

Multiple change-point analysis

A Bayesian approach to change-point analysis for point processes (y_1, y_2, \dots, y_n) : combines

- Poisson-process likelihood:
 $p(y|x) \propto \exp\{\sum_{i=1}^n \log x(y_i) - \int_0^L x(t) dt\}$
- prior model for step function $x(t)$, $0 \leq t < L$, representing intensity

Prior model: represent step function by

$$(k, \{s_j\}_{j=1}^k, \{h_j\}_{j=0}^k): x(t) = \sum_j h_j I_{[s_j, s_{j+1})}(t)$$

- number of steps k : Poisson(λ),
- step heights h_j : Gamma(α, β),
- step positions s_j : $p(s|k) \propto \prod_j (s_{j+1} - s_j)$,

all independent.

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MCMC for step functions

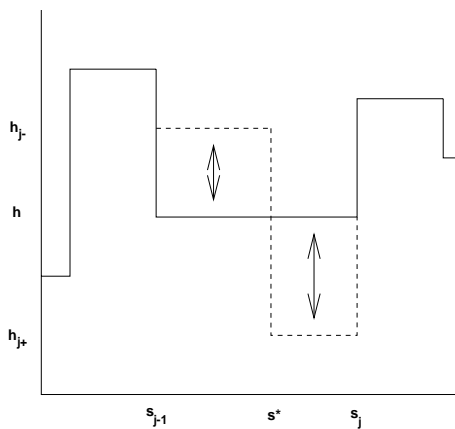
$$\theta = (k, \{s_j\}_{j=1}^k, \{h_j\}_{j=0}^k, \alpha, \beta)$$

I will use four moves:

- Metropolis change to a randomly chosen step height h_j .
- Metropolis change to a randomly chosen step position s_j .
- Jump move: birth/death of steps
 - birth: choose new step position s^* at random, split current step height h into two: (h_-, h_+)
 - death: choose step at random to kill, combine current step heights (h_-, h_+) into one: h
- Update hyperparameters α, β

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Birth and death of steps



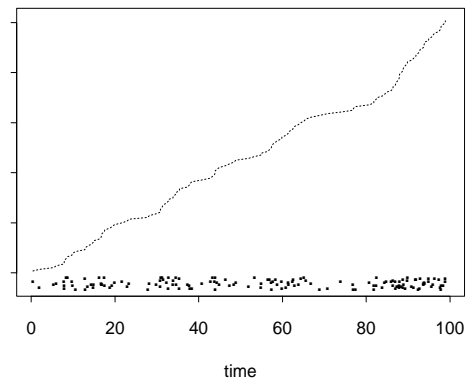
$$h_-^w h_+^{w_+} = h^{w_- + w_+}$$

$$(h, w, s^*, u) \leftrightarrow (h_-, h_+, w_-, w_+)$$

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Example: cyclones hitting the Bay of Bengal

141 cyclones over a period of 100 years
(a cyclone is a storm with winds $> 88 \text{ km h}^{-1}$).

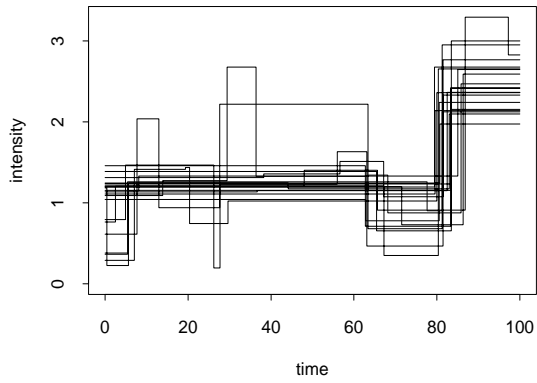


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Choices of hyperparameters:

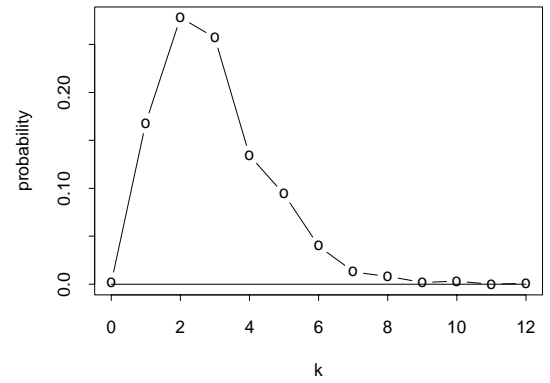
- Prior on k : Poisson(λ), with $\lambda = 3$.
- Prior on h_j : Gamma(α, β), with
 - $\alpha \sim \Gamma(2, 2)$
 - $\beta \sim \Gamma(1, n/L)$

Sample of step functions from the posterior:



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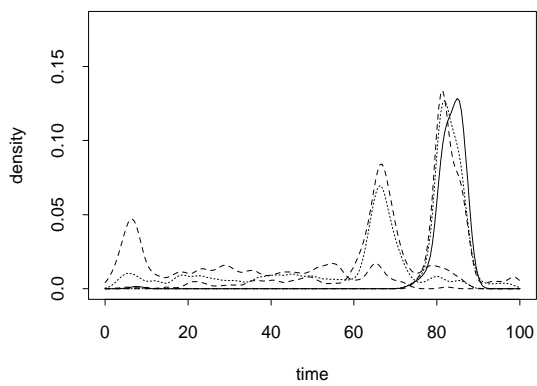
Posterior for the number of change points k



Zero change points is ruled out; $k = 1$ or 2 more probable than under the prior.

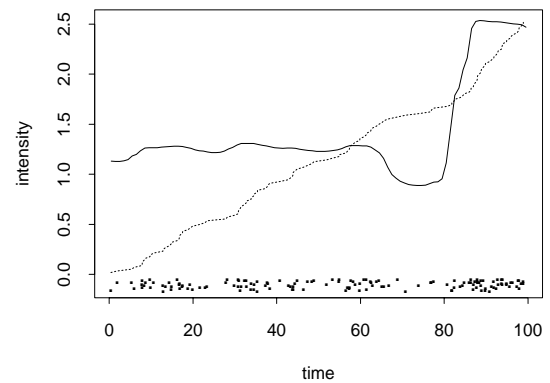
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Posterior density estimates for change-point positions



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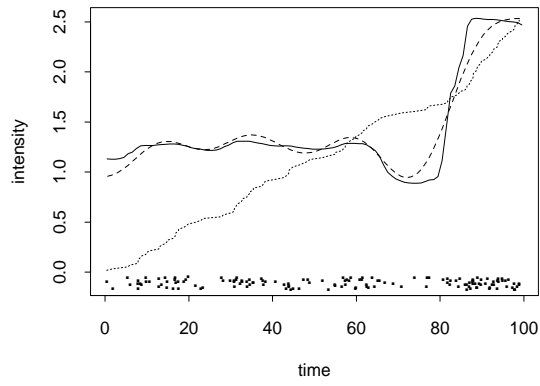
Model-averaged estimate: $E(x(\cdot)|y)$



(the expectation of a random step function is not a step function).

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Ordinary smoothing methods (in this case a kernel smoother) can't match that mean curve



– fixed-bandwidth smoothers either over-smooth the steps, or under-smooth the plateaux.