

A double branching model for earthquake forecasting

earthquake forecasting

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Outline of the presentation

1. The model proposed: rationale and general features
2. Applications of the model: some scientific results
3. Setting and testing the model for Italy
4. Points to take home

- ❖ It is based on a **Stepwise Branching process**. The data are analyzed at different steps, in order to get different aspects of the earthquake generation processes (see the **Boosting approach**). This works well when different physical processes are in play (e.g., co- and post-seismic triggering).
- ❖ The method deals with **regions** not **single faults**; this implies limits in the spatial resolution, but we do not mind about possible incompleteness of the faults catalog.
- ❖ We have used the model to explore **different spatial-time-magnitude window** (different seismic catalogs) in order to check how the model parameters vary.
- ❖ The model has been submitted to different CSEP testing regions (Global, 1yr; Western Pacific, 1 yr.; Italy, 3 months; **Japan, 1 yr.?**)

Double Branching Model

Double Branching Model (1°step)

SHORT TERM INTERACTIONS

$$\lambda_1(t, x, y) = \mathbf{v}_1 \cdot \mathbf{u}_1(x, y) + \sum_{t_i < t} \left[\frac{\mathbf{k}_1}{(t - t_i + \mathbf{c})^p} e^{\alpha_1 (M_i - M_0)} \frac{\mathbf{c}_{d_1 q_1}}{(r^2 + \mathbf{d}_1^2)^{q_1}} \right]$$

Epidemic Type Aftershocks Model
(Ogata, 1988, 1998)

Double Branching Model (Declustering)

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Declustering procedure

Zhuang et al., 2002

$$\pi_i = \frac{\mathbf{v}_1 \cdot \mathbf{u}_1(x_i, y_i)}{\lambda(t_i, x_i, y_i)}$$

Double Branching Model (2° step)

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$$\lambda_2(t, x, y) = \mathbf{v}_2 \cdot \mathbf{u}_2(x, y) + \sum_{t_i < t} \left[\mathbf{k}_2 e^{-\left(\frac{t-t_i}{\tau}\right)} e^{\alpha_2 (M_i - M_0)} \frac{c_{d_2 q_2}}{(r^2 + \mathbf{d}_2^2)^{q_2}} \right]$$

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Poisson Model

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Zhuang et al., 2002

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Branching Model

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τ relaxation time

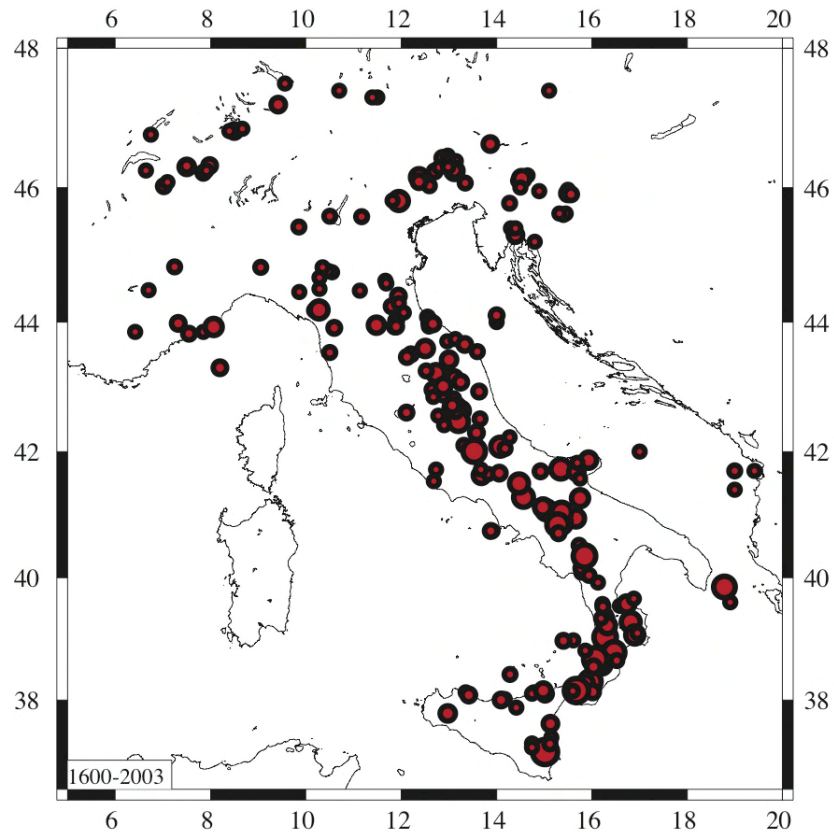
$e^{-t/\tau}$ model proposed by

[Piersanti 1991; Piersanti et al., 1995; 1997]

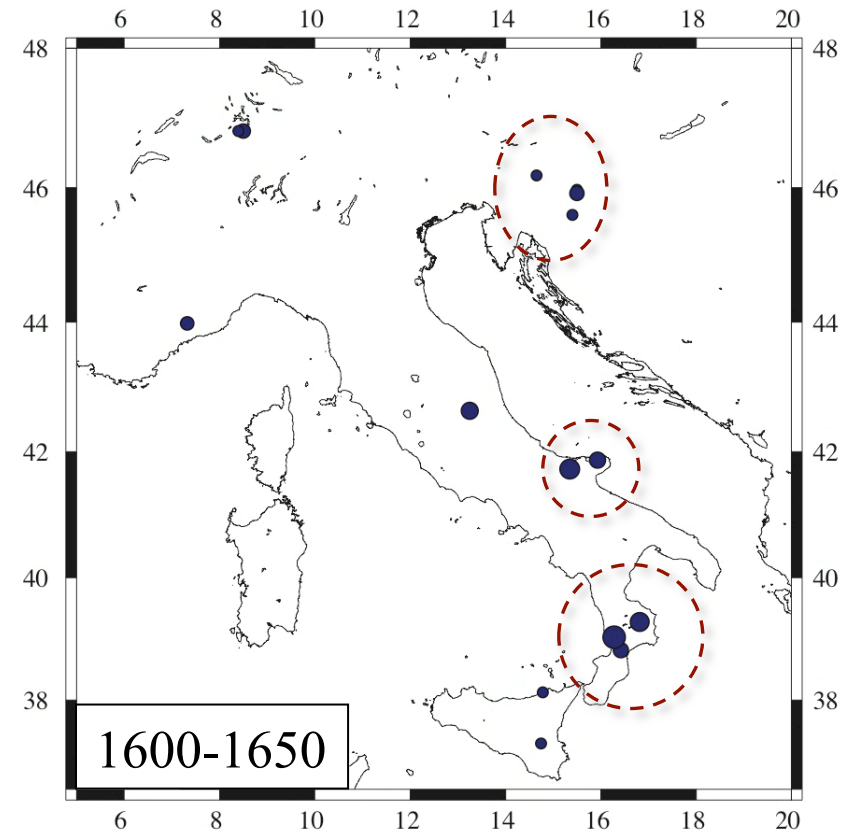
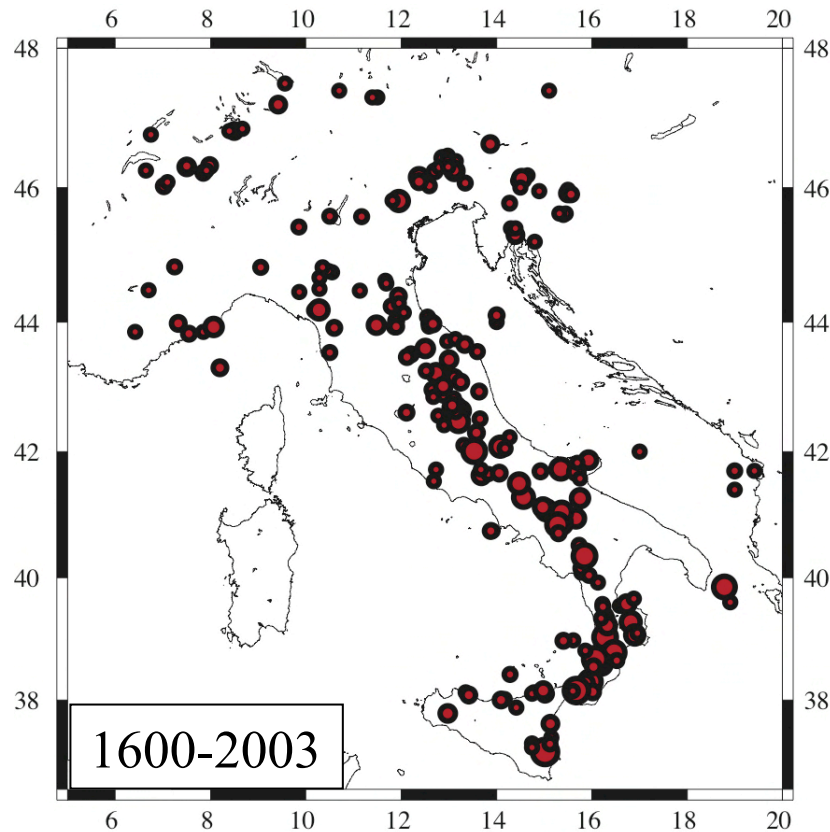
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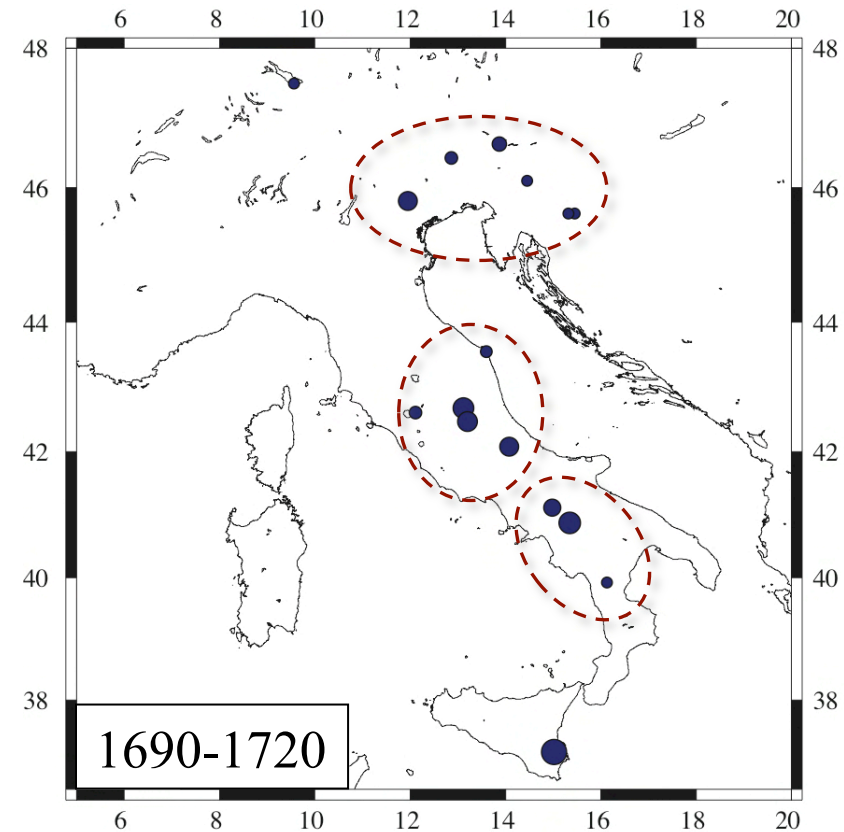
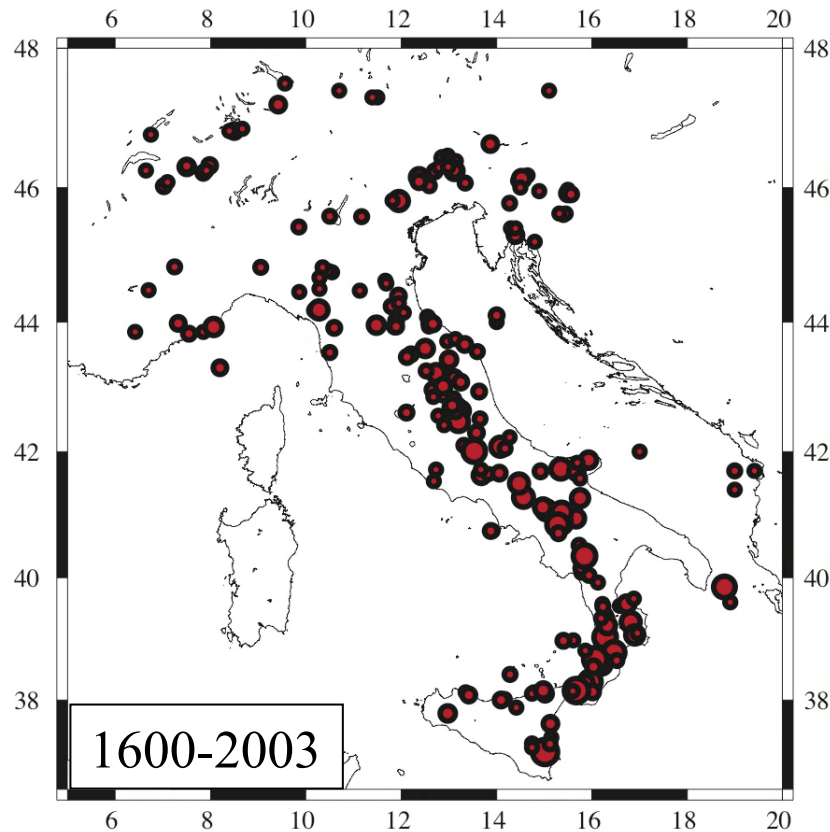
Double Branching Model: evidence of long-term clustering



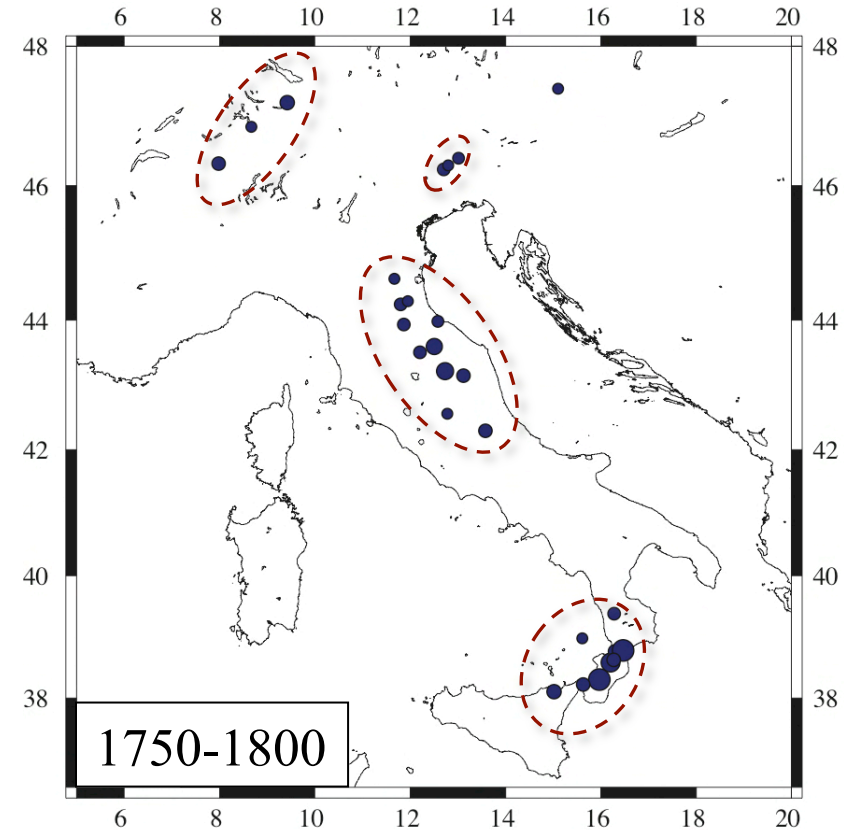
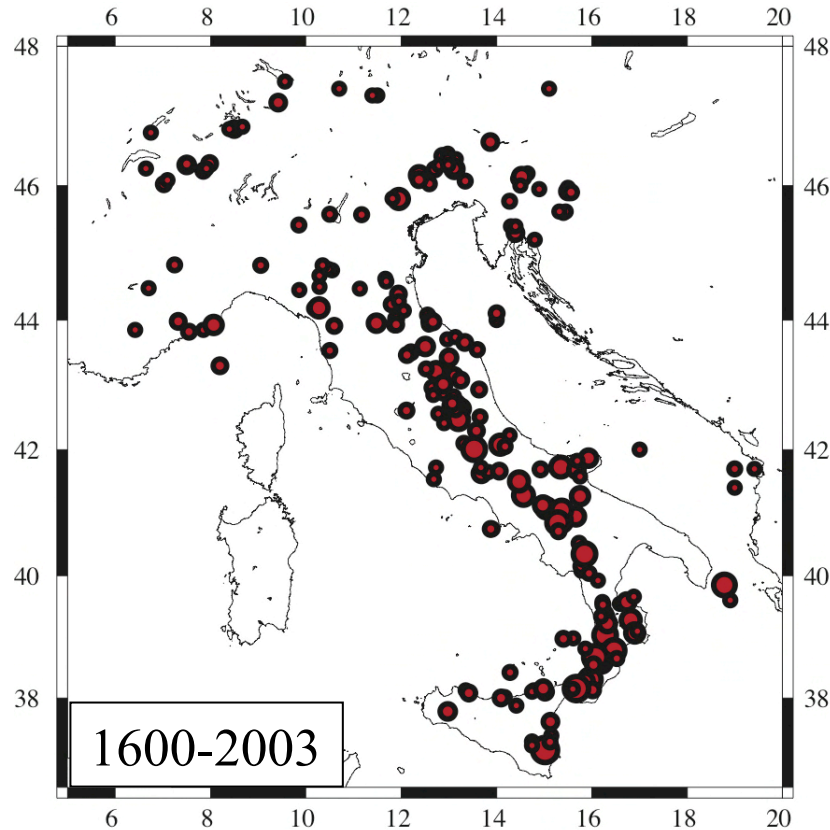
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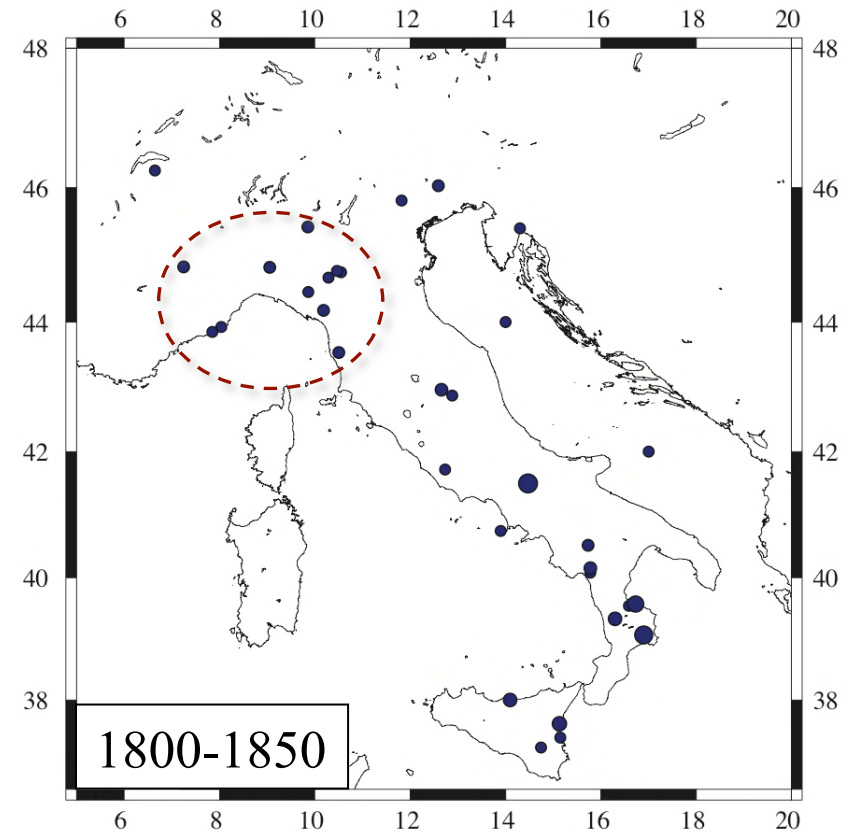
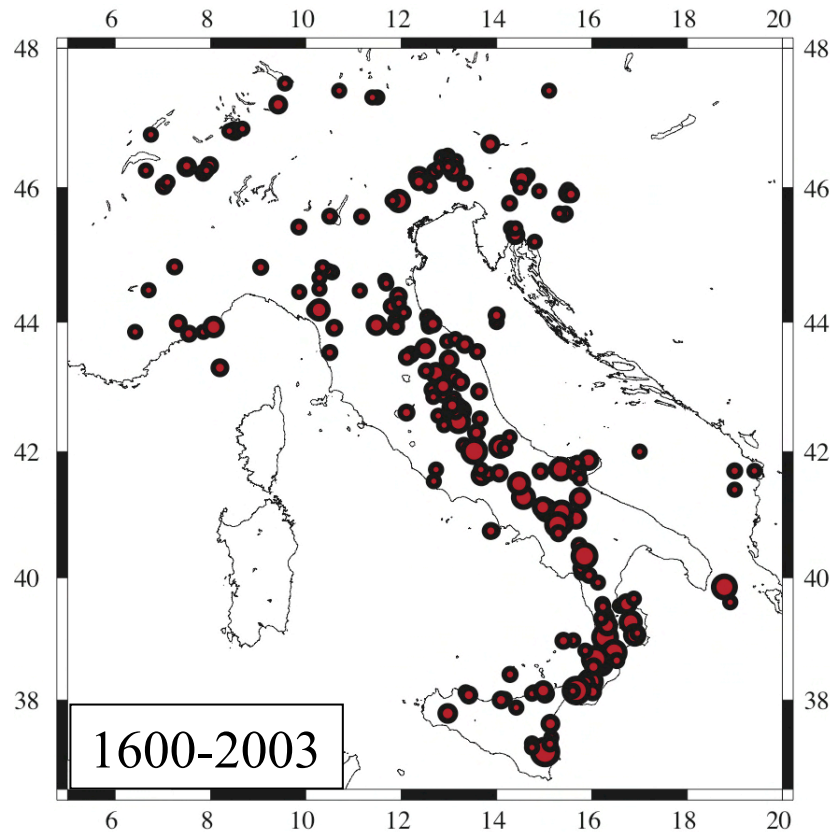
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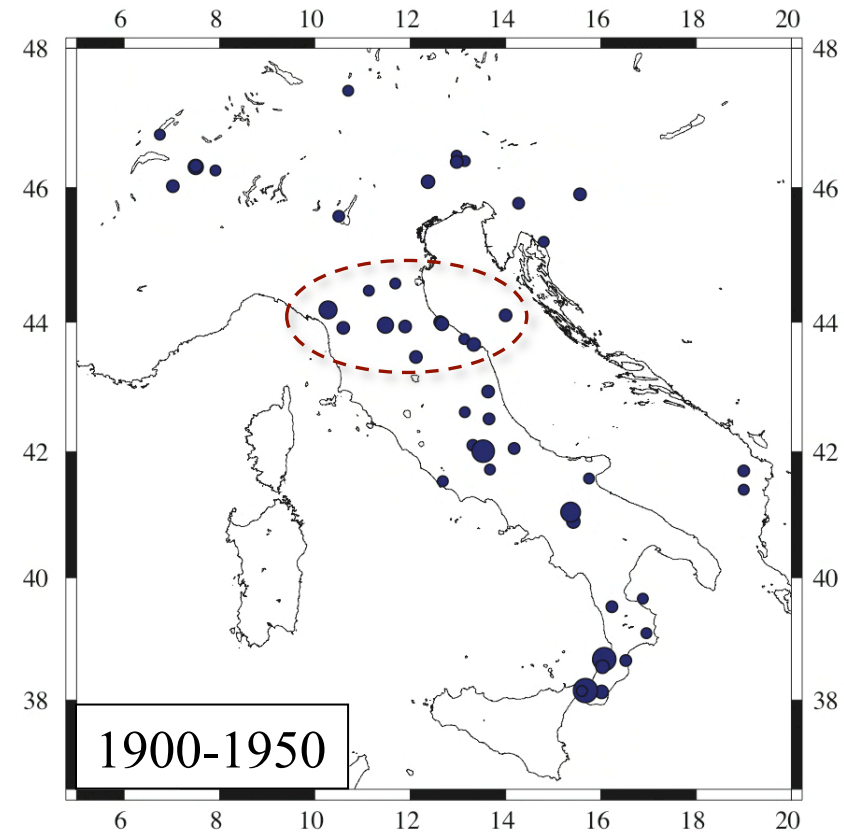
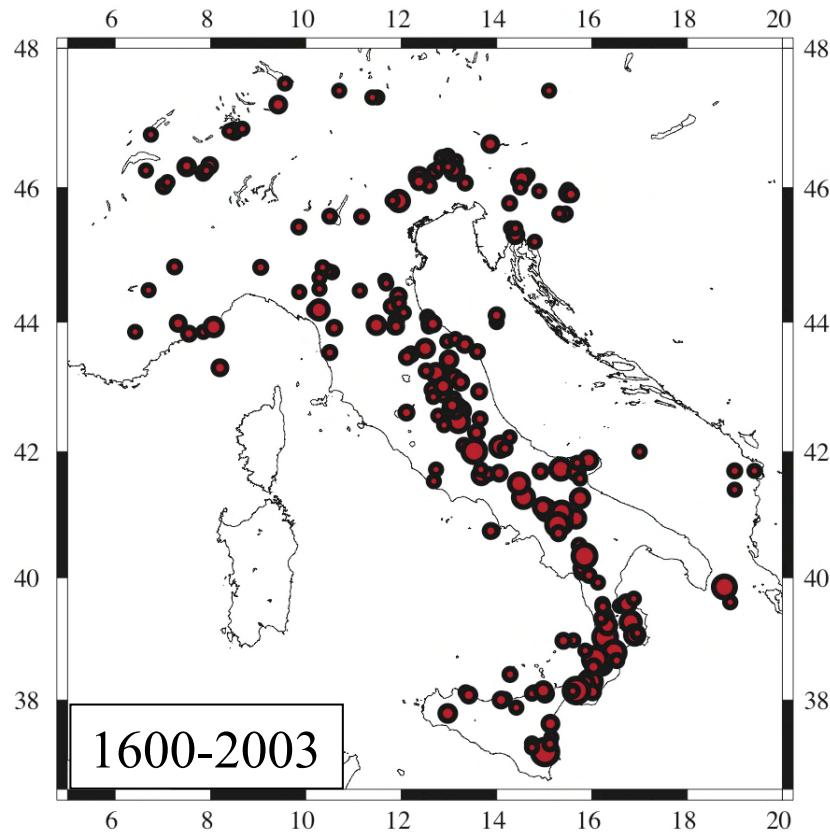
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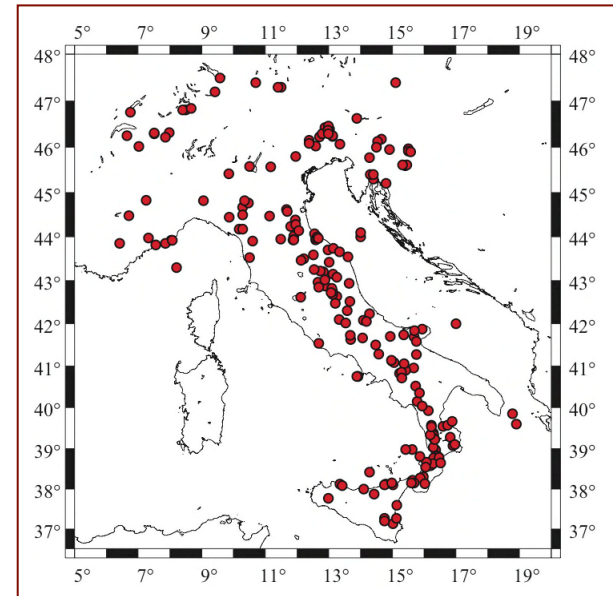
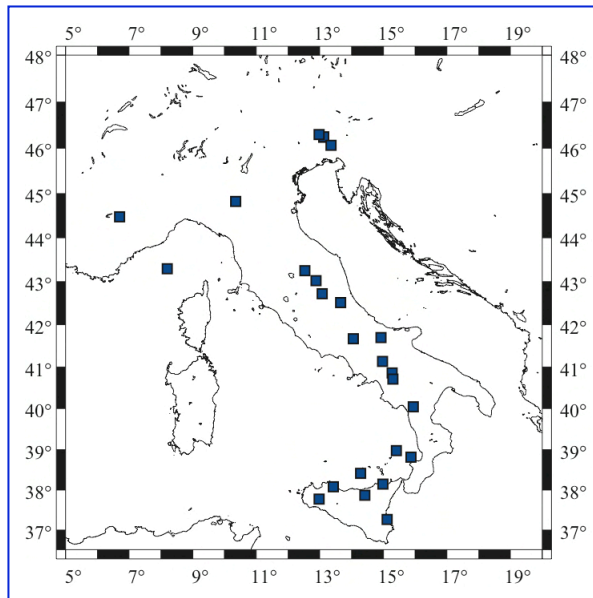


- ❖ The model has been applied to the **Pacheco and Sykes (1992)** catalog: Ms 7.0+, max depth 70 km, 698 earthquakes in the time interval 1900-1989. **NEIC** catalog; Ms 6.0+, max depth 70 km, 3590 earthquakes in the time interval 1974-2006. Italian **CPTI04** catalog: Mw 5.5+; 203 earthquakes in the time interval 1600-2003.
- ❖ The model always **perform better** than a stationary **ETAS** model and **Poisson**
- ❖ There is always a significant **time variation of the “background”** (K_2 is always significantly different from zero)
- ❖ The time variation has always a **characteristic time τ of about 30 years** (post-seismic relaxation?)
- ❖ The other parameters of the model are similar to values obtained for aftershock sequences (**universality?**)

Historical seismicity in Italy: CPTI catalog

Catalogo *P*arametrico *T*erremoti *I*taliani **catalog**
1600-2003, $M_w \geq 5.5$, 203 events

CPTI *Learning* database
1600-1950
176 events

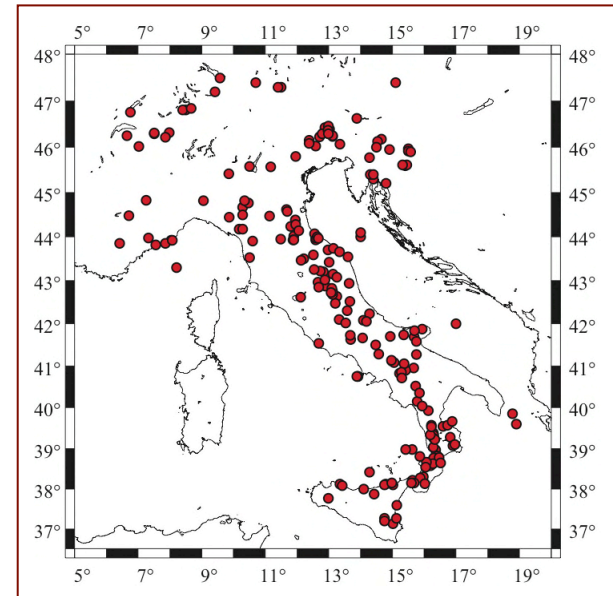
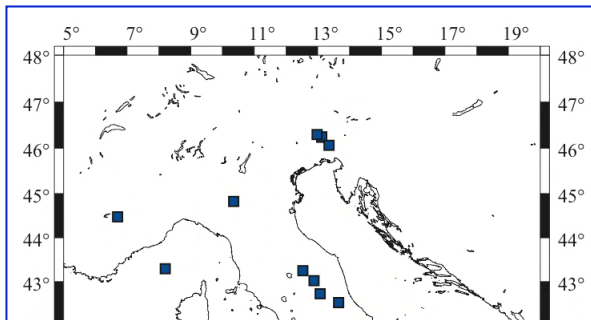


CPTI *testing* database
1950-2003
27 events

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CPTI *L*earning database
1600-1950
176 events



The CPTI catalog is just declustered

We skip the first step (ETAS modeling) of our procedure!

CPTI catalog: Modeling

long-term clustering

$$\lambda_2(t, x, y) = \nu_2 \cdot u_2(x, y) + \sum_{t_i < t} \left[k_2 e^{-\left(\frac{t-t_i}{\tau}\right)} e^{\alpha_2 (M_i - M_0)} \frac{c_{d_2 q_2}}{(r^2 + d_2^2)^{q_2}} \right]$$

tectonic loading

Parameter	Poisson Model	Second Branching Model
ν_2	$0.50 \pm 0.04 \text{ year}^{-1}$	$0.23 \pm 0.04 \text{ year}^{-1}$
K_2		0.019 ± 0.004
τ		$34 \pm 6 \text{ year}$
α_2		~ 0.0
d_2		$21 \pm 7 \text{ km}$
q_2		1.7 ± 0.2
Log-lik	-2725.6	-2660.5

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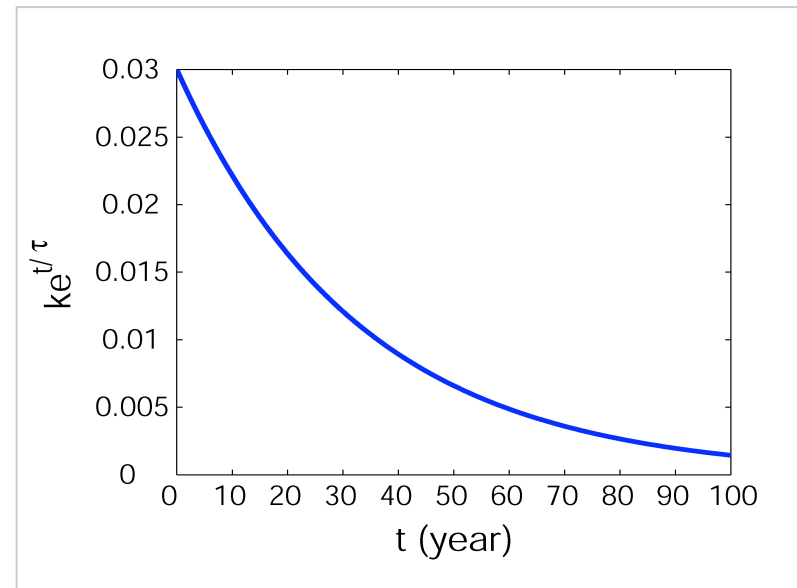
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Branching Model better than Poisson Model?

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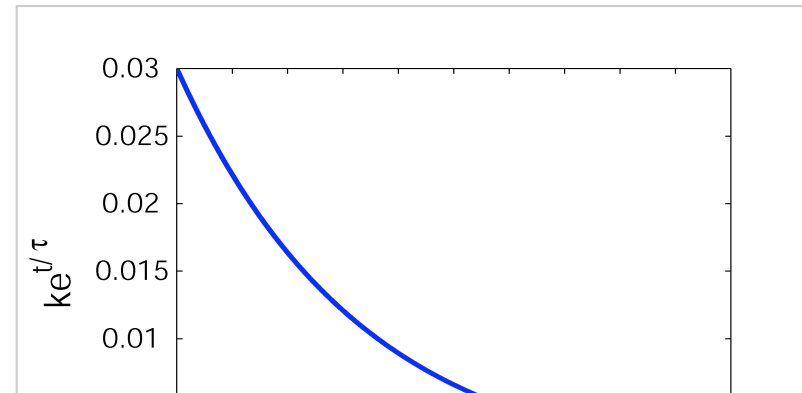
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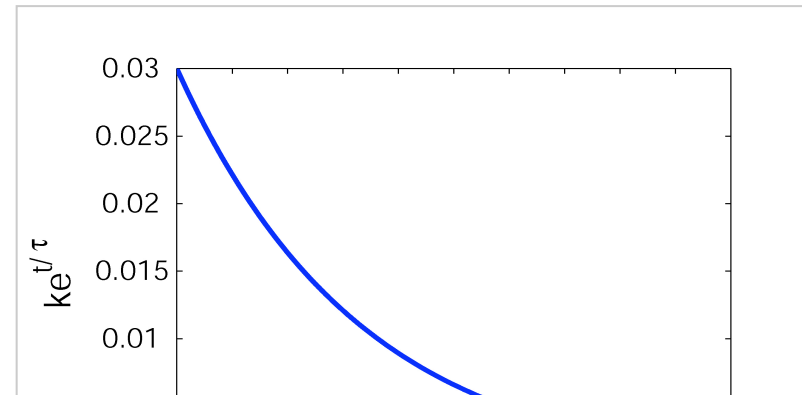


We have a not negligible probability that an event has a triggering effect after one century from its occurrence

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Long-term Clustering: universal law?

Pacheco and Sykes catalog (1900-1990, Mw ≥ 7.0) $\tau = 36 \pm 7 \text{ year}$

NEIC catalog (1974-2006, Mw ≥ 6.0) $\tau = 30 \pm 5 \text{ year}$

(Marzocchi and Lombardi, 2008)

CPTI catalog: Testing

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\mathcal{H}_0 : Poisson Process

\mathcal{H}_1 : Branching model

CPTI catalog: Testing

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Information gain

$$IG_{pe} = \frac{\text{Log}L_1 - \text{Log}L_0}{N}$$

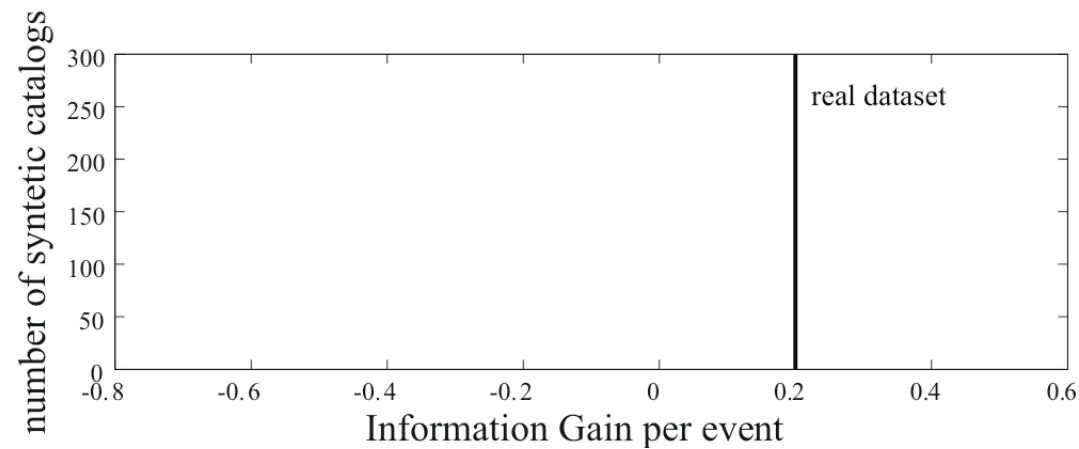
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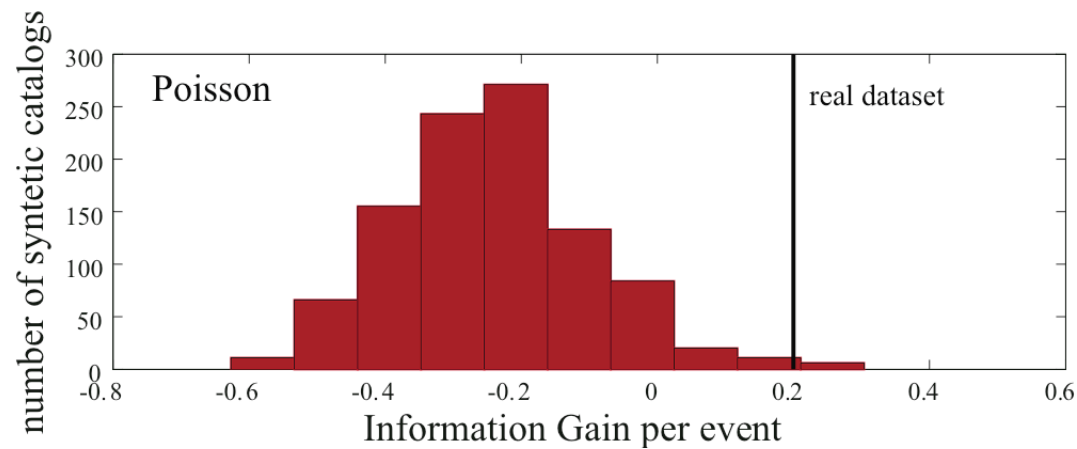
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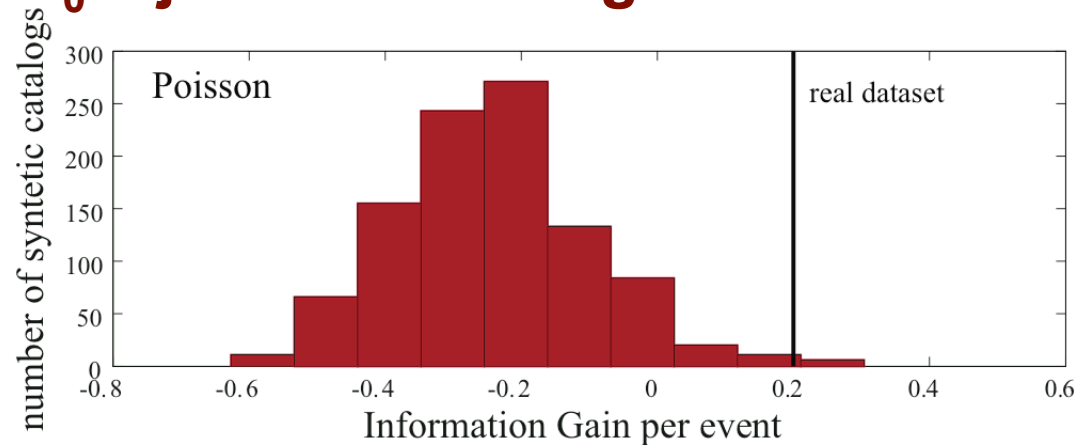
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\mathcal{H}_0 rejected at 1% significance level!



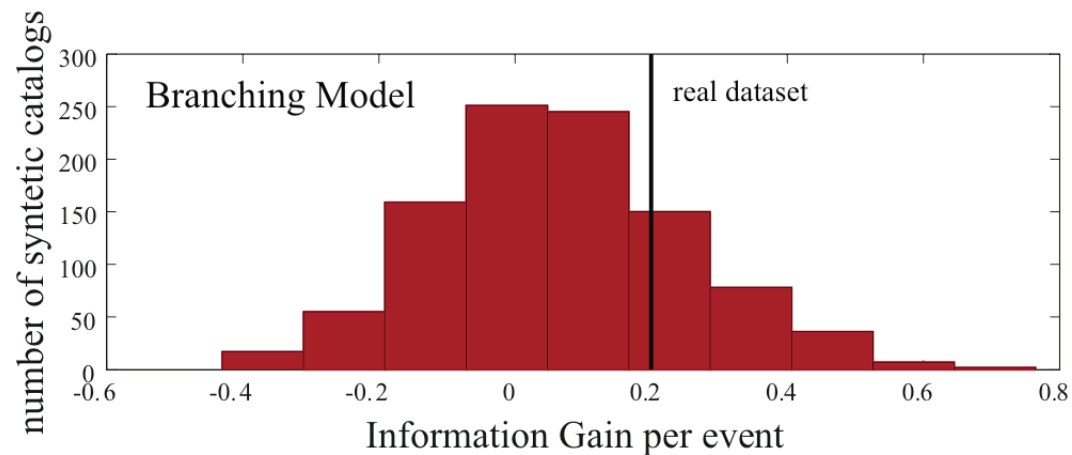
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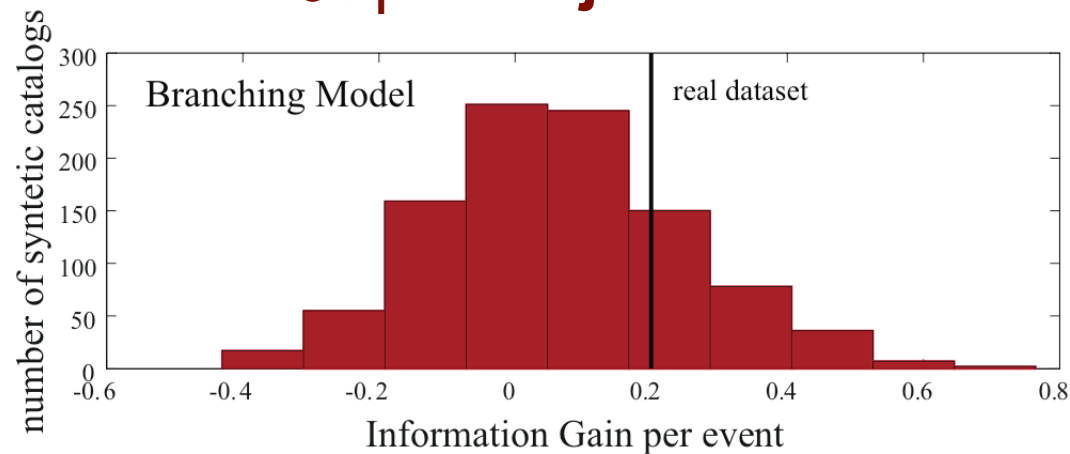
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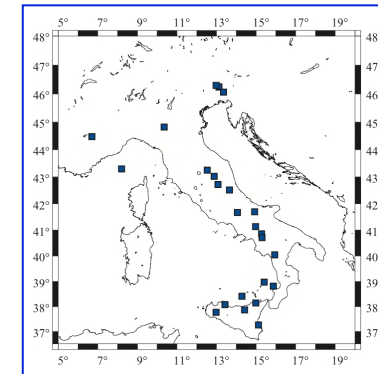
\mathcal{H}_1 not rejected!



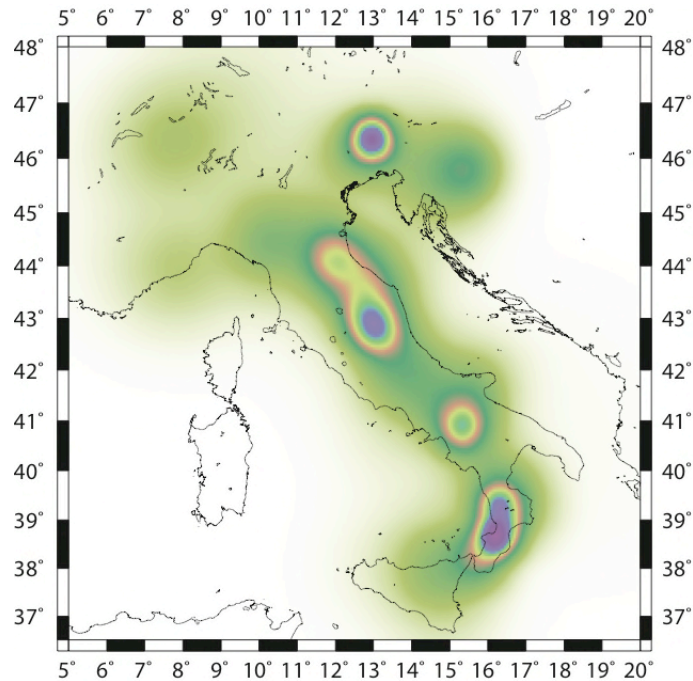
CPTI catalog: Forecasting

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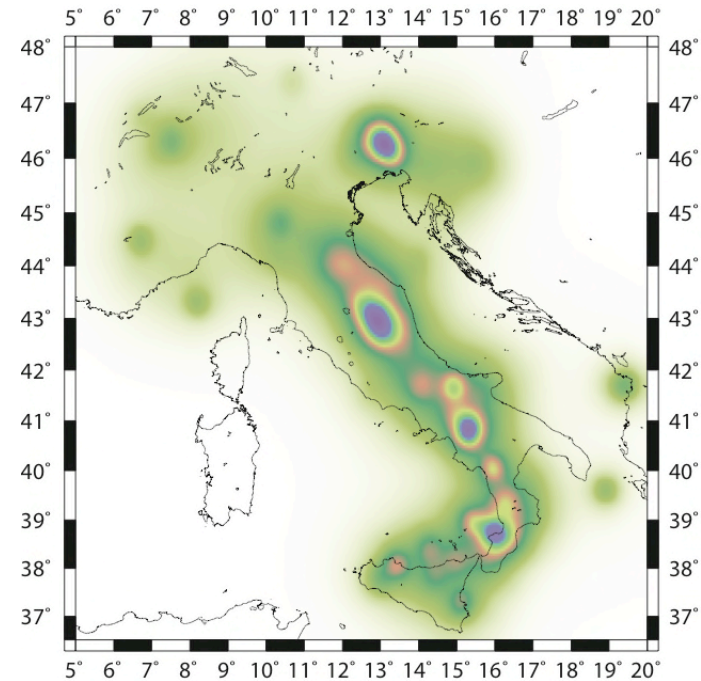
50 years: 2008-2057



Poisson Model



Branching Model



Points to bring home

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- **universality** (independence by magnitude-space-time range)
- one of the most reliable physical explanations could be the **postseismic stress transfer**
- implication for **seismic hazard assessment**

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- the historical seismicity occurred in Italy shows a **long-term clustered** behaviour
- **universality** (independence by magnitude-space-time range)
- one of the most reliable physical explanations could be the **postseismic stress transfer**
- implication for **seismic hazard assessment**
- the model is set and running for **global test, Western Pacific, and Italy** (soon). Aim: to submit it for **Japan** (1 yr forecast).