



Heriot-Watt University &
The University of Edinburgh



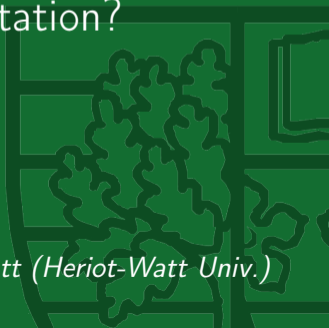
How Does Long-Range Dispersal Affect Pattern Formation in Semi-Arid Vegetation?

Equadiff 2019

11 July 2019

Lukas Eigentler

joint work with Jamie Bennett (Ben Gurion Univ.), Jonathan Sherratt (Heriot-Watt Univ.)



Vegetation Patterns

Vegetation patterns are a classic example of a **self-organisation principle** in ecology.



Bushes in Niger.



More water reaches the top edge of a stripe.¹

- Plants increase water infiltration into the soil and thus induce a **positive feedback loop**.
- On sloped ground, stripes grow **parallel to the contours**.
- Stripes either **move uphill** or are **stationary**.

¹Dunkerley, D.: *Desert* 23.2 (2018).

Klausmeier Model

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill

One of the most basic phenomenological models is the Klausmeier reaction-advection-diffusion model.²

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

²Klausmeier, C. A.: *Science* 284.5421 (1999).

Klausmeier Model

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill

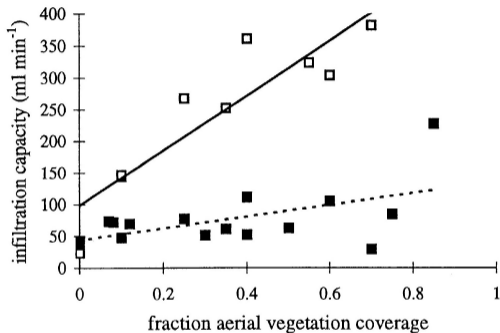
One of the most basic phenomenological models is the Klausmeier reaction-advection-diffusion model.

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

Water Uptake

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill



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.

Infiltration capacity increases with plant density³

³Rietkerk, M. et al.: *Plant Ecol.* 148.2 (2000)

Local Model

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill

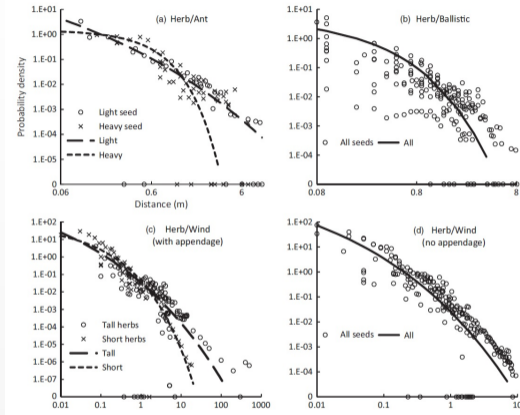
The Klausmeier model models plant dispersal by a diffusion term, i.e. a local process.

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

Nonlocal Seed Dispersal

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill



More realistic: **Include effects of nonlocal processes**, such as dispersal by wind or large mammals.

Data of long range seed dispersal ⁴

⁴Bullock, J. M. et al.: *J. Ecol.* 105.1 (2017)

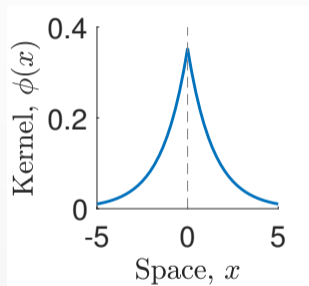
Nonlocal Model

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 C - dispersal rate

Diffusion is replaced by a **convolution** of the **plant density** u with a **dispersal kernel** ϕ . The scale parameter a controls the width of the kernel.

$$\frac{\partial u}{\partial t} = \underbrace{u^2 w}_{\text{plant growth}} - \underbrace{Bu}_{\text{plant loss}} + \underbrace{C(\phi(\cdot; a) * u(\cdot, t) - u)}_{\text{nonlocal plant dispersal}},$$

$$\frac{\partial w}{\partial t} = \underbrace{A}_{\text{rainfall}} - \underbrace{w}_{\text{evaporation}} - \underbrace{u^2 w}_{\text{water uptake by plants}} + \underbrace{\nu \frac{\partial w}{\partial x}}_{\text{water flow downhill}} + \underbrace{d \frac{\partial^2 w}{\partial x^2}}_{\text{water diffusion}}.$$



Laplacian Kernel

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill, $1/a$ - kernel width

C - dispersal rate

If ϕ decays exponentially as $|x| \rightarrow \infty$, and $C = 2/\sigma(a)^2$, then the nonlocal model tends to the local model as $\sigma(a) \rightarrow 0$.

E.g. Laplace kernel

$$\phi(x) = \frac{a}{2} e^{-a|x|}, \quad a > 0, \quad x \in \mathbb{R}.$$

Useful because

$$\hat{\phi}(k) = \frac{a^2}{a^2 + k^2}, \quad k \in \mathbb{R}.$$

and **allows transformation into a local model**. If $v(x, t) = \phi(\cdot; a) * u(\cdot; t)$, then

$$\frac{\partial^2 v}{\partial x^2}(x, t) = a^2(v(x, t) - u(x, t))$$

Spatially Constant Equilibria

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill, $1/a$ - kernel width

C - dispersal rate

Desert steady state:

$$(\bar{U}_d, \bar{W}_d) = (0, A) \text{ stable.}$$

If $A \geq 2B$, there are two additional spatially uniform equilibria:

$$(\bar{U}_-, \bar{W}_-) = \left(\frac{2B}{A - \sqrt{A^2 - 4B^2}}, \frac{A - \sqrt{A^2 - 4B^2}}{2} \right) \text{ stable if } B < 2,$$

$$(\bar{U}_+, \bar{W}_+) = \left(\frac{2B}{A + \sqrt{A^2 - 4B^2}}, \frac{A + \sqrt{A^2 - 4B^2}}{2} \right) \text{ unstable.}$$

Travelling Waves

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate

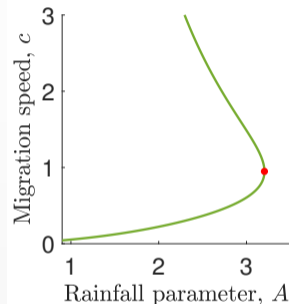
- On sloped ground (some) patterns slowly move uphill.
- Travelling wave ansatz $u(x, t) = U(z)$, $w(x, t) = W(z)$, $z = x - ct$ gives the corresponding first-order travelling waves integro-ODE system

$$\begin{aligned}\frac{dU}{dz} &= -\frac{1}{c} (U^2W - BU + C (\phi(\cdot; a) * U(\cdot) - U(z))), \\ \frac{dW}{dz} &= W_1, \\ \frac{dW_1}{dz} &= -\frac{1}{d} (A - W - U^2W + (c + \nu)W_1).\end{aligned}$$

Travelling Waves

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate

- Patterns correspond to **limit cycles** of the travelling wave integro-ODEs.
- Local model: **Patterns emanate from a Hopf bifurcation**. In the A - c plane, the parameter region supporting patterns is bounded above by a **Hopf bifurcation**.⁵



Location of the Hopf bifurcation in A - c plane.

⁵Sherratt, J. A. and Lord, G. J.: *Theor. Popul. Biol.* 71.1 (2007)

Onset of Patterns

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill, $1/a$ - kernel width

c - migration speed, C - dispersal rate

The growth rate $\lambda \in \mathbb{C}$ of perturbations $\tilde{U}(z)$, $\tilde{W}(z)$, $\tilde{W}_1(z)$ to $(\bar{U}_-, \bar{W}_-, \bar{W}_{1-})$, satisfies (after linearisation)

$$\lambda^5 + \alpha\lambda^4 + \beta\lambda^3 + \gamma\lambda^2 + \delta\lambda + \varepsilon = 0,$$

A **Hopf bifurcation** requires $\lambda = i\omega$, $\omega \in \mathbb{R}$. This yields

$$\alpha\omega^4 - \gamma\omega^2 + \varepsilon = 0,$$

$$\omega^5 - \beta\omega^3 + \delta\omega = 0.$$

Solving for, and eliminating ω^2 gives

$$\frac{\gamma \pm \sqrt{\gamma^2 - 4\alpha\varepsilon}}{2\alpha} = \frac{\beta \pm \sqrt{\beta^2 - 4\delta}}{2}.$$

Onset of Patterns

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill, $1/a$ - kernel width

c - migration speed, C - dispersal rate

$$\alpha = \frac{d(B - C) + c(c + \nu)}{cd},$$

$$\beta = \frac{-2B^2 (a^2 cd - (B - C)(c + \nu)) - Ac (A + \sqrt{A^2 - 4B^2})}{2B^2 cd},$$

$$\gamma = \frac{-2B^2 a^2 (d + c(c + \nu)) + A(B + C) (A + \sqrt{A^2 - 4B^2}) - 4B^3}{2B^2 cd},$$

$$\delta = \frac{a^2 (-2B^3(c + \nu) + Ac (A + \sqrt{A^2 - 4B^2}))}{2B^2 cd},$$

$$\varepsilon = \frac{a^2 (-A (A + \sqrt{A^2 - 4B^2}) + 4B^2)}{2B^2 cd}.$$

Onset of Patterns

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate

Using that $\nu \gg 1$,

$$A_{\max} = \left(\frac{3C - B - 2\sqrt{2C}\sqrt{C - B}}{(B + C)^2} \right)^{\frac{1}{4}} a^{\frac{1}{2}} B^{\frac{5}{4}} \nu^{\frac{1}{2}},$$

to leading order in ν as $\nu \rightarrow \infty$.

- Note that $A_{\max} = O(\sqrt{\nu})$.
- Decrease in a (i.e. increase in kernel width) causes decrease of A_{\max} .
- Increase in dispersal rate C causes decrease of A_{\max} .

Onset of Patterns

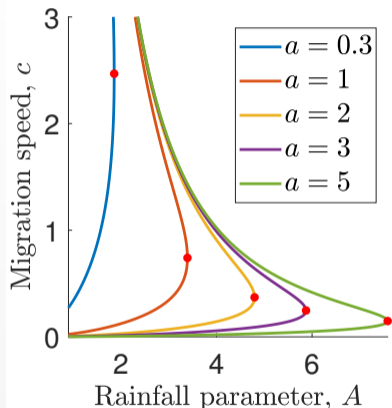
A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate

Using that $\nu \gg 1$,

$$A_{\max} = \left(\frac{3C - B - 2\sqrt{2C}\sqrt{C - B}}{(B + C)^2} \right)^{\frac{1}{4}} a^{\frac{1}{2}} B^{\frac{5}{4}} \nu^{\frac{1}{2}},$$

to leading order in ν as $\nu \rightarrow \infty$.

- Note that $A_{\max} = O(\sqrt{\nu})$.
- Decrease in a (i.e. increase in kernel width) causes decrease of A_{\max} .
- Increase in dispersal rate C causes decrease of A_{\max} .



Locus of Hopf bifurcation for fixed C and varying a .

Other Kernel Functions

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate

Gaussian kernel:

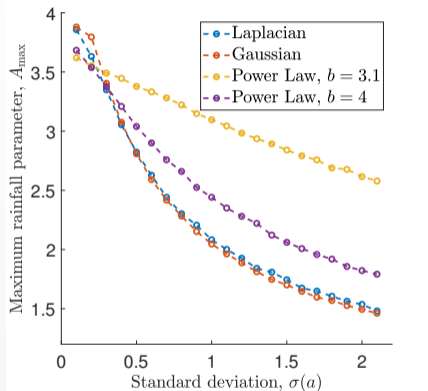
$$\phi(x) = \frac{a_g}{\sqrt{\pi}} e^{-a_g^2 x^2}, \quad x \in \mathbb{R}, a_g > 0.$$

Power law distribution:

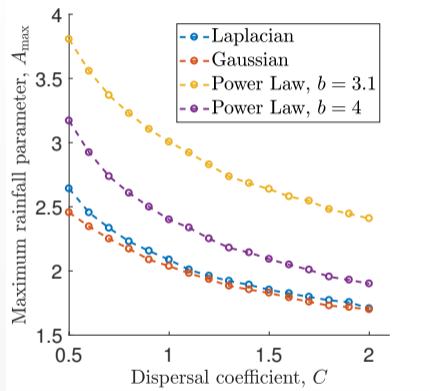
$$\phi(x) = \frac{(b-1)a_p}{2(1+a_p|x|)^b}, \quad x \in \mathbb{R}, a_p > 0, b > 3.$$

Numerical Simulations

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate



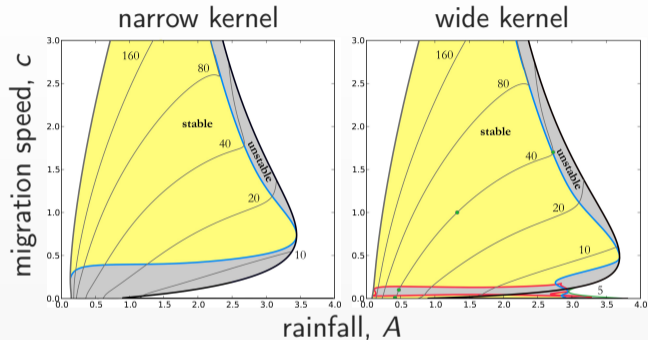
Maximum rainfall parameter under changes to kernel width a .



Maximum rainfall parameter under changes to the dispersal rate C .

Stability of Patterns

A - rainfall, B - plant loss, d - w. diffusion
 ν - w. flow downhill, $1/a$ - kernel width
 c - migration speed, C - dispersal rate



Stability of patterns in the A - c plane ⁶.

Stability of patterns is calculated using the numerical continuation method by Rademacher et al. ⁷

For wide kernels, the stability boundary towards the desert state changes from Eckhaus to Hopf-type. This yields

- **increased resilience** due to oscillating vegetation densities in peaks,

⁶Bennett, J. J. R. and Sherratt, J. A.: *J. Theor. Biol.* (2018)

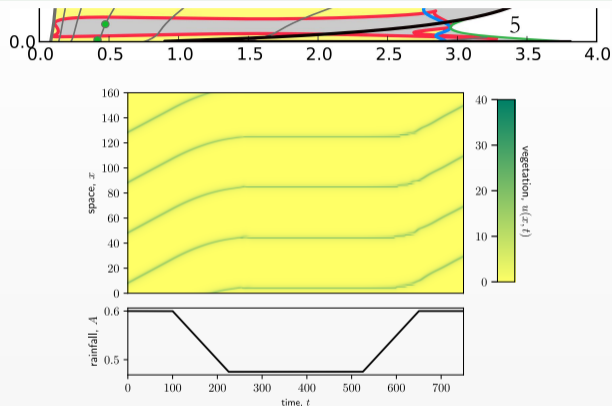
⁷Rademacher, J. D., Sandstede, B. and Scheel, A.: *Physica D* 229.2 (2007)

Stability of Patterns

A - rainfall, B - plant loss, d - w. diffusion

ν - w. flow downhill, $1/a$ - kernel width

c - migration speed, C - dispersal rate



Stability of patterns is calculated using the numerical continuation method by Rademacher et al. ⁹

For wide kernels, the stability boundary towards the desert state changes from Eckhaus (sideband) to Hopf-type. This yields

- **increased resilience** due to oscillating vegetation densities in peaks,
- existence of **stable patterns with small migration speed** ($c \ll 1$).

Existence of stable (almost) stationary patterns ⁸.

⁸Bennett, J. J. R. and Sherratt, J. A.: *J. Theor. Biol.* (2018)

⁹Rademacher, J. D., Sandstede, B. and Scheel, A.: *Physica D* 229.2 (2007)

Conclusions

- Wider kernels and higher dispersal rates inhibit pattern onset.
- But plants develop a narrow dispersal kernel \Rightarrow possible trade-off?
- Mathematically motivated form of trade-off: $C = 2/\sigma(a)^2$. Model tends to the local reaction-advection-diffusion system as $\sigma(a) \rightarrow 0$.
- Tendency to form patterns depends on the kind of decay of the dispersal kernel.
- Long-range seed dispersal increases the resilience of a pattern and stabilises (almost) stationary patterns.

References

Slides are available on my website.

<http://www.macs.hw.ac.uk/~le8/>



Bennett, J. J. R. and Sherratt, J. A.: 'Long-distance seed dispersal affects the resilience of banded vegetation patterns in semi-deserts'. *J. Theor. Biol.* (2018), in press.



Eigentler, L. and Sherratt, J. A.: 'Analysis of a model for banded vegetation patterns in semi-arid environments with nonlocal dispersal'. *J. Math. Biol.* 77.3 (2018), pp. 739–763.