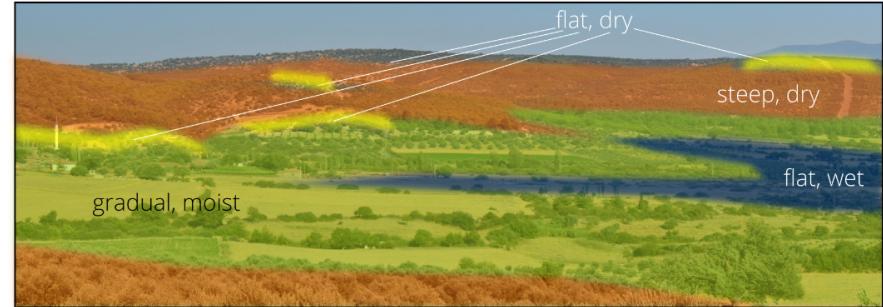
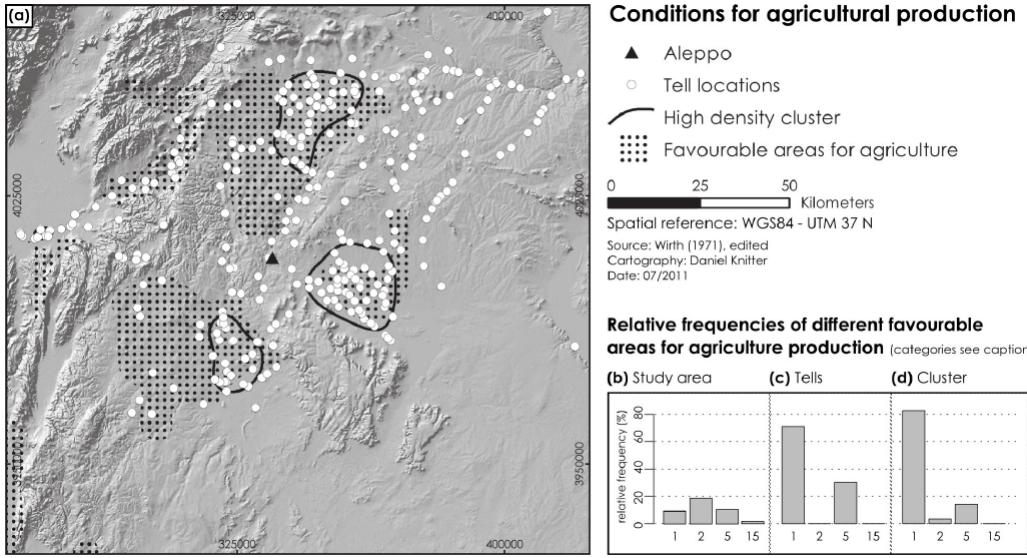
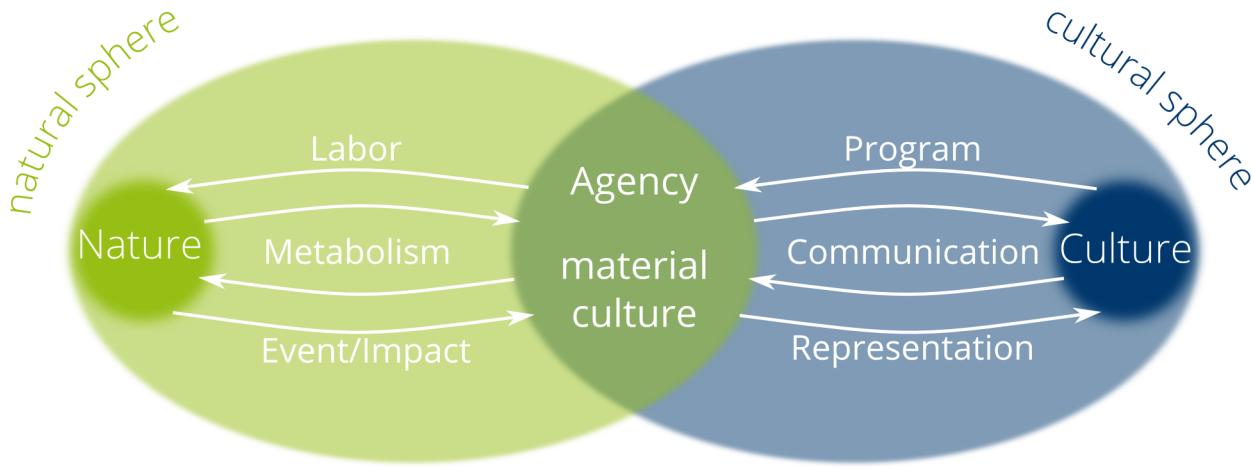


V/S Modellieren in der Landschaftsarchäologie

Freie Universität Berlin
M.Sc. Landschaftsarchäologie

Dr. Daniel Knitter

About me



Terminliches

03.12.2015 10-14

10.12.2015 10-14

17.12.2015 10-14

~~07.01.2016 10-14~~

14.01.2016 10-14

~~21.01.2016~~

~~28.01.2016~~

04.02.2016 10-14

11.02.2016 10-14

Blocktermin als Ersatz...wann?

Aus der Studienordnung

Qualifikationsziele:

Die Studentinnen und Studenten können zwischen Black-Box-, Grey-Box- und White-Box-Modellen differenzieren und deren Vor- und Nachteile benennen. Sie sind in der Lage Modelle unter Berücksichtigung der vorhandenen Datenlage und Fragestellung auszuwählen und anzuwenden.

Inhalte:

In der Landschaftsarchäologie gibt es unterschiedlichste Fragestellungen, für deren Beantwortung sich eine Modellbildung bzw. Modellierung anbietet:
Wasserhaushaltsmodellierungen, Predictive Modelling, Least-Cost-Path-Analysis, Erstellung von Modellen zur Landschaftsentwicklung. Ausgewählte Modellansätze werden in der Vorlesung zunächst vorgestellt und im Rahmen des Seminars anschließend erprobt.

Aus der Prüfungsordnung

Modul 302: Modellieren in der Landschaftsarchäologie		
Zugangsvoraussetzungen: Keine		
Lehr- und Lernformen	Modulprüfung	Pflicht zu regelmäßiger Teilnahme
Vorlesung		Teilnahme wird empfohlen
Übung	Hausarbeit (ca. 2 000 Wörter)	Ja
Leistungspunkte: 5		



Abgabe eines ausformulierten R-Skripts, welches ein selbst gewähltes Modellierungsproblem untersucht
(inkl. Problemstellung/State of the art/Literatur/...)

+ Lesen von Texten, Hausaufgaben

Modellierung...?

Modell – lat. modulus = Gebäude in verkleinertem Maßstab

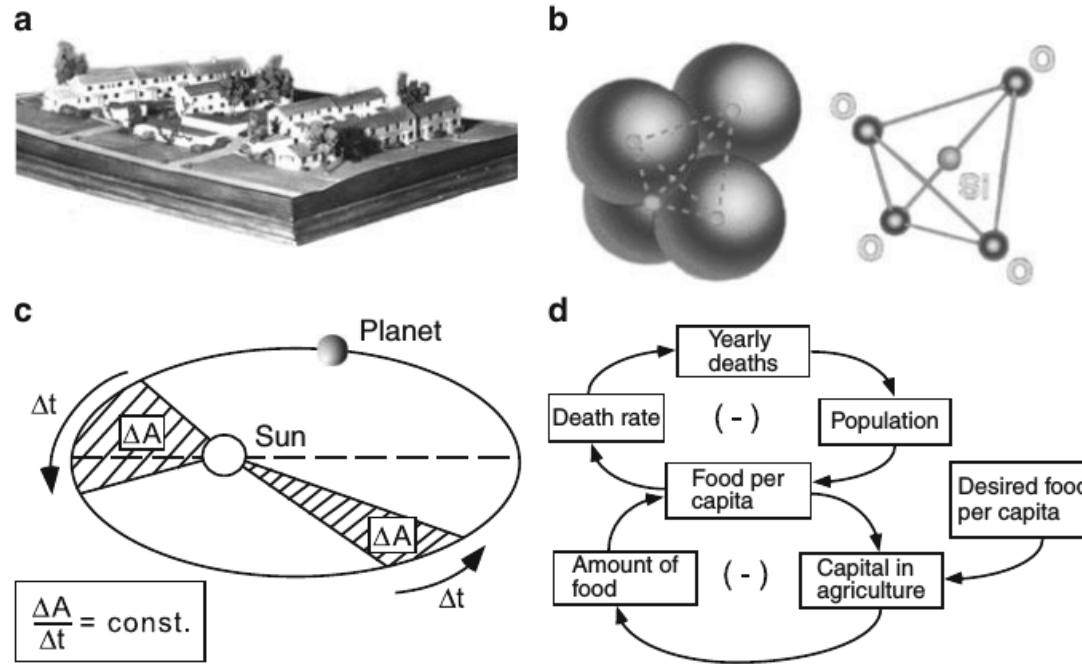


Fig. 1.2: Examples of models: (a) Architectural model: Scaled-down execution of a planned or existing building. (b) Physical model: Model of chemical molecules. (c) Mathematical model: Kepler's second law of planetary motion. (d) Conceptional model: Excerpt from the World model by Forrester and Meadows (see [Meadows et al. 1972](#))

Modellierung...?

“Ein Modell ist ein Konzept zur vereinfachten Darstellung eines komplexen Systems. Es dient dazu, die wichtigen Eigenschaften eines Systems darzustellen und die nebensächlichen Eigenschaften außer Acht zu lassen”

(Imboden & Koch 2003, 9).

Hausaufgabe 1 zum nächsten Mal

Nakoinz, Oliver, and Martin Hinz. 2015. "Modelle in der Archäologie."

In Wissenschaft und Kunst der Modellierung - Kieler Zugang zur Definition, Nutzung und Zukunft, edited by Bernhard Thalheim and Ivor Nissen. Berlin, Boston: De Gruyter.

→ gibt es per Mail von mir [oder als E-Ressource bei der Stabi]

11 Modelle in der Archäologie

Oliver Nakoinz und Martin Hinz

11.1 Einleitung und Forschungsgeschichte

In den letzten Jahren haben Modelle und Modellierung in der Archäologie erheblich an Popularität gewonnen. Dennoch ist es eine eher kleine Gruppe, die aktiv und bewusst Modellierung betreibt und den Modellbegriff gezielt einsetzt. Neben dieser überschaubaren expliziten Nutzung von Modellen spielt jedoch die implizite Modellnutzung in der Archäologie, wie in allen anderen Disziplinen, eine große Rolle. Diese steht im Vordergrund des vorliegenden Beitrages. Es wird versucht die wichtigsten latenten Modelle zu identifizieren und in ihrem forschungsgeschichtlichen Kontext darzustellen. Anschließend werden beispielhaft einige explizite Modelle von Raumstrukturen vorgestellt.

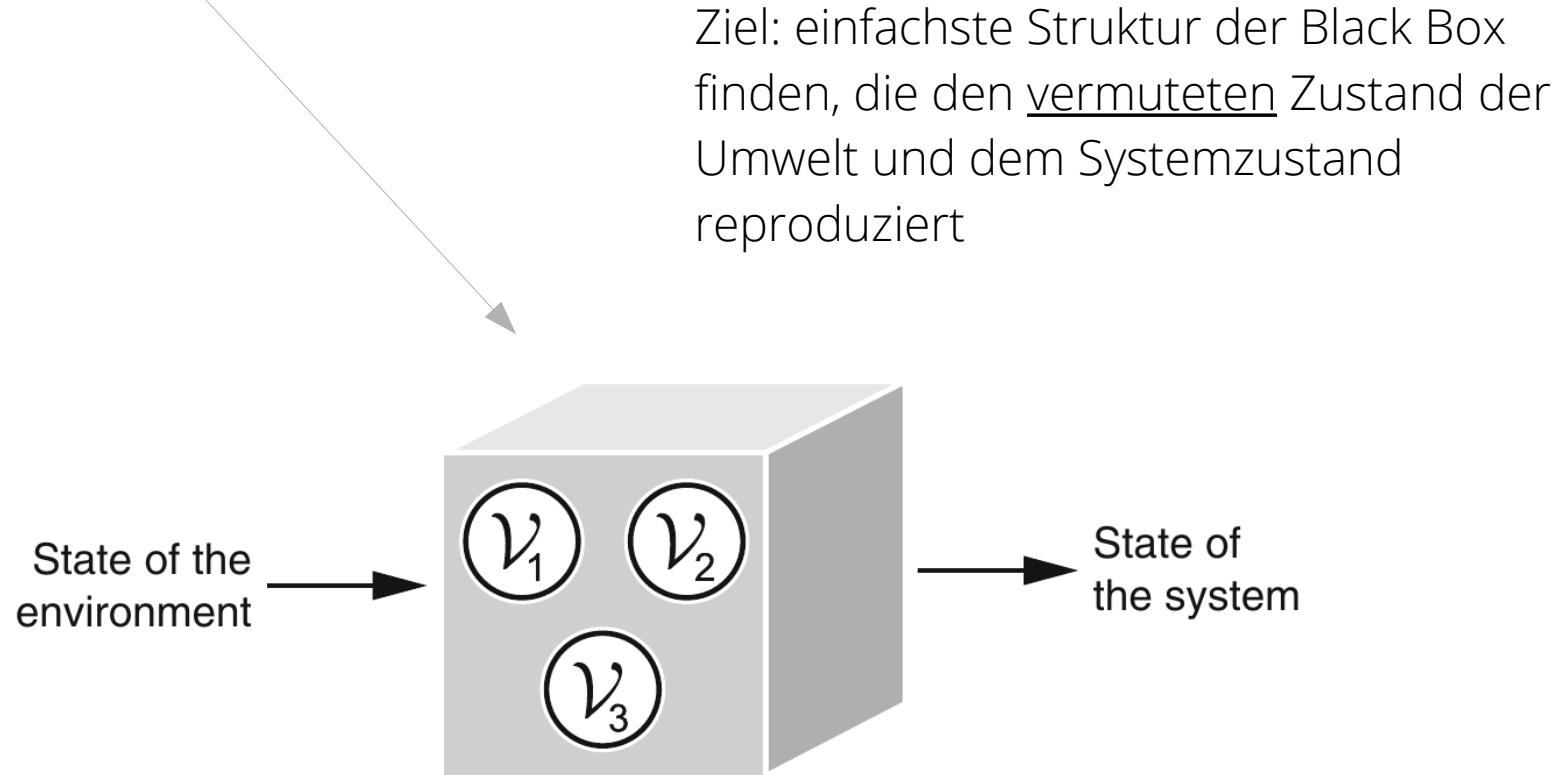
11.1.1 Archäologie zwischen Geistes- und Naturwissenschaften

Ebenso wie die Geographie und einige andere Fächer nimmt die Archäologie eine besondere Stellung zwischen Geistes- und Naturwissenschaften ein. Die Archäologie versteht sich einerseits als historische Wissenschaft, die historische Fragestellungen zu klären hat. Die Besonderheit der archäologischen Quellen führen zu einem spezifischen Schwerpunkt der Fragestellungen. Während Ereignisgeschichte anhand archäologischer Quellen meist nur sehr schwer zu schreiben ist, so ist die Strukturgeschichte ein natürliches Anwendungsfeld der Archäologe. Die archäologischen Quellen unterscheiden sich andererseits sehr deutlich von den historischen Quellen und machen die Archäologie zu einem nativen Mitglied der Naturwissenschaften. Archäologische Funde und Befunde sind zunächst nicht mit Bedeutung ausgestattet, wie es historische Quellen sind. Das hermeneutische Ausdeuten archäologischer Quellen und deren Interpretation erscheint demnach oft eher als wilde Spekulation denn als Wissenschaft. *Per se* sind archäologische Quellen dieser geisteswissenschaftlichen Herangehensweise nicht zugänglich. Sie erlauben jedoch die Anwendung naturwissenschaftlicher Konzepte. Während die Geisteswissenschaften im Wesentlichen auf dem Deuten und Aushandeln prädefinierter Bedeutungen gegebener Gegenstände beruhen, setzt die Naturwissenschaft mit einer Strukturanalyse zunächst bedeutungsloser Gegenstände ein und stattet diese im Laufe des Forschungsprozesses mit Bedeutung aus. Anders

Black-, Grey-, White-box Models

?

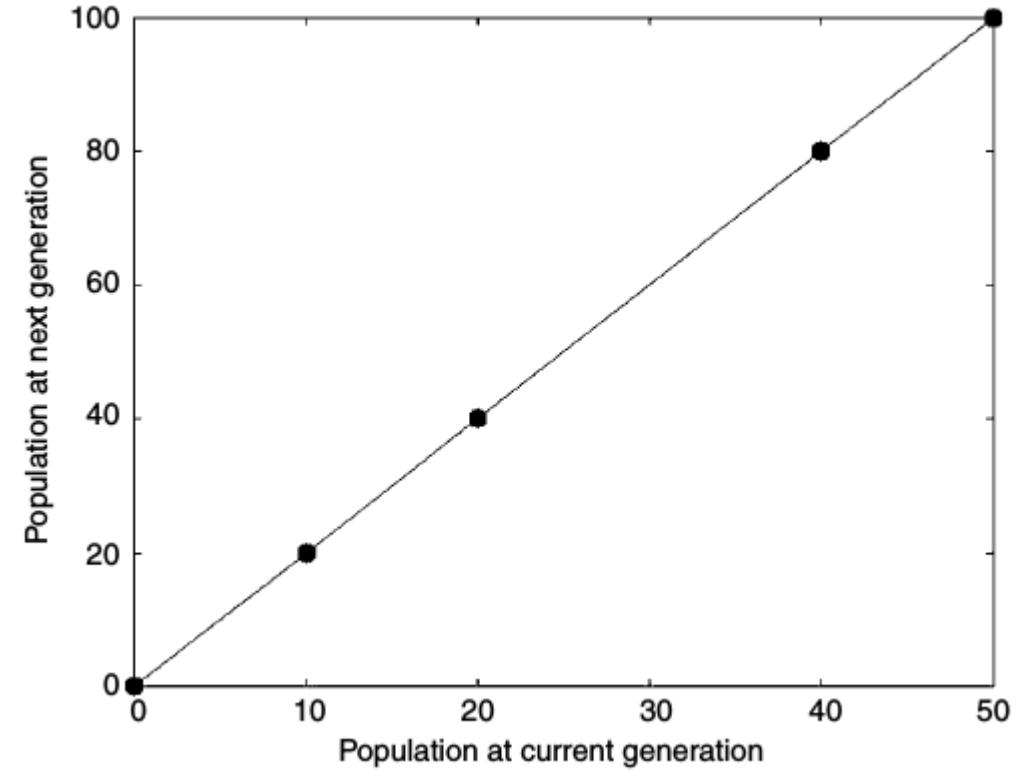
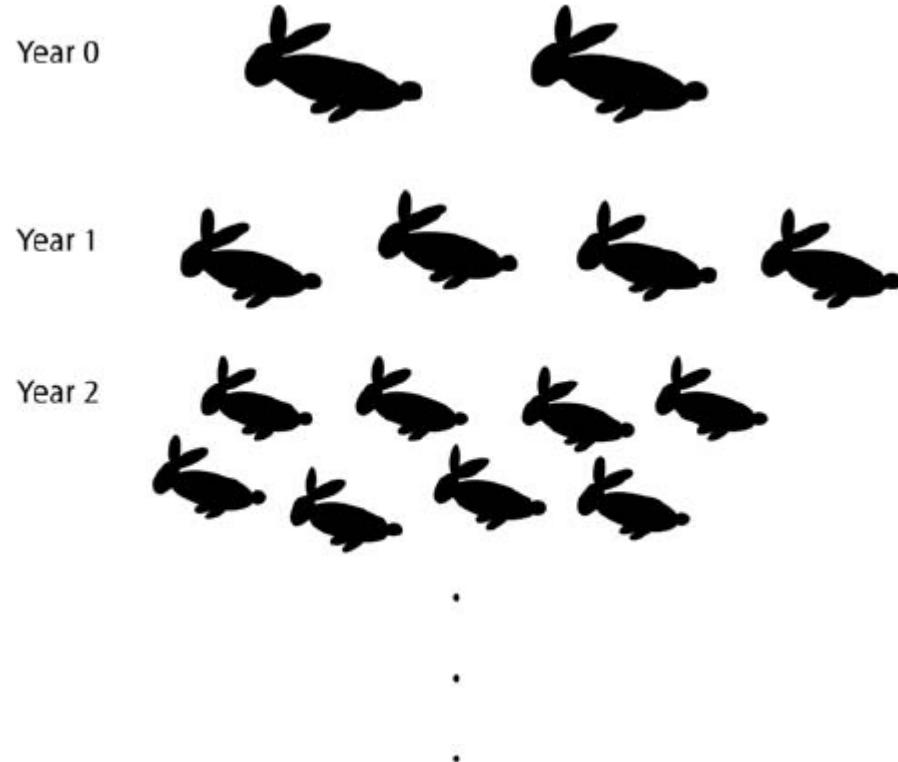
Black-, Grey-, White-box Models



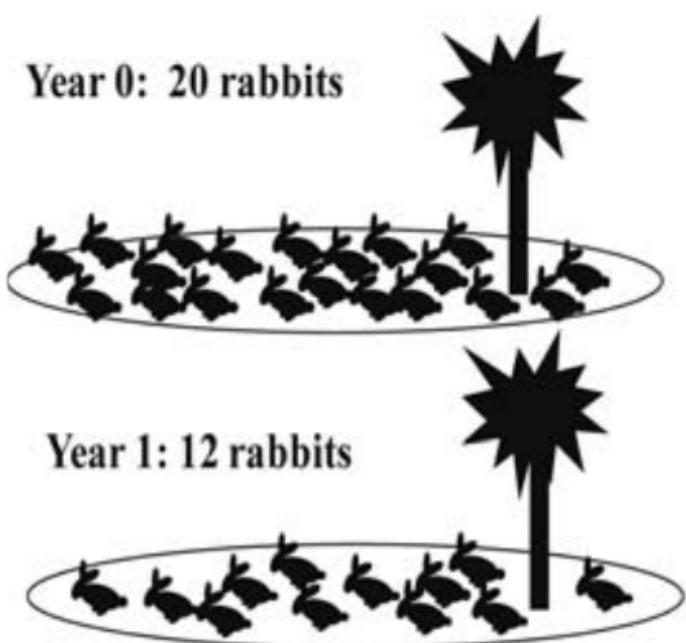
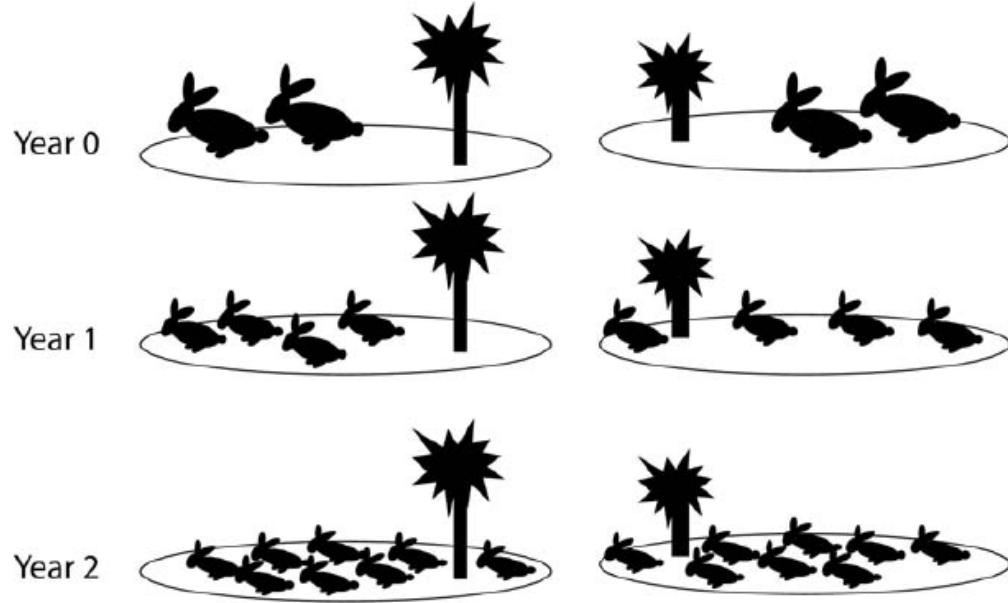
Black-, Grey-, White-box Models

Black-box models are completely nonmechanistic. We cannot investigate interactions of subsystems of such a non-transparent model. A white-box model of complex dynamic systems has ‘transparent walls’ and directly shows underlying mechanisms – all events at micro-, meso- and macro-levels of a modeled dynamic system are directly visible at all stages. Logical deterministic cellular automata is the only known approach, which allows to create white-box models of complex dynamic systems (3). A micro-level is modeled by a lattice site (cellular automata cell). A meso-level of local interactions of micro-objects is modeled by a cellular automata neighbourhood. A macro-level is modeled by the entire cellular automata lattice. Unfortunately, this simple approach is commonly used in the overloaded form, what makes it less transparent. This is achieved by adding differential equations and stochasticity. Grey-box models are intermediate and combine black-box and white-box approaches. Basic ecological models are of black-box type, e.g. Malthusian, Verhulst, Lotka-Volterra models. These models are not individual-based and cannot show features of local interactions of individuals of competing species. That is why they principally cannot provide a mechanistic insight into interspecific competition.

Black-, Grey-, White-box Models



Black-, Grey-, White-box Models

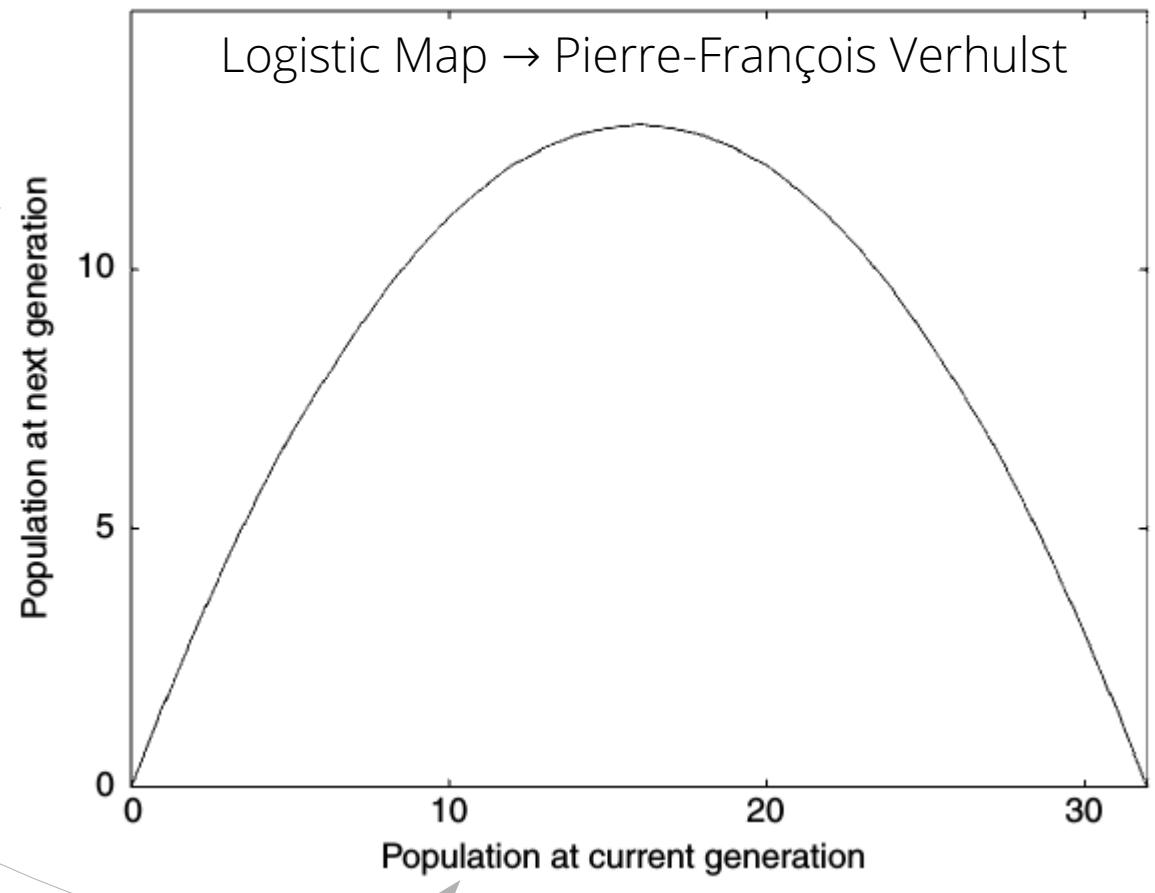


Black-, Grey-, White-box Models

$$\frac{dP}{dt} = P \cdot r$$

$$x_{t+1} = R x_t (1 - x_t)$$

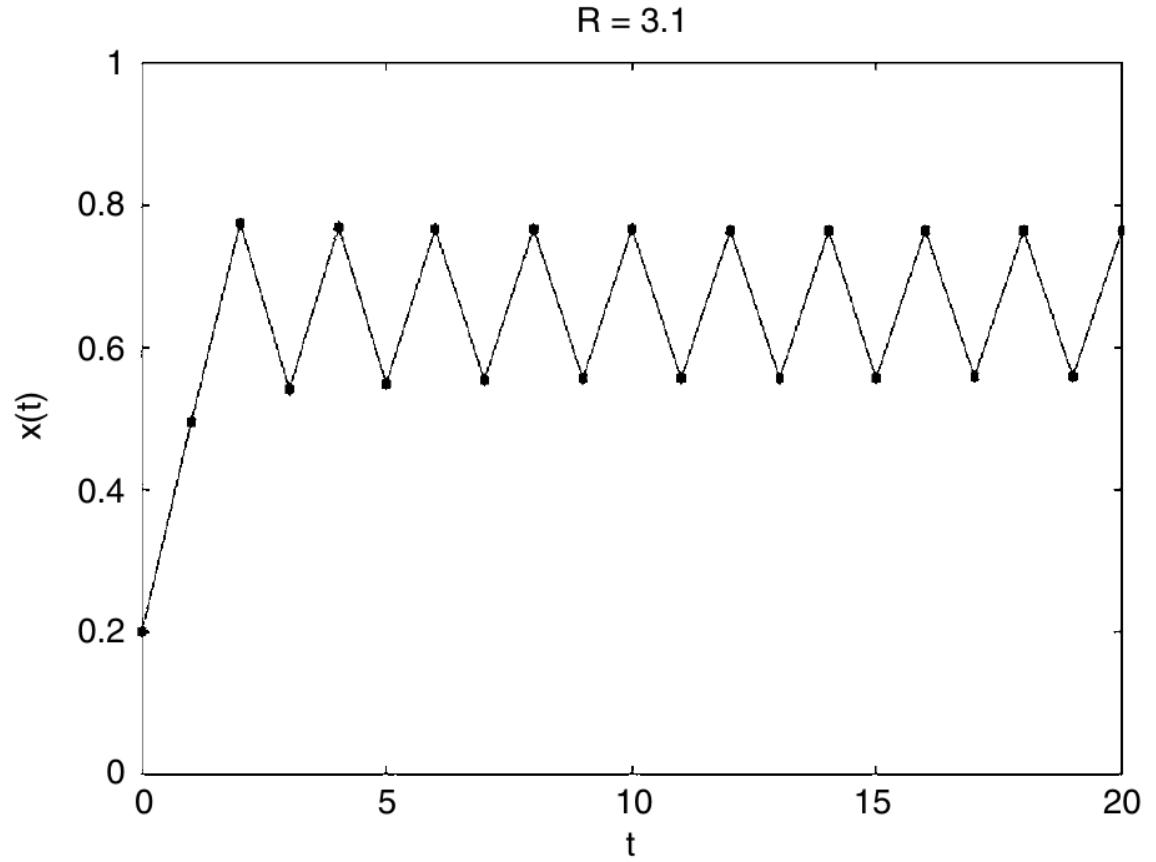
Birth rate/death rate
+ carrying capacity



Black-, Grey-, White-box Models

$$\frac{dP}{dt} = P \cdot r$$

$$x_{t+1} = R x_t (1 - x_t)$$



Black-, Grey-, White-box Models

$$\frac{dP}{dt} = P \cdot r$$

$$x_{t+1} = R x_t (1 - x_t)$$

“deterministic chaos”

→ Sensitivity to initial conditions

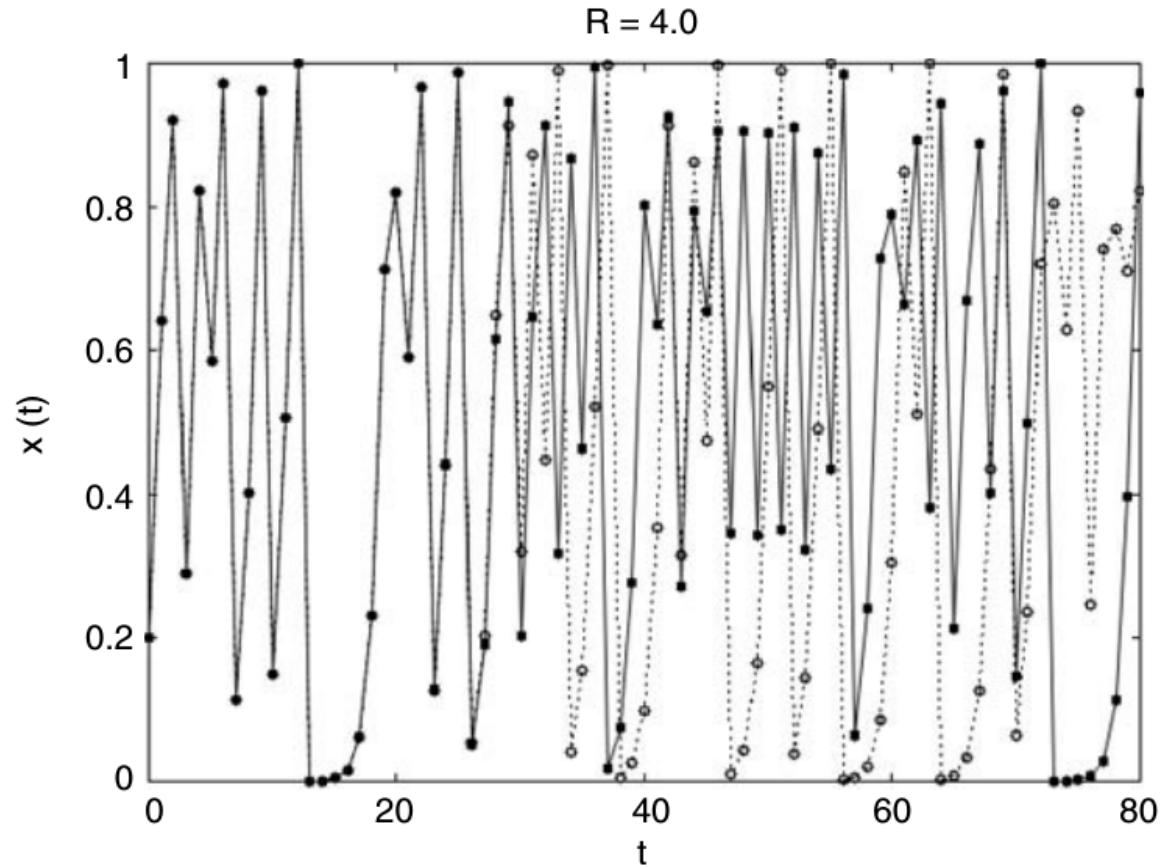


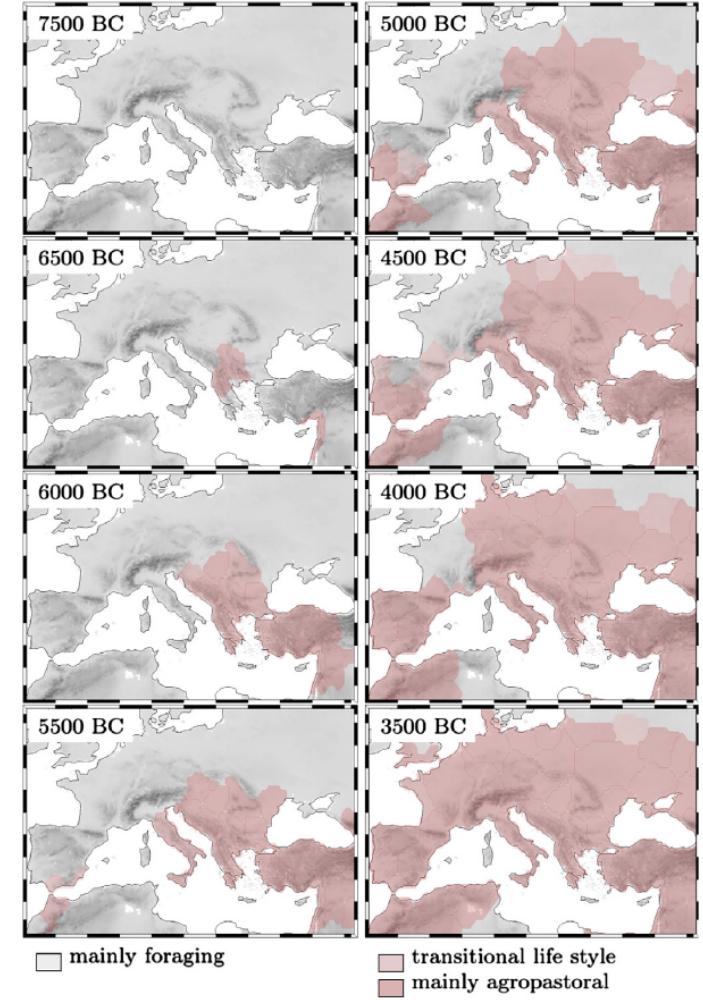
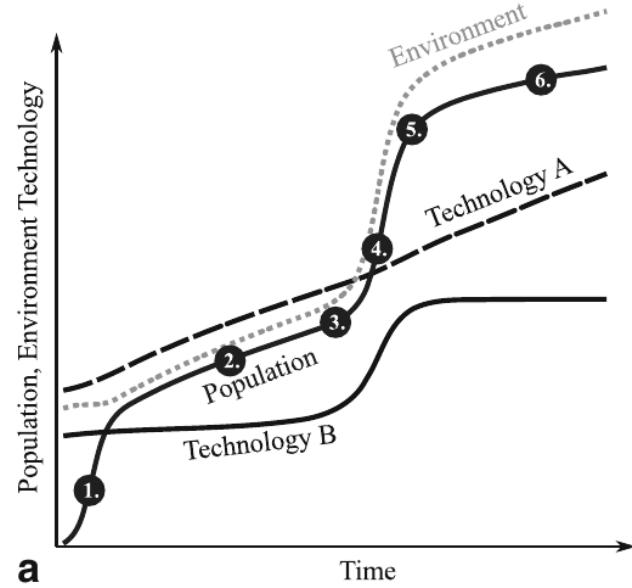
FIGURE 2.10. Two trajectories of the logistic map for $R = 4.0 : x_0 = 0.2$ and $x_0 = 0.2000000001$.

Black-, Grey-, White-box Models

$$\frac{dP}{dt} = P \cdot r$$

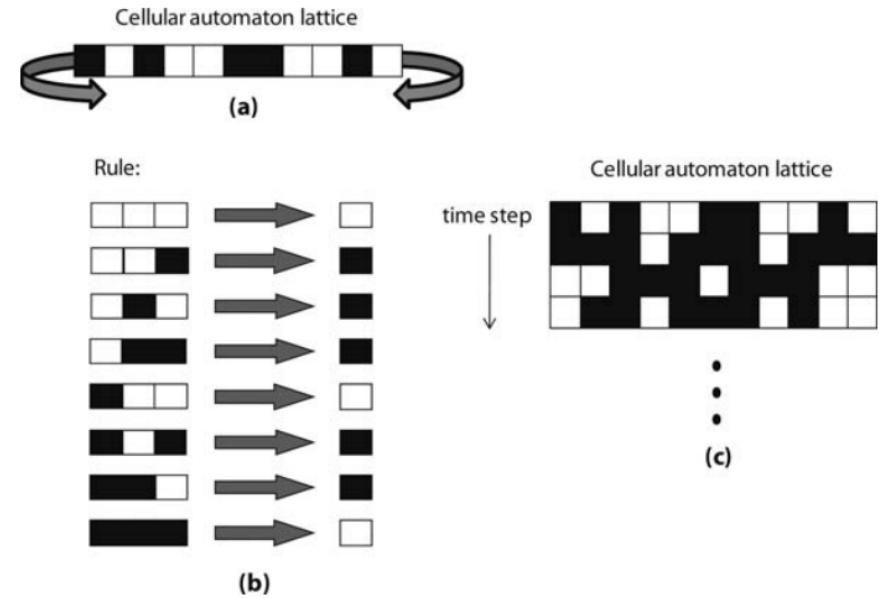
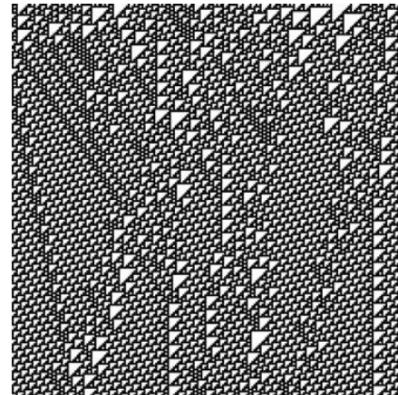
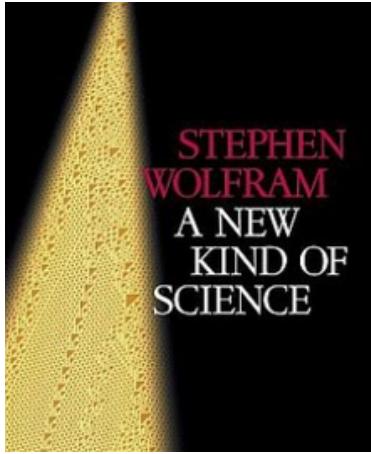
$$r = \mu \cdot (1 - \omega T_A) \cdot \left(1 - \gamma \sqrt{T_A} P\right) \cdot SI - \rho \cdot T_A^{-1} \cdot P$$

PET trajectories



(Lemmen 2011, 2014)

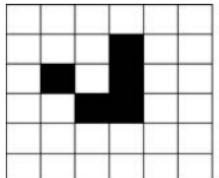
Black-, Grey-, White-box Models



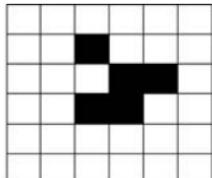
Black-, Grey-, White-box Models

Conway's Game of life

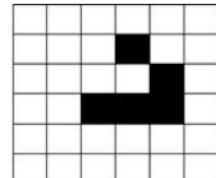
- Any live cell with fewer than two live neighbours dies (loneliness)
- Any live cell with two or three live neighbours lives on to the next generation.
- Any live cell with more than three live neighbours dies (too crowded)
- Any dead cell with exactly three live neighbours becomes a live cell, as if by reproduction.



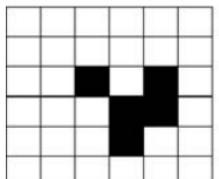
$t = 0$



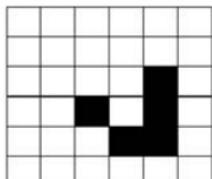
$t = 1$



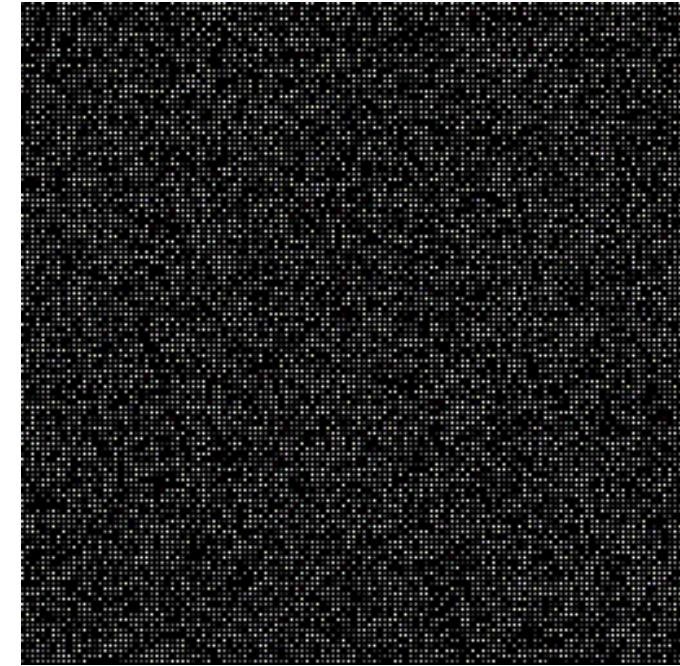
$t = 2$



$t = 3$



$t = 4$



Black-, Grey-, White-box Models

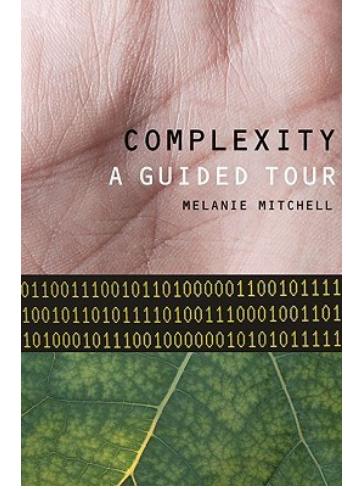


Chaos and Fractals
An Elementary Introduction

David P. Feldman

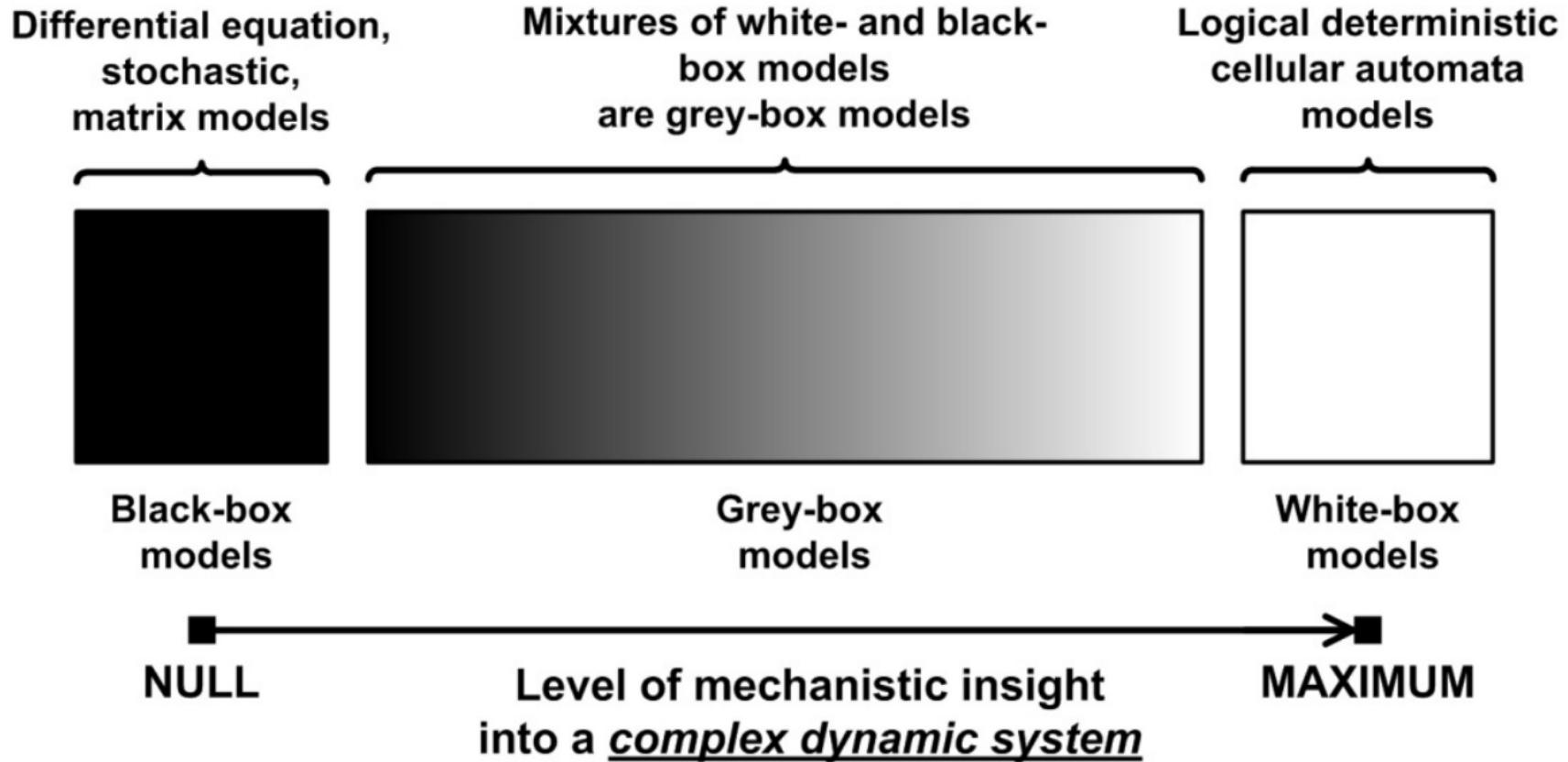
OXFORD

Copyrighted Material



<https://www.complexityexplorer.org/>

Black-, Grey-, White-box Models



Where to start?

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Spatial data are special

Pitfalls of spatial data:

- spatial autocorrelation
- modifiable areal unit problem
- scale
- ecological fallacy
- non-uniformity of space
- edge effects

Spatial autocorrelation

The Earth is not an isotropic plate

Spatial data is not random

Data from locations near to each other are usually more similar than data far away from each other

Space and location are important because of Spatial autocorrelation.

Without spatial autocorrelation, spatial analyses would be pointless

Spatial autocorrelation



Low



High



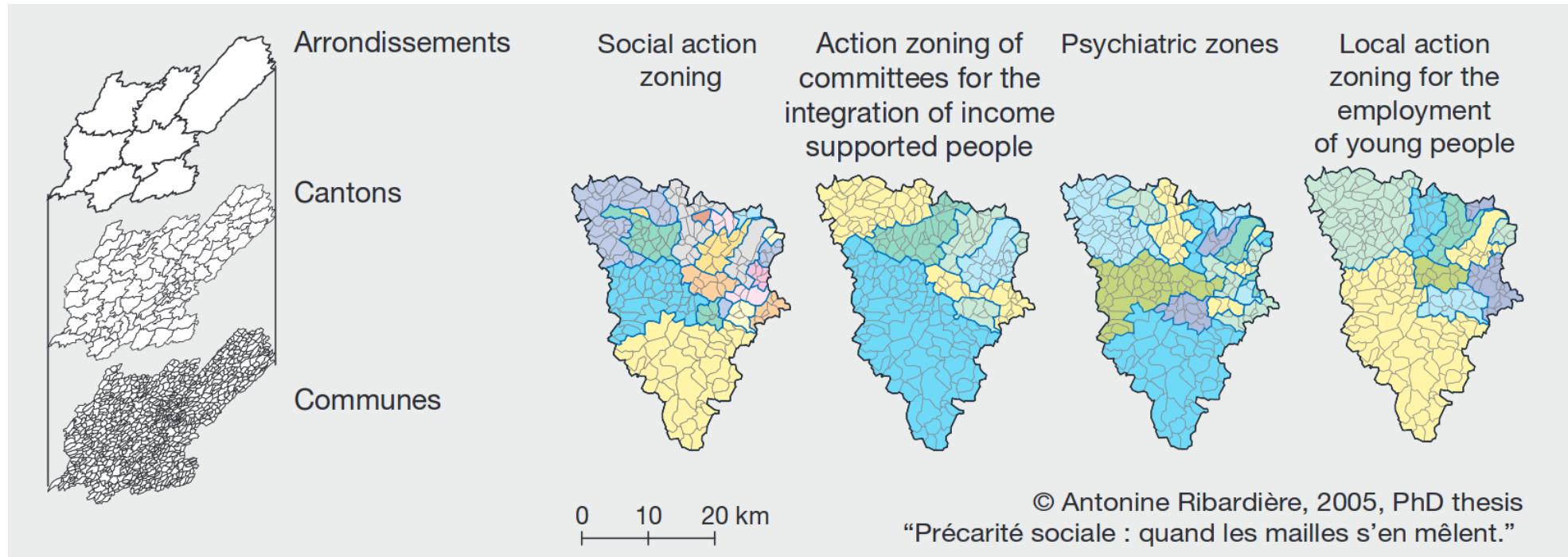
Maximum

Modifiable Areal Unit Problem (MAUP)

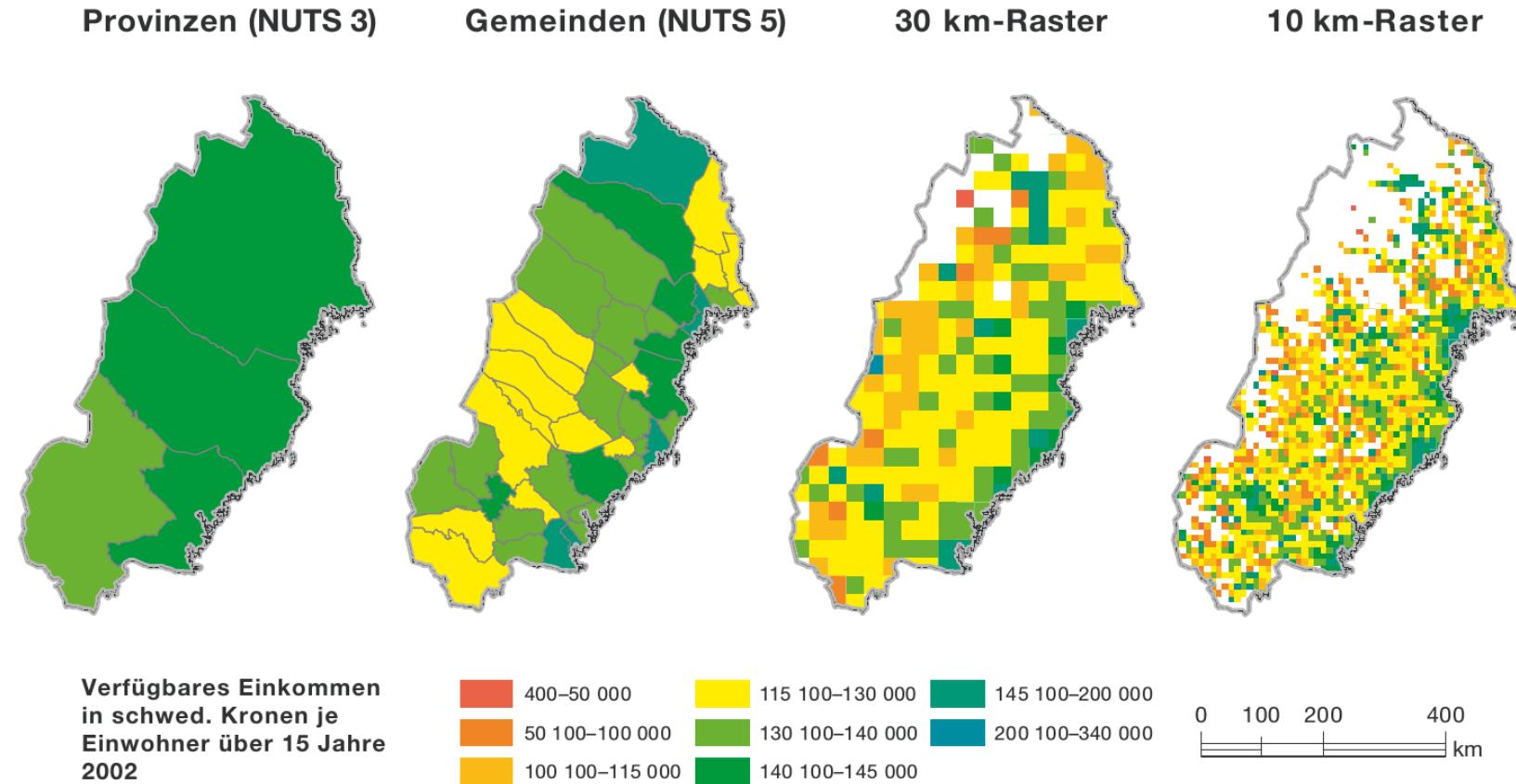
- aggregation units used are arbitrary with respect to the phenomena under investigation
- If spatial units are specified differently, one might get very different patterns

→ Openshaw, Stan (1984): The modifiable areal unit problem

Modifiable Areal Unit Problem (MAUP)

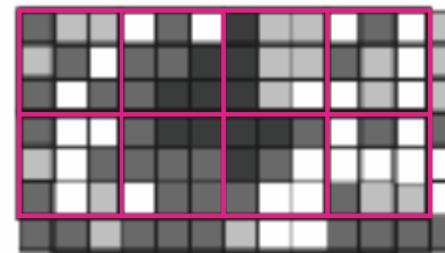
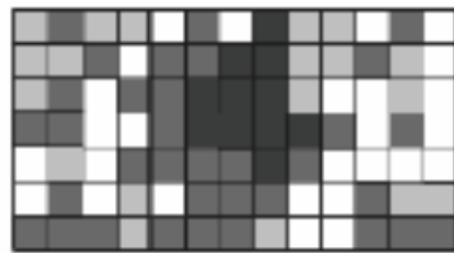


Modifiable Areal Unit Problem (MAUP)

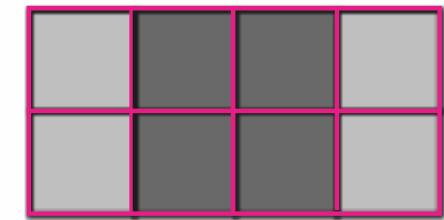


© M. Strömgren, K. Holme, E. Holm, S.M.C., Umeå University, Sweden

Modifiable Areal Unit Problem (MAUP)



Zone A



Mean of zones

Modifiable Areal Unit Problem (MAUP)

Independent variable Dependent variable

87	95	72	37	44	24
40	55	55	38	88	34
41	30	26	35	38	24
14	56	37	34	8	18
49	44	51	67	17	37
55	25	33	32	59	54

72	75	85	29	58	30
50	60	49	46	84	23
21	46	22	42	45	14
19	36	48	23	8	29
38	47	52	52	22	48
58	40	46	38	35	55

Aggregation scheme 1

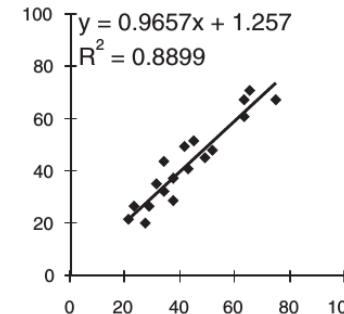
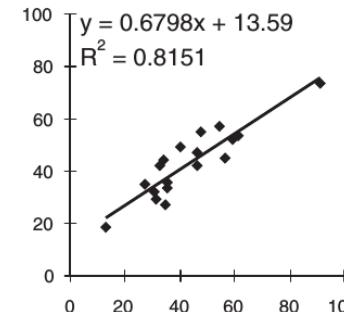
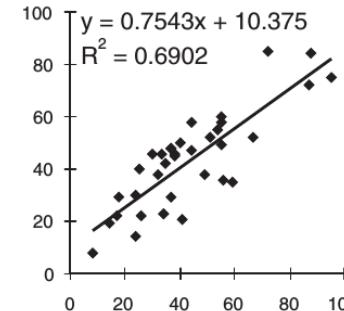
91	54.5	34
47.5	46.5	61
35.5	30.5	31
35	35.5	13
46.5	59	27
40	32.5	56.5

73.5	57	44
55	47.5	53.5
33.5	32	29.5
27.5	35.5	18.5
42.5	52	35
49	42	45

Aggregation scheme 2

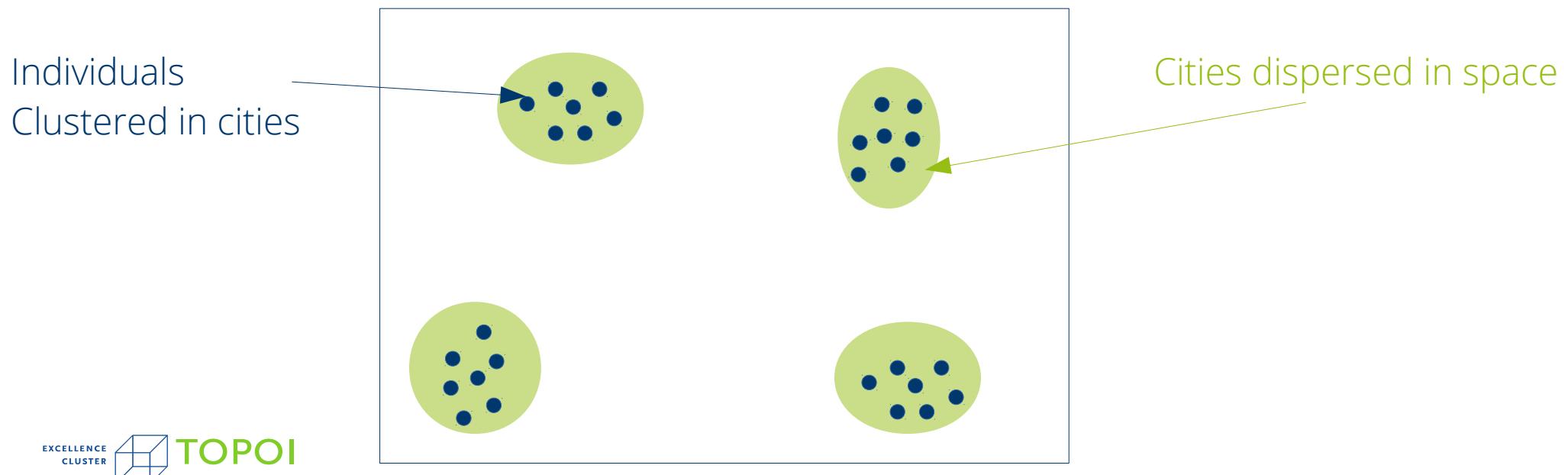
52	27.5	63.5
34.5	43	75
42	31.5	63.5
49.5	34.5	37.5
38	23	66
45.5	21	29

48	20	61
43.5	41	67.5
49	35	67
45	32.5	37.5
28.5	26.5	71
51.5	21.5	26.5



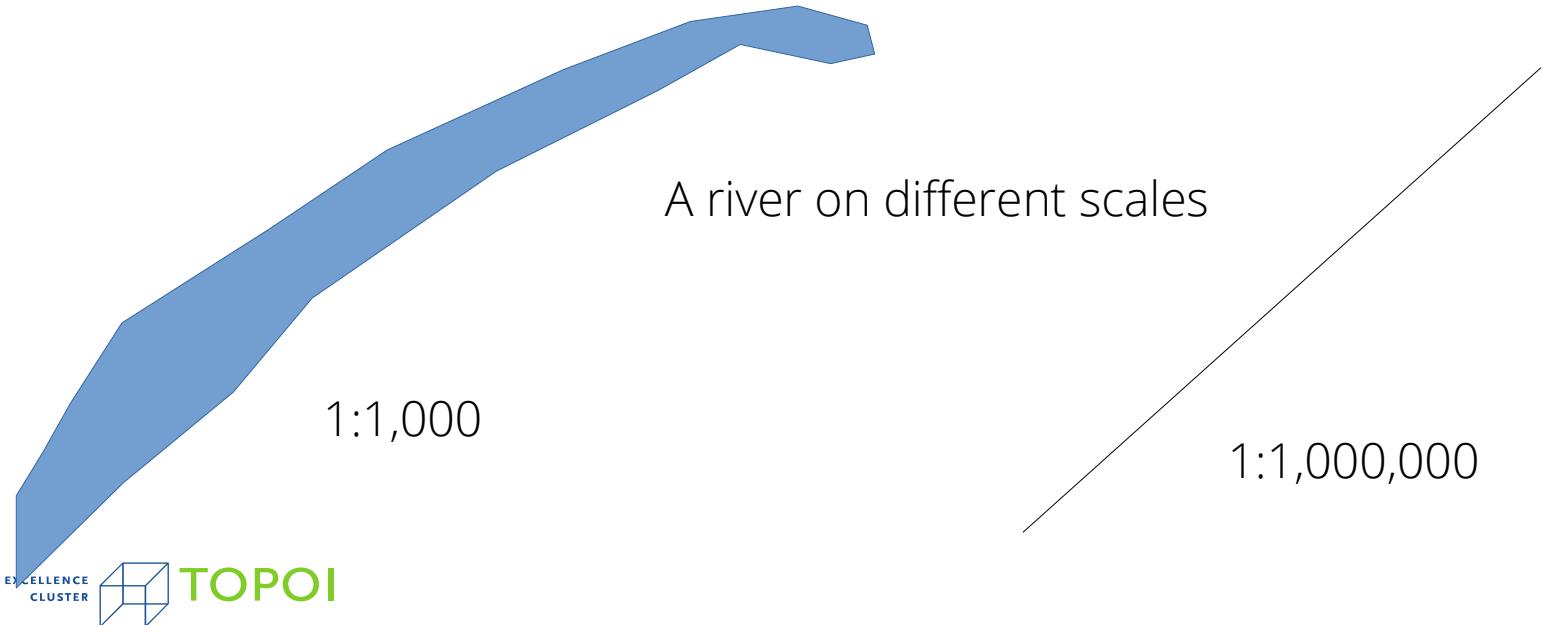
Scale effects

- Scale affects results and representativity of data
- e.g. cities may be represented as points or polygons
- results depend on the scale of analysis, e.g. province or county → MAUP



Scale effects

- Scale affects results and representativity of data
- e.g. cities may be represented as points or polygons
- results depend on the scale of analysis, e.g. province or county → MAUP



Ecological fallacy

Arises when:

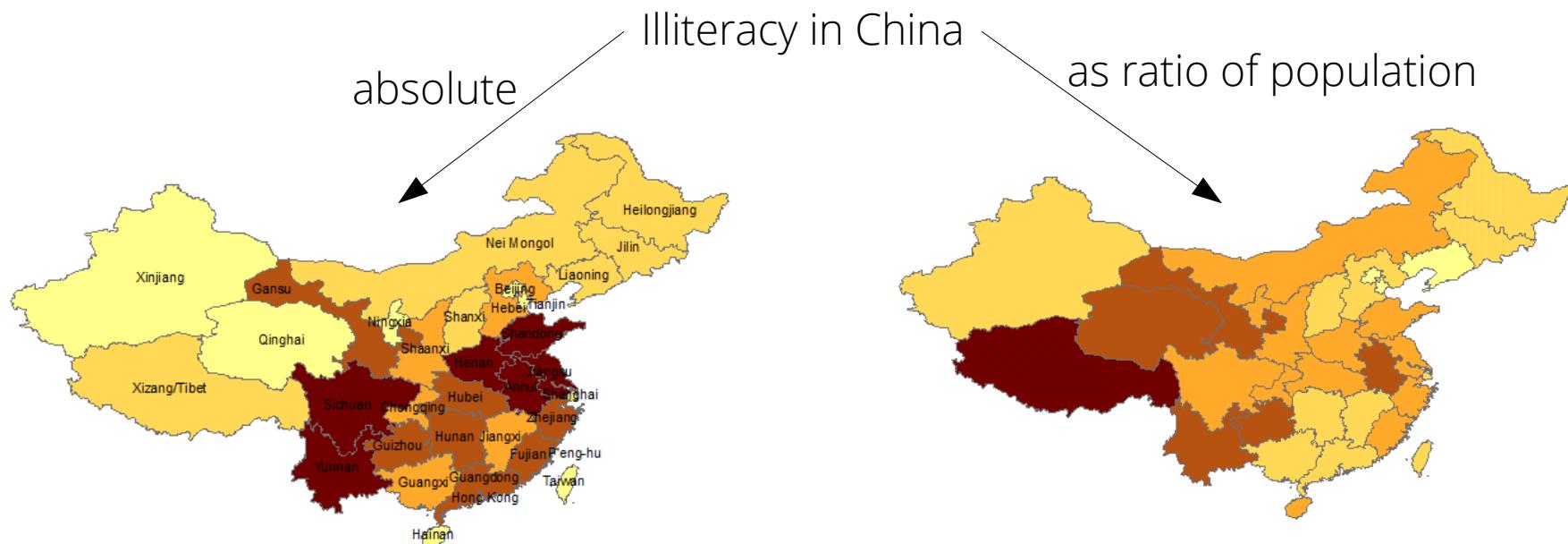
- Statistical relationship at one level of aggregation...
- ...is assumed to hold true because it holds at a more detailed level

For example, we might observe a strong relationship between income and crime at the county level, with lower-income counties being associated with higher crime rates. If from this we conclude that lower-income individuals are more likely to commit a crime, then we are falling for the ecological fallacy. In fact, it is only valid to say exactly what the data say: that lower-income counties tend to experience higher crime rates. What causes the observed effect may be something entirely different

Non-uniformity of space

Bank robberies are clustered
→ because banks are clustered

Diseases due to bad air are clustered
→ because factories are clustered



Edge effects

- unless you study the entire world: Every study region has a boundary
- You do not have data for outside your study region
- But: the “outside” data may/do affect the data of your study region (if there is spatial autocorrelation...)

Spatial data are special

Pitfalls of spatial data:

- spatial autocorrelation
- modifiable areal unit problem
- scale
- ecological fallacy
- non-uniformity of space
- edge effects

Potentials of spatial data:

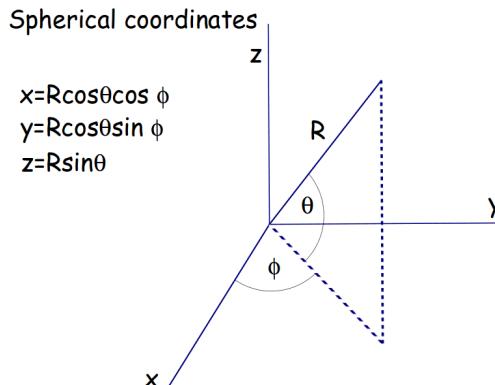
- distance
- adjacency
- interaction
- neighborhood

Distance

$$d_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$$

Euclidean distance as calculated using
Pythagoras' formula

→ for *projected* data on the local/regional scale



R = radius of earth, ca. 6378 km

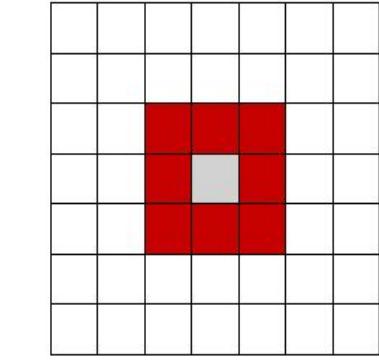
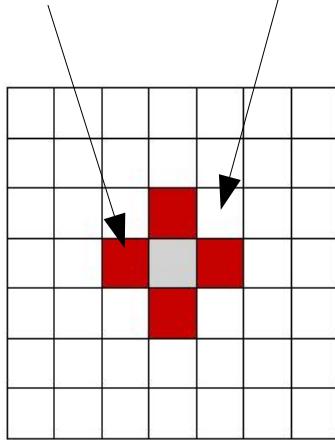
Spherical distance via spherical coordinates
and a 3D version of *Pythagoras'* formula

$$d^2 = (x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2$$

→ for *unprojected* data on a global scale

Binary

→ things are adjacent or not

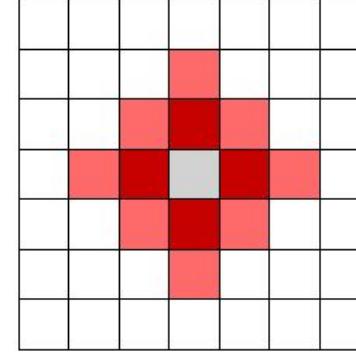


1st order neighbor

Adjacency

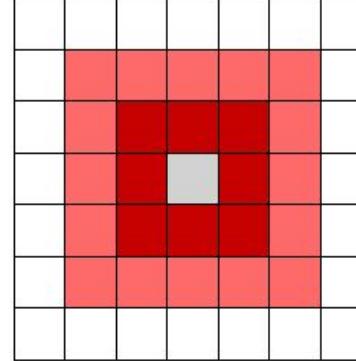
Rook case

→ sharing a boundary



Queen case

→ sharing a boundary or point

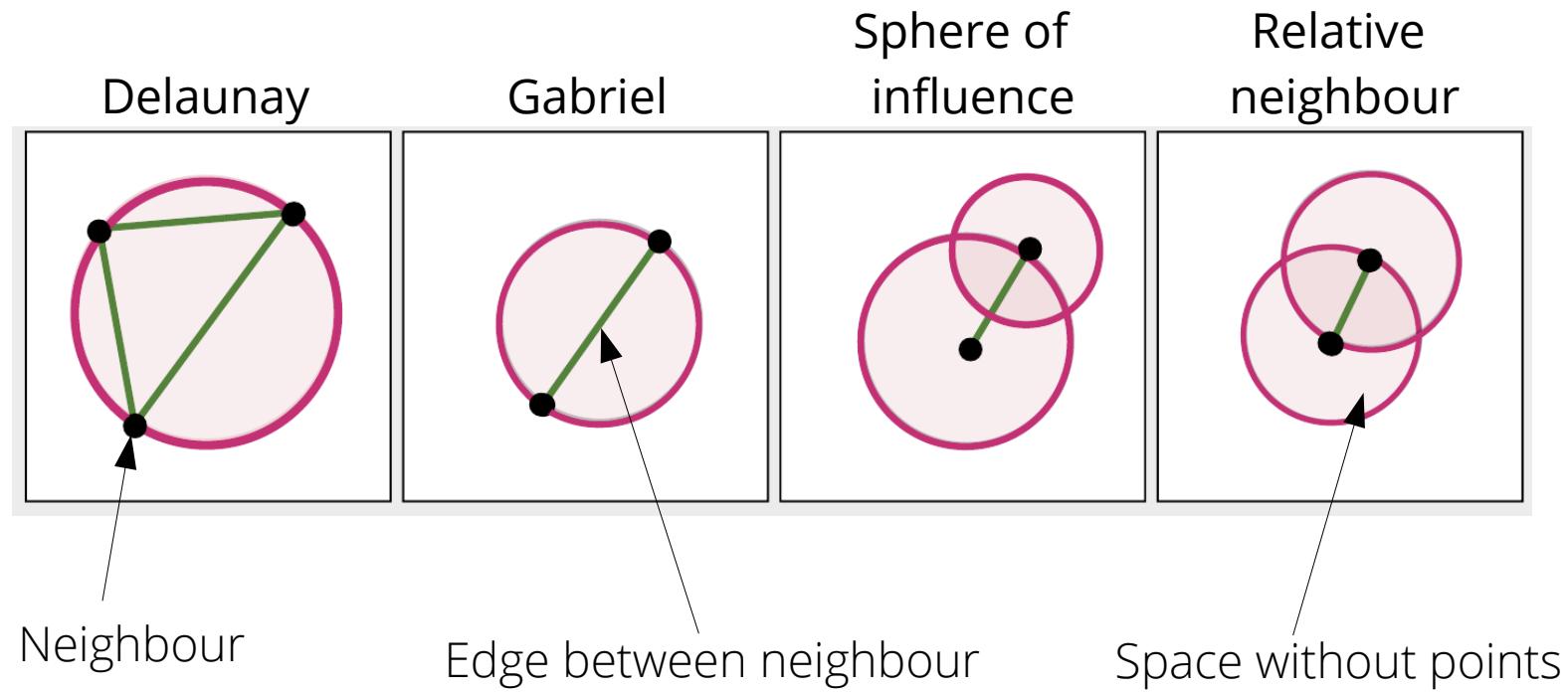


2nd order neighbor

and also...distance based neighbors...

Adjacency

...or graph based neighbors.



Interaction

- A combination of distance and adjacency
- related to Tobler's first law of geography:

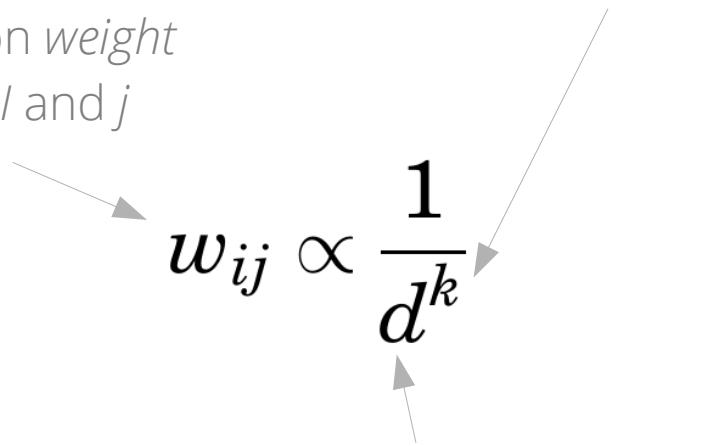
*"(...) everything is related to everything else,
but near things are more related
than distant things"* (Tobler 1970, 236)

interaction weight
between i and j

rate of decline

$$w_{ij} \propto \frac{1}{d^k}$$

distance between i and j



...strength of interaction between i and j is proportional to their inverse distance

Neighborhood

Neighborhood of X is:

- the set of adjacent cells/areas
- the set of points/cells/areas within a certain distance
- the set of points/cells/areas that share some specific attribute or that are alike

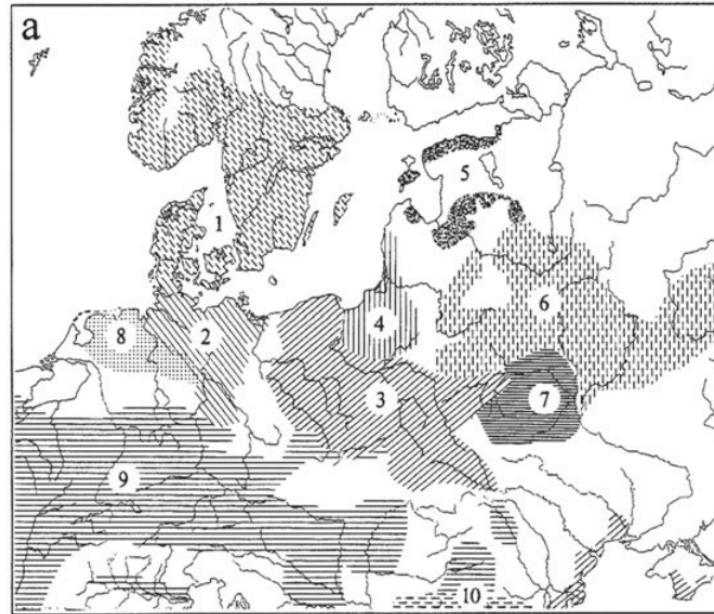
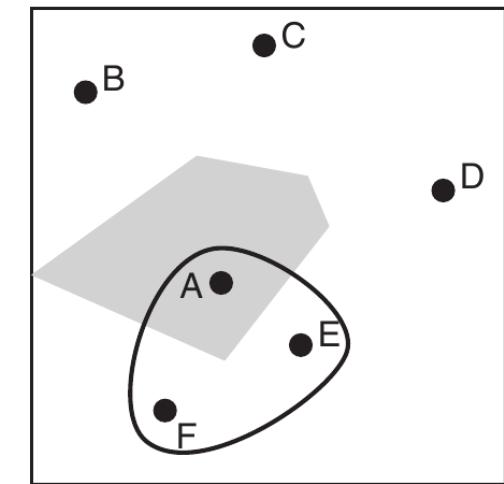
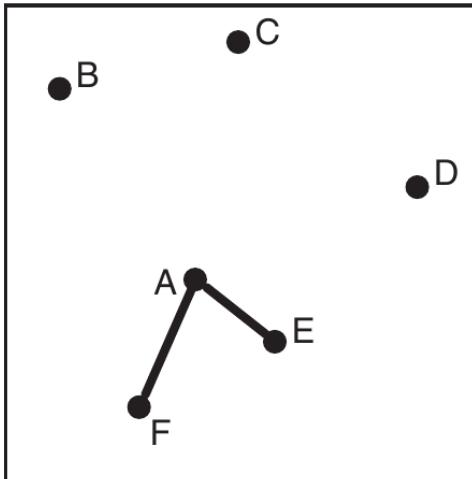
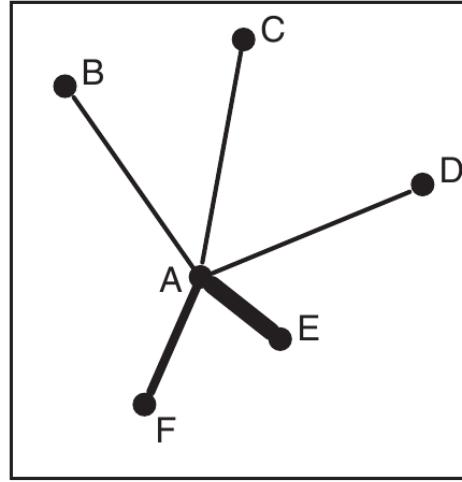
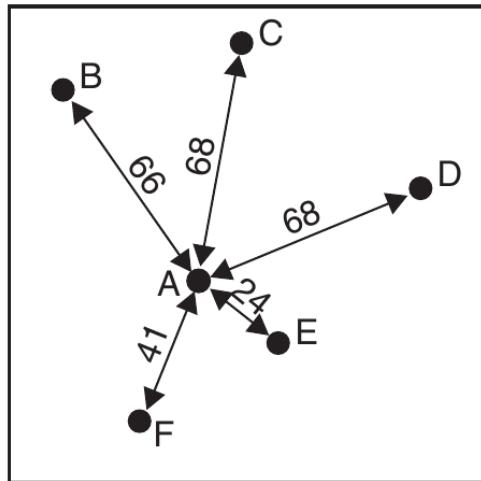


Abb. 11.4. Karte ältereisenzeitlicher Kulturen (Mey10, Abb. 1), die Nummern verweisen auf einzelne eisenzeitliche Kulturen, die mit unterschiedlichen Signaturen angezeigt sind.

Distance – Adjacency – Interaction - Neighborhood



Distance – Adjacency – Interaction - Neighborhood

$$\mathbf{D} = \begin{bmatrix} 0 & 66 & 68 & 68 & 24 & 41 \\ 66 & 0 & 51 & 110 & 99 & 101 \\ 68 & 51 & 0 & 67 & 91 & 116 \\ 68 & 110 & 67 & 0 & 60 & 108 \\ 24 & 99 & 91 & 60 & 0 & 45 \\ 41 & 101 & 116 & 108 & 45 & 0 \end{bmatrix}$$

Distance – Adjacency – Interaction - Neighborhood

$$\mathbf{A}_{k=3} = \begin{bmatrix} * & 1 & 0 & 0 & 1 & 1 \\ 1 & * & 1 & 0 & 1 & 0 \\ 1 & 1 & * & 1 & 0 & 0 \\ 1 & 0 & 1 & * & 1 & 0 \\ 1 & 0 & 0 & 1 & * & 1 \\ 1 & 1 & 0 & 0 & 1 & * \end{bmatrix}$$

$$\mathbf{A}_{d \leq 50} = \begin{bmatrix} * & 0 & 0 & 0 & 1 & 1 \\ 0 & * & 0 & 0 & 0 & 0 \\ 0 & 0 & * & 0 & 0 & 0 \\ 0 & 0 & 0 & * & 0 & 0 \\ 1 & 0 & 0 & 0 & * & 1 \\ 1 & 0 & 0 & 0 & 1 & * \end{bmatrix}$$

Distance – Adjacency – Interaction - Neighborhood

$$\mathbf{W} = \begin{bmatrix} \infty & 0.0152 & 0.0147 & 0.0147 & 0.0417 & 0.0244 \\ 0.0152 & \infty & 0.0196 & 0.0091 & 0.0101 & 0.0099 \\ 0.0147 & 0.0196 & \infty & 0.0149 & 0.0110 & 0.0086 \\ 0.0147 & 0.0091 & 0.0149 & \infty & 0.0167 & 0.0093 \\ 0.0417 & 0.0101 & 0.0110 & 0.0167 & \infty & 0.0222 \\ 0.0244 & 0.0099 & 0.0086 & 0.0093 & 0.0222 & \infty \end{bmatrix}$$

row totals:

0.1106					
0.0639					
0.0688					
0.0646					
0.1016					
0.0744					

$\rightarrow W[1,2] = W[1,:] / \text{sum}(W[1,:])$

...

$$\mathbf{W} = \begin{bmatrix} \infty & 0.1370 & 0.1329 & 0.1329 & 0.3767 & 0.2205 \\ 0.2373 & \infty & 0.3071 & 0.1424 & 0.1582 & 0.1551 \\ 0.2136 & 0.2848 & \infty & 0.2168 & 0.1596 & 0.1252 \\ 0.2275 & 0.1406 & 0.2309 & \infty & 0.2578 & 0.1432 \\ 0.4099 & 0.0994 & 0.1081 & 0.1640 & \infty & 0.2186 \\ 0.3279 & 0.1331 & 0.1159 & 0.1245 & 0.2987 & \infty \end{bmatrix}$$

column totals:

1.4161	0.7949	0.8949	0.7805	1.2510	0.8626
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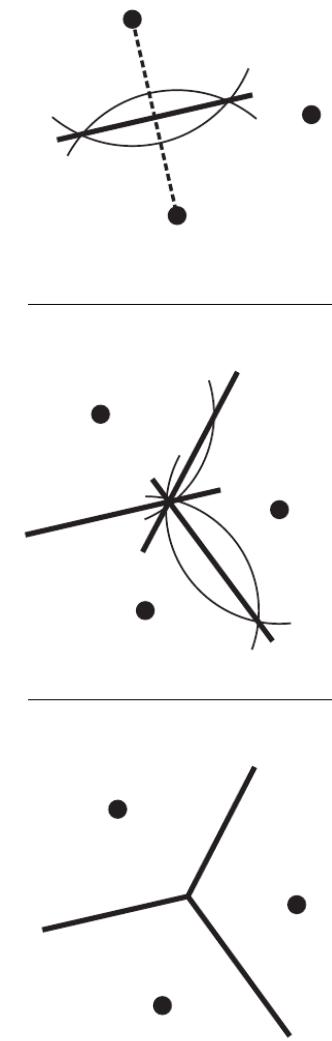
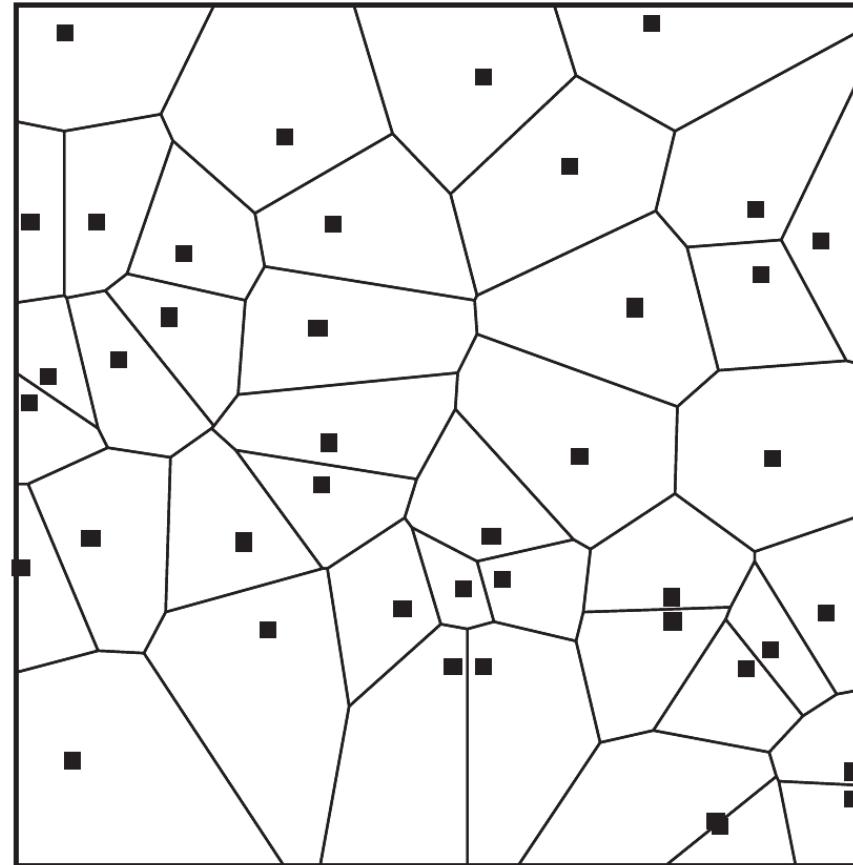
Voronoi/Thiessen polygons

Adjacency

Interaction

Distance

Neighborhood



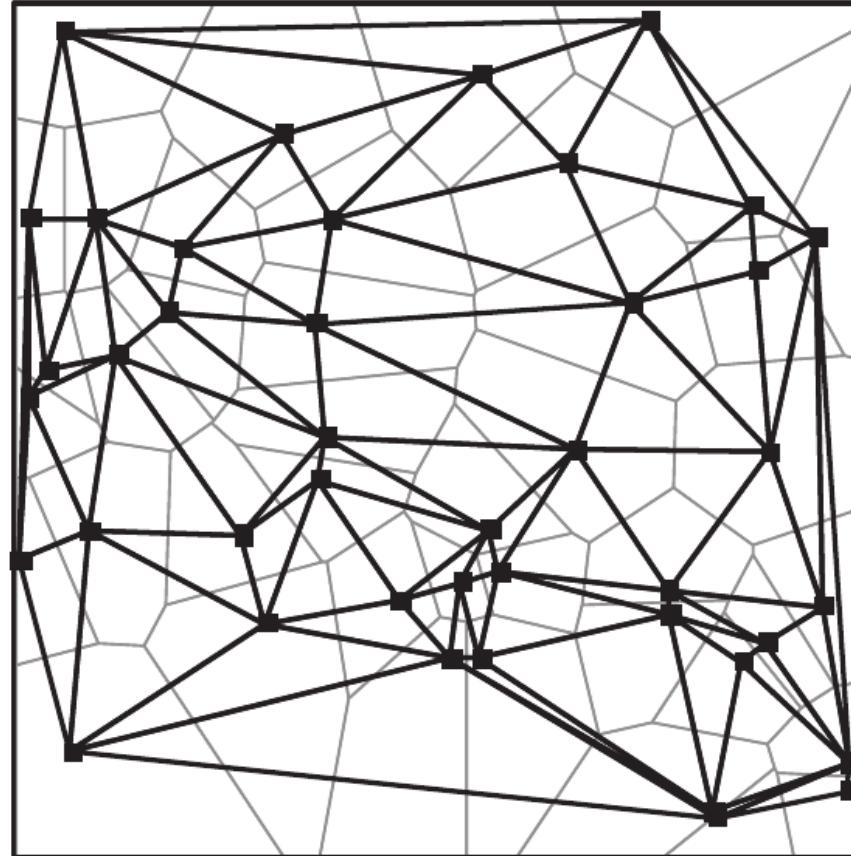
Voronoi/Thiessen polygons

Adjacency

Interaction

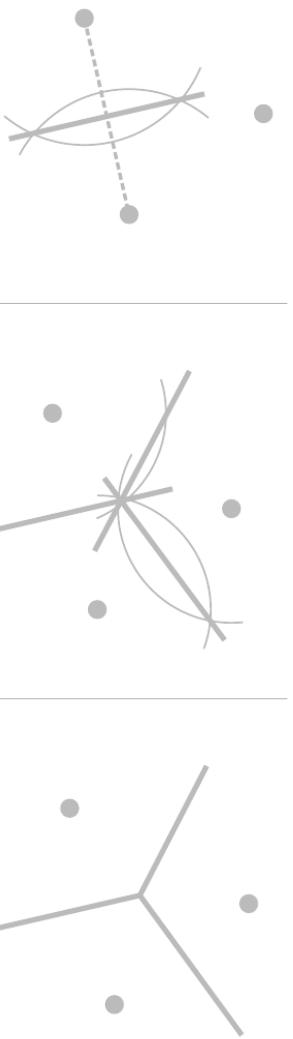
Distance

Neighborhood



TOPOI

Proximity polygon sharing an edge....Delaunay triangulation



(O'Sullivan & Unwin, 2010)

Spatial data are special

Pitfalls of spatial data:

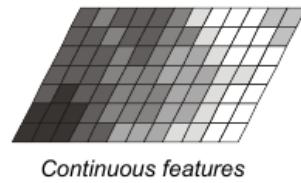
- spatial autocorrelation
- modifiable areal unit problem
- scale
- ecological fallacy
- non-uniformity of space
- edge effects

Potentials of spatial data:

- distance
- adjacency
- interaction
- neighborhood

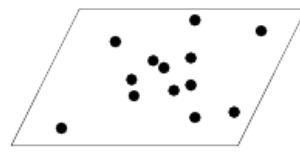
...it's all about processes

GEOSTATISTICS



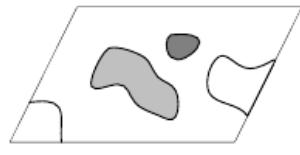
Continuous features

POINT PATTERN ANALYSIS



Point objects

LATTICE STATISTICS



Areal objects (polygons)

However you call it and focus on...

Everything is about processes...

Because they are the cause of spatial patterning,
Autocorrelation, ...

Fig. 1.2: Spatial statistics and its three major subfields after
Cressie (1993).

https://userpage.fu-berlin.de/~dare211/VS_Modellierung/Intro_R.html