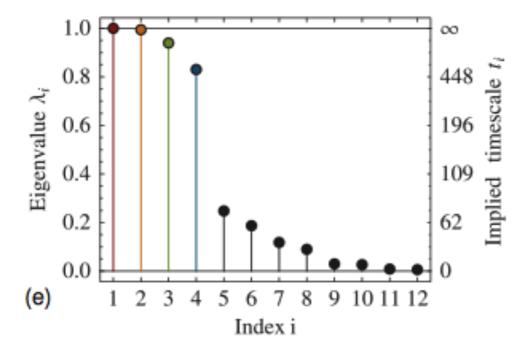
Simon Olsson

2020 PyEMMA Workshop FU Berlin **Tuesday, Feb 18th**

Discretization

Backward propagator

$$\rho_{\tau} = \mathcal{T}(\tau)\rho_0$$



Spectral decomposition

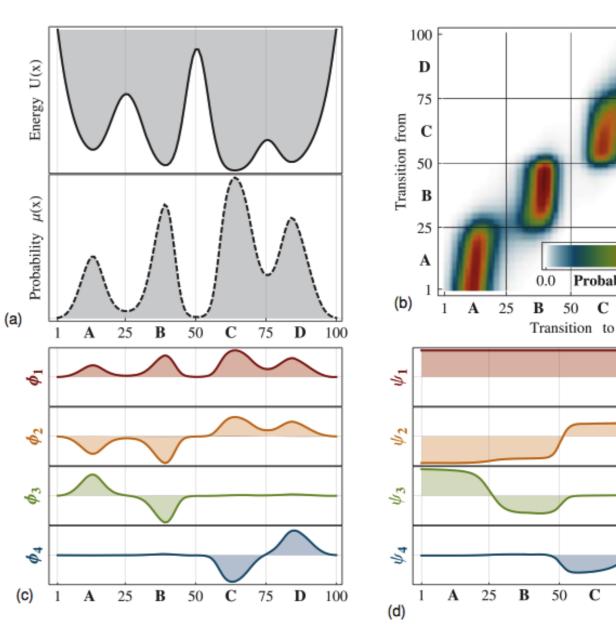
$$\rho_{\tau} = \sum_{i=1}^{\infty} e^{-\tau \kappa_i} \langle \psi_i \mid \rho_0 \rangle \psi_i$$

Probabilities

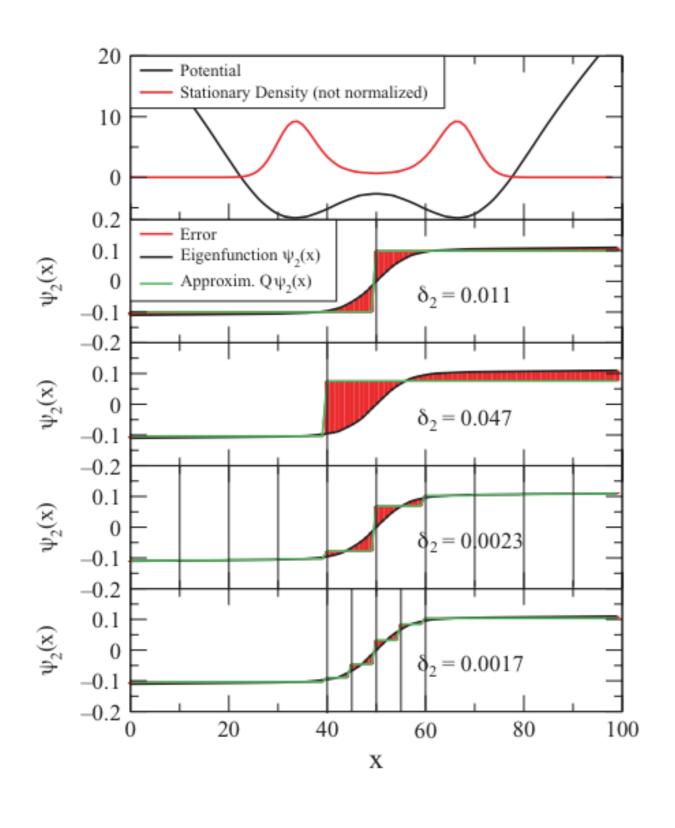
C

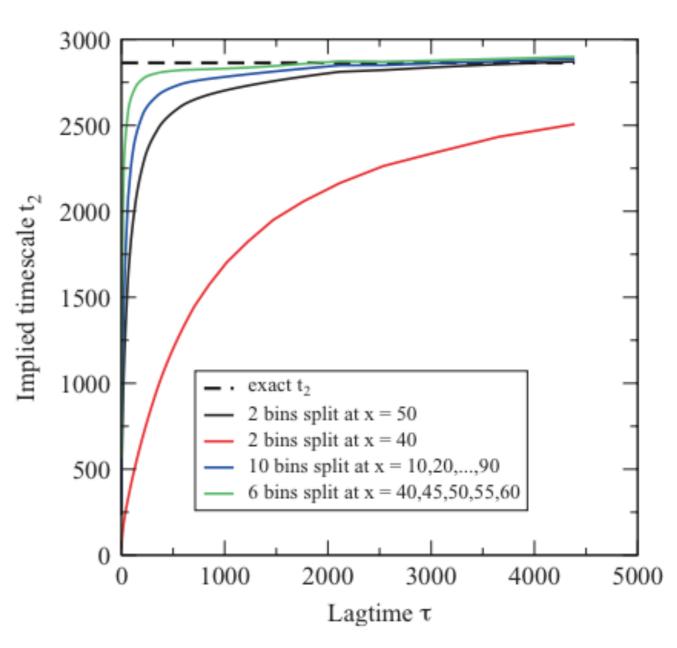
75

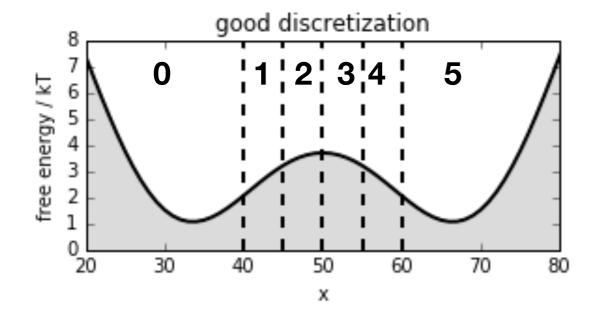
75 **D** 100

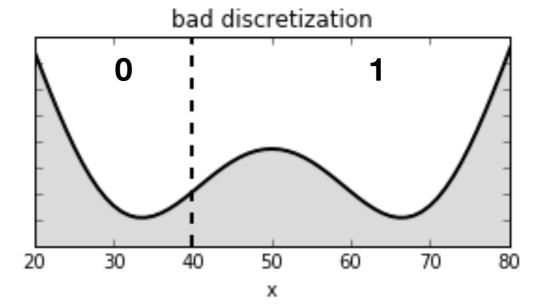


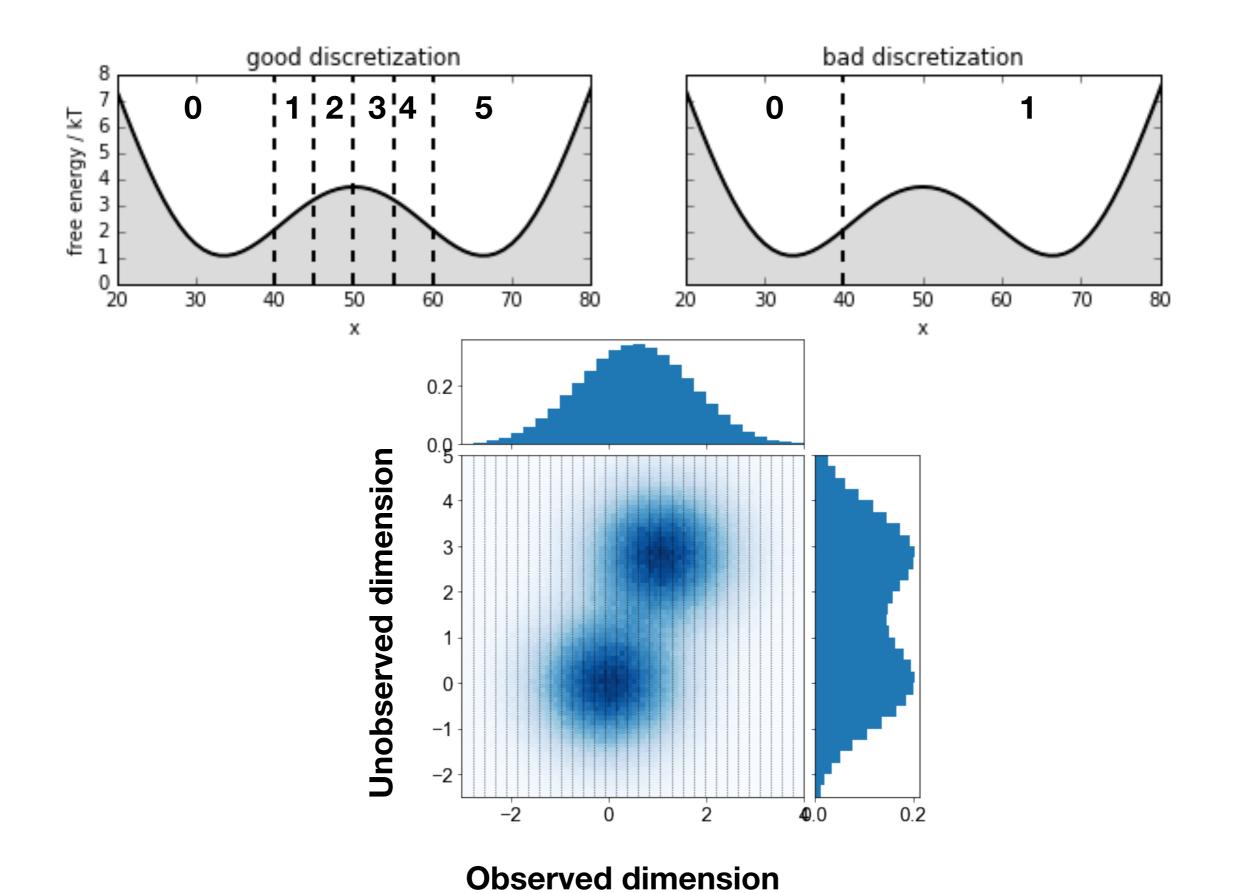
Discretization

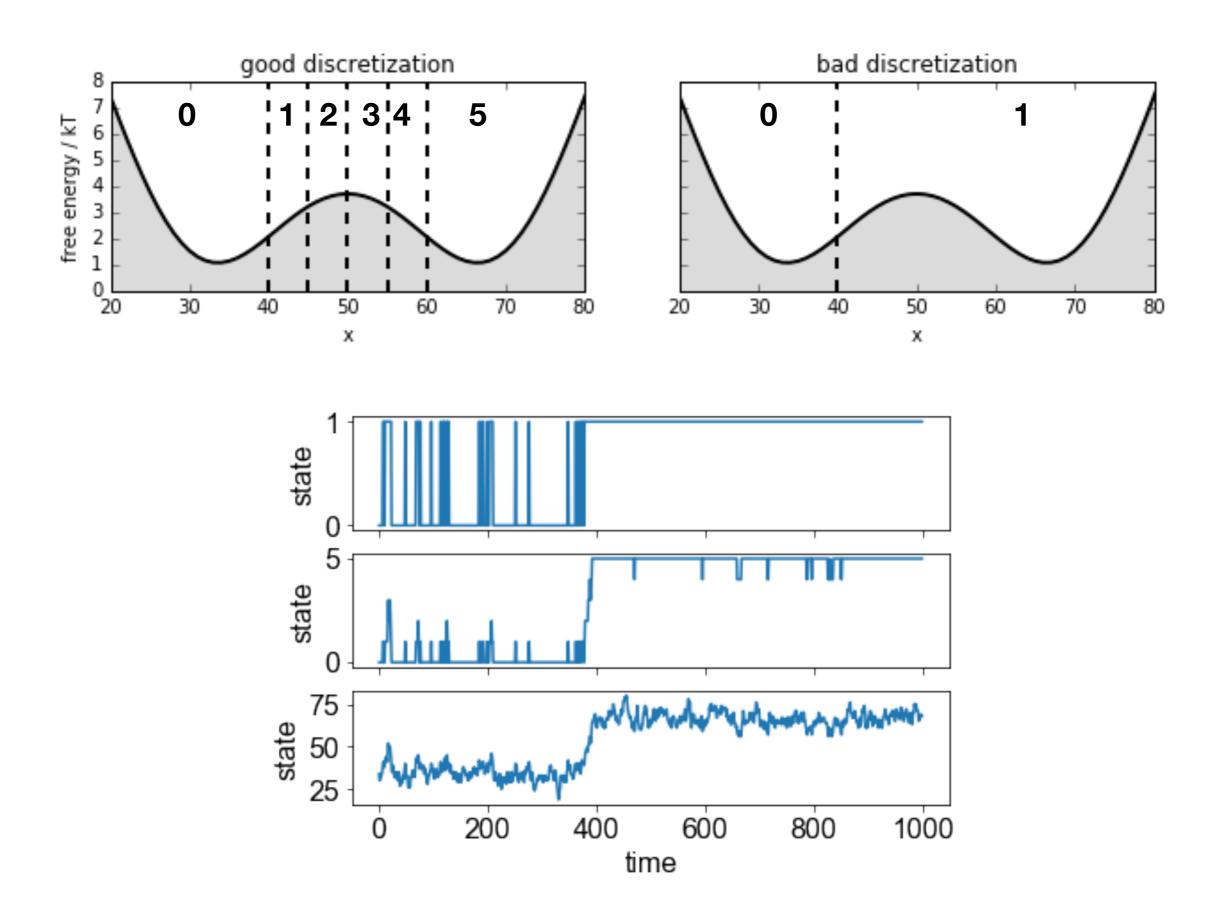


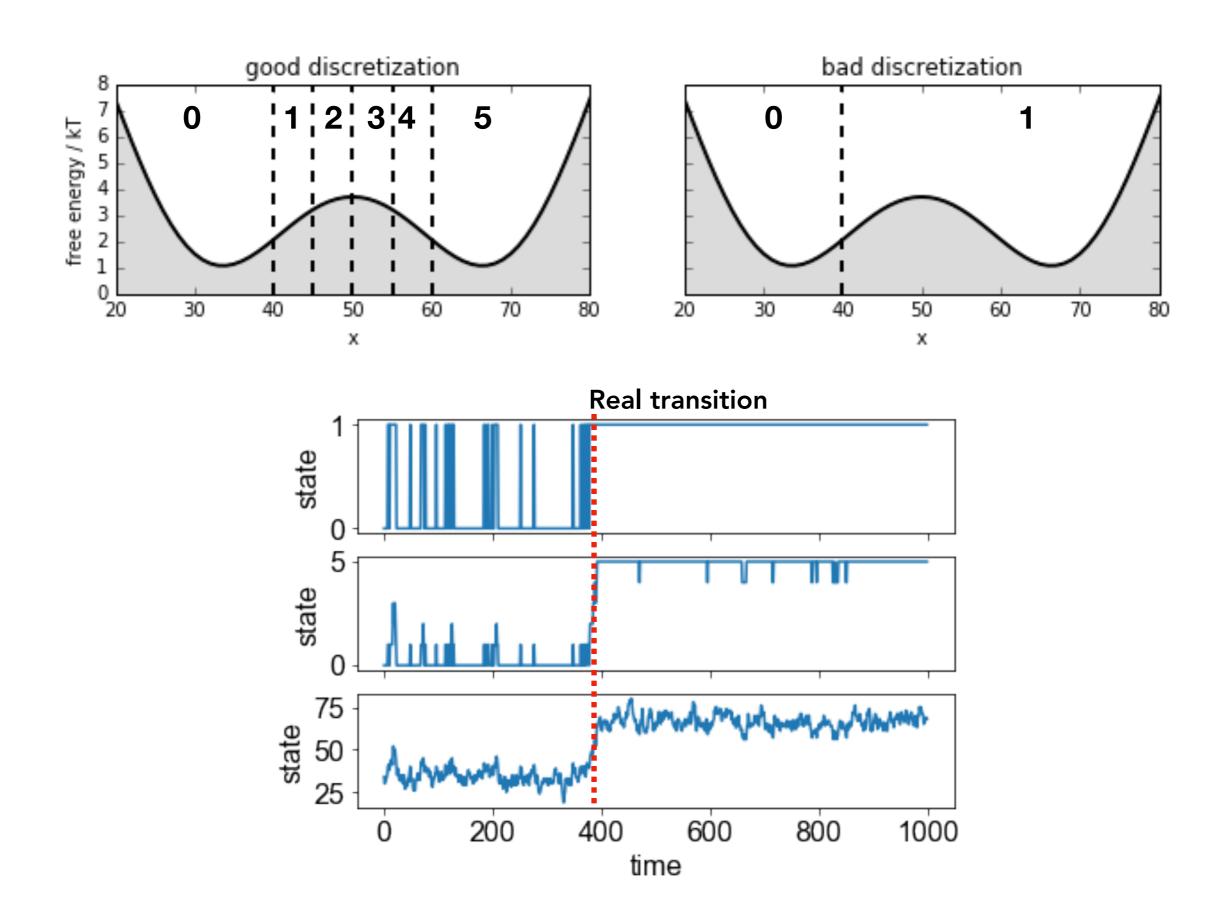


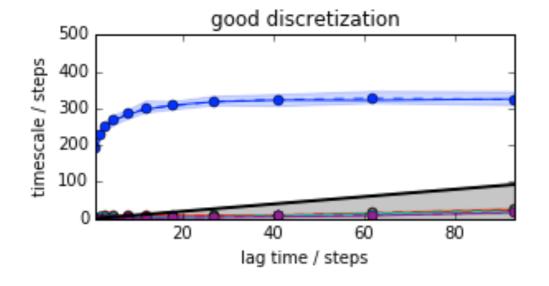


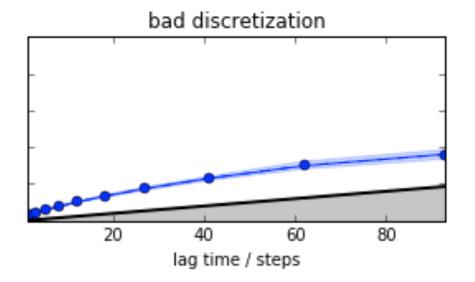


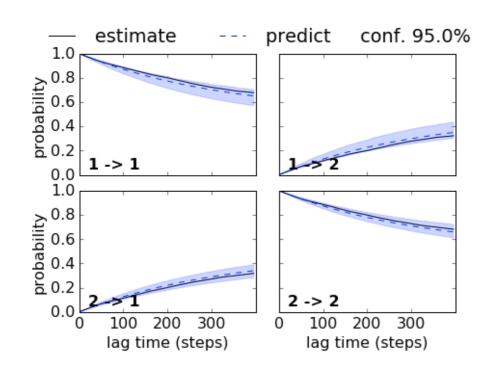


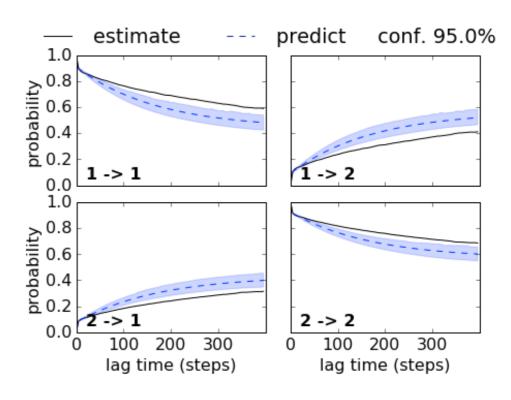












Projection/discretization error leads to systematic errors

- Discretization and projection errors hampers our ability to distinguish between meta-stable states
- Apparent non-Markovian behavior of the dynamics.

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Remedies:

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- Improve featurization and clustering

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However we know that the underlying dynamics is Markovian, can we exploit fact in some way?

We assume the existence of an underlying (hidden) Markovian dynamics described by the transition probabilities $\mathbf{P} = \{p_{ij}\}$

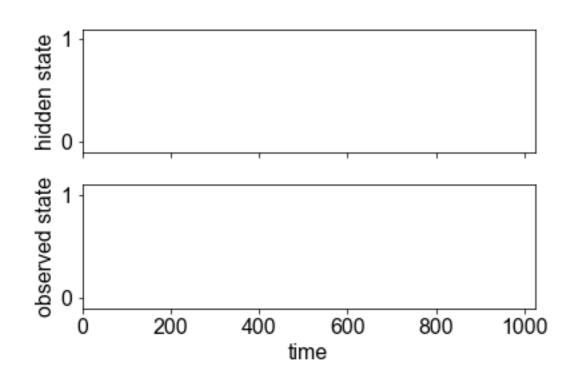
Instead of observing the state h_i directly we observe some distorted representation, s_i with a probability $\chi_{h_i s_i}$ — the emission probabilities.

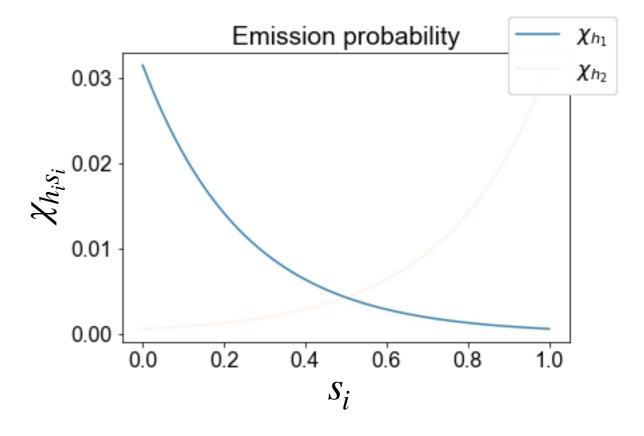
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Hidden states
$$h_1 \longrightarrow h_2 \longrightarrow \cdots \longrightarrow h_T$$

 $\downarrow \chi_{h_1 s_1} \longrightarrow \chi_{h_s s_s} \longrightarrow \cdots \longrightarrow h_T$
Observed states $s_1 \longrightarrow s_2 \longrightarrow \cdots \longrightarrow s_T$



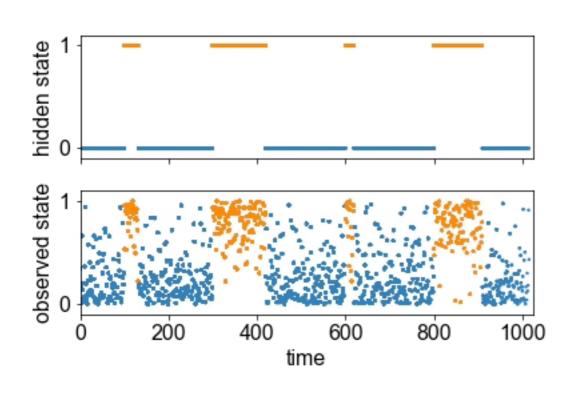


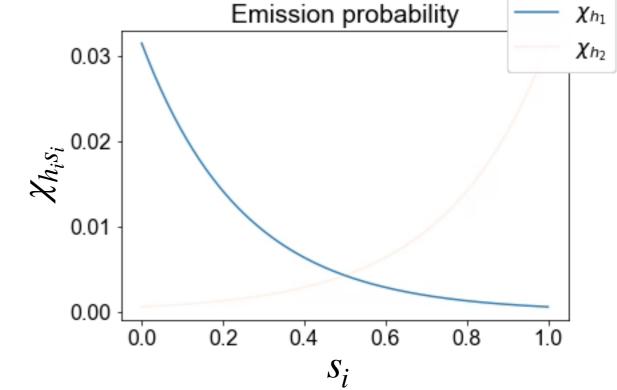
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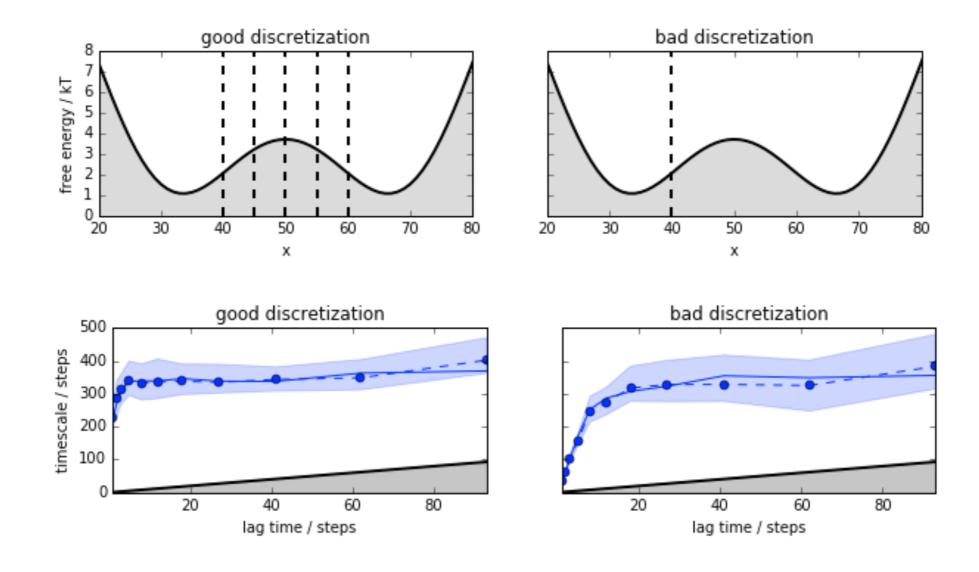
Maximum-Likelihood and Bayesian estimators are available:

Rabiner Proc IEEE (1989) 77,2, pp.257 Noé et al. JCP (2013) 139, 184114 Chodera et al. arxiv:1108:1430

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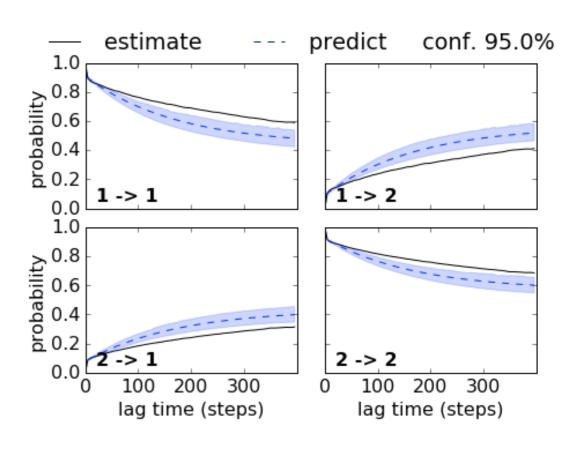
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Good discretization

estimate predict conf. 95.0% 1.0 probability 6.0 8.0 8.0 8.0 1 -> 1 0.0 1.0 probability 8.0 8.0 8.0 8.0 2 -> 2 200 200 100 300 lag time (steps) lag time (steps)

Bad discretization



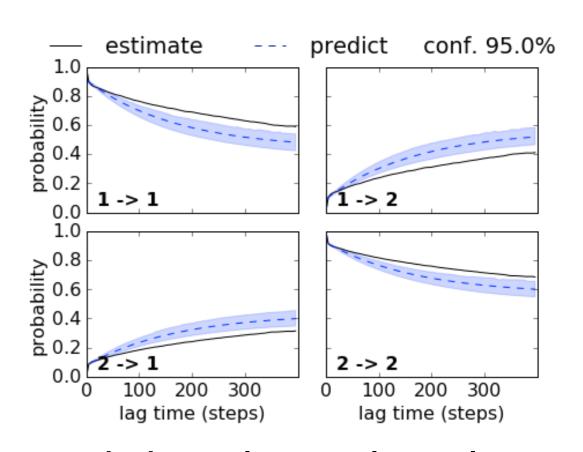
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We get a robust model of the dynamics which simultaneously resolves meta-stable states.

Questions?