

Slides are available on my website.
<http://lukaseigentler.github.io>

Can we predict wavelength changes of patterned ecosystems?

Sheffield Mathematical and Statistical Modelling Seminar

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joint work with Mattia Sensi (University of Trento, Italy)

Stripe patterns

Banded vegetation patterns and intertidal mussel beds are classic examples of **self-organisation principles** in ecology.

Vegetation stripes in Ethiopia.



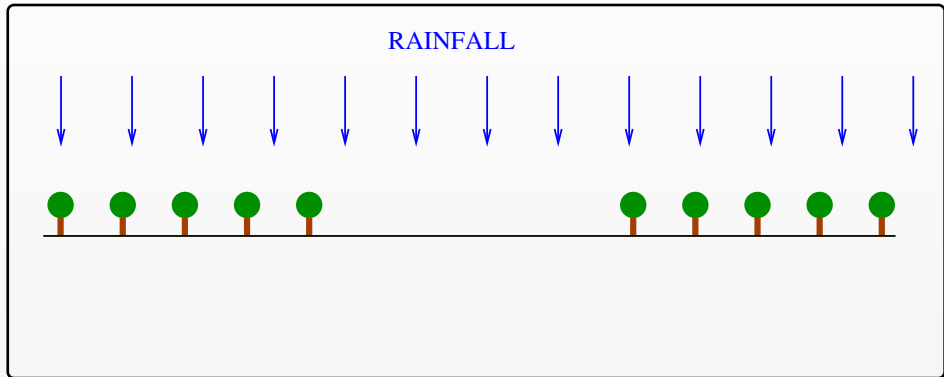
Intertidal mussel beds in the Wadden Sea.



- Parallel to topographic contours and shoreline.
- Caused by a **scale-dependent feedback loop** comprising long-range competition for a limiting resource and short-range facilitation.

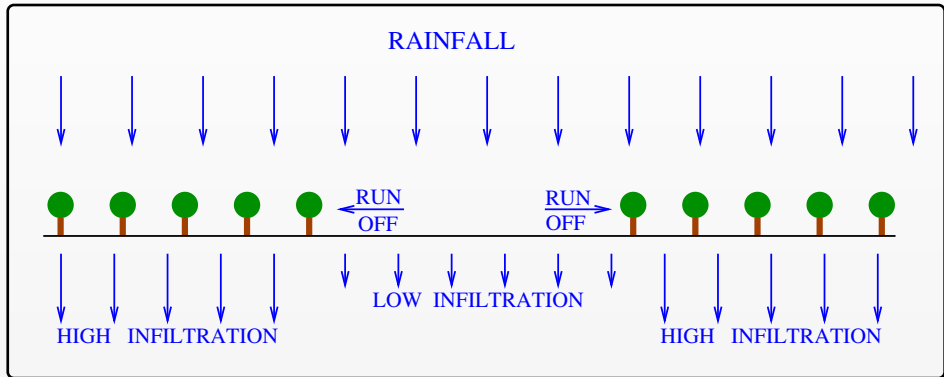
Local facilitation in vegetation patterns

Positive feedback loop: Water infiltration into the soil depends on local plant density $\Rightarrow$ redistribution of water towards existing plant patches $\Rightarrow$ further plant growth.



Local facilitation in vegetation patterns

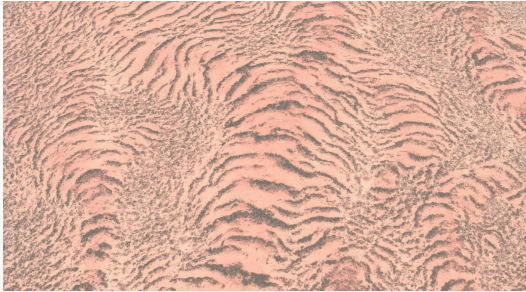
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Klausmeier model for vegetation patterns

One of the most basic phenomenological models for vegetation patterns is the **extended Klausmeier reaction-advection-diffusion model**.¹

$$\begin{aligned}\frac{\partial u}{\partial t} &= \underbrace{u^2 w}_{\text{plant growth}} - \underbrace{Bu}_{\text{plant loss}} + \underbrace{\frac{\partial^2 u}{\partial x^2}}_{\text{plant dispersal}}, \\ \frac{\partial w}{\partial t} &= \underbrace{A}_{\text{rainfall}} - \underbrace{w}_{\text{evaporation}} - \underbrace{u^2 w}_{\text{water uptake by plants}} + \underbrace{\nu \frac{\partial w}{\partial x}}_{\text{water flow downhill}} + \underbrace{d \frac{\partial^2 w}{\partial x^2}}_{\text{water diffusion}}.\end{aligned}$$

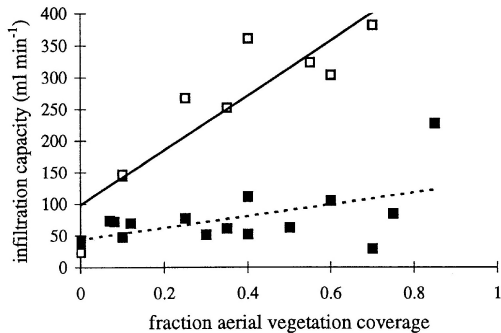
¹Klausmeier, C. A.: *Science* 284.5421 (1999).

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Water uptake



Infiltration capacity increases with plant density²

²Rietkerk, M. et al.: *Plant Ecol.* 148.2 (2000)

The nonlinearity in the water uptake and plant growth terms arises because plants increase the soil's water infiltration capacity.

⇒ Water uptake = Water density × plant density × infiltration rate.

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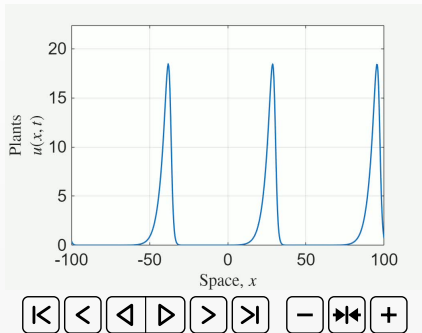
Sediment accumulation model for mussel beds

A very similar model, the **sediment accumulation model** describes pattern formation in intertidal mussel beds⁴

$$\begin{aligned}
 \frac{\partial m}{\partial t} &= \overbrace{\frac{\delta a m (s + \eta)}{s + 1}}^{\text{mussel growth}} - \underbrace{\frac{m}{m}}_{\text{mussel death}} + \overbrace{\frac{\partial^2 m}{\partial x^2}}^{\text{mussel dispersal}}, \\
 \frac{\partial s}{\partial t} &= \underbrace{\frac{m}{m}}_{\text{sediment build-up}} - \underbrace{\frac{\theta s}{\theta s}}_{\text{sediment erosion}} + \underbrace{D \frac{\partial^2 s}{\partial x^2}}_{\text{sediment dispersal}}, \\
 \frac{\partial a}{\partial t} &= \underbrace{\frac{1 - \varepsilon a}{1 - \varepsilon a}}_{\text{transport from upper water layers}} - \underbrace{\frac{\beta a m (s + \eta)}{s + 1}}_{\text{algae consumption}} + \underbrace{\nu \frac{\partial a}{\partial x}}_{\text{algae flow with tide}}.
 \end{aligned}$$

⁴Liu, Q.-X. et al.: *Proc. R. Soc. Lond. B.* 279.1739 (2012).

Periodic travelling waves

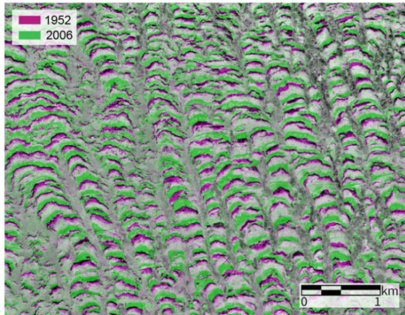


[alternative video link.](#)

- Model represents vegetation patterns as **periodic travelling waves (PTWs)**.

Uphill movement in ecology

Timeseries data.⁵



- Data shows that vegetation stripes can **move uphill** ($< 1m$ per year).

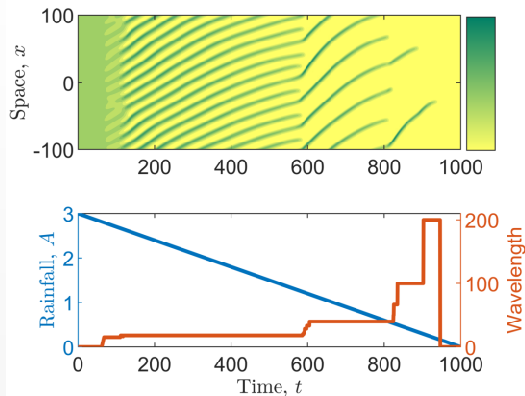
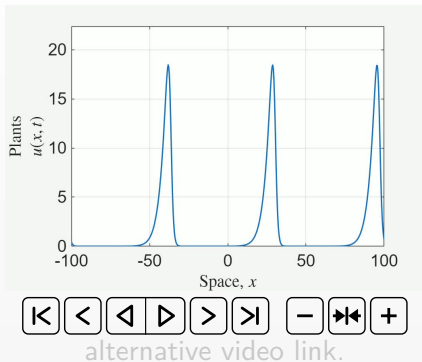
Uphill migration due to water gradient.⁶



⁵Gandhi, P. et al.: *Dryland ecohydrology*. Springer International Publishing, 2019, pp. 469–509.

⁶Dunkerley, D.: *Desert 23.2* (2018).

Periodic travelling waves

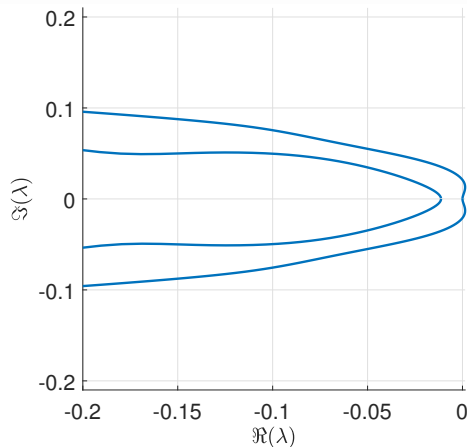


- Model represents vegetation patterns as **periodic travelling waves (PTWs)**.
- Along rainfall gradient, transition from uniform vegetation to desert occurs via several pattern transitions.

Wavelength changes

- State-of-the-art: predict wavelength changes through PTW stability properties.
- PTW linear stability is determined by their **essential spectra**.
- Calculated using numerical continuation.^a

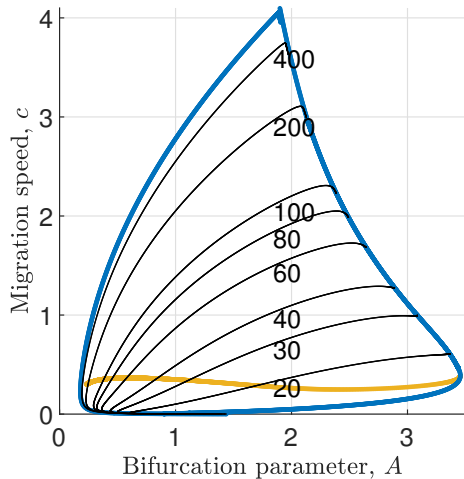
^aRademacher, J. D., Sandstede, B. and Scheel, A.: *Physica D* 229.2 (2007).



Wavelength changes

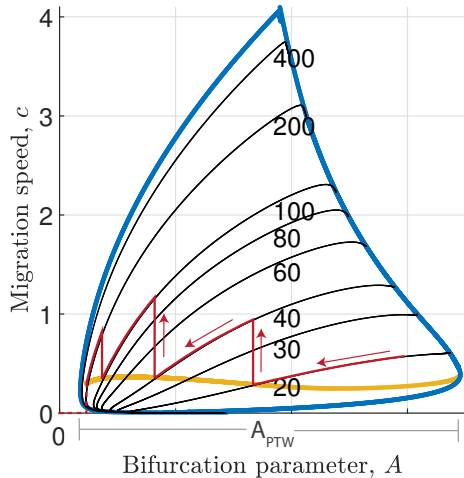
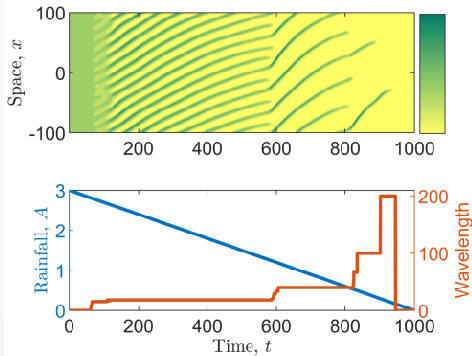
- State-of-the-art: predict wavelength changes through PTW stability properties.
- PTW linear stability is determined by their **essential spectra**.
- Calculated using numerical continuation.^a
- Wavelengths changes are typically predicted through the **Busse balloon**: parameter space of stable PTWs.

^aRademacher, J. D., Sandstede, B. and Scheel, A.: *Physica D* 229.2 (2007).

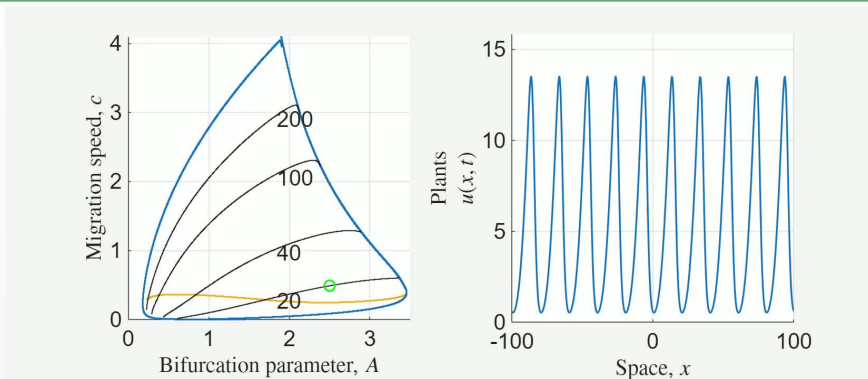


Wavelength changes

- Wavelengths are preserved, provided they remain stable.
- Upon destabilisation a wavelength change occurs.



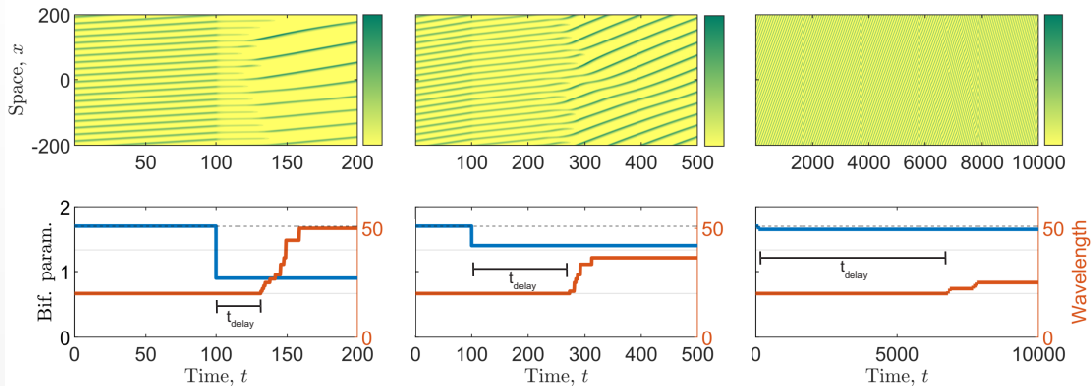
Wavelength changes



[alternative video link.](#)

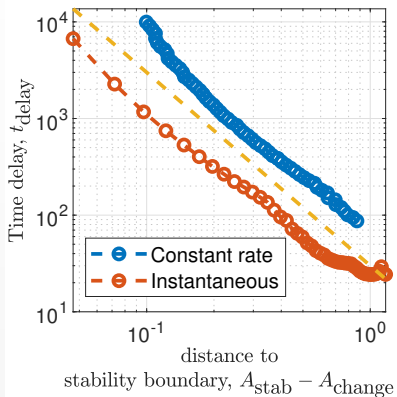
- Significant delays between crossing a stability boundary and observing wavelength changes occur.

Delays to wavelength changes



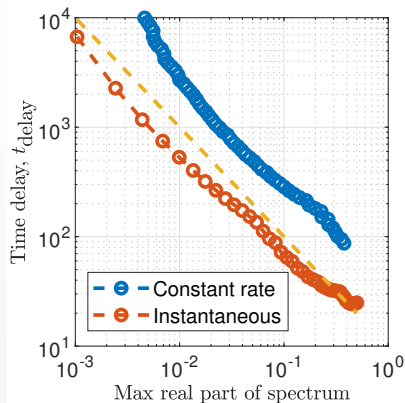
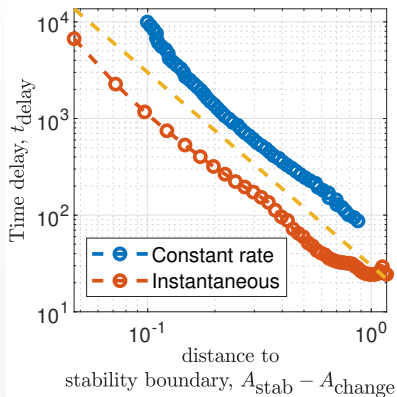
- Significant delays between crossing a stability boundary and observing wavelength changes occur.
- **Order of magnitude differences** in delay depending on parameter values.

Predicting delays



There are clear trends between delay and bifurcation parameter

Predicting delays



There are clear trends between delay and bifurcation parameter and delay and max real part of the spectrum. **no predictive power**

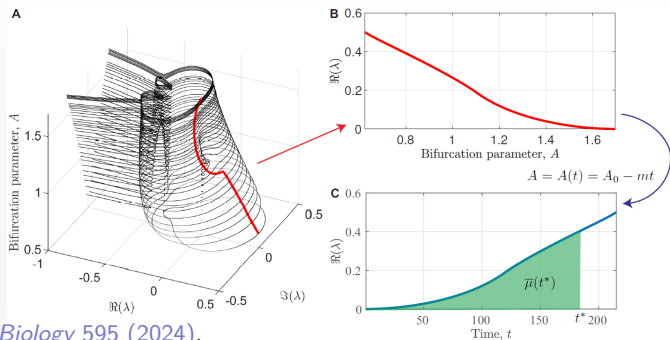
Predicting delays

Can predict the **order of magnitude of the delay** through the **accumulated maximal instability**⁷

$$\bar{\mu}(A(t)) = \int_{t_{\text{stab}}}^t \mu(\tau) d\tau, \quad t \geq t_{\text{stab}}.$$

t_{stab} is the time of the last crossing of the stability boundary.

$\mu(t)$ is the max real part of the spectrum at time t .



⁷EL and Sensi, M.: *Journal of Theoretical Biology* 595 (2024).

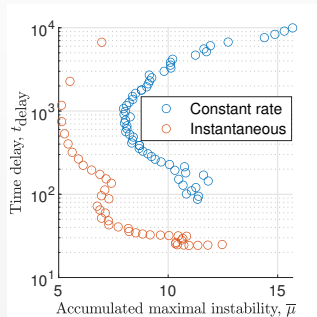
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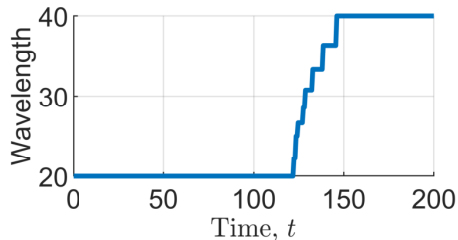
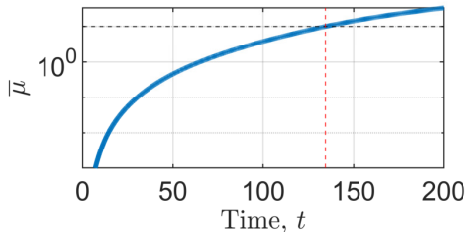
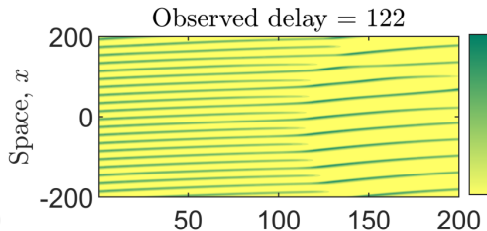
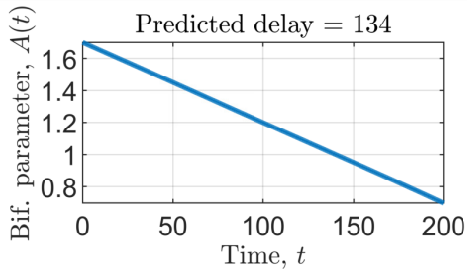
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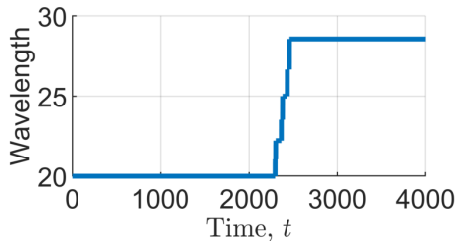
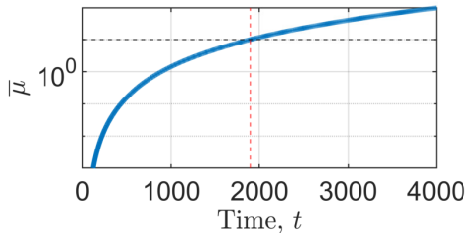
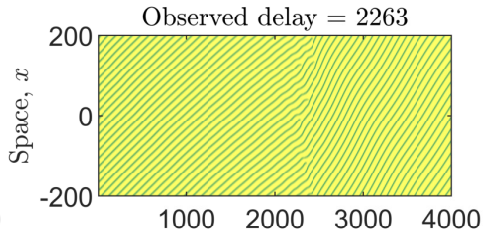
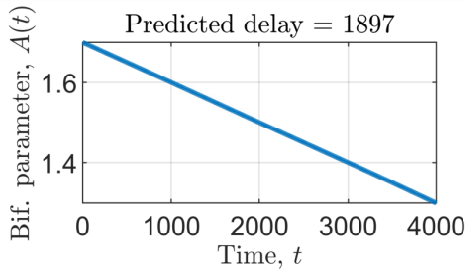
Wavelength change occurs when $\bar{\mu} \approx 10$

⁷EL and Sensi, M.: *Journal of Theoretical Biology* 595 (2024).

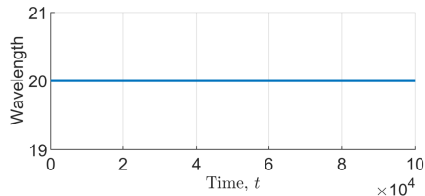
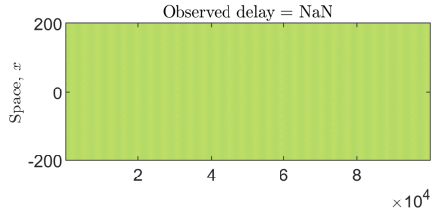
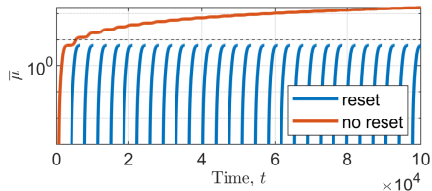
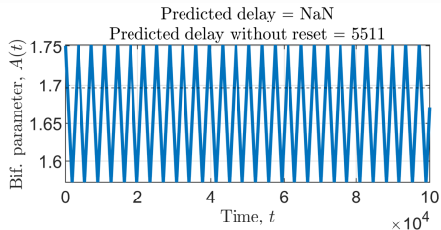
Delay prediction in practice



Delay prediction in practice

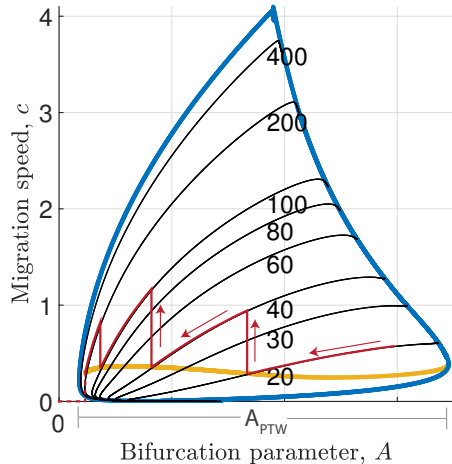


Delay prediction reset in stable regions



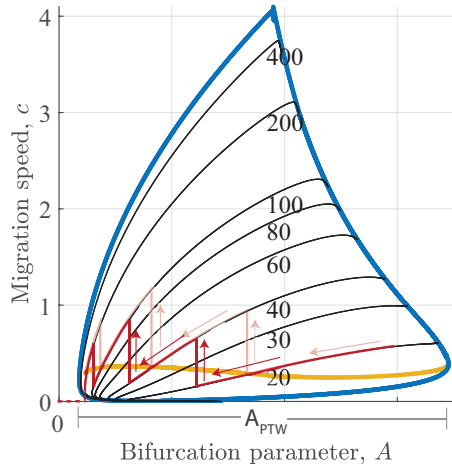
Conclusions

- Wavelength changes that occur after crossing a stability boundary are subject to a delay.



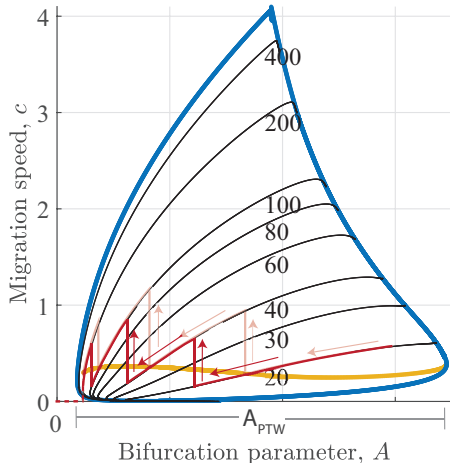
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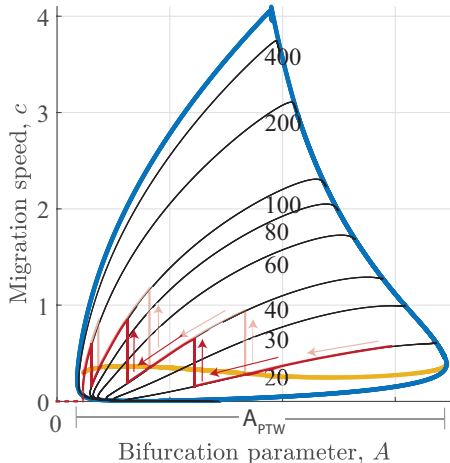
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- Order of magnitude of the delay can be predicted by tracking the maximum real part of the spectrum of the destabilised pattern over time.



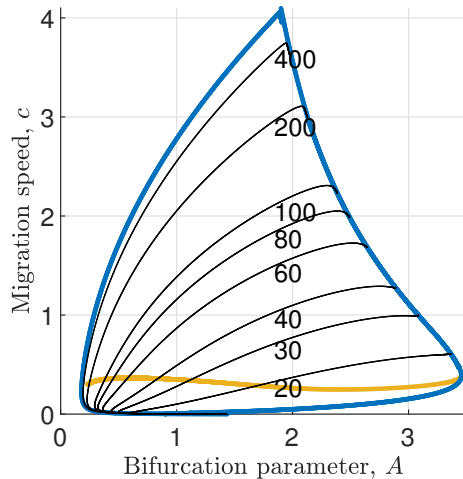
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- Order of magnitude of the delay can be predicted by tracking the maximum real part of the spectrum of the destabilised pattern over time.
- Open question: What new wavelength is chosen?



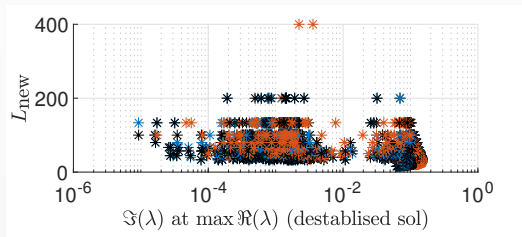
Wavelength changes

- Open question: What new wavelength is chosen?
- For fixed PDE parameters, there is multistability of different periodic travelling waves.



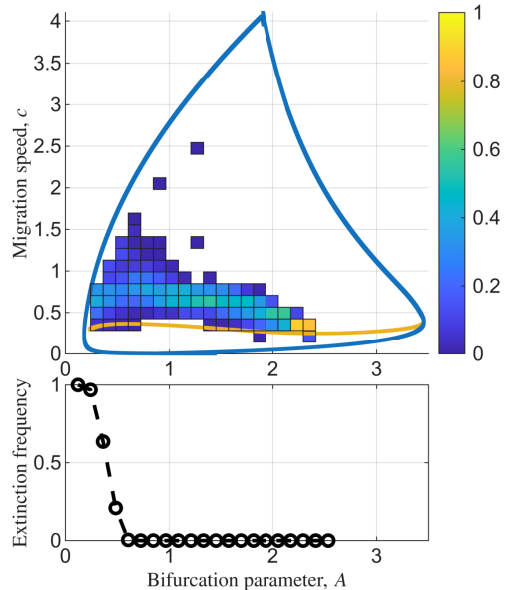
Linear analysis insufficient

- Created a large dataset of wavelength changes through simulations.
- Compared wavelength change dynamics with features of essential spectra.
- Suggests that **linear analysis is insufficient to characterise wavelength changes**.



Some selection data

- Large areas of Busse balloon remain unselected.
- Extinction does not necessarily occur at the edge of the Busse balloon.



What next?

- New methods needed to understand periodic travelling wave wavelength selection.
- Take inspiration from known results on λ - ω systems?
- Similar trends observed for mussel model $\Rightarrow$ possible to derive principles applicable to a wider class of models?
- Need evidence of wavelength changes in empirical systems.

References

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- [1] [Eigentler, L. and Sensi, M.:](#) 'Delayed loss of stability of periodic travelling waves: insights from the analysis of essential spectra'. *Journal of Theoretical Biology* 595 (2024), p. 111945.