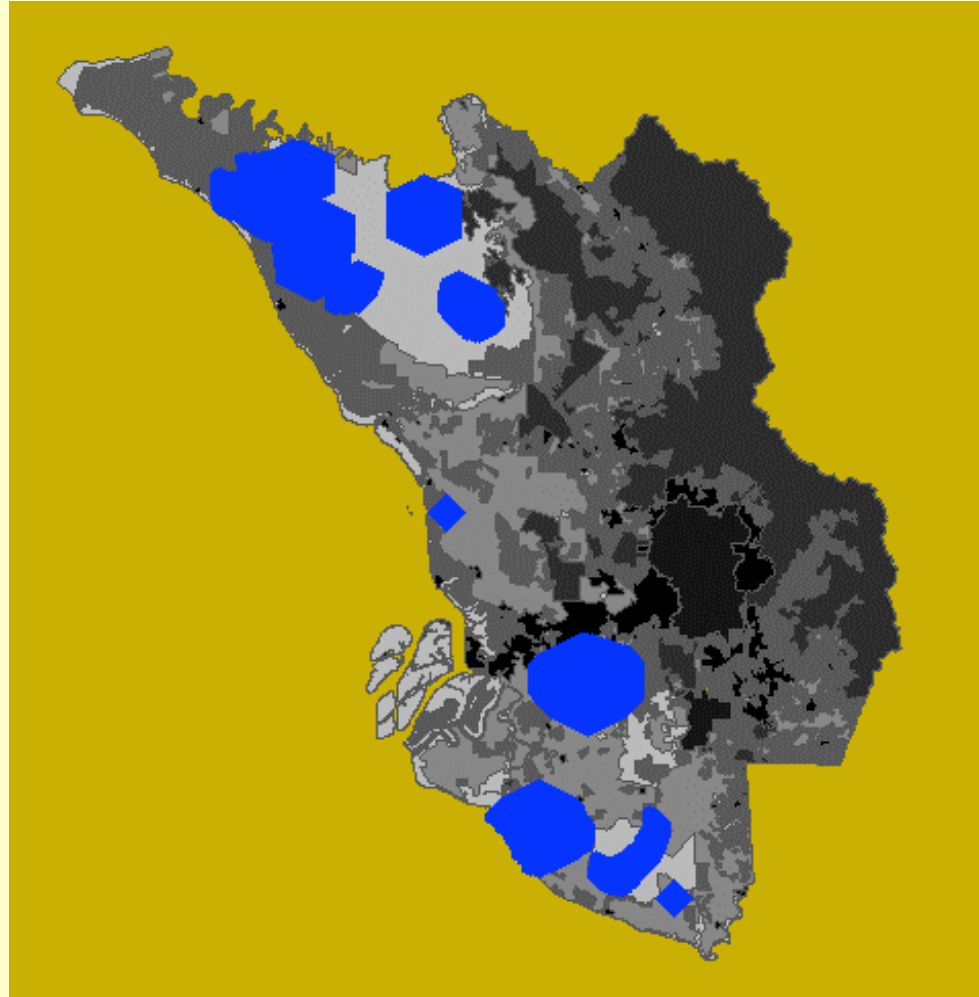


The spread map based Pde



The spread map based Pde

When the seats expand according to the spread map, then they progressively *invade the whole space*



A theoretical tool

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 - ⇒ daily evolution of the fire;
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- It results in *predictions* for the
 - ⇒ daily evolution of the fire;
 - ⇒ the possible natural extinction,
 - ⇒ location of the scar regions.
- Finally, it also leads to *simulations* of the fire propagation.

Random Closed Sets (or RACS)

Let \mathbb{R}^d be the Euclidean space of dimension d ,

$\mathcal{F} = \mathcal{F}(\mathbb{R}^d)$ denotes the family of all closed sets of \mathbb{R}^d ,

$\mathcal{K} = \mathcal{K}(\mathbb{R}^d)$ the family of all compact sets.

- **σ -algebra** : Given an element $K \in \mathcal{K}$ consider the class $\mathcal{F}(K)$ of all closed sets that miss the compact set K . As K spans the family \mathcal{K} , the classes $\{ \mathcal{F}(K), K \in \mathcal{K} \}$ generate a σ -algebra.
- **RACS** : Moreover, as \mathcal{F} is a compact space, one can weight σ by probabilities P . Then each triplet (\mathcal{F}, σ, P) *defines a RACS*.

This abstract definition of a RACS goes back to G.Matheron and D.G.Kendall. However, these authors made their approach more tractable by proving the following result.

The Matheron-Kendall theorem

- **Characteristic Theorem** : Every RACS X is characterized by the datum of the probabilities

$$Q(K) = \Pr \{ K \subseteq X^c \} \quad K \in \mathcal{K}.$$

Conversely, a family $\{Q(K), K \subseteq \mathcal{K}\}$ defines a unique RACS if and only if $1 - Q(K)$ is a **Choquet capacity** such that

$$0 \leq Q \leq 1 \quad \text{and} \quad Q(\emptyset) = 1.$$

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- **Choquet' capacity** : numerical function Q on \mathcal{K} such that

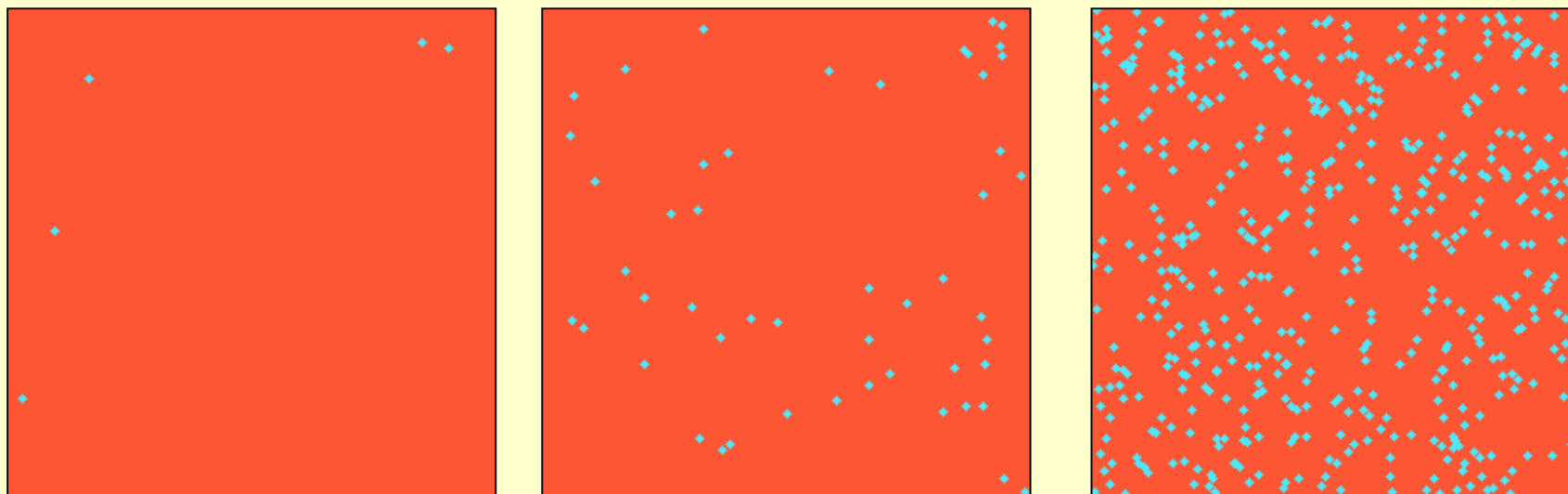
$$1- S_1(K; K_1) = Q(K) - Q(K \cup K_1)$$

$$S_n(K; K_1 \dots K_n) = S_{n-1}(K; K_1 \dots K_{n-1}) - S_n(K \cup K_n; K_1 \dots K_{n-1})$$

$$2- K_n \downarrow K \quad \text{implies} \quad Q(K_n) \uparrow Q(K)$$

Poisson points

- Here are usual simulations of Poisson points (slightly dilated by a rhomb for the display)



They are «*usual*» in that the intensity $\theta(x)$ is *constant*

Poisson points

A basic random set is that of the Poisson points, defined as follows

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The functional $Q(K)$ of Poisson points θ is

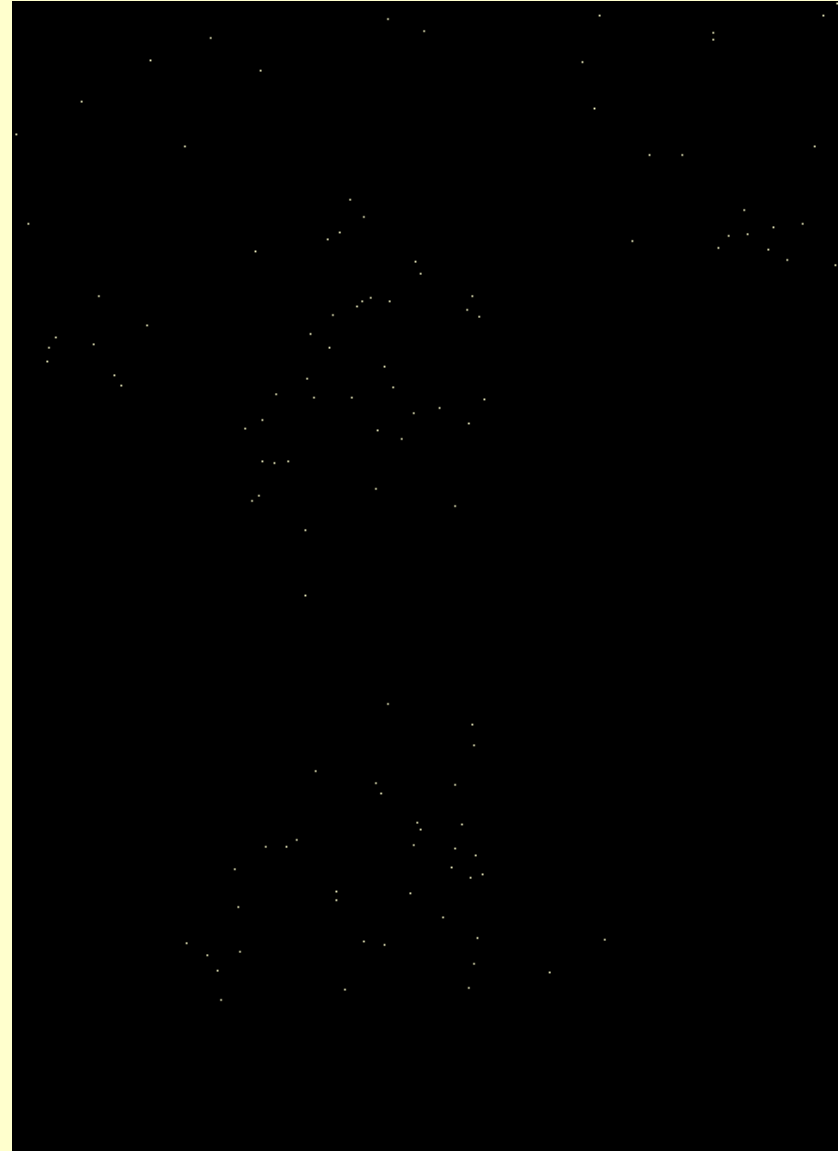
$$Q(K) = \exp\left\{-\int_K \theta(dx)\right\} = \exp\{-\theta(K)\}$$

Regionalized Poisson points

In some cases the intensity of the Poisson points can also *vary* through the space ...

We still have the probability $\theta(x) dx$ of one point in dx ,

but θ is now an underlying function of the space



Regionalized Poisson points

As θ is multiplied by
a constant factor,
the number of points
increases



Regionalized Poisson points

And more again...

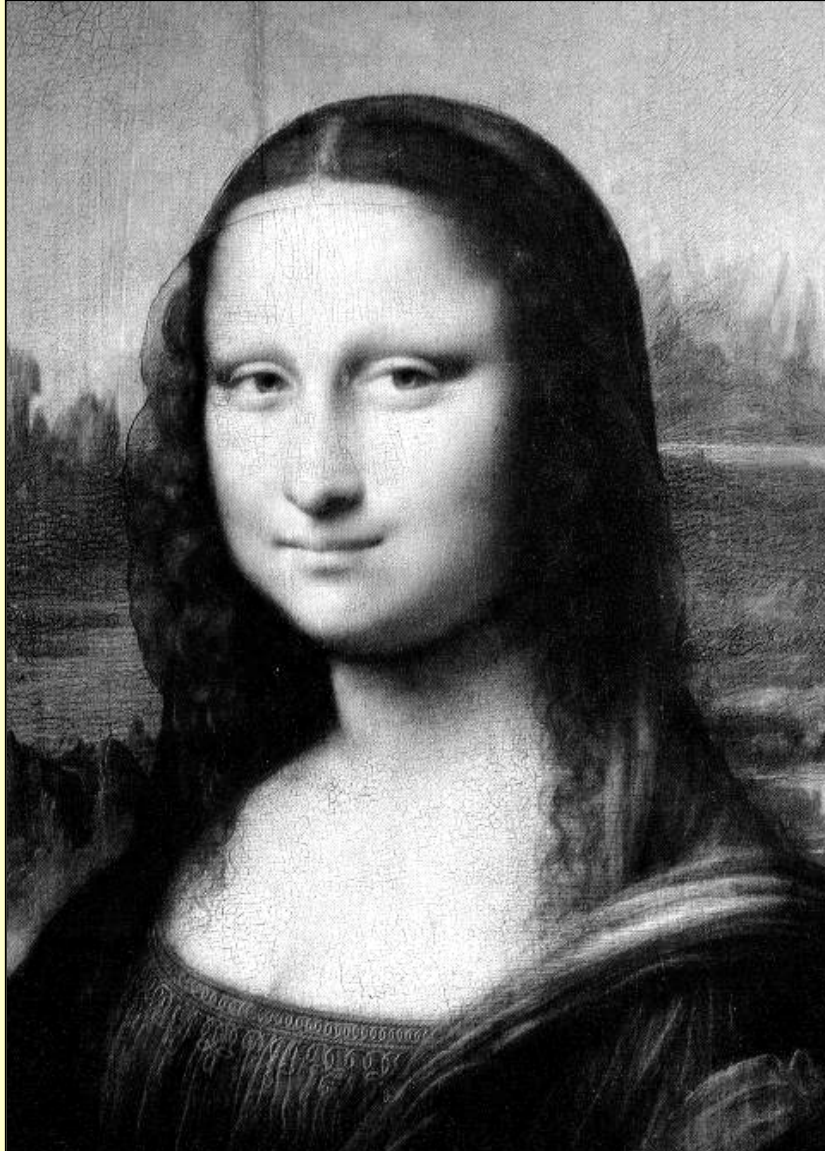


Regionalized Poisson points

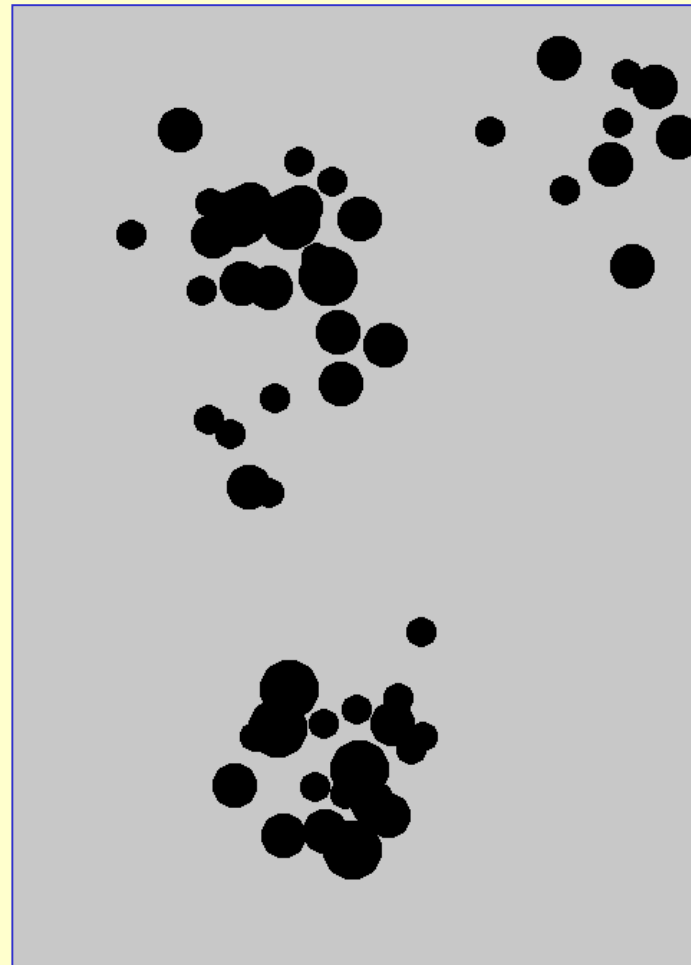
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Regionalized Poisson points



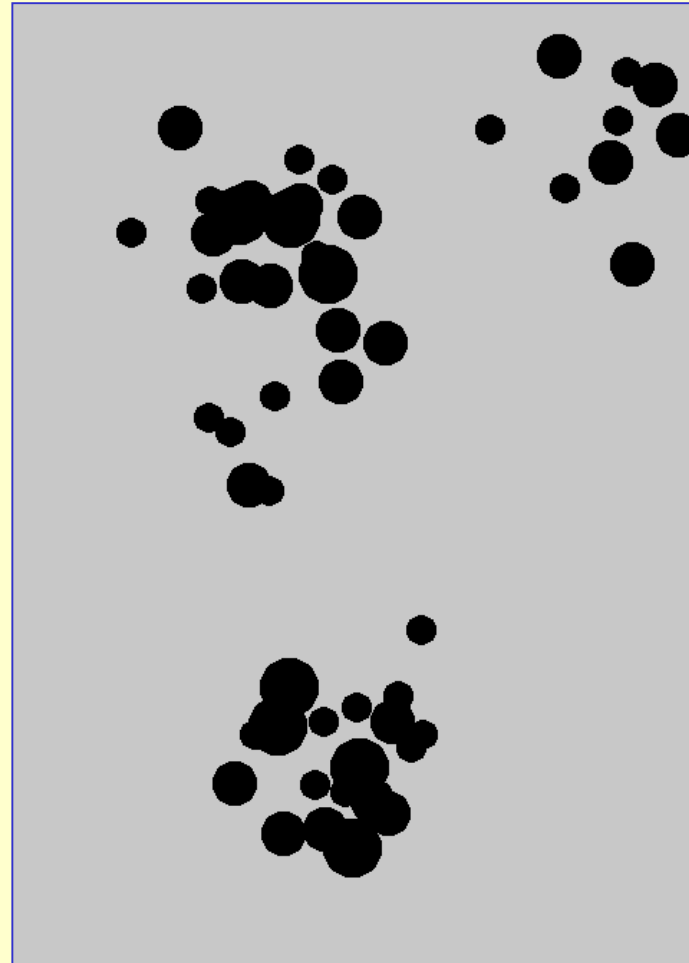
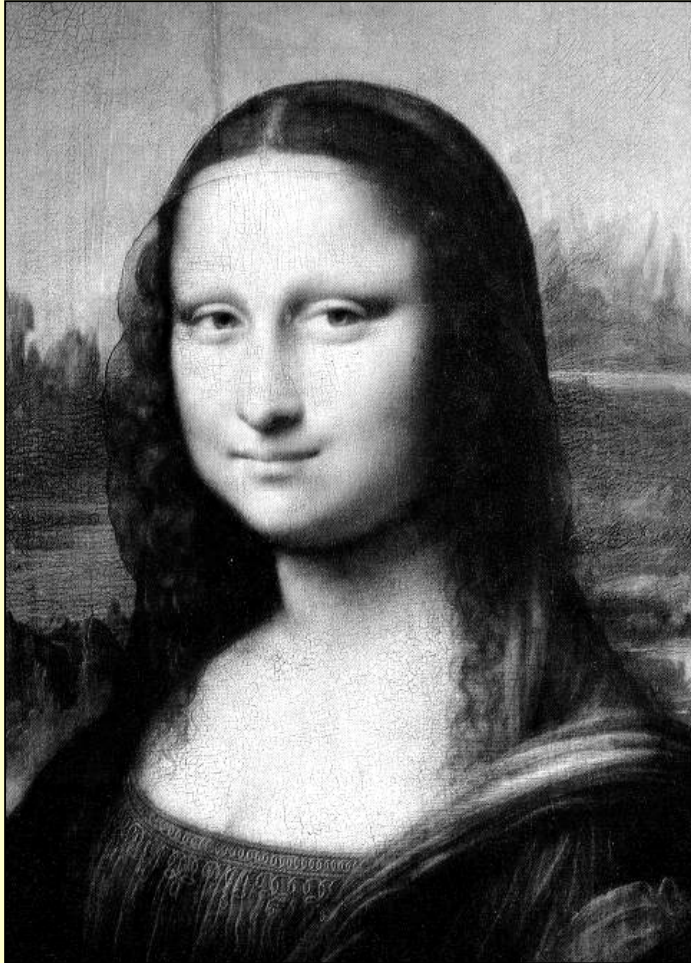
Boolean random set



*variable
intensity*
 $\theta(x)$

*variable
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Two parameters

Just as a Boolean random set , a random spread depends on

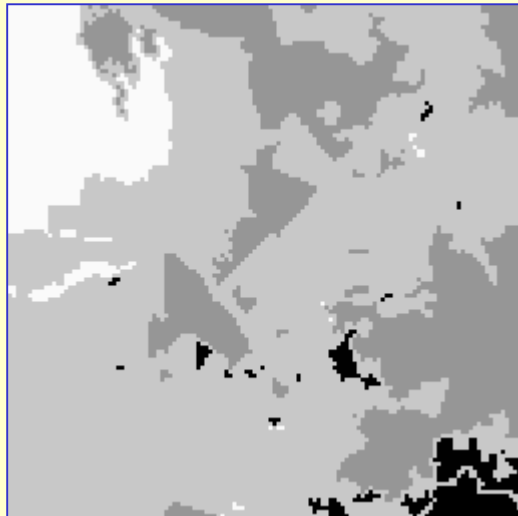
- The *the intensity θ , non negative numerical function*
- The *dilation δ , a set function $\mathbb{R}^d \rightarrow \mathcal{P}(\mathbb{R}^d)$*

Two parameters

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- The *the intensity θ , non negative numerical function*
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Here we take for set $\delta(x)$ the disc of radius ρ at point x



Point intensity θ

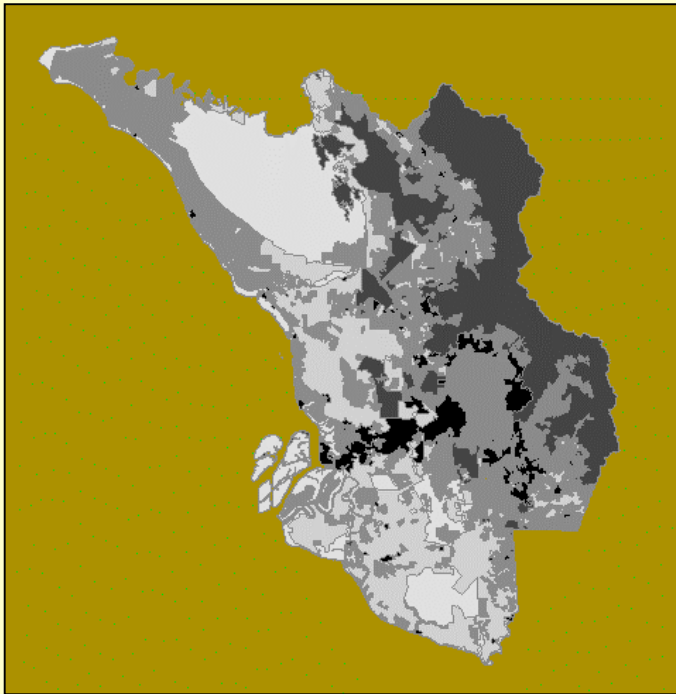


Dilation radius ρ

Parameters maps

The two previous maps are details of the following risk maps

Fuel amount : θ



Relative Scale :
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Fire spread : ρ

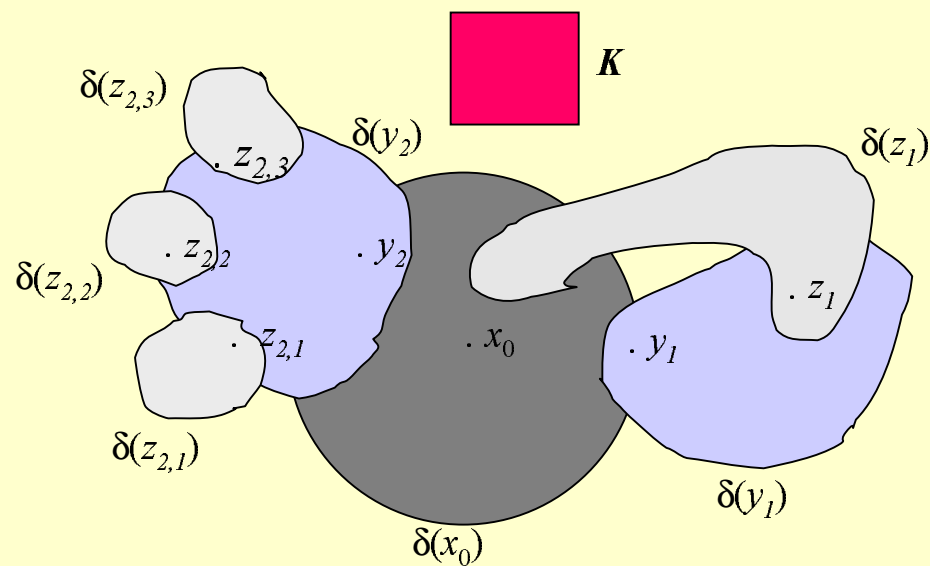


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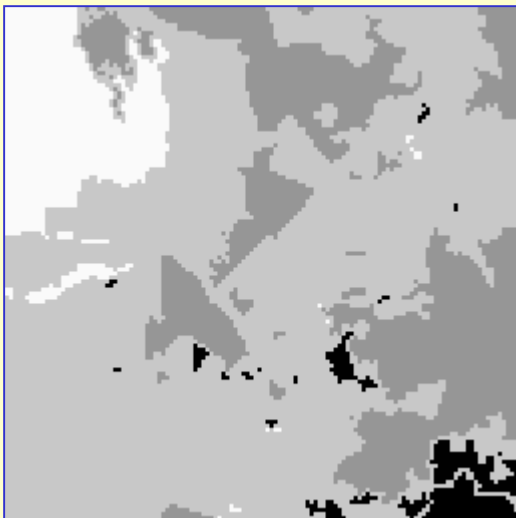
(Selangor State, Malaysia)

Random Spread

The idea of a random spread is the following



Iterated spread



Point intensity



Dilation radius

Fire $X_n = \delta(I_{n-1}) = \delta \circ [\beta]^{n-1} (I_0)$

Seat $I_n = \beta^n(I_0) = \cup \{ \delta(x_i) \cap J_i, x_i \in I_1 \}$

Examples of iterated Spread

Functional of the Boolean set

The Boolean Random set $X(\theta, \delta)$ is characterized by the probabilities $Q(K)$ that K misses the RACS, for all compact sets $K \subset \mathbb{R}^d$ (Choquet characteristic). We have that

$$Q(K) = \exp \left\{ - \int_{\zeta(K)} \theta(d\mathbf{x}) \right\} = \exp \left\{ - \theta[\zeta(K)] \right\}$$

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where ζ is the *reciprocal* of δ , i.e.

$$\mathbf{x} \cap \zeta(K) \neq \emptyset \quad \Leftrightarrow \quad \delta(\mathbf{x}) \cap K \neq \emptyset$$

Functional of iterated spread

Let us calculate the functionals $Q_1 \dots Q_n$ of spreads $X_1 \dots X_n$.

- The first step is just Boolean, so that the Choquet characteristic

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- Now, to say that K misses the n^{th} fire starting from x is equivalent to saying that K misses the $(n-1)^{\text{th}}$ fire from y , cond. upon $y \in \delta(x_0)$.

This results in the induction relation

$$Q_n(K) = \exp \left[1 - \int_{\zeta(x)} \theta(dy) Q_{n-1}(K | y) \right]$$

Reciprocal dilation

- *Reciprocal dilation*: Again we meet the reciprocal dilation ζ of δ i.e. such that

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- **Variable δ** But in the application to forest fires, δ varies from 1 to 5 from place to place. Which conditions must we demand to δ to get a non trivial expression for

$$\exp \left[- \int_{\zeta(\mathbf{K})} \theta(dz) g(z) \right] ?$$

Compact dilation

Here, the convenient class that of the *compact dilations* δ .

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- 1- the structuring function $x \rightarrow \delta(x)$ is u.s.c. from \mathbb{R}^d into \mathcal{K}
- 2- the union $\bigcup \{\delta_{-x}(x), x \in \mathbb{R}^d\}$ has a compact closure.

The second axiom implies that when x is far away enough, then $\delta(x)$ surely misses K

Compact dilation

- When δ is compact, then
 - ζ also is compact,
 - δ and ζ are u.s.c. mappings from \mathcal{F} to \mathcal{F} and from \mathcal{K} to \mathcal{K}

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The following result shows that compact dilations model the *geographical maps*, with their discontinuities (fires that stop at a river, for example)

- Let $\delta(x)$ be the disc of centre x and radius $r(x)$. When

$$\mathbf{x} \rightarrow \mathbf{r}(\mathbf{x}) \text{ is u.s.c.} \quad \text{and} \quad \mathbf{r}(\mathbf{x}) < \mathbf{r}_{\max} < \infty$$

then both δ and ζ are compact.

Scars

- Does the random spread model fit with actual fires data ?
- We can match the « *scars* » left by the fires union Y_n of all spreads X_i from steps 1 to n

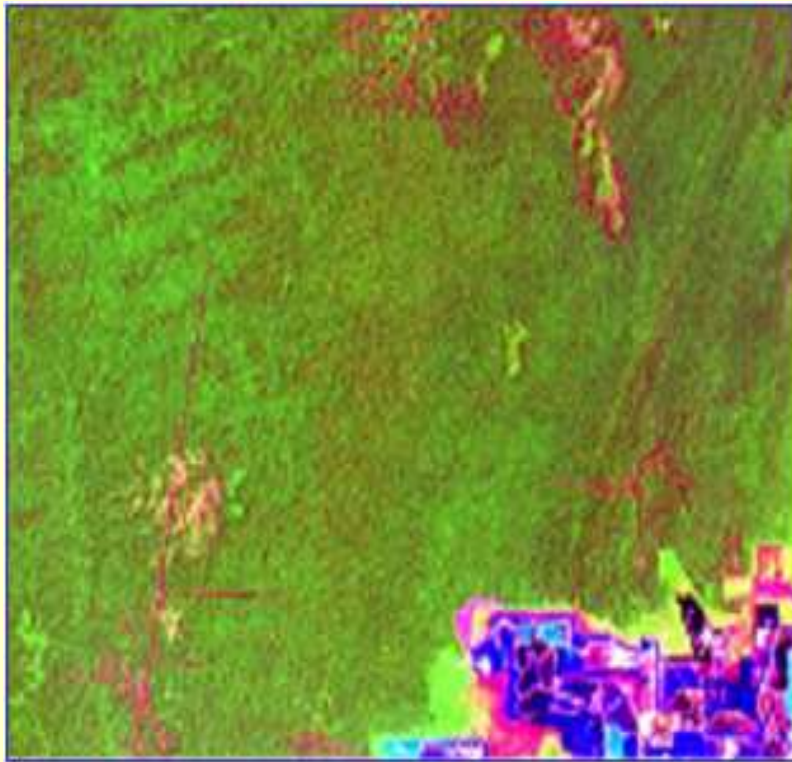
$$Y_n = \cup \{X_i, 1 \leq i \leq n\}$$

- But what happens after a long time, for Y_∞ ?

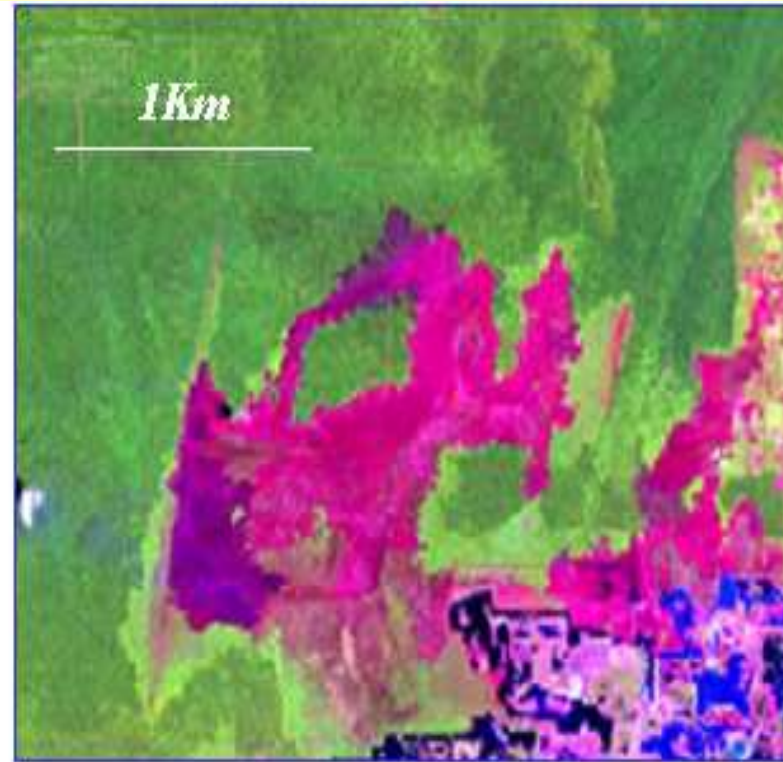
Does the fire stop ? Does it expand indefinitely?

Scars

Example of a scar : A same region in 2000 and in 2004



a)



b)

Upper bounds

For finding an upper bound the scar Y_n , introduce the parameter

$$s(\mathbf{x}) = \int \delta(\mathbf{x}) \theta(d\mathbf{x})$$

- When $s(\mathbf{x}) < s_{\max} < 1$ then the scar Y_n is upper bounded by the Boolean RACS of primary grain $\delta(\mathbf{x})$ and of intensity

$$\theta(\mathbf{x}) / 1 - s_{\max}$$

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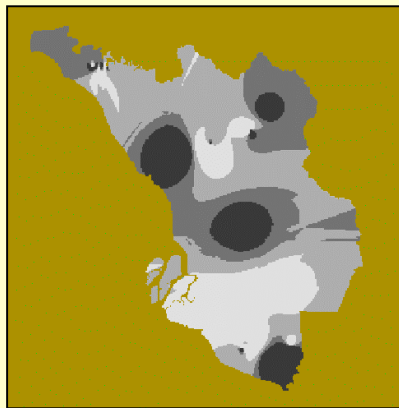
- When not, the scar can expand indefinitely.

This suggests to compare the map of $s(\mathbf{x})$ with the actual scars.

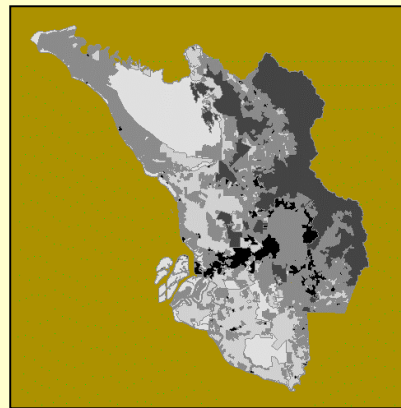
Scar function

The scar function is the product of our two input maps

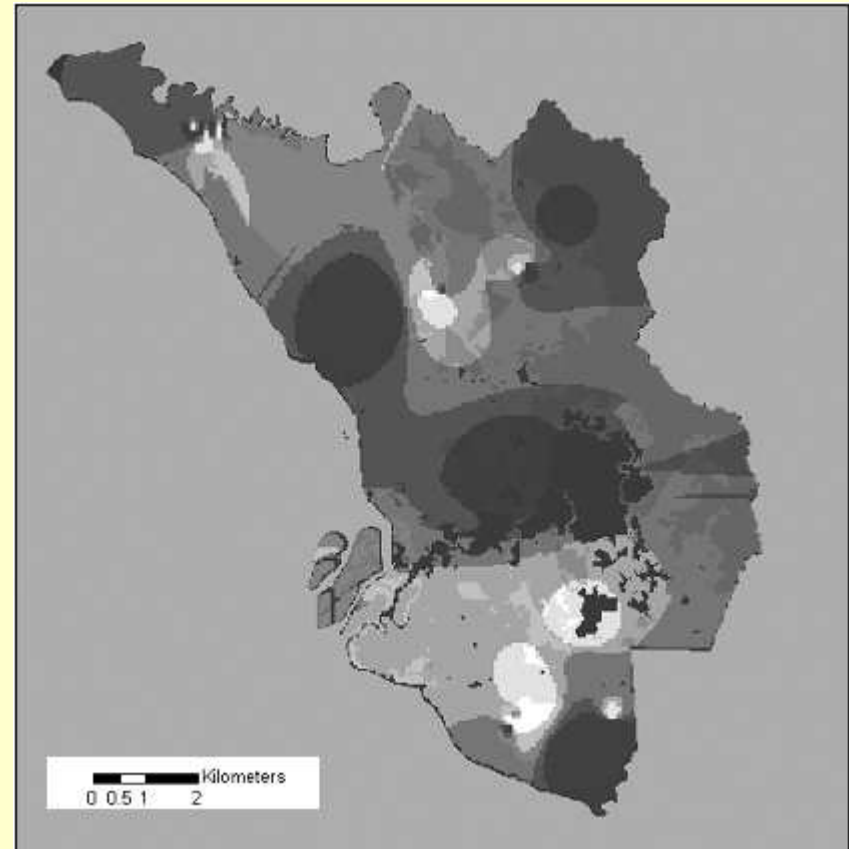
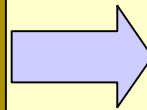
$$s(\mathbf{x}) = 2\pi \cdot \rho(\mathbf{x}) \theta(\mathbf{x})$$



Spread radius ρ



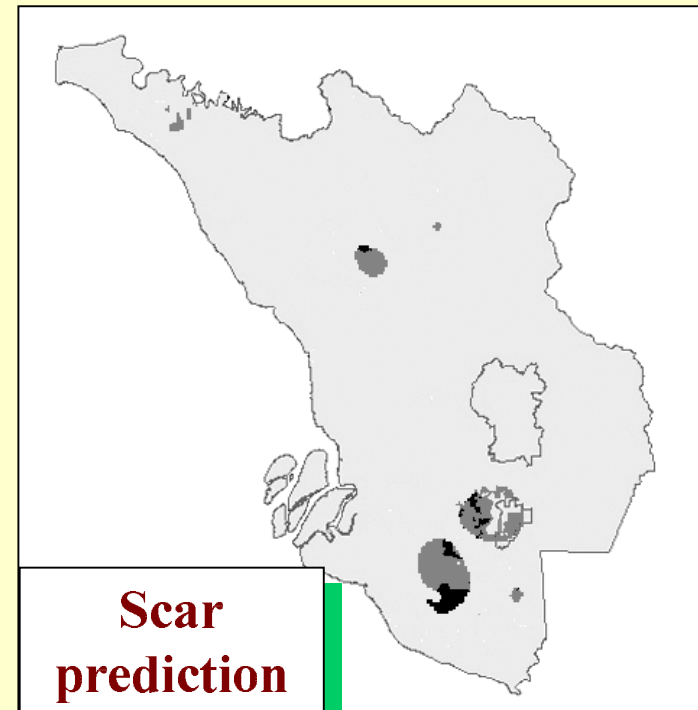
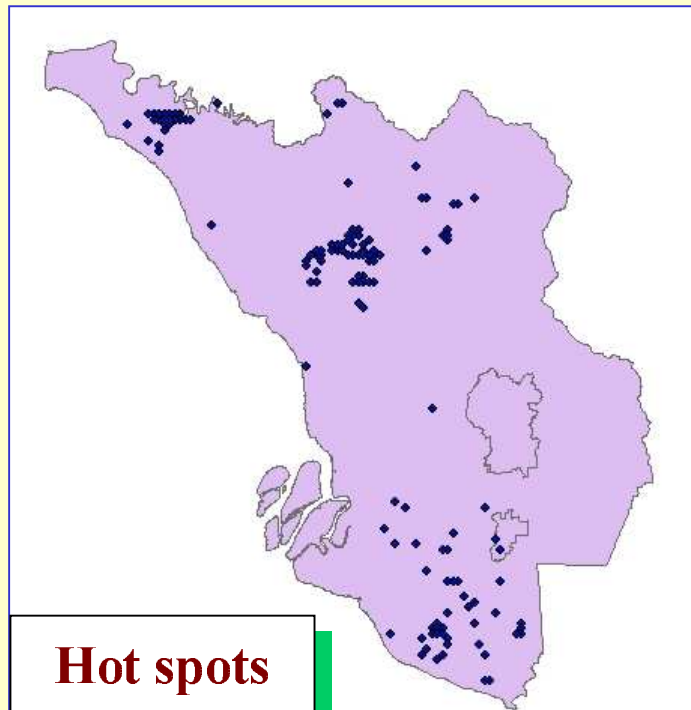
Fuel amount θ/k



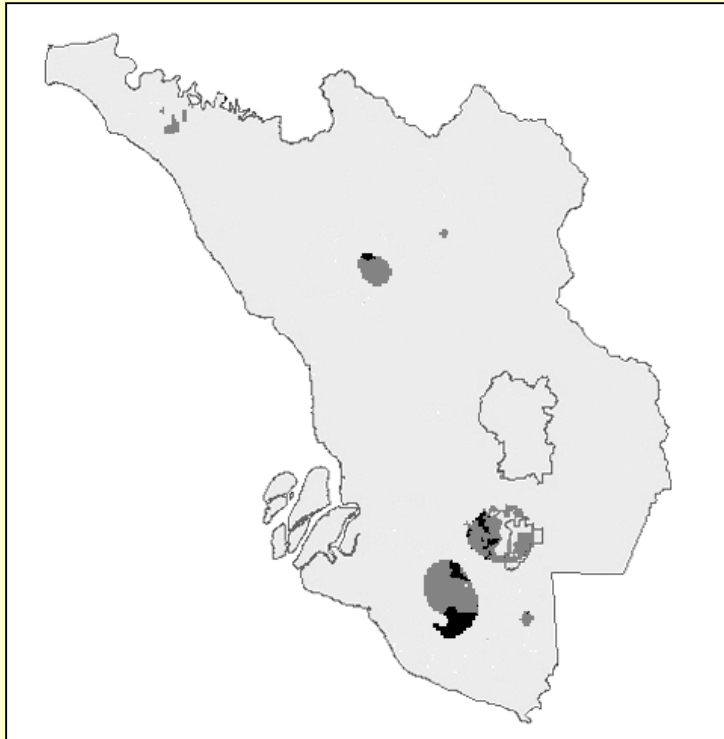
Scar function of Selangor

Hot spots

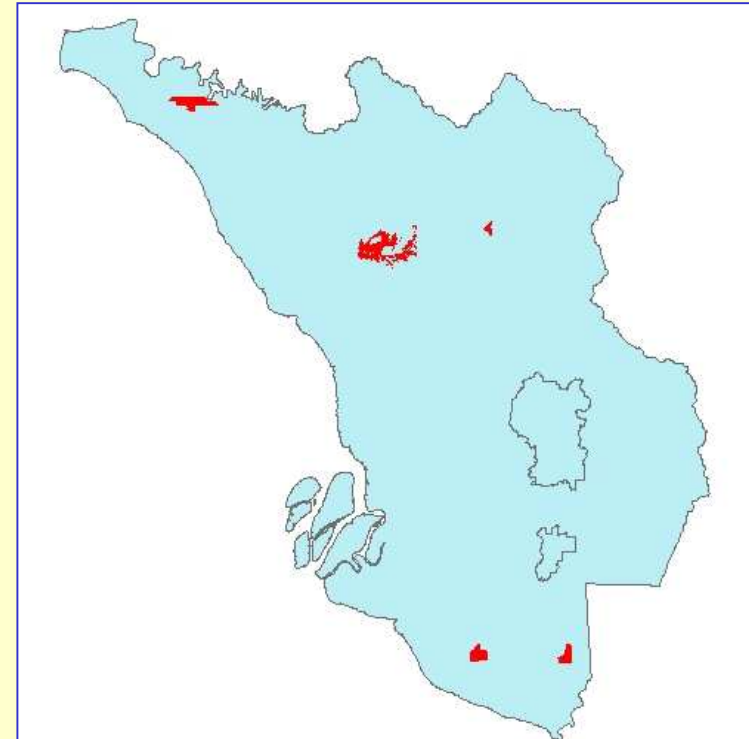
- We obtain a predictor of the scars by *thresholding* the scar function s above k ,
- The seasonal parameter k is estimated by the *hot spots number*



Results



*Scars
From model prediction*



*Scars
from satellite detection*

Period 2001-2004

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- This approach relies on the stochastic model of *Random Spread* , which generalizes *Boolean random set*.
- For forest fires, it results in correct predictions of the *scars*.
- The model is currently tested on the daily spreads.

Thank you very much

for your attention !

References

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