





Bridging the gap between structurally realistic models and viability theory in savanna ecosystems

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PATRES: Savanna case study

Overview:

- 1) Start with a detailed, site-specific savanna model
- 2) Identify essential pattern dynamics of the model
- 3) Build a low dimensional, mathematically tractable approx. capturing key pattern dynamics
- 4) Apply viability theory to approximate model



Savannas



Defined by:

- 1) Continuous grass layer
- 2) Discontinuous tree layer

Cover ~20% of Earth's terrestrial surface area

Harbor considerable biodiversity

Economically significant as grazing lands



Current threats to savannas

Human population growth

Global climate change

Changing land use patterns

Overexploitation by humans -Specifically overgrazing





Savanna management

Must balance system integrity against economic goals

Price of failure is high. Bush encroached savannas effectively useless for grazing

Manage by manipulating fire and/or grazing regimes





"Structurally realistic" savanna models

Relatively complex models used to integrate knowledge on particular savannas

Typically stochastic. Include many parameters and variables

Often reproduce key features of focal site

Seldom used for management purposes



The Jeltsch model

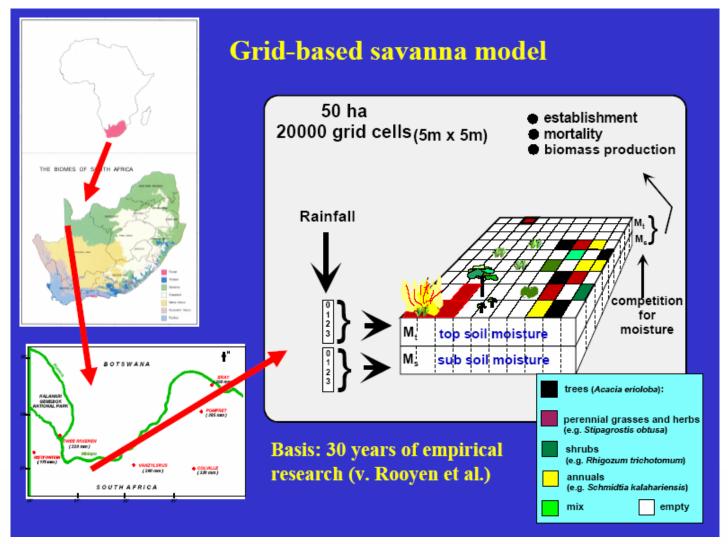
Developed by Jeltsch and colleagues in 1990's

Focuses on a South African savanna

Integrates 30+ years of detailed empirical data from this site

Widely cited and still widely used







Complex model, but complex dynamics?

Dynamics of macroscopic state variables might not be complicated

This appears to be the case for the Jeltsch model

Idea: Use the relatively simple dynamics of a few key patterns to reduce model dimensionality



Targets of analysis

Focus on trees (and grasses):

1) Population density time series
$$\left(\rho_1 = \frac{\#trees}{\#sites}\right)$$

2) Distance-dependent spatial pattern (g statistic)



Logistic-like population growth

$$\frac{dN}{dt} = rN\left(1 - \frac{N}{K}\right)$$

$$= \frac{0.12}{0.08}$$

$$= \frac{0.08}{0.06}$$

$$= \frac{0.08}{0.04}$$

$$= \frac{0.08}{0.04}$$

$$= \frac{0.08}{0.04}$$

$$= \frac{0.08}{0.04}$$

$$= \frac{0.08}{0.06}$$

$$= \frac{0.08}{0.09}$$

$$= \frac{0.09}{0.09}$$

0.14



Point-pattern analysis

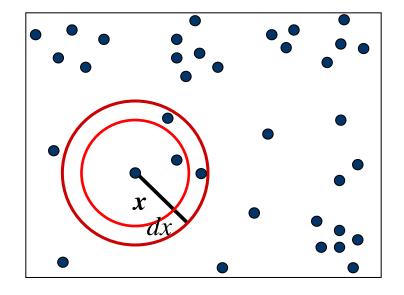
Normalized pair correlation statistic Stoyan & Stoyan 1994; Condit et al. 2000

$$g_{x} = \frac{\sum N_{x}}{\lambda \sum A_{x}}$$

$$g=1 \longrightarrow \text{Random}$$

$$g>1 \longrightarrow \text{Clustered}$$

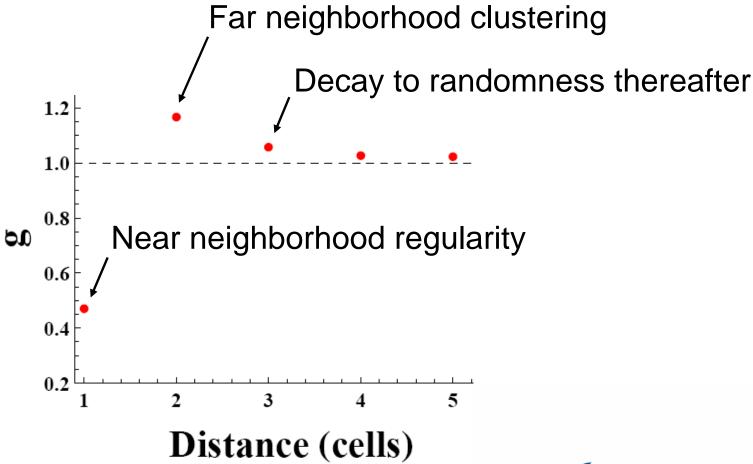
$$g<1 \longrightarrow \text{Regular}$$



Where λ is the density of the point pattern, and N_x A_x are the # of neighbors and area of the annulus, respectively, at dist. x.



Strong signatures of competition and short dispersal





Key patterns

Logistic-like population growth

Spatial pattern:

- 1) Near neighborhood regularity
- 2) Far neigh. clustering
- 3) Decays to randomness thereafter

Suggests a **spatial logistic model** with 2 interaction scales

Model overview

Square lattice, periodic boundary conds., two states: tree (1) or grass (0)

Prop. of sites in state 1 is ρ_1 and in state 0 is $\rho_0 = 1 - \rho_1$

Model is an extension of the contact process

Tree dispersal & competition occur within defined (Moore) neighborhoods



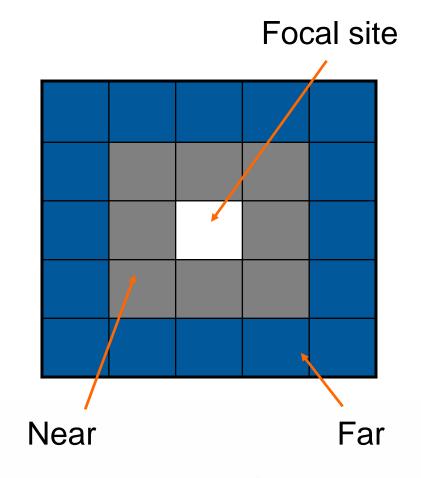
Interaction neighborhoods

Define two neighborhoods—"near" and "far"

Assume:

- establishment comp.
 occurs only over near neigh.
- birth occurs in both near & far neighborhoods
- 3) Neighborhood status is symmetric

Disp. scale > Comp. scale





Birth

Trees reproduce at constant rate *b*

Each site within birth (n + f) neighborhood of a tree receives offspring at rate:

 $\beta = \frac{b}{z_n + z_f}$

If an seed lands on a tree occupied site, nothing happens

If it lands on a grass occupied site, it has a chance to establish

Establishment

Given birth, the seed establishes with probability

$$P_E = P_C^{Surv} P_F^{Surv}$$

where:

$$P_C^{Surv} = e^{-\delta C}$$
 and $P_F^{Surv} = 1$ (for now)



Moment equations

An open hierarchy of equations of the form

$\frac{d\rho_1}{dt} = f(\rho_1, \rho_{11})$	Mean Density
$\frac{d\rho_{11}}{dt} = f(\rho_1, \rho_{11}, \rho_{111})$	Most Spatial Structure
$\frac{ai}{do}$	Spatial Structure
$\frac{d\rho_{111}}{dt} = f(\rho_1, \rho_{11}, \rho_{111}, \rho_{1111})$:	Increasingly Minute Spatial Detail

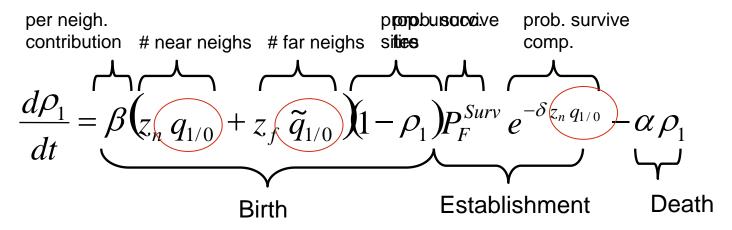
By itself, this is useless, but mean-field and pair approximations make this approach tractable

Multi-scale pair approximation, Ellner 2001



Moment approximations: Mean dynamics

Eqn. for mean is a simple balance equation



Local density terms carry spatial information



Moment approximations: Density

 $q_{1/0}$ and $\tilde{q}_{1/0}$ are the near & far neighborhood local densities

Pair:

$$\frac{d\rho_1}{dt} = \beta \left(z_n \ q_{1/0} + z_f \ \widetilde{q}_{1/0} \right) \left(1 - \rho_1\right) P_F^{Surv} \ e^{-\delta z_n q_{1/0}} - \alpha \rho_1$$
 Mean field:

$$\frac{d\rho_{1}}{dt} = \beta \left(z_{n} \rho_{1} + z_{f} \rho_{1} \right) (1 - \rho_{1}) P_{F}^{Surv} e^{-\delta z_{n} \rho_{1}} - \alpha \rho_{1}$$

Multiscale pair approximation: Ellner 2001



Pair Approx.: Pair correlations

Local dens. can be written in terms of pair and singlet probs.

$$q_{1/0} = \frac{\rho_1 - \rho_{11}}{1 - \rho_1}$$
 & $\tilde{q}_{1/0} = \frac{\rho_1 - \tilde{\rho}_{11}}{1 - \rho_1}$

Each int. neighborhood requires a pair dens. eqn.

Near:

$$\frac{1}{2}\frac{d\rho_{11}}{dt} = \beta \left(1 + (z_n - 1)q_{1/0} + z_f \ \widetilde{q}_{1/0}\right) \left(\rho_1 - \rho_{11}\right) P_F^{Surv} e^{-\delta \left[1 + (z_n - 1)q_{1/0}\right]} - \alpha \rho_{11}$$

Far:

$$\frac{1}{2} \frac{d\tilde{\rho}_{11}}{dt} = \beta \left(z_n \, q_{1/0} + 1 + (z_f - 1) \tilde{q}_{1/0} \right) \left(\rho_1 - \tilde{\rho}_{11} \right) P_F^{Surv} e^{-\delta z_n \, q_{1/0}} - \alpha \, \tilde{\rho}_{11}$$

Full analysis of PA savanna model in Calabrese et al. (In press, American Naturalist)

Back to the Jeltsch model

We have an equation for ρ_1

g statistics can be derived from PA:
$$g_n = \frac{\rho_{11}}{\rho_1^2}$$
 $g_f = \frac{\rho_{11}}{\rho_1^2}$

PA has two free parameters: b and δ (plus α)

They can be estimated by fitting the PA to the Jeltsch model



Calculating the death rate

Only "background" mortality important for adult trees

Jeltsch et al. 1996 describe the (background) mortality process:

- 1) Established adults survive to age 120
- 2) Linear increase in the mortality rate from 120 to a max. age of 250.

Can use this info to calculate the mean lifespan \overline{a} , and then calculate the death rate in the PA as:

$$\alpha = 1/\overline{a}$$



Average life span

$$m_1 = d_1$$
 $m_2 = (1 - d_1)d_2$ $m_3 = (1 - d_1)(1 - d_2)d_3$

$$d_1 = \frac{1}{k} \qquad d_2 = \frac{2}{k} \qquad d_n = \frac{n}{k} \quad \text{for } n \le k$$

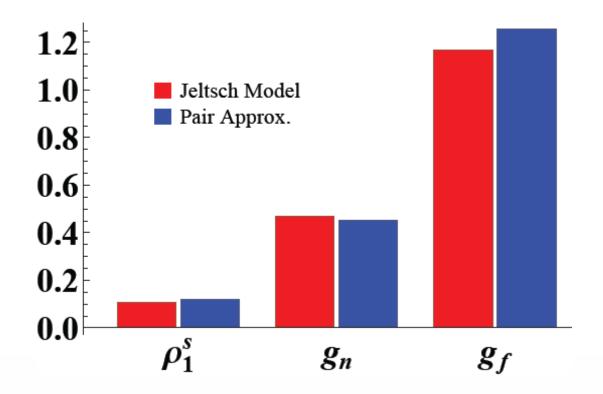
$$m_n = \frac{n}{k} \prod_{i=1}^{n-1} 1 - \frac{i}{k}$$
 $\overline{a} = \sum_{i=1}^{k} m_i i$

$$k = 130$$
 $\overline{a} = 120 + \sum_{i=1}^{130} m_i i$ $\overline{a} = 133.9656$

$$\alpha = 1/133.9656 = 0.00746$$



Fit PA to 3 patterns

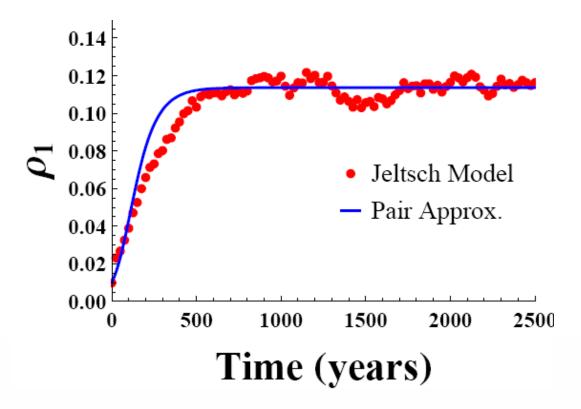


$$b = 4.4$$

$$\delta$$
=1.3



Transient dynamics





Adding fire

Negatively affects trees by killing primarily juveniles

Fire regimes can be manipulated to control trees

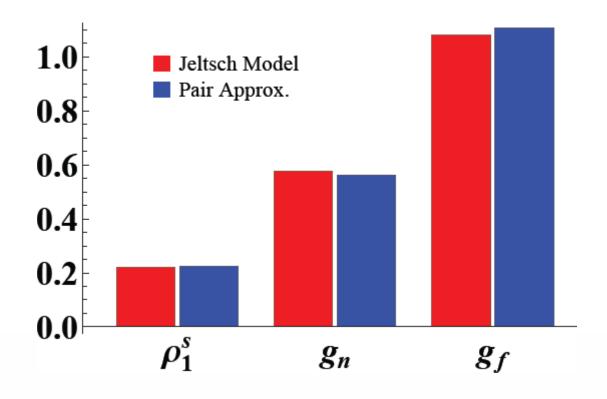
Direct: Prescribed burns

Indirect: Varying grazing pressure

Fire and grazing represent key control actions



First fit competition model w/o fire

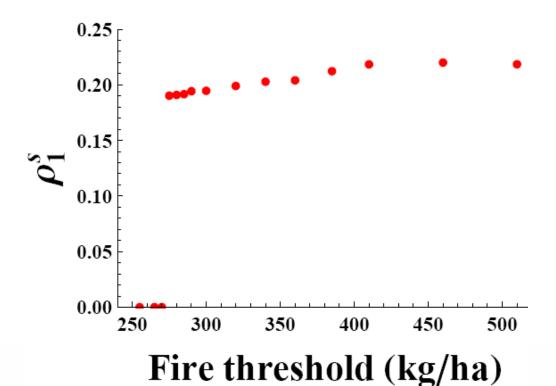


$$b = 11.7$$
 $\delta = 1.1$

$$\delta$$
=1.1



Fire in the Jeltsch model



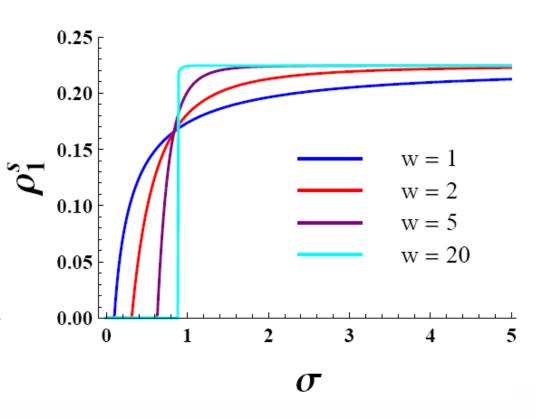


Fire in pair approx.

$$P_F^{Surv} = 1 - \frac{(c(1-\rho_1))^w}{(c\sigma)^w + (c(1-\rho_1))^w}$$

 σ affects fire freq.

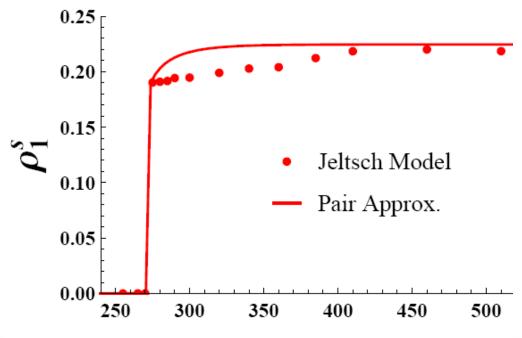
w determines abruptness and location of trans.



c rescales x-axis



Comparing the transitions



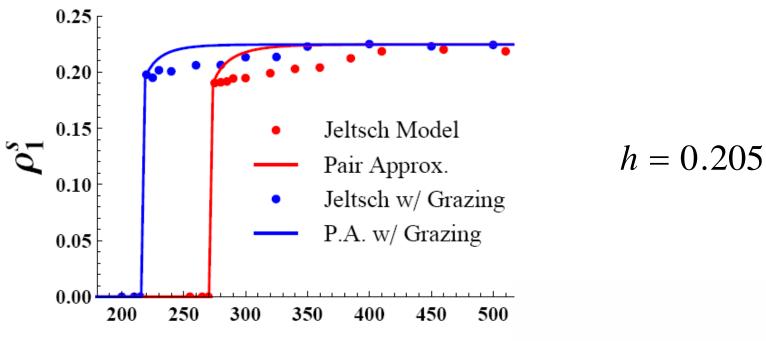
Fire threshold (kg/ha)

$$c = 322$$
 $w = 14$



Grazing

$$P_F^{Surv} = 1 - \frac{((1-h)c(1-\rho_1))^w}{(c\sigma)^w + ((1-h)c(1-\rho_1))^w}$$



Fire threshold (kg/ha)



Summary of Approximation

$$\frac{d\rho_1}{dt} = \frac{b}{3} \left(q_{1/0} + 2 \tilde{q}_{1/0} \right) \left(1 - \rho_1 \right) P_F^{Surv} e^{-8\delta q_{1/0}} - \alpha \rho_1$$

$$\frac{1}{2}\frac{d\rho_{11}}{dt} = \frac{b}{24}\left(1 + 7q_{1/0} + 16\tilde{q}_{1/0}\right)\left(\rho_{1} - \rho_{11}\right)P_{F}^{Surv}e^{-\delta\left[1 + 7q_{1/0}\right]} - \alpha\rho_{11}$$

$$\frac{1}{2}\frac{d\tilde{\rho}_{11}}{dt} = \frac{b}{24} \left(8q_{1/0} + 1 + 15\tilde{q}_{1/0}\right) \left(\rho_1 - \tilde{\rho}_{11}\right) P_F^{Surv} e^{-8\delta q_{1/0}} - \alpha \tilde{\rho}_{11}$$

$$q_{1/0} = \frac{\rho_1 - \rho_{11}}{1 - \rho_1}$$

$$\tilde{q}_{1/0} = \frac{\rho_1 - \tilde{\rho}_{11}}{1 - \rho_1}$$

$$q_{1/0} = \frac{\rho_1 - \rho_{11}}{1 - \rho_1} \qquad \tilde{q}_{1/0} = \frac{\rho_1 - \tilde{\rho}_{11}}{1 - \rho_1} \qquad P_F^{Surv} = 1 - \frac{\left((1 - h)c(1 - \rho_1)\right)^w}{\left(c\sigma\right)^w + \left((1 - h)c(1 - \rho_1)\right)^w}$$

$$b = 11.7$$

$$\alpha = 0.00746$$

 σ = user determined

$$\delta = 1.1$$

$$c = 322$$

g = control parameter

$$w = 14$$



A viability problem based on MF approx.

$$\frac{d\rho_{1}}{dt} = be^{-\delta z_{n}\rho_{1}} \frac{\sigma}{\sigma + 1 - \rho_{1}} (\rho_{1} - \rho_{1}^{2}) - \alpha \rho_{1} = \phi(\rho_{1}, h)$$

where $\hat{\sigma} = \sigma/(1-h)$

Redefine system to include dynamics of control:

$$\rho_1(t+dt) = \rho_1(t) + \phi(\rho_1(t), h(t))dt$$
$$h(t+dt) = h(t) + u(t)dt$$

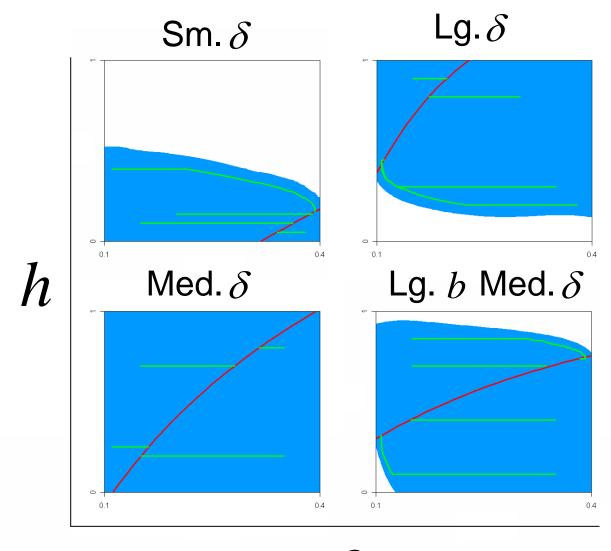
Subject to constraints:

$$\rho_{1Min} \le \rho_1(t) \le \rho_{1Max}$$

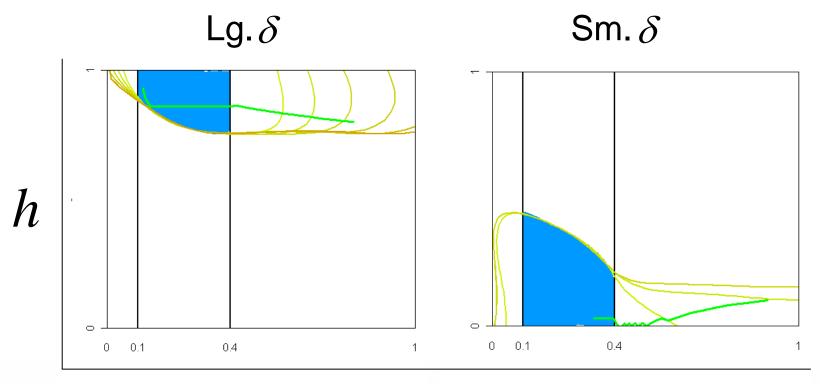
$$|u(t)| \leq h'_{Max}$$



Viability kernels



Quantifying resilience







Conclusions

Path from Jeltsch mod. to viab. analysis is long & winding

Formal approximation frameworks (e.g., moment equations) cannot be applied directly to models of arbitrary complexity

Main value of this exercise is as a proof of concept

Next step is to apply a control policy identified using the approx. mod. to the full Jeltsch mod.



Thanks!





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