

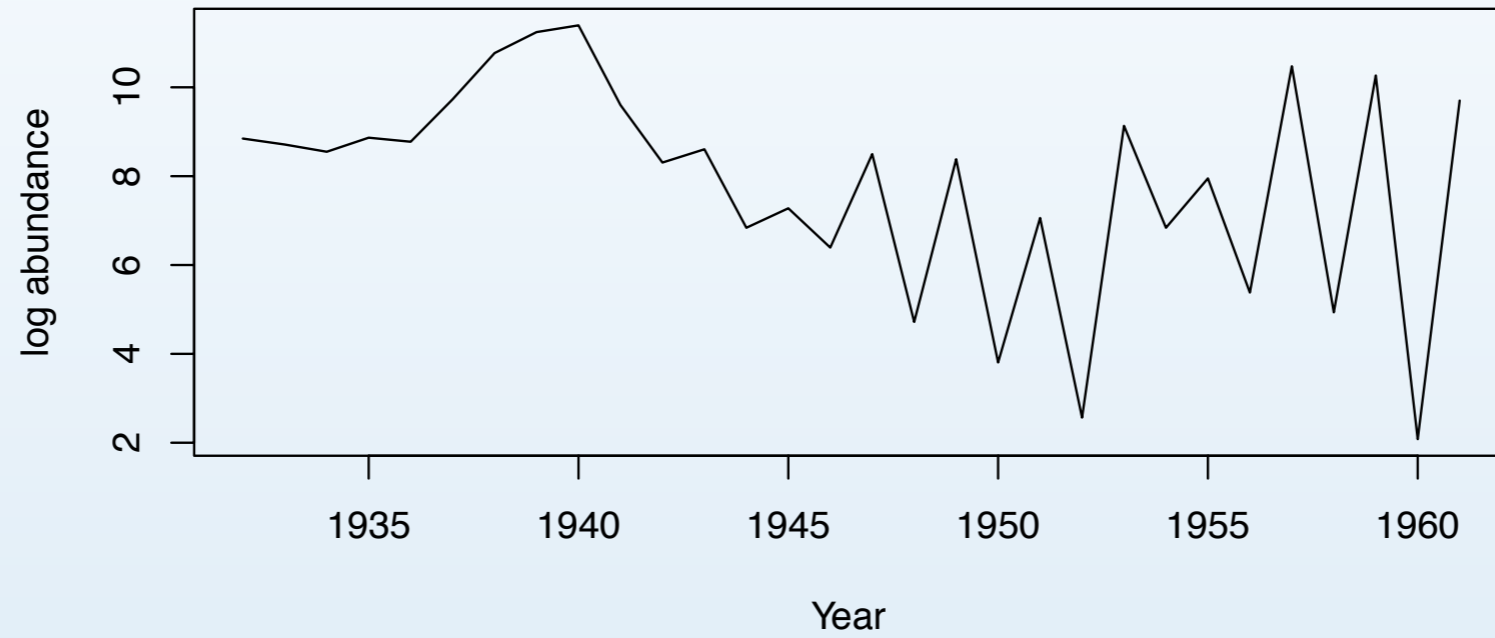
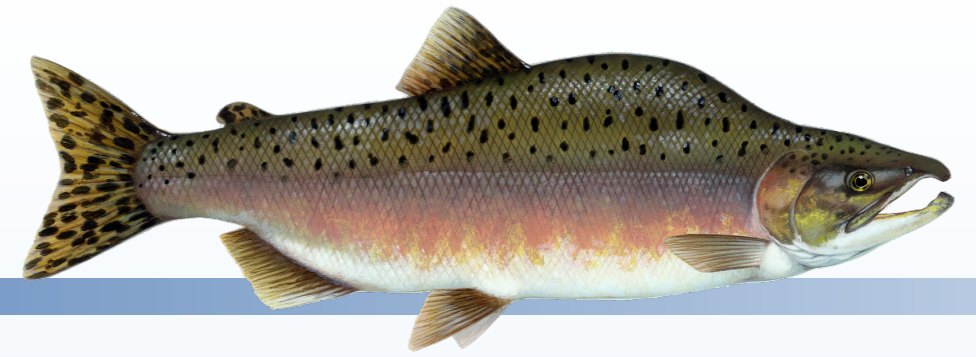
Bayesian model selection for Markov chains using sparse probability vectors

Matthew Heiner¹, Athanasios Kottas¹, and Stephan Munch²

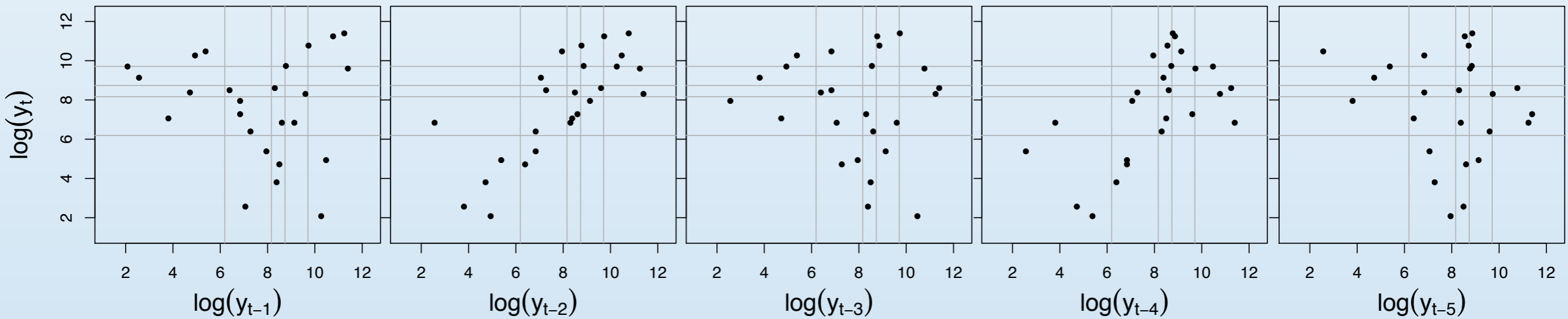
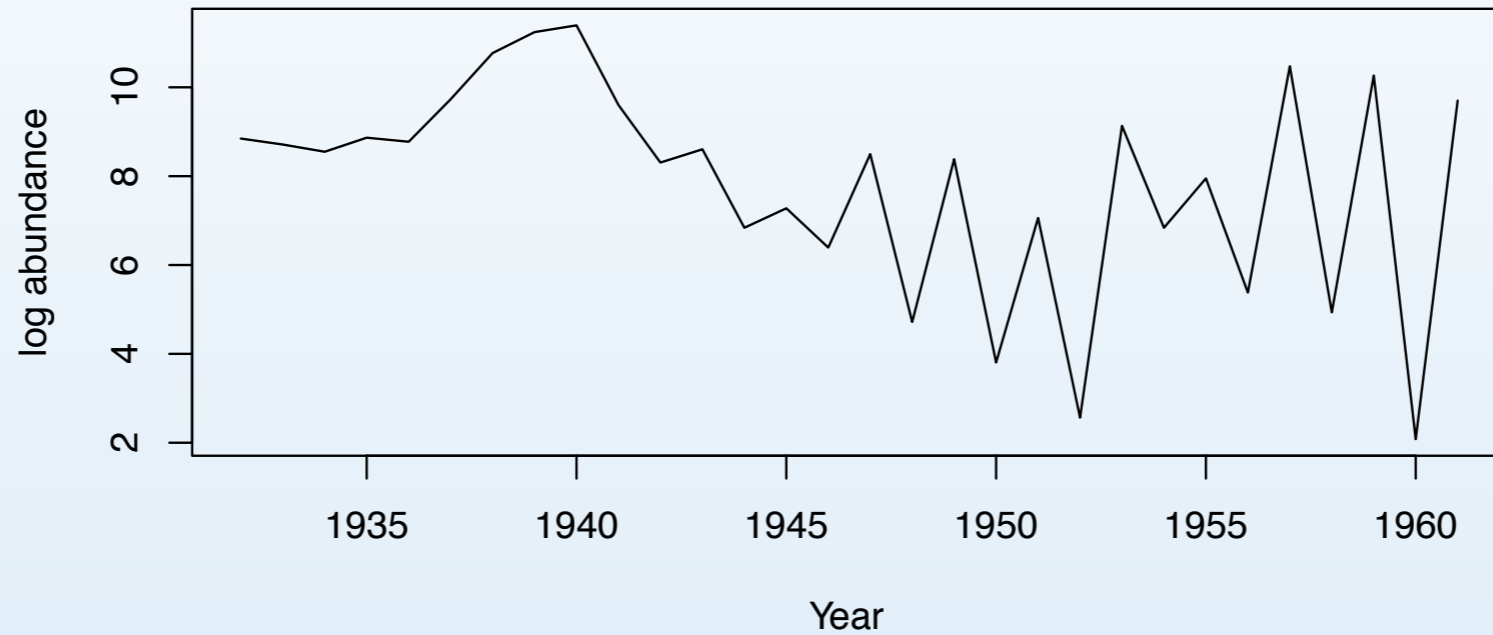
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Pink Salmon time series



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Mixture transition distribution

Raftery (1985)

Example with two states and three lags:

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first order
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Sparse probability vectors

Sparse Dirichlet mixture (SDM) prior

$$p(\boldsymbol{\theta}) \propto \text{Dir}(\boldsymbol{\theta}; \boldsymbol{\alpha}) \times \sum_{k=1}^K \theta_k^\beta, \quad \beta > 1$$

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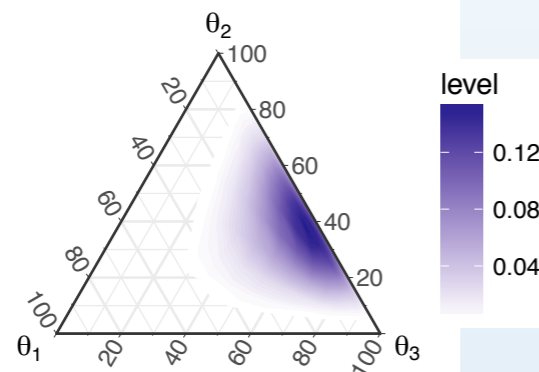
$$p(\boldsymbol{\theta} \mid \mathbf{n}) \propto \prod_{k=1}^K \theta_k^{n_k} \times \text{Dir}(\boldsymbol{\theta}; \boldsymbol{\alpha}) \times \sum_{k=1}^K \theta_k^\beta \propto \text{Dir}(\boldsymbol{\theta}; \boldsymbol{\alpha} + \mathbf{n}) \times \sum_{k=1}^K \theta_k^\beta$$

Sparse probability vectors

Multinomial data example

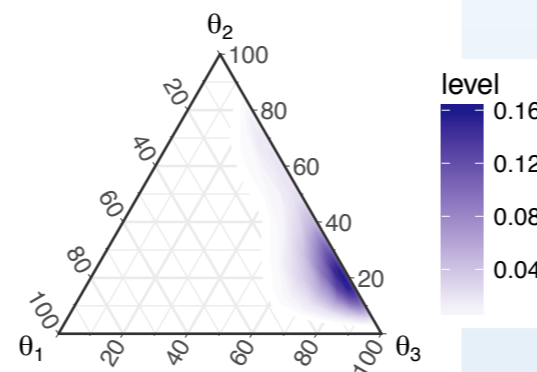
Density of $\theta|x$, Dirichlet prior

$n = (0,3,5)$ $\alpha = 0.333$



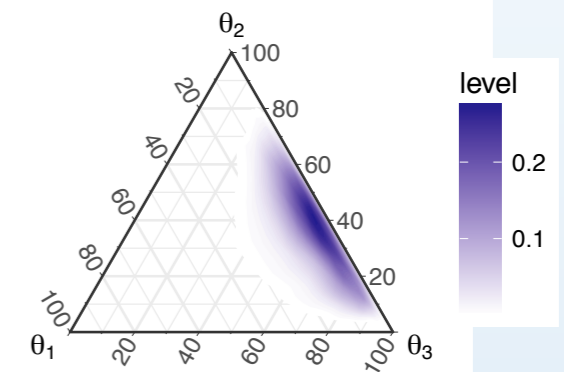
Density of $\theta|x$, SDM prior

$n = (0,3,5)$ $\alpha = 0.333$ $\beta = 8$



Density of $\theta|x$, SBM prior

$n = (0,3,5)$ $\eta = 16$ $\pi = 0.65$ $\gamma = 1.5$ $\delta = 1$

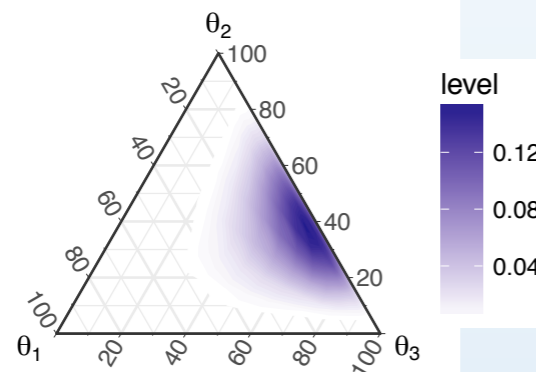


Sparse probability vectors

Multinomial data example

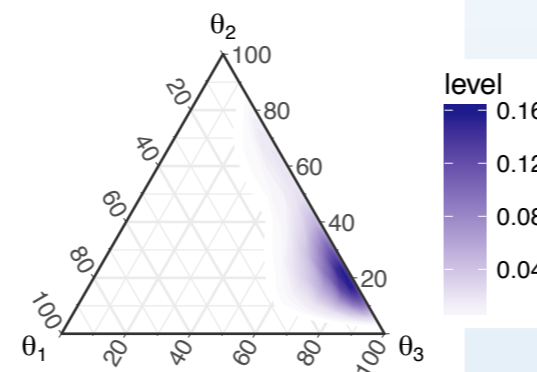
Density of θ_{1x} , Dirichlet prior

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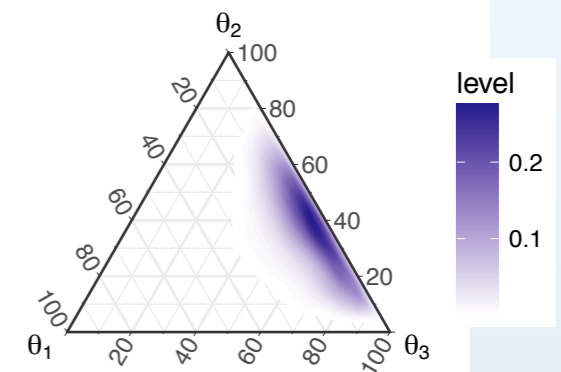
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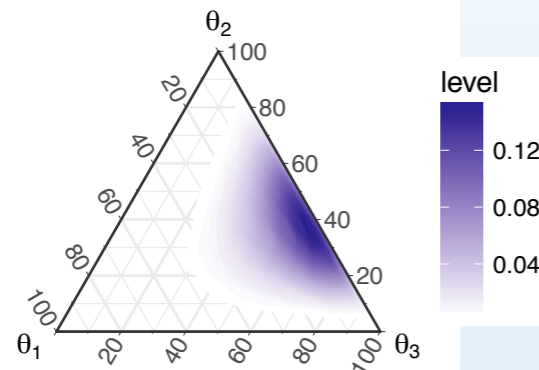


Lag inclusion inference, salmon data

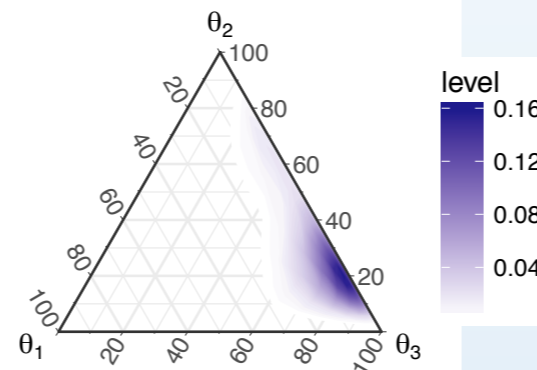
Sparse probability vectors

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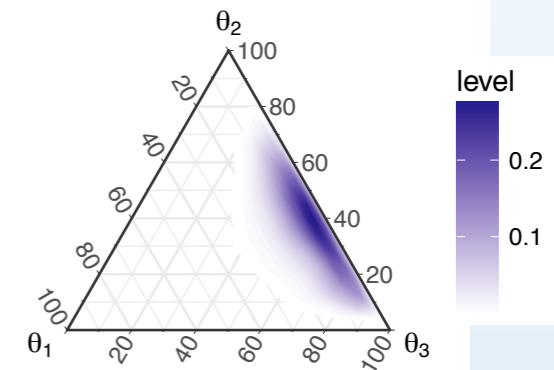


Density of $\theta|x$, SDM prior
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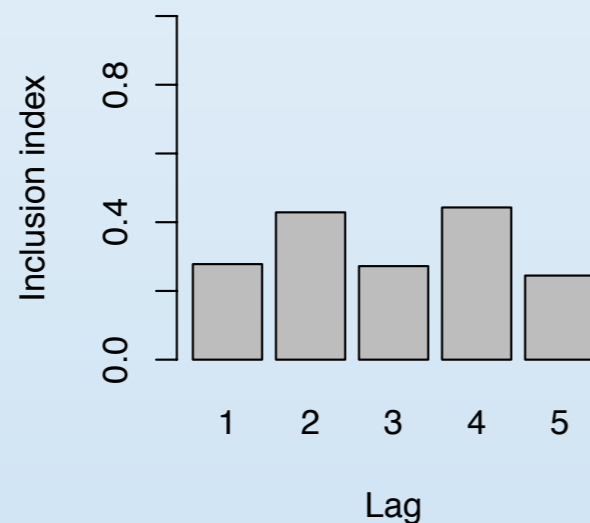


Density of $\theta|x$, SBM prior

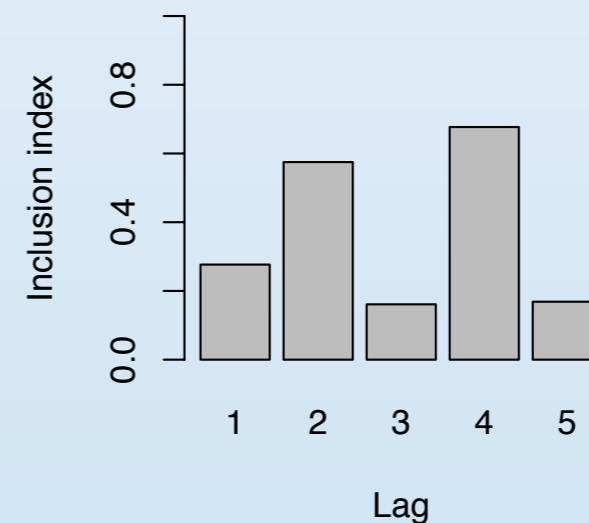
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Dirichlet prior



SDM prior

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References:

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