

Multistability in electric power grids on meshed, complex networks

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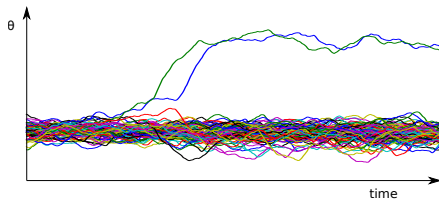
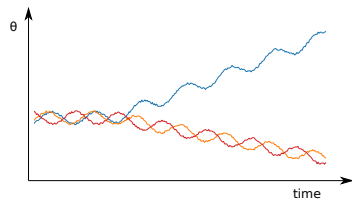
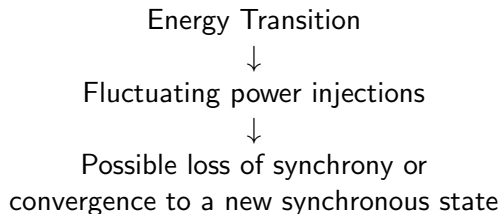
Joint work with

M. Tyloo, T. Coletta, and Ph. Jacquod

R. Delabays, M. Tyloo, and Ph. Jacquod, *Chaos* **27** (2017)

T. Coletta, R. D., I. Adagideli, and Ph. Jacquod, *New J. Phys* **18** (2016)

Noisy injection



K. Schmietendorf, J. Peinke, and O. Kamps, *Euro. Phys. J. B* **90** (2017)

B. Schäfer et al., *Phys. Rev. E* **95** (2017)

Questions

How likely is it for a system to lose synchrony?

Can we quantify this probability?

What are the relevant parameter of the fluctuations?

The model

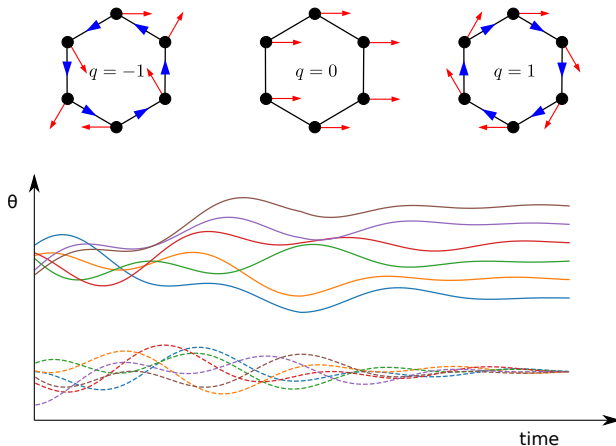
First-order, lossless approximation of the Swing Equations.

The Kuramoto model:

$$\dot{\theta}_i = (P_i + \delta P_i(t)) - K \sum_{j=1}^n a_{ij} \sin(\theta_i - \theta_j), \quad i \in \{1, \dots, n\}.$$

- ▶ $\theta_i \in \mathbb{S}^1$: Voltage angle,
- ▶ $P_i \in \mathbb{R}$: Power injection,
- ▶ K : Identical coupling,
- ▶ a_{ij} : Adjacency matrix.

Multistability

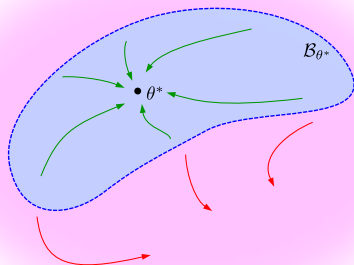


R. D., T. Coletta, and Ph. Jacquod, *J. Math. Phys.* **57** (2016)

R. D., T. Coletta, and Ph. Jacquod, *J. Math. Phys.* **58** (2017)

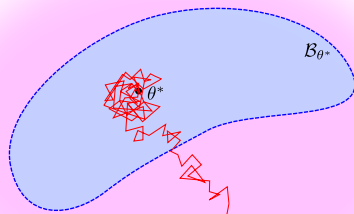
Basins of attraction

$$\mathcal{B}_{\theta^*} := \{\theta^{\circ} \in \mathbb{T}^n : \theta(0) = \theta^{\circ} \implies \theta(t \rightarrow \infty) = \theta^*\} .$$



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Simulations

Gaussian time-correlated noise: $\delta P_i: [0, +\infty) \rightarrow \mathbb{R}$ such that

$$\langle \delta P_i(t) \cdot \delta P_j(t') \rangle = \delta_{ij} \cdot \delta P_{i,0} \cdot \delta P_{j,0} \cdot \exp(-|t - t'|/\tau_0),$$

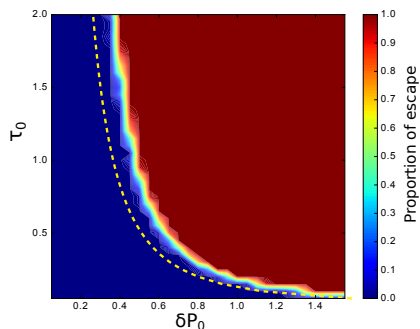
where $\tau_0 > 0$ is the correlation time.

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where $\tau_0 > 0$ is the correlation time.



- ▶ Cycle $n = 83$,
- ▶ Identical frequencies:
 $P_i \equiv 0$,
- ▶ Observation time:
8e5 iterations,
- ▶ Dashed yellow line:
our prediction.

Analytical prediction

We consider

$$\theta_i(t) = \theta_i^* + \delta\theta_i(t), \quad \mathcal{J}(\boldsymbol{\theta}^*) \cdot \mathbf{u}_\alpha = \lambda_\alpha \mathbf{u}_\alpha.$$

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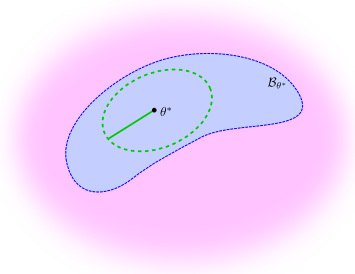
Linearizing the dynamics around the stable fixed point:

$$\begin{aligned} \langle \delta\theta^2 \rangle &= \left\langle \left(\sum_{\alpha} c_{\alpha} \mathbf{u}_{\alpha} \right)^2 \right\rangle \\ &= \sum_{\alpha \geq 2, i} (\delta P_{i,0} u_{\alpha,i})^2 \left(\frac{1}{\lambda_{\alpha}(\lambda_{\alpha} + \tau_0^{-1})} + \frac{e^{-2\lambda_{\alpha}t}}{\lambda_{\alpha}(\lambda_{\alpha} - \tau_0^{-1})} - \frac{2e^{-(\lambda_{\alpha} + \tau_0^{-1})t}}{\lambda_{\alpha}^2 - \tau_0^{-2}} \right) \end{aligned}$$

Analytical prediction

For $\delta P_{i,0} \equiv \delta P_0$ and $t \rightarrow \infty$,

$$\langle \delta \theta^2 \rangle = \delta P_0^2 \sum_{\alpha \geq 2} \frac{1}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1})}.$$

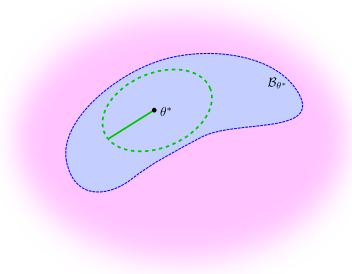


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Can we derive a condition for loss of synchrony?



Escape from the basin

Gradient dynamics:

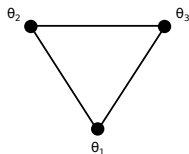
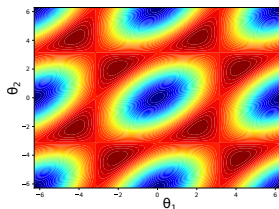
$$\dot{\boldsymbol{\theta}} = -\nabla V(\boldsymbol{\theta}), \quad V(\boldsymbol{\theta}) = -\sum_i P_i \theta_i + \sum_{i < j} a_{ij} (1 - \cos(\theta_i - \theta_j)) .$$

Escape from the basin

Gradient dynamics:

$$\dot{\theta} = -\nabla V(\theta), \quad V(\theta) = -\sum_i P_i \theta_i + \sum_{i < j} a_{ij} (1 - \cos(\theta_i - \theta_j)).$$

Almost surely, escape occurs through a 1-saddle.



Single cycle

Distance between stable sync state $\theta^{(0)}$ and the closest 1-saddle $\varphi^{(1)}$ is

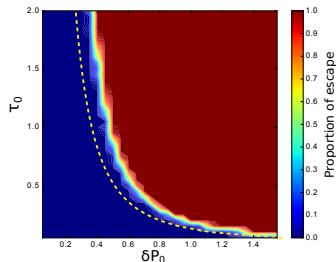
$$\|\theta^{(0)} - \varphi^{(1)}\|_2^2 = \frac{n(n^2 - 1)}{12(n - 2)^2} \pi^2 =: \Delta.$$

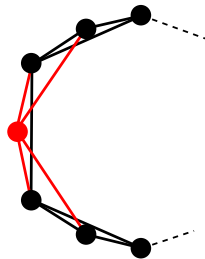
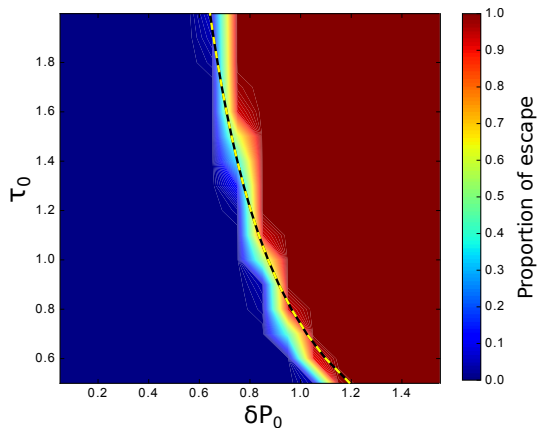
Single cycle

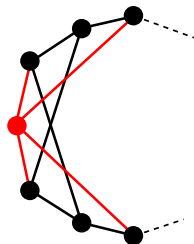
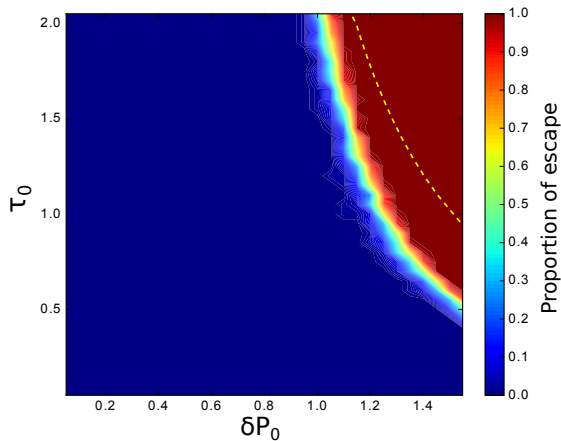
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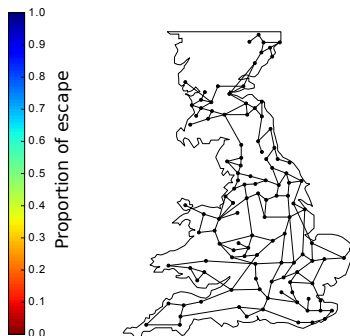
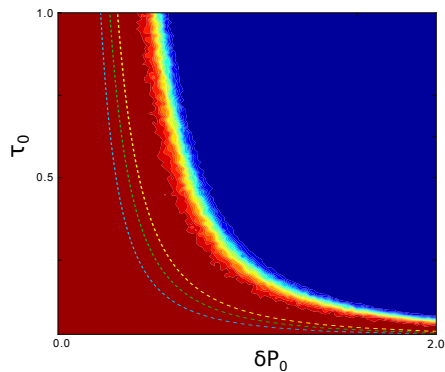
$$\delta P_0^2 \sum_{\alpha \geq 2} \frac{1}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1})} = \Delta.$$

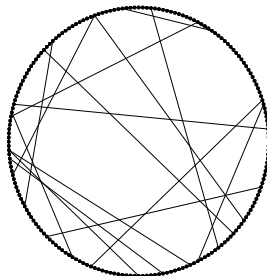
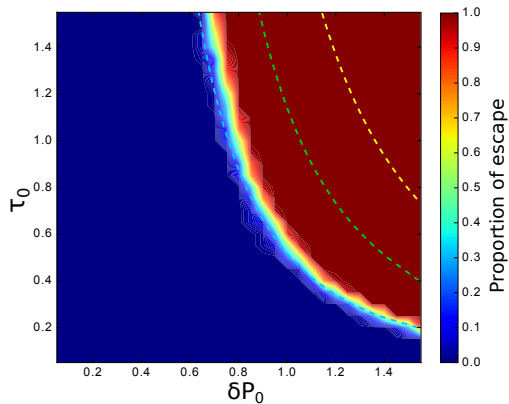


Cycle with 2nd neighbor, $n = 83$ 

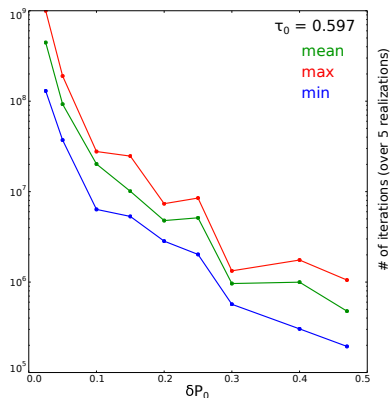
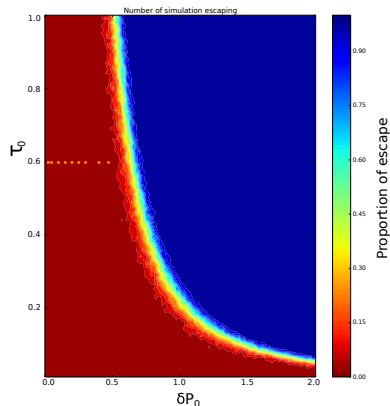
Cycle with 3rd neighbor, $n = 83$ 

Pseudo-UK, $n = 120$

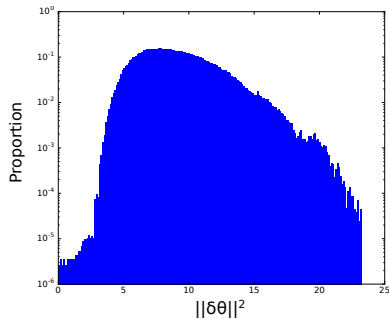
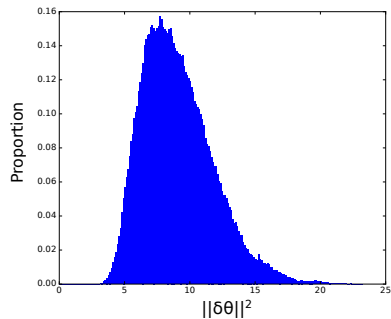


Small-world, $n = 200$ 

Caveat – UK network



Caveat – UK network



Conclusion

- ▶ Qualitatively the same for the second order model

$$\ddot{\theta}_i + \gamma \dot{\theta}_i = (P_i + \delta P_i(t)) - K \sum_j a_{ij} \sin(\theta_i - \theta_j).$$

- ▶ Where are PV and Wind on this diagram?

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- ▶ Where are PV and Wind on this diagram?

Thank you!

Histograms of 2-norms

