

Significant improvements of the space-time ETAS model for forecasting of accurate baseline seismicity

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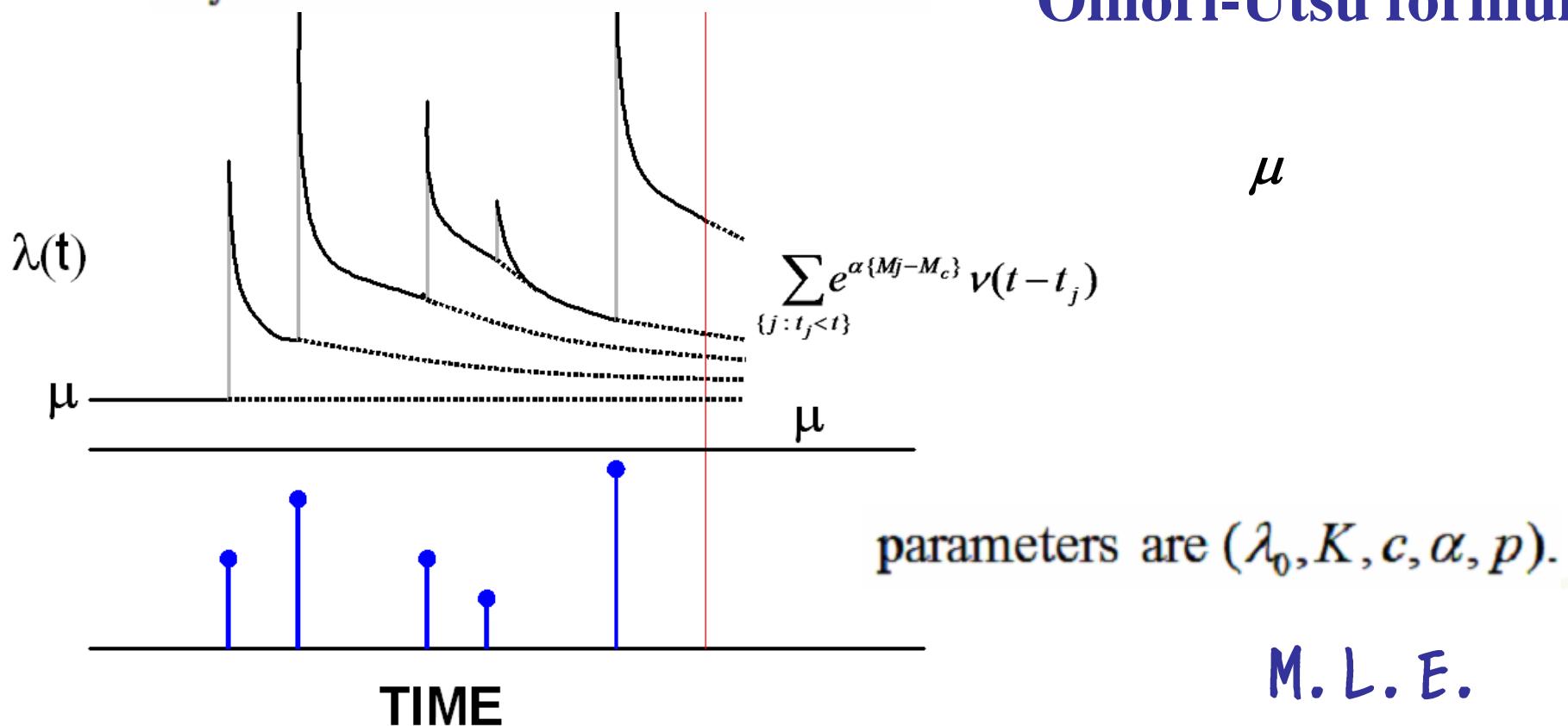
ETAS model: Ogata (1988, JASA)

$$\lambda(t) = \mu + \sum e^{\alpha\{M_j - M_c\}} v(t - t_j)$$

where $\{j : t_j < t\}$
 t_j is occurrence time of j th event;
 M_j is magnitude of j th event;

$$v(t) = \frac{K}{(t + c)^p}$$

Omori-Utsu formula



M. L. E.

Space-Time ETAS model

$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

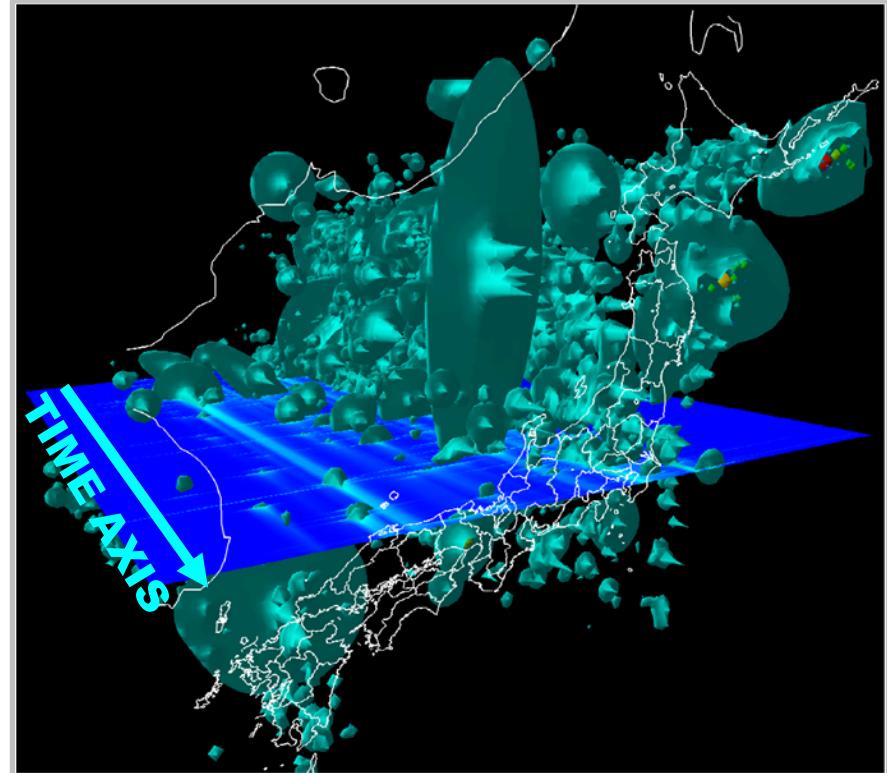
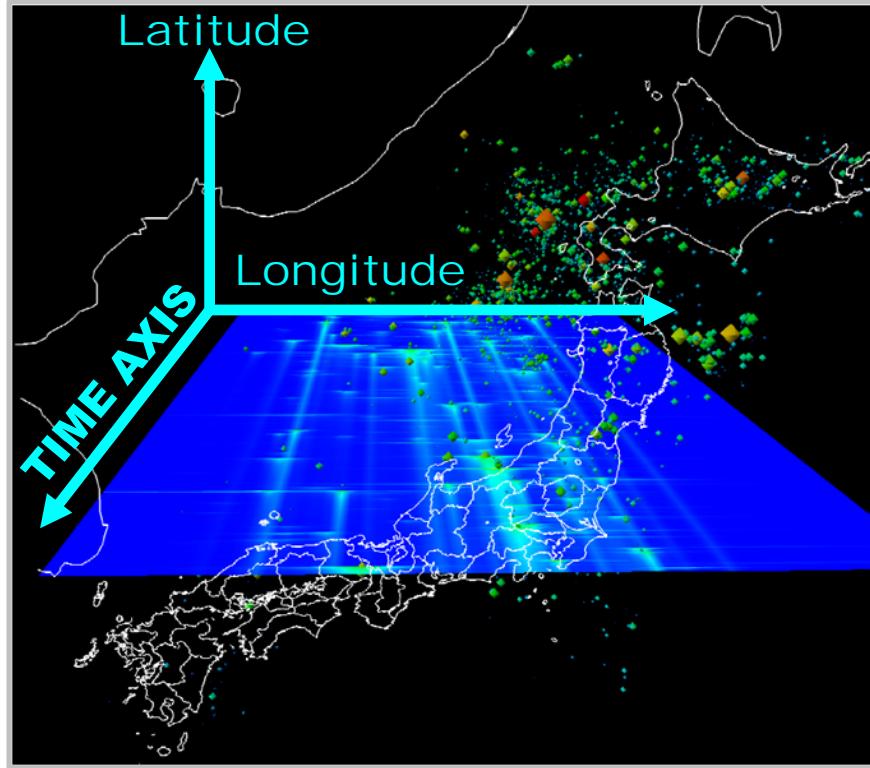
where

$$Q_j(x, y) = (x - x_j, y - y_j) S_j \begin{pmatrix} x - x_j \\ y - y_j \end{pmatrix}$$

Earthquakes (particles)

In and around Japan; depth<100km;
1926 - 1995; magnitude ≥ 5.0

Space-time ETAS model
Isosurface of an occurrence rate
 $\lambda(t, x, y)$



Space-Time ETAS model

$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

Space-Time ETAS model

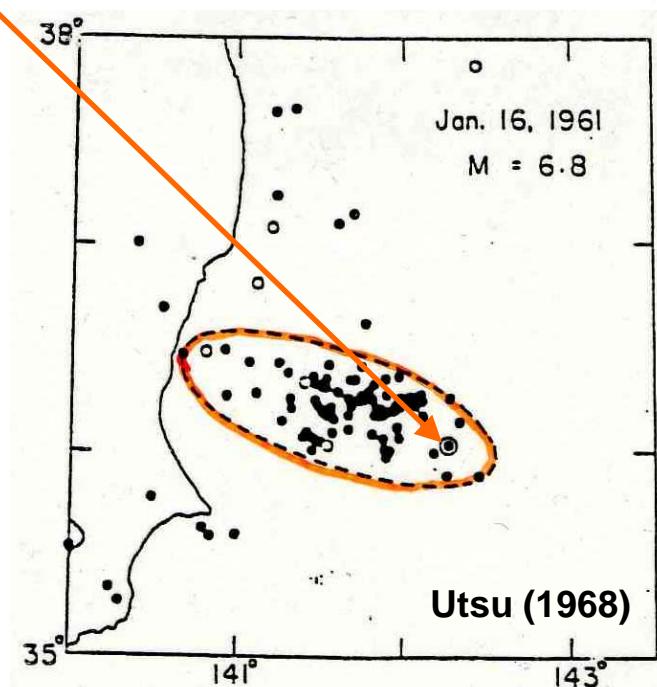
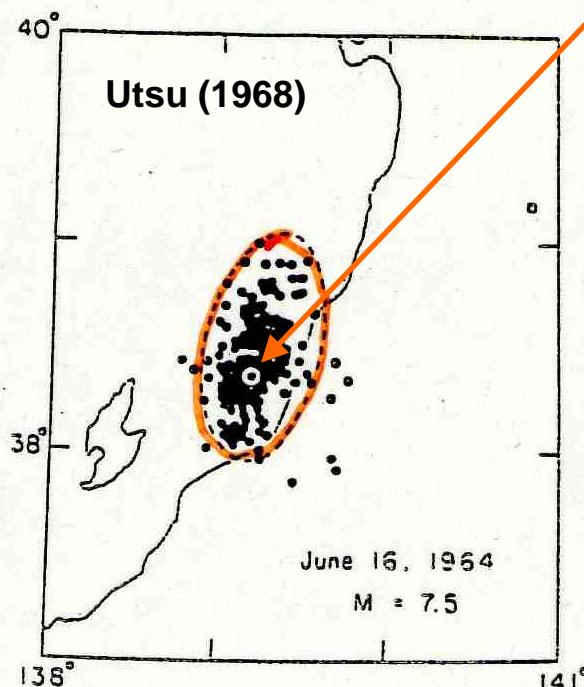
$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

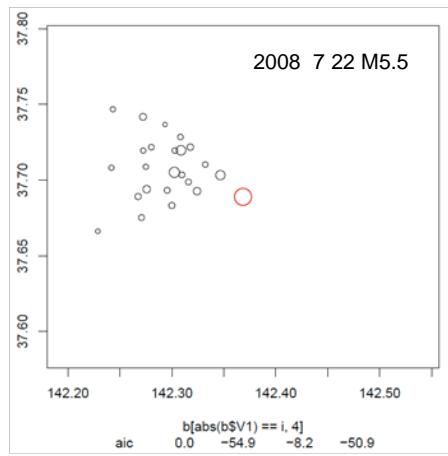
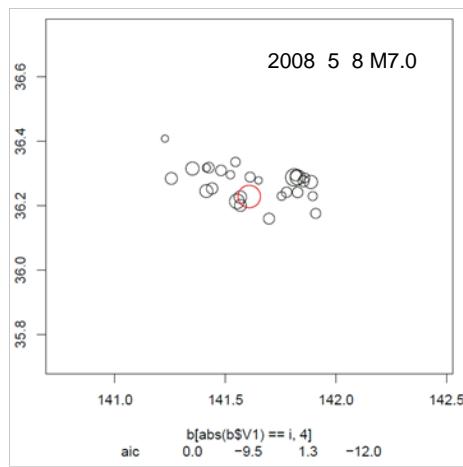
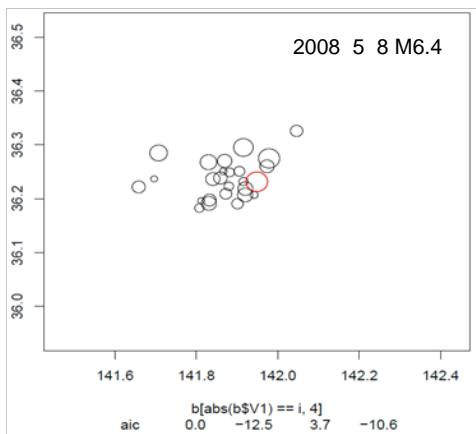
where

$$Q_j(x, y) = (x - \bar{x}_j, y - \bar{y}_j) S_j \begin{pmatrix} x - \bar{x}_j \\ y - \bar{y}_j \end{pmatrix}$$

for **ANISOTROPY**:

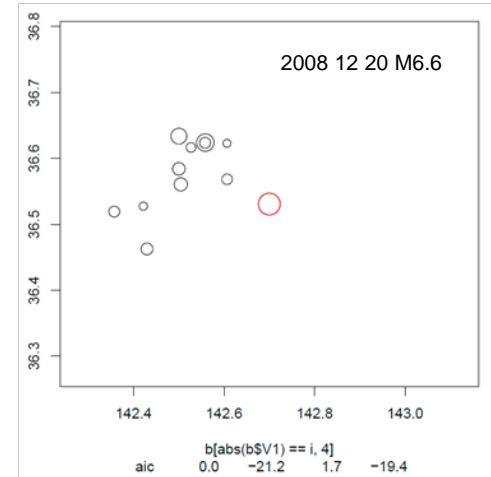
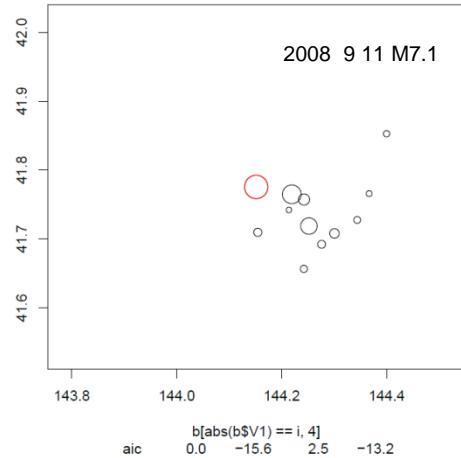
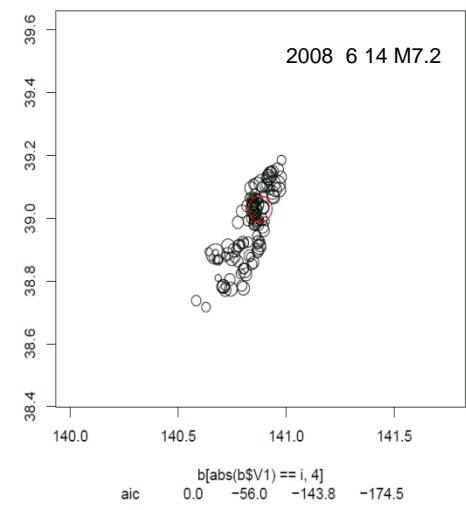
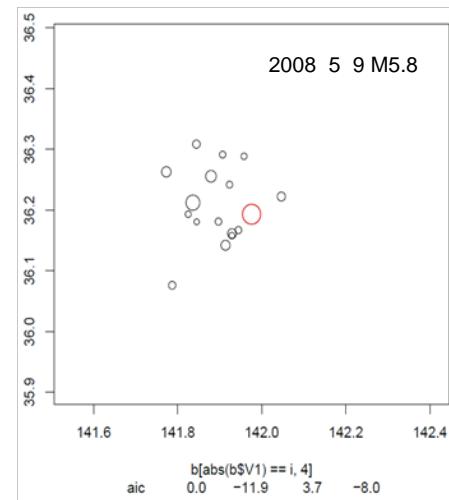
Epicenter





Spatial distributions of the first 1 hour aftershocks

(Red circle = mainshock)



$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

Immediately after a large event, we adopt:

Isotropic kernel within 1 hour

$$Q_j(x, y) = (x - x_j)^2 + (y - y_j)^2 = (x - x_j, y - y_j) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x - x_j \\ y - y_j \end{pmatrix}$$

Anisotropic kernel after 1 hour

$$\begin{aligned} Q_j(x, y) &= \frac{1}{\sqrt{1-\rho^2}} \left(\frac{\sigma_2^2}{\sigma_1^2} (x - \bar{x}_j)^2 - 2\rho(x - \bar{x}_j)(y - \bar{y}_j) + \frac{\sigma_1^2}{\sigma_2^2} (y - \bar{y}_j)^2 \right) \\ &= \frac{1}{\sigma_1^2 \sigma_2^2 \sqrt{1-\rho^2}} (x - \bar{x}_j, y - \bar{y}_j) \begin{pmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{pmatrix}^{-1} \begin{pmatrix} x - \bar{x}_j \\ y - \bar{y}_j \end{pmatrix} \end{aligned}$$

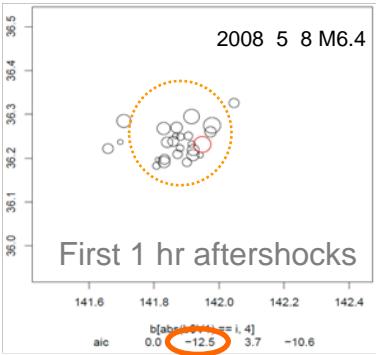
$$Q_j(x, y) = \frac{1}{\sqrt{1-\rho^2}} \left(\frac{\sigma_2^2}{\sigma_1^2} (x - x_j^0)^2 - 2\rho(x - x_j^0)(y - y_j^0) + \frac{\sigma_1^2}{\sigma_2^2} (y - y_j^0)^2 \right) = \frac{1}{\sigma_1^2 \sigma_2^2 \sqrt{1-\rho^2}} (x - \bar{x}_j, y - \bar{y}_j) \begin{pmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{pmatrix}^{-1} \begin{pmatrix} x - \bar{x}_j \\ y - \bar{y}_j \end{pmatrix}$$

Model selection using the AIC

Models

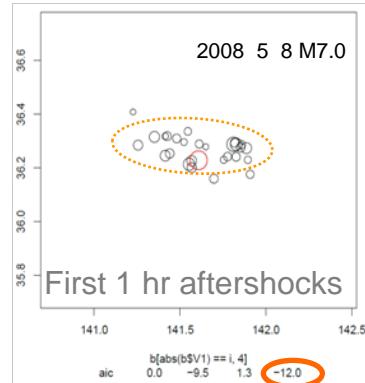
(1)

$$N\left(\begin{pmatrix} x_1 \\ y_1 \end{pmatrix}, \begin{pmatrix} \hat{\sigma}^2 & 0 \\ 0 & \hat{\sigma}^2 \end{pmatrix}\right)$$



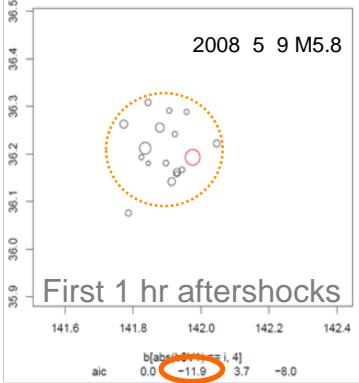
(2)

$$N\left(\begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix}, \begin{pmatrix} \hat{\sigma}^2 & 0 \\ 0 & \hat{\sigma}^2 \end{pmatrix}\right)$$



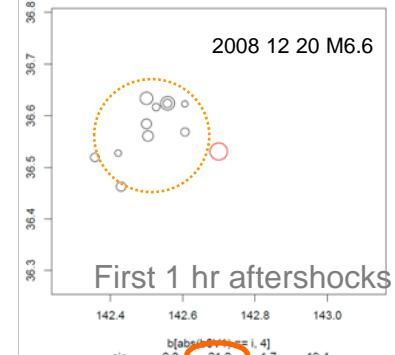
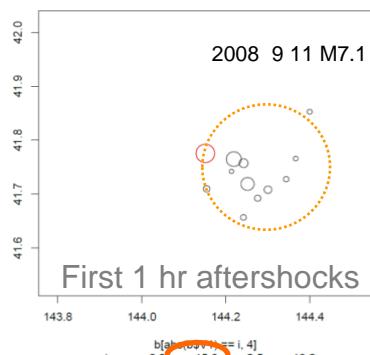
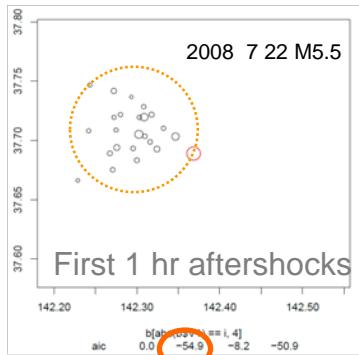
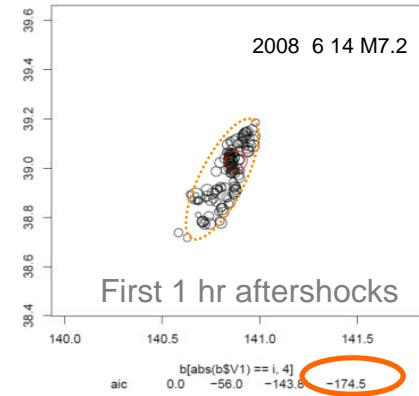
(3)

$$N\left(\begin{pmatrix} x_1 \\ y_1 \end{pmatrix}, \begin{pmatrix} \hat{\sigma}_1^2 & \hat{\rho} \hat{\sigma}_1 \hat{\sigma}_2 \\ \hat{\rho} \hat{\sigma}_1 \hat{\sigma}_2 & \hat{\sigma}_2^2 \end{pmatrix}\right)$$



(4)

$$N\left(\begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix}, \begin{pmatrix} \hat{\sigma}_1^2 & \hat{\rho} \hat{\sigma}_1 \hat{\sigma}_2 \\ \hat{\rho} \hat{\sigma}_1 \hat{\sigma}_2 & \hat{\sigma}_2^2 \end{pmatrix}\right)$$



Location Dependent Space-Time ETAS model



Space-Time ETAS model

$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

where

$$Q_j(x, y) = (x - x_j, y - y_j) S_j \begin{pmatrix} x - x_j \\ y - y_j \end{pmatrix}$$

Heterogeneity in Space

$$\mu \Rightarrow \mu(x, y);$$

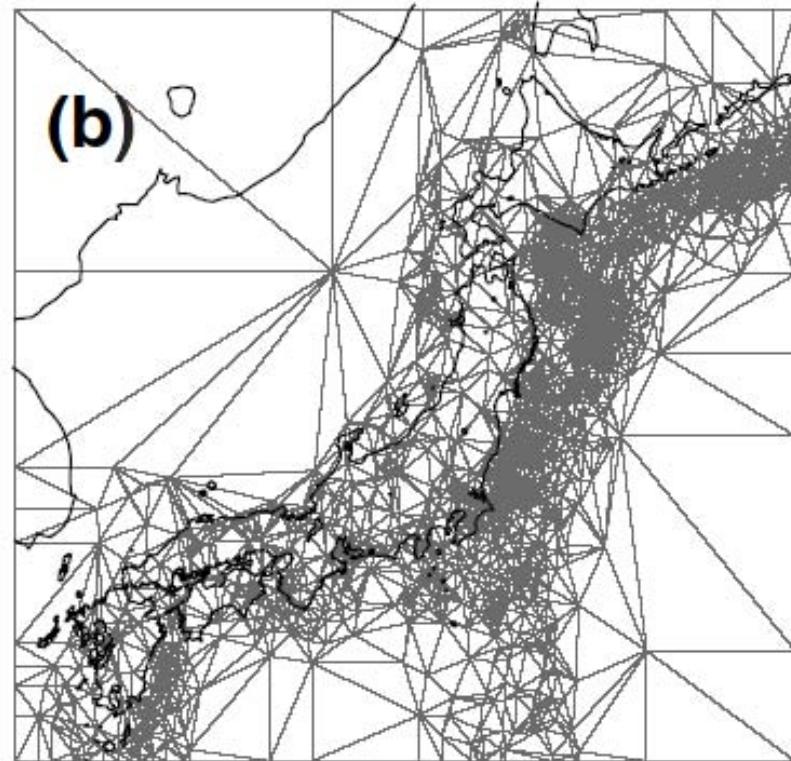
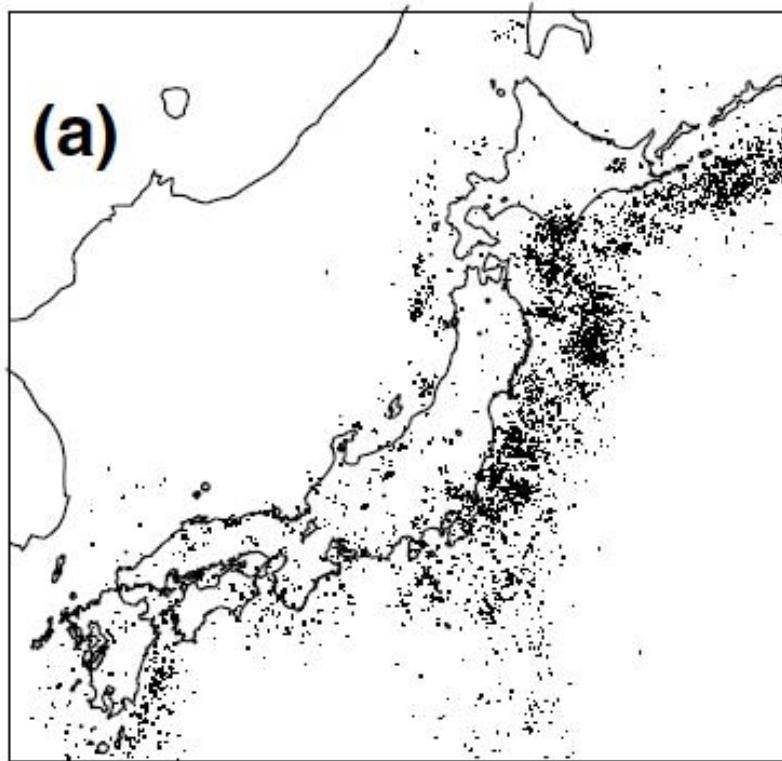
$$K \Rightarrow K(x_j, y_j); \quad \alpha \Rightarrow \alpha(x_j, y_j);$$

$$p \Rightarrow p(x_j, y_j); \quad q \Rightarrow q(x_j, y_j)$$

Hierarchical Space-Time (HIST) ETAS model

estimated an Objective Bayesian method with smoothness constraints

1926 - 1995, $M \geq 5.0$, depth < 100km



Space-Time ETAS model

$$\lambda(t, x, y) = \underline{\mu} + \sum_{\{j: t_j < t\}} \frac{\underline{K}}{(t - t_j + c)\underline{p}} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-\underline{q}}$$

Location Dependent Space-Time

ETAS model and occurrence data

$\{(t_i, x_i, y_i, M_i); i = 1, \dots, n\}$ in $[0, T] \times A$

are given. Then **Log Likelihood** is

$$\begin{aligned}\log L(\theta) &= \log L(\mu_{\theta_1}, K_{\theta_2}, \alpha_{\theta_4}, p_{\theta_5}, q_{\theta_7}) \\ &= \sum_{i=1}^n \log \lambda_\theta(t_i, x_i, y_i) - \int_0^T \iint_A \lambda_\theta(t, x, y) dt dx dy.\end{aligned}$$

where $\theta = (\theta_1, \theta_2, \theta_4, \theta_5, \theta_7)$

Penalized Log Likelihood

$$\begin{aligned}Q(\theta | w_\mu, w_K, w_\alpha, w_p, w_q) \\ = \log L(\theta) - \text{penalty}(\theta | w_\mu, w_K, w_\alpha, w_p, w_q)\end{aligned}$$

where the *penalty* is

$$\begin{aligned}\int \int_A dx dy \left\{ w_1(\mu_x^2 + \mu_y^2) + w_2(K_x^2 + K_y^2) + \right. \\ \left. w_3(\alpha_x^2 + \alpha_y^2) + w_4(p_x^2 + p_y^2) + w_5(q_x^2 + q_y^2) \right\}\end{aligned}$$

$$\rho = (w_\mu, w_K, w_\alpha, w_p, w_q)$$

$$\text{posterior}(\theta | \rho) = \frac{L(\theta) \cdot \text{prior}(\theta | \rho)}{\Lambda(\rho)}$$

Normalizing factor of Posterior

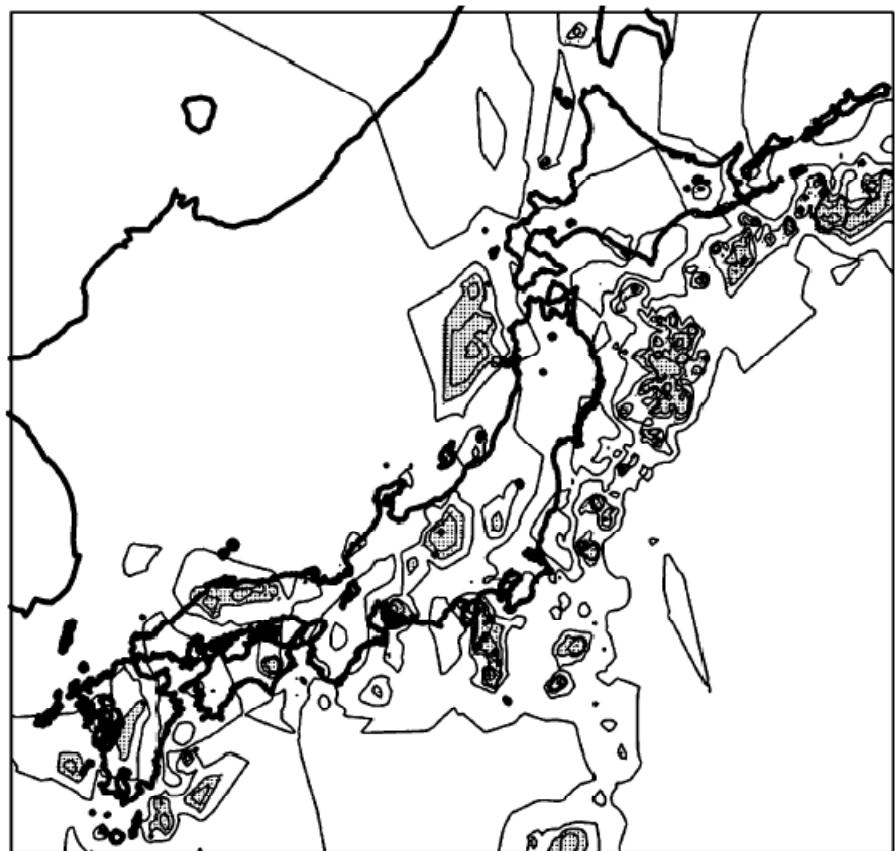
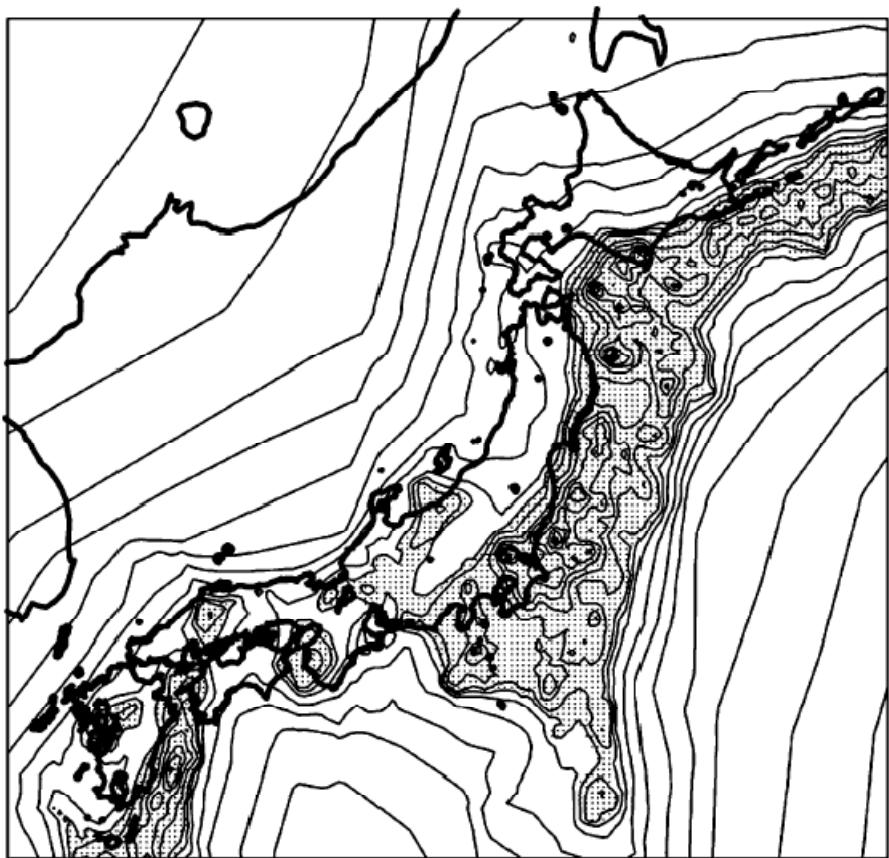
$$\Lambda(\rho) = \int \dots \int L(\theta) \cdot \text{prior}(\theta | \rho) d\theta$$

(Likelihood of a Bayesian model)

Choose ρ that maximize the $\Lambda(\rho)$,
and then minimize

$$AIC = (-2) \max_{\rho} \{\log \Lambda(\rho)\} + 2 \times \dim(\rho)$$

Akaike Bayesian information criterion (Akaike, 1980)



Space-Time ETAS model

$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

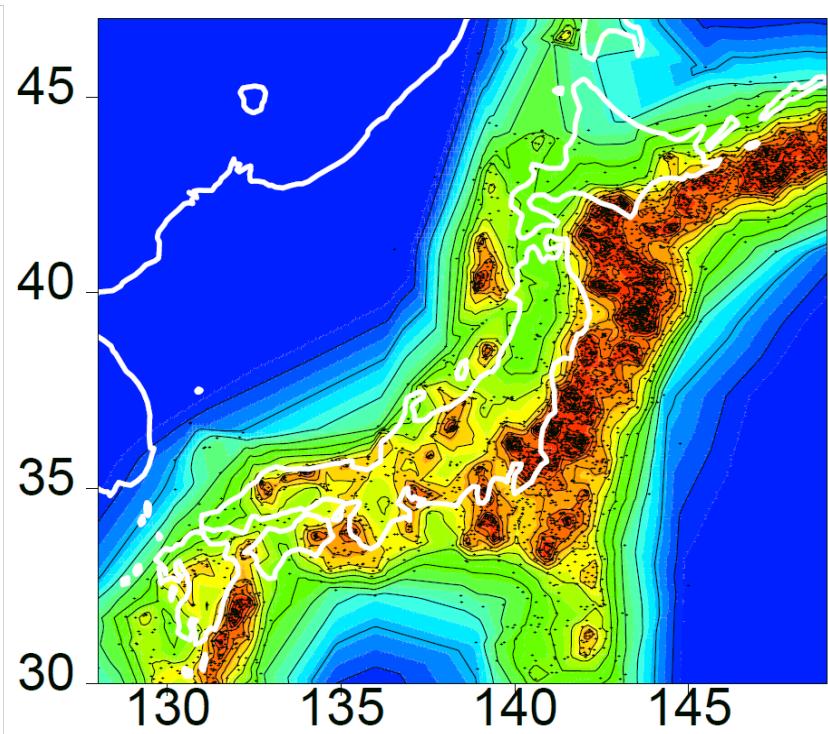
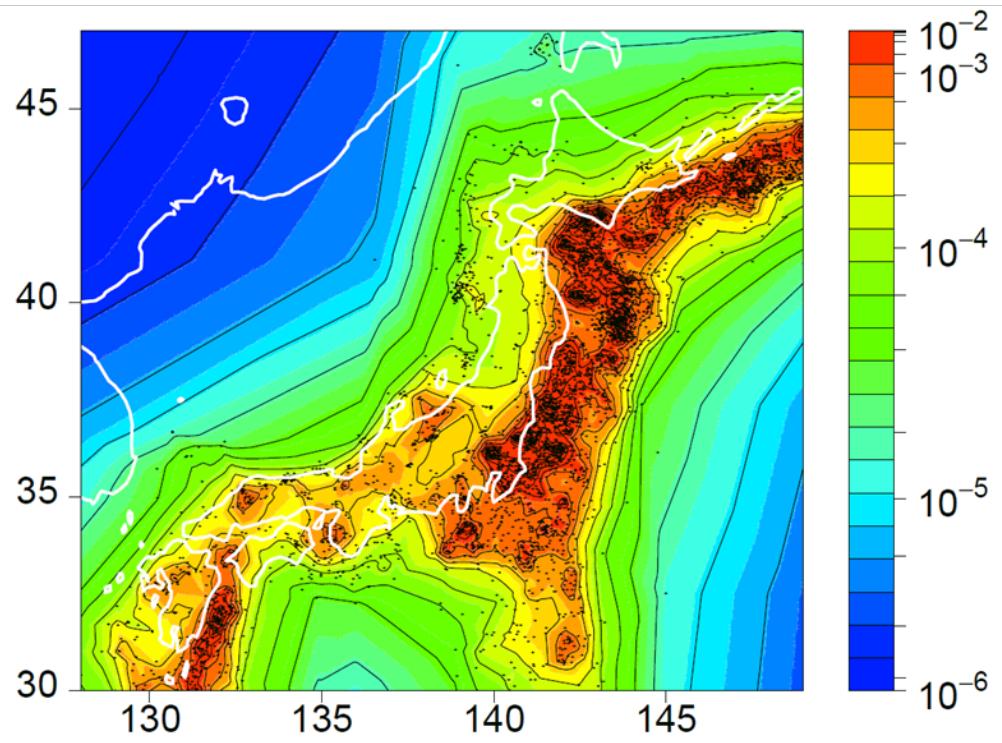
Long term forecast of large earthquakes, in and around Japan



1926-1995 $M \geq 5.0$

$\mu(x,y)$

$$\lambda(x,y) = \frac{1}{70} \int_{1926}^{1995} \lambda(t, x, y) dt$$



Estimated from
 $M \geq 5.0$
for 1926-1995

$$\mu(x,y)$$

45

40

35

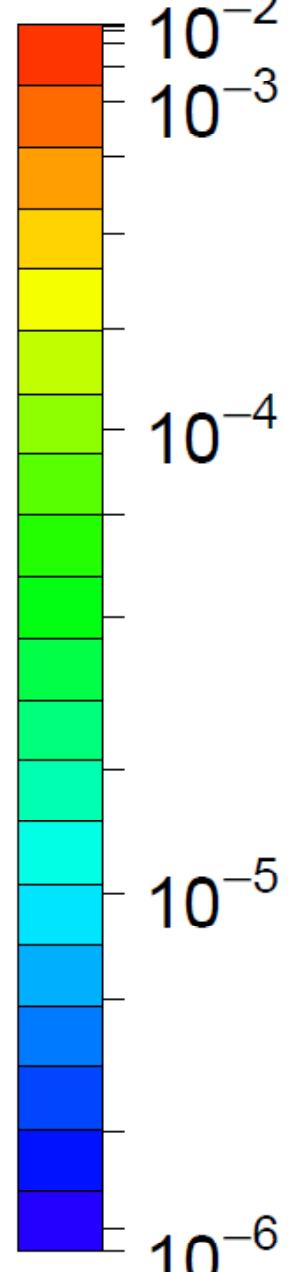
30

130

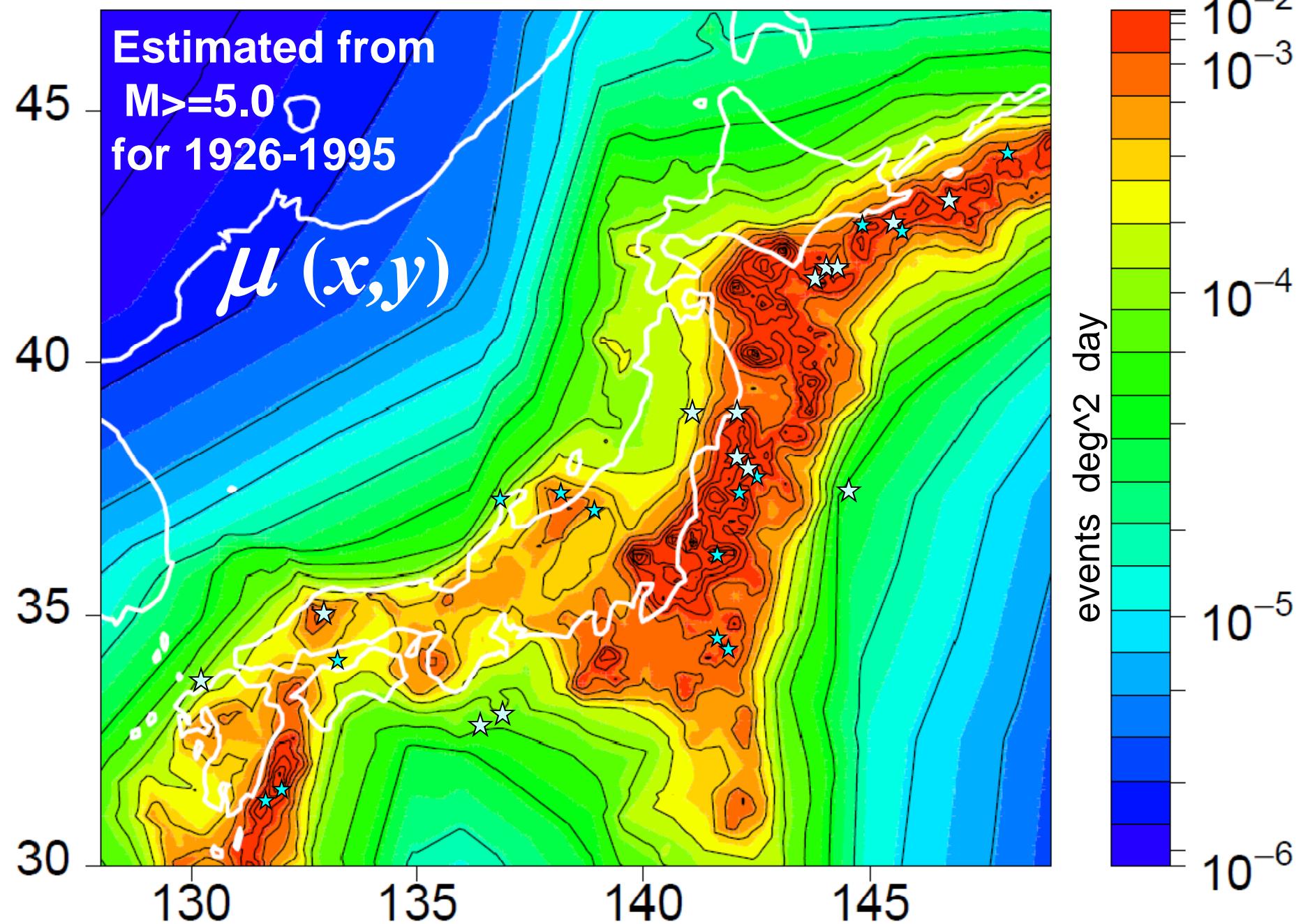
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140

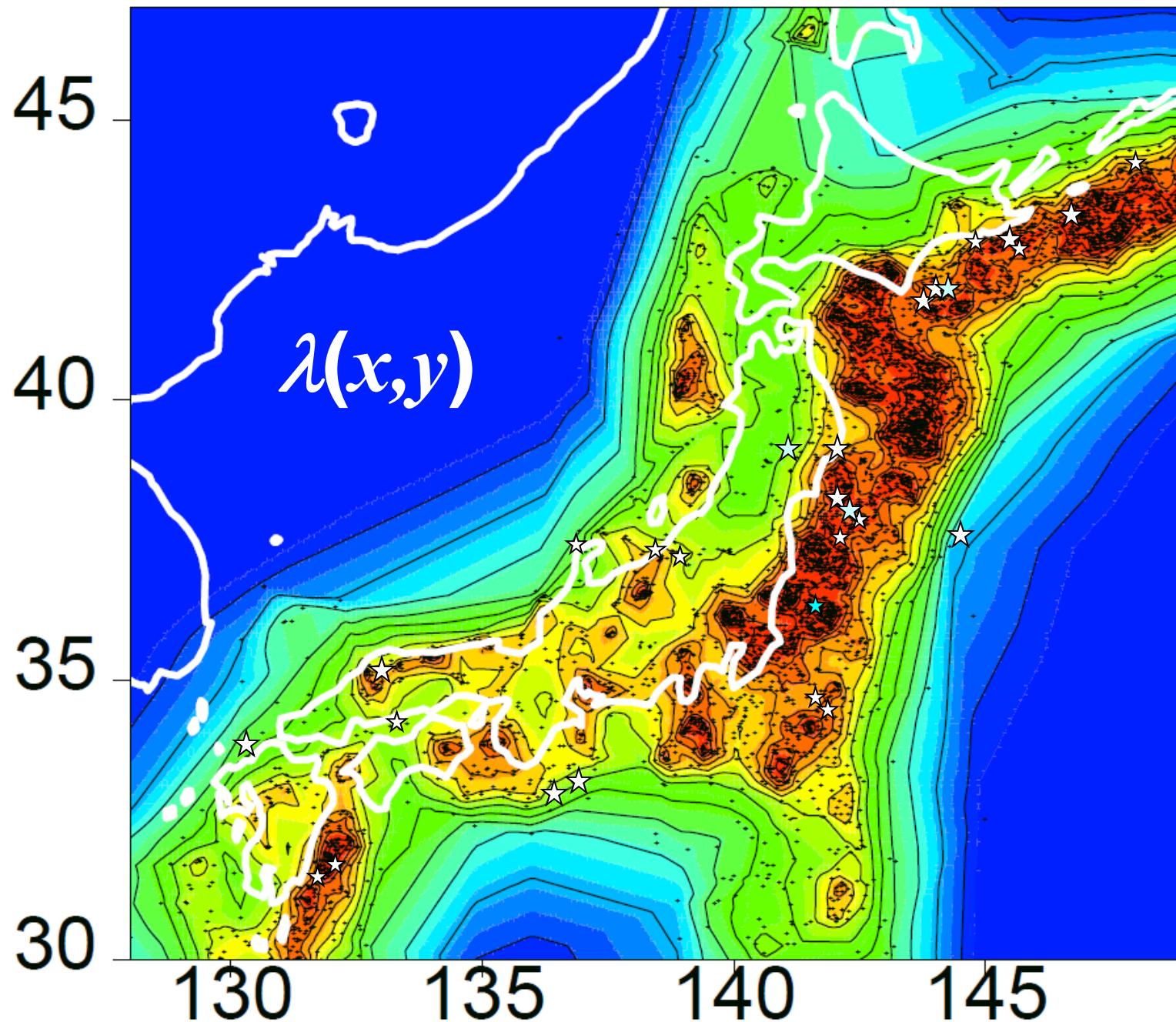
145



 = earthquakes of $M \geq 6.7$ occurred during 1996 - 2009



Earthquakes of $M \geq 6.7$ occurred during 1996 - 2009



Earthquakes of $M \geq 6.7$ occurred during 1996 - 2009

45

$K(x, y)$

40

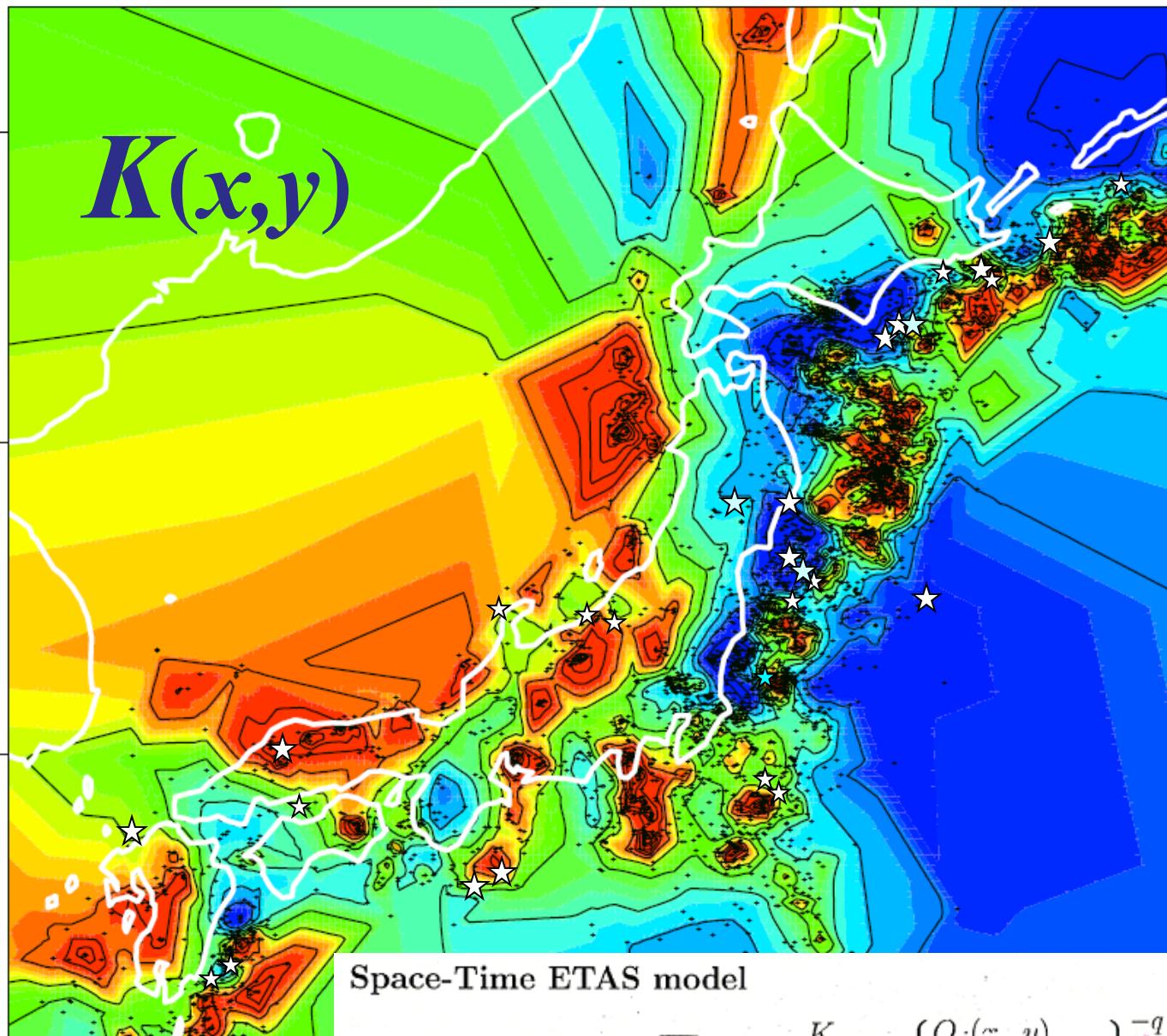
10^{-3}

35

events $\text{deg}^2 \text{ day}$

30

$10^{-3.5}$



Space-Time ETAS model

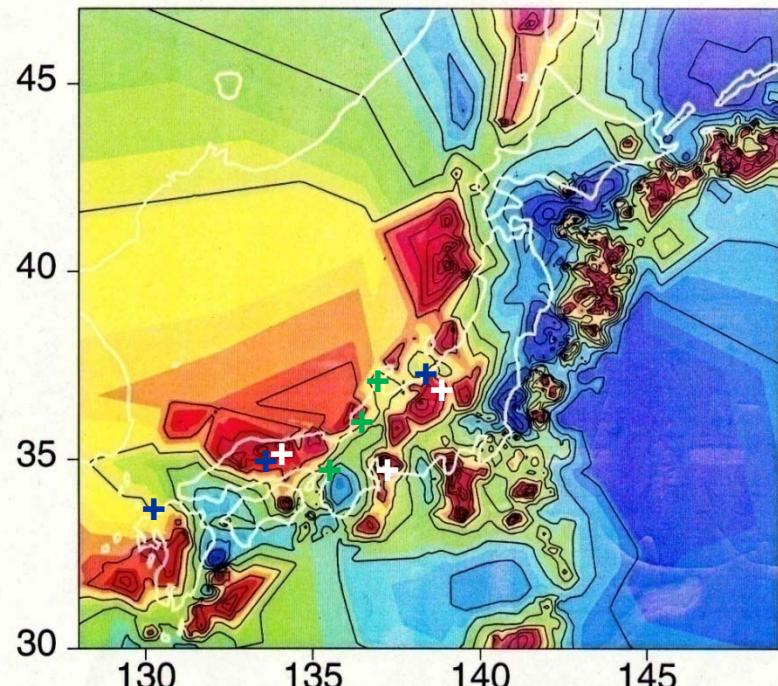
$$\lambda(t, x, y) = \underline{\mu} + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

130

Short term forecast of earthquakes, in and around Japan



$$\lambda(t, x, y) = \mu + \sum_{\{j: t_j < t\}} \frac{K}{(t - t_j + c)^p} \left\{ \frac{Q_j(x, y)}{e^{\alpha M_j}} + d \right\}^{-q}$$

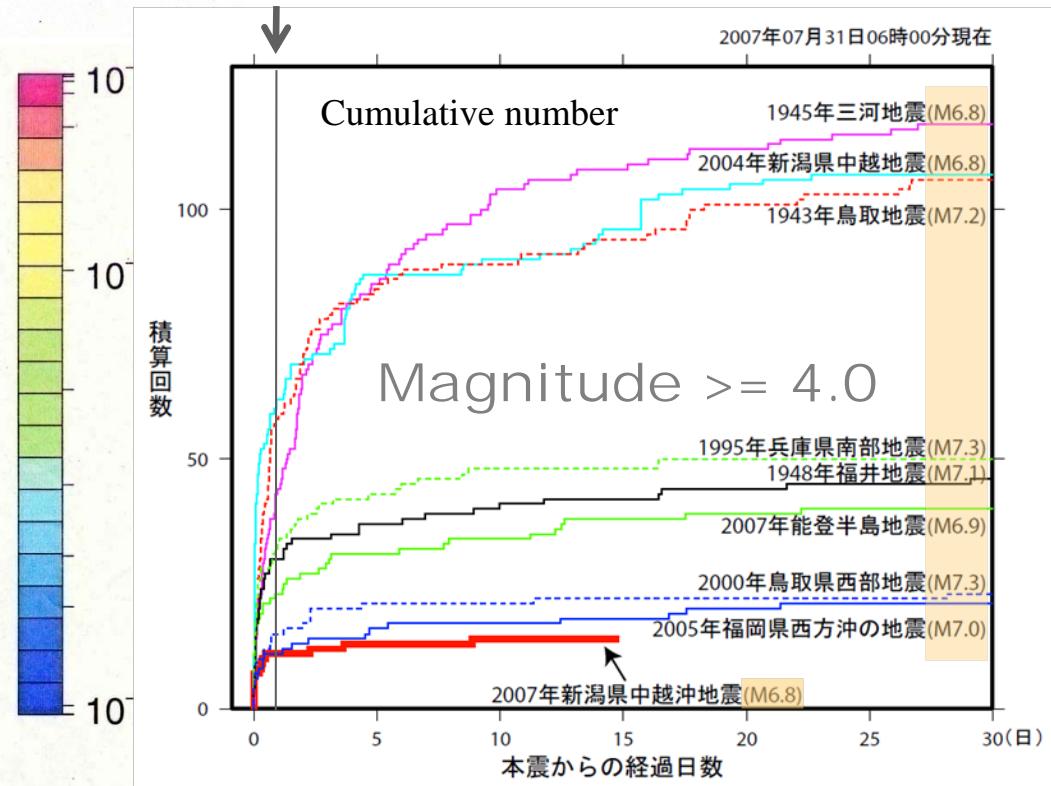


K-values

Estimated from $M \geq 5.0$
for 1926-1995

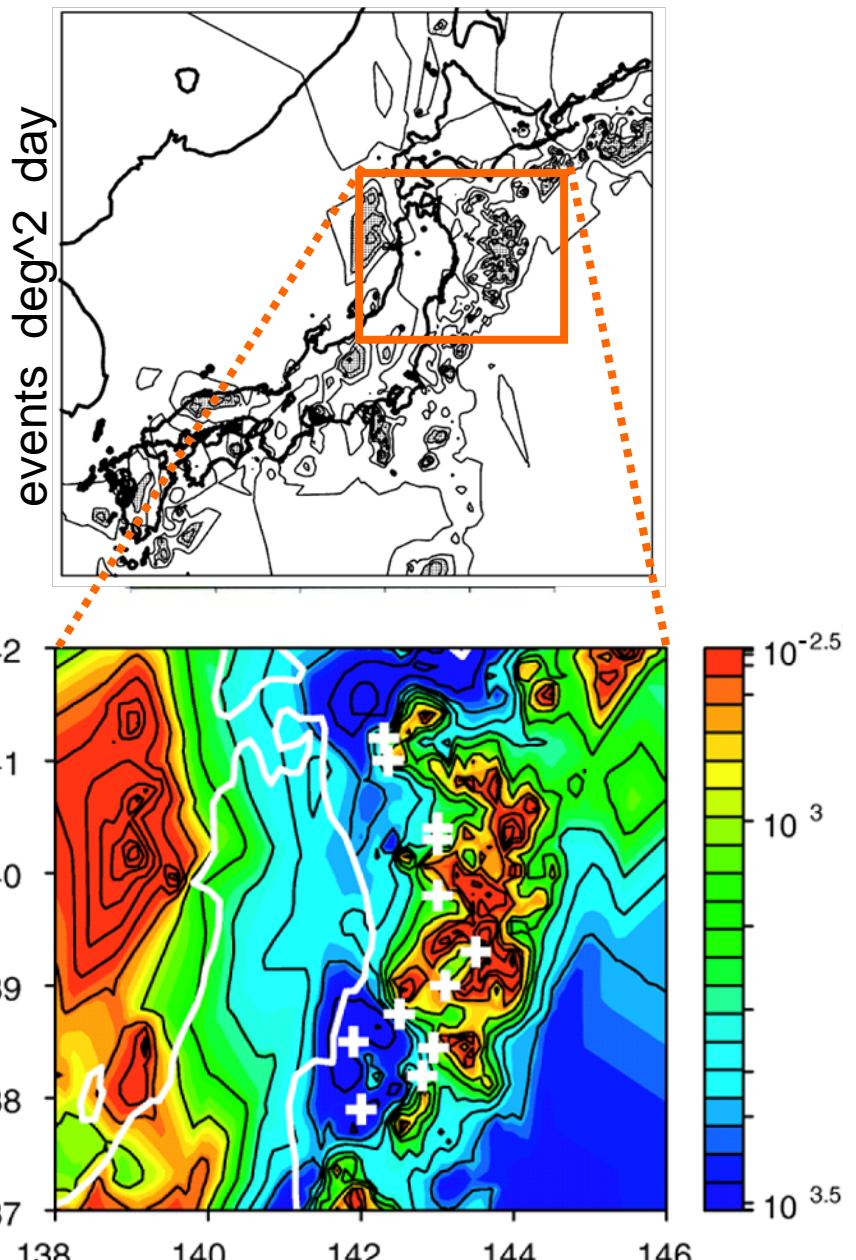
Mainshock M6.8~7.3
Due to the JMA report.

24 hours

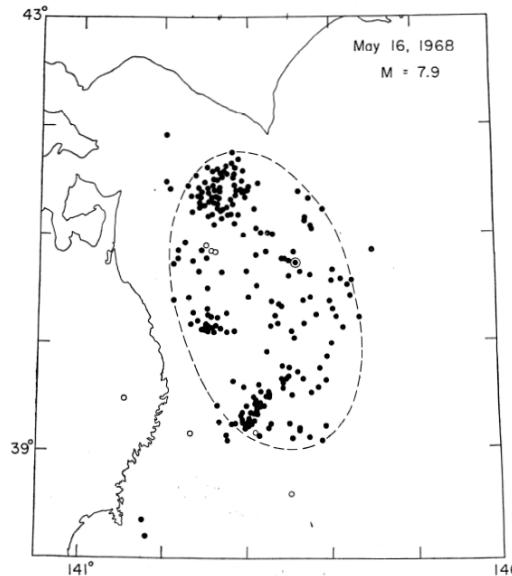


Elapsed time after main shock (days)

K-values

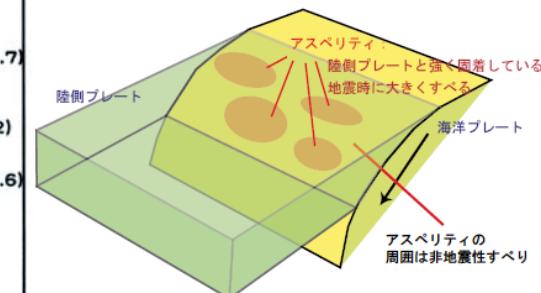
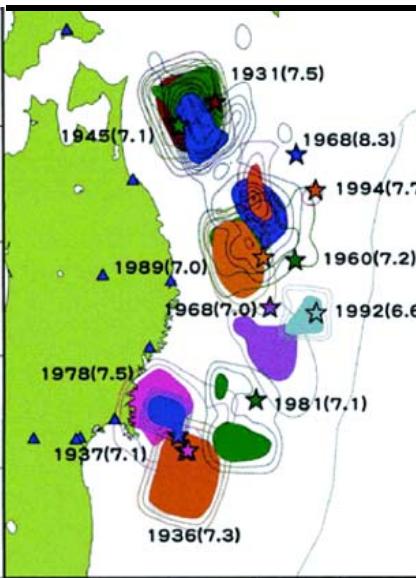


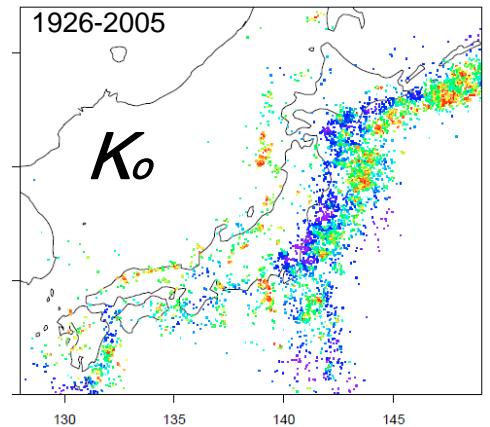
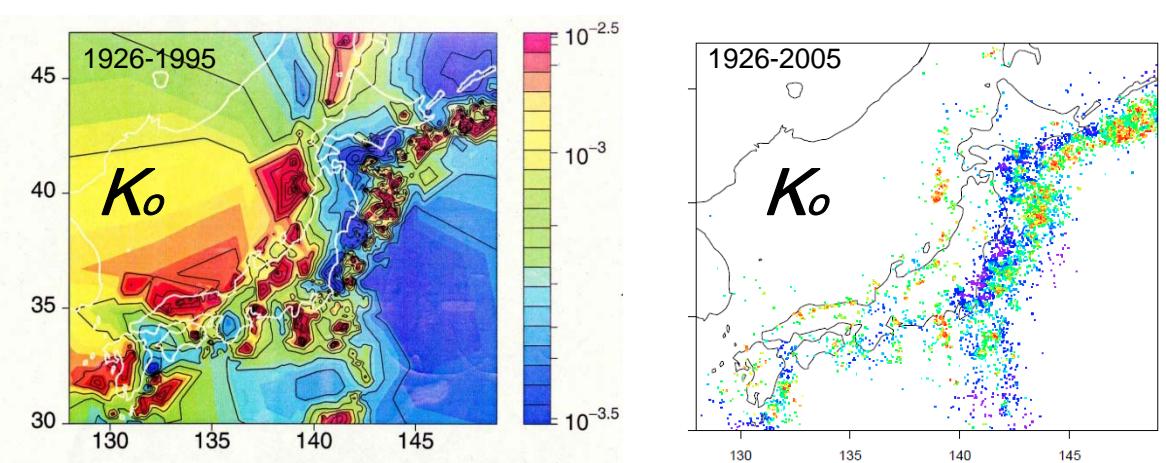
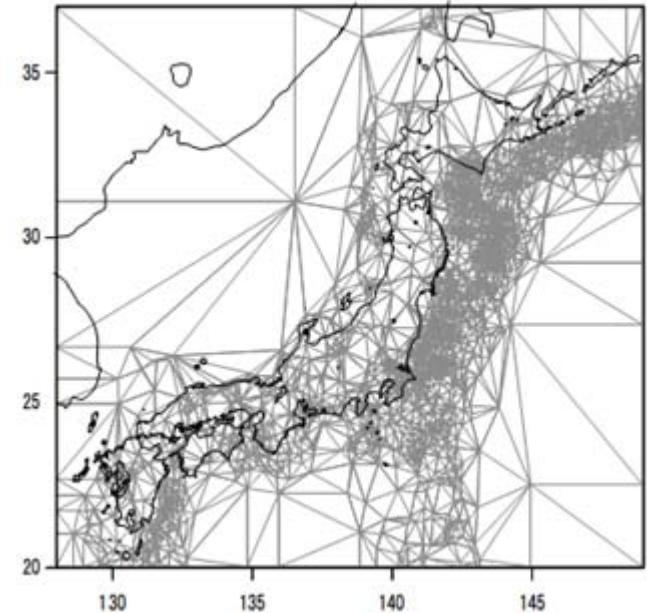
1968 M7.9 Tokachi aftershocks



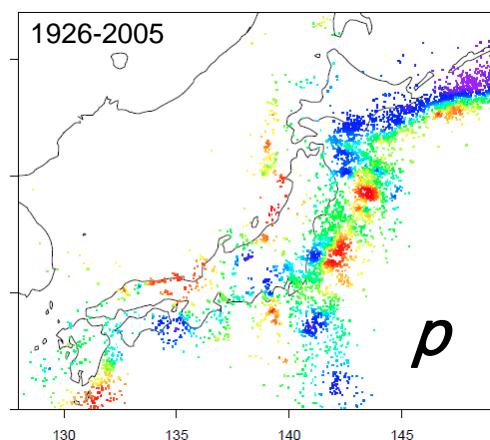
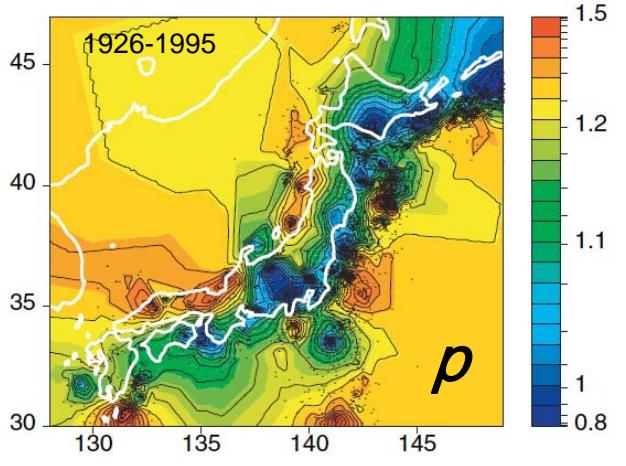
Utsu(1969)

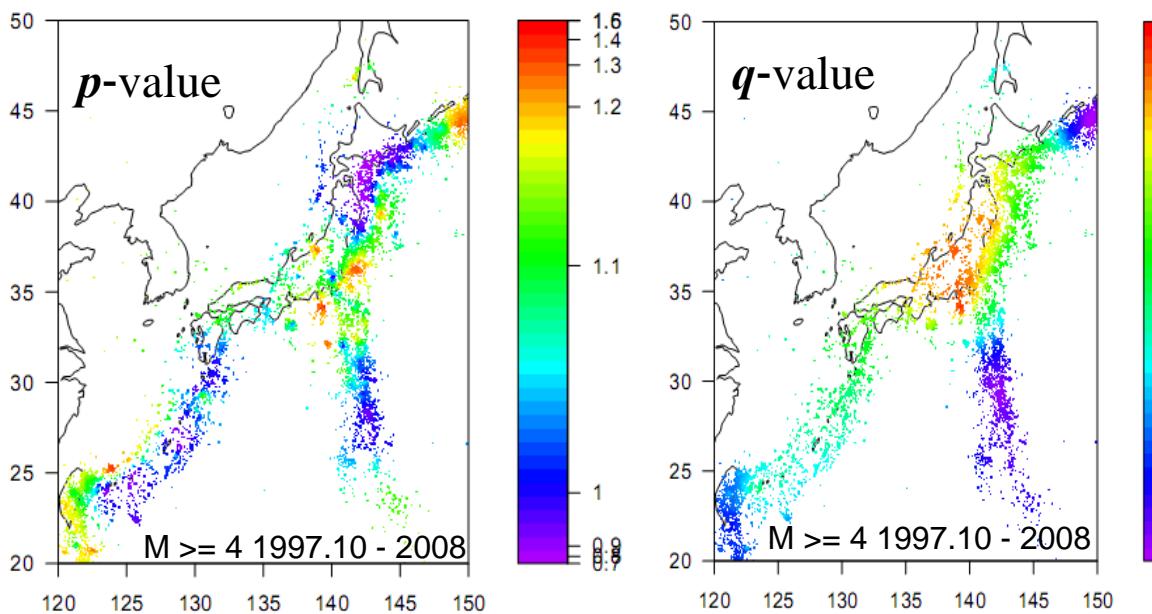
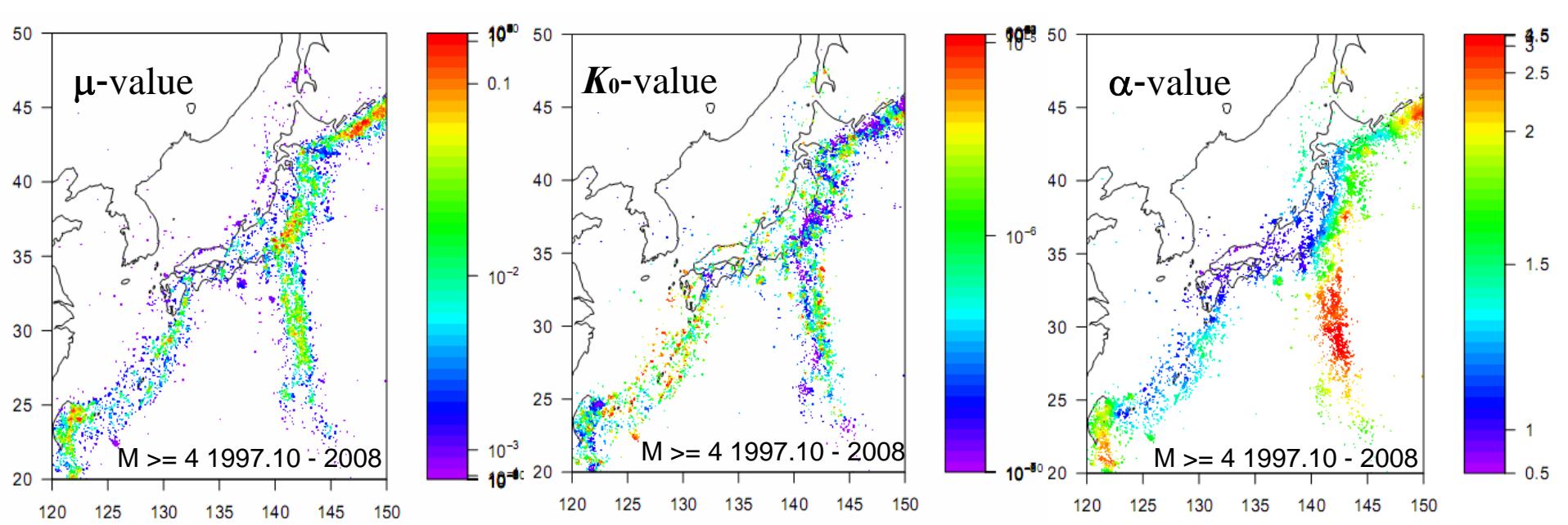
Asperities
Yamanaka & Kikuchi (2001)

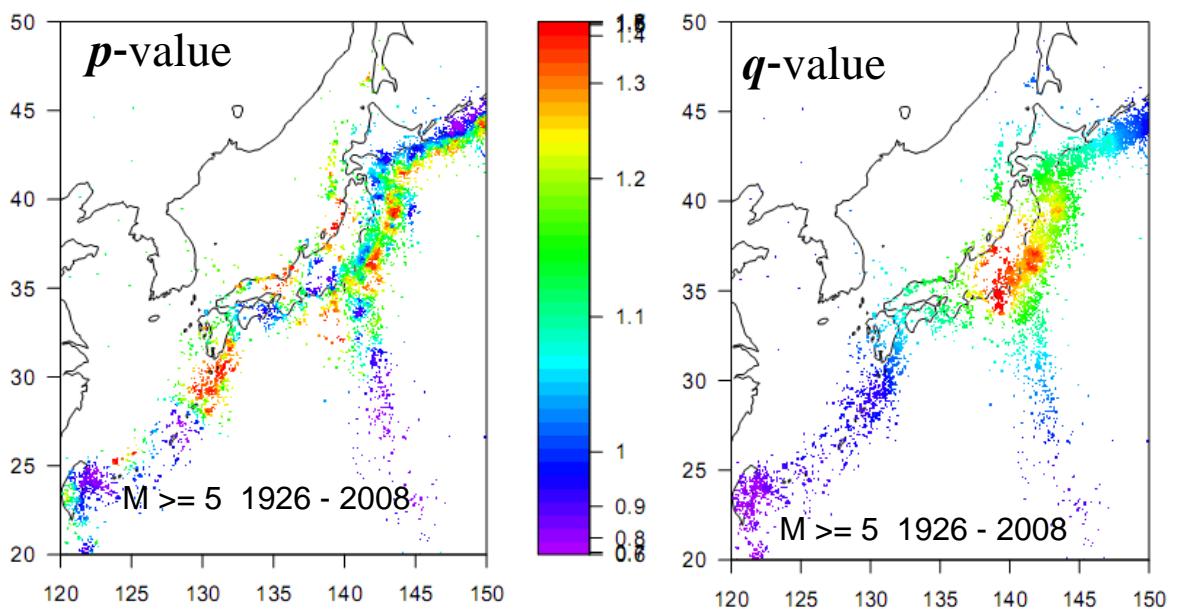
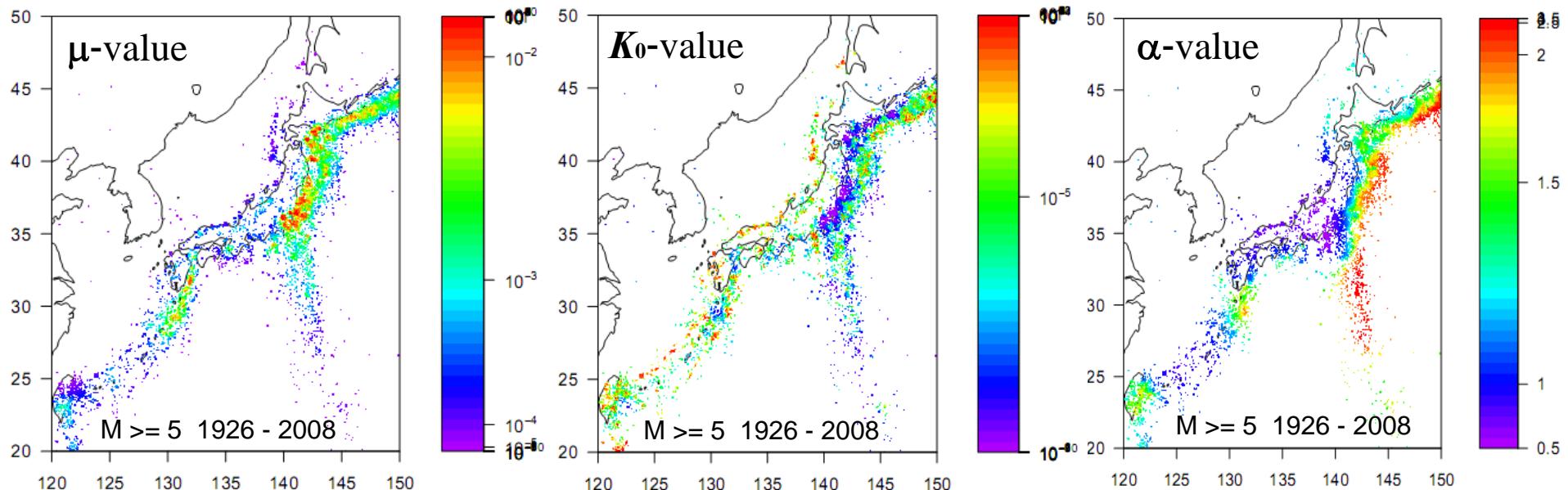




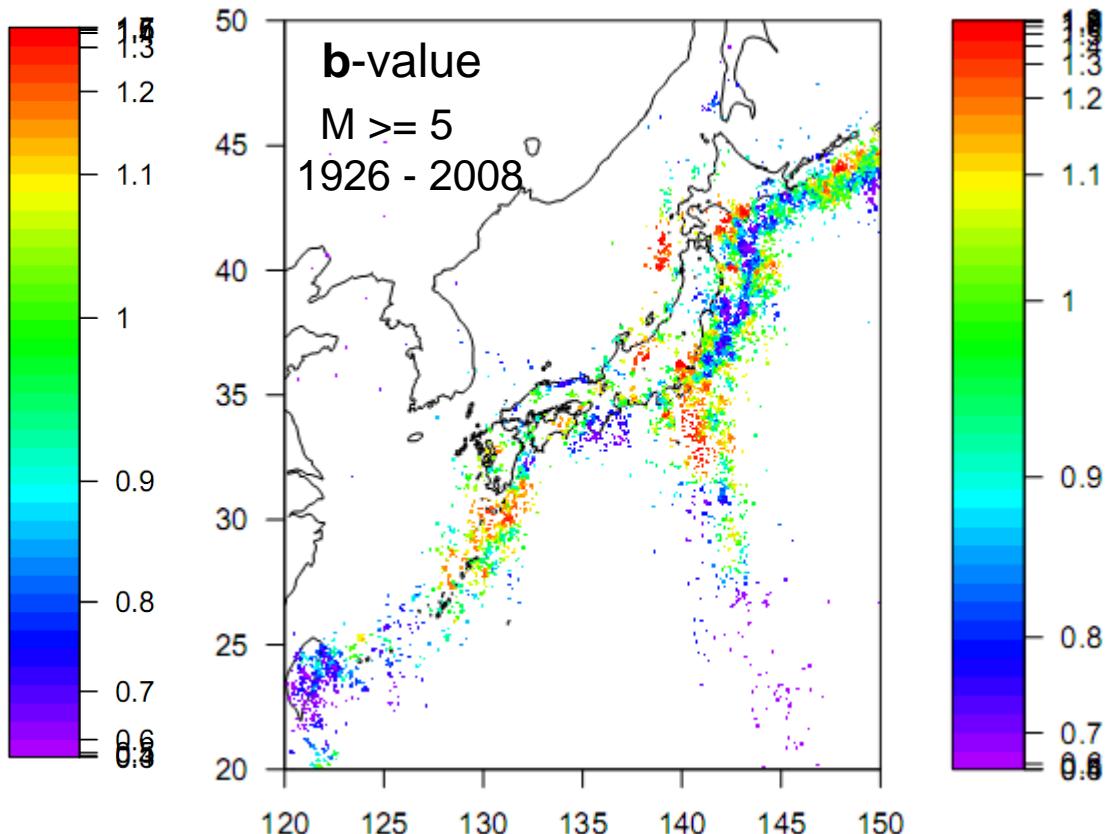
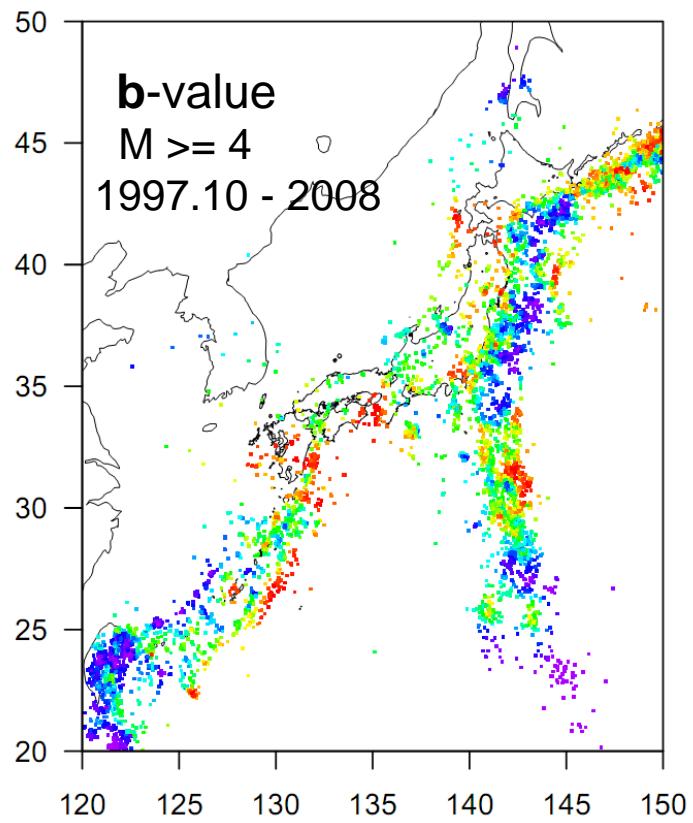
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Magnitude frequency



これで終わりです。お疲れさまでした。

Thank you very much for your attention

