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Modelling Socio-Economic Processes with Minimalistic Agents

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Schedule

1. Complex vs. Minimalistic Agents
2. Communication Field
3. Model of Economic Aggregation
4. Computer Simulations
5. Conclusions

Complex System

“By complex system, it is meant a system comprised of a (usually large) number of (usually strongly) interacting entities, processes, or agents, the understanding of which requires the development, or the use of, new scientific tools, nonlinear models, out-of equilibrium descriptions and computer simulations.”

Journal “Advances in Complex Systems”

Multi-Agent Systems (MAS)

agent:

- ◇ subunit with “intermediate” complexity

⇒ may represent local processes, individuals, species, agglomerates, components, ...

multi-agent system:

- ◇ *large number / different types* of agents
- ◇ interactions between agents:
 - on *different spatial and temporal scales*
 - local / direct interaction
 - global / indirect interactions (coupling via resources)

Complex Agent

- ◇ autonomous; knowledge based / behavior based rules
- ◇ performs complex actions: BDI, rational choices, ...
- ◇ specialization, learning, genetic evolution, ...

1. *Problem*: information flow

“rational agent” (economics):

- ◇ complete knowledge of all possible actions and their outcomes (or known probability distribution over outcomes)
- ◇ common knowledge assumption

2. *Problem*: combinatoric explosion of the state space

- ◇ 1000 Agents with 10 rules $\Rightarrow 10^{13}$ possibilities
- ◇ *Solution*:
restrict interactions \Rightarrow control of information flow
personally addressed interaction instead of “broadcasting”
- ◇ *freedom*: define rules *and* interactions \Rightarrow *pitfall*

Minimalistic Agent

- ◇ possible simplest set of rules \Rightarrow “sufficient” complexity
(depends on the system considered)
- ◇ *functional information*:
simple *algorithm*, which is steadily repeated
- ◇ *structural information*:
external information (*data*) received by the agent
- ◇ *pragmatic information*: \Rightarrow effective information
 - emerges from the processing of the data by the algorithm
 - specific for each agent \Rightarrow enables actions, decisions

Characteristics:

- ◇ *internal degree of freedom*: θ_i (e.g. $-1, +1$)
 \Rightarrow enables different actions
- ◇ mobility (active walker, active Brownian particle)
- ◇ deterministic and stochastic influences (Langevin equation):

$$\frac{dr_i}{dt} = \alpha_i \left. \frac{\partial h^e(r, t)}{\partial r} \right|_{r_i} + \sqrt{2\varepsilon_i} \xi(t)$$

- local response of agent i to *structural information*
- α_i : weights the importance of the information received
- ε_i : individual sensitivity
- $h_e(r, t)$: effective communication field \Rightarrow components:

$$\nabla_i h^e(r, t) = \nabla_i h^e(\dots, h_\theta(r, t), h_\theta(r, t), \dots)$$

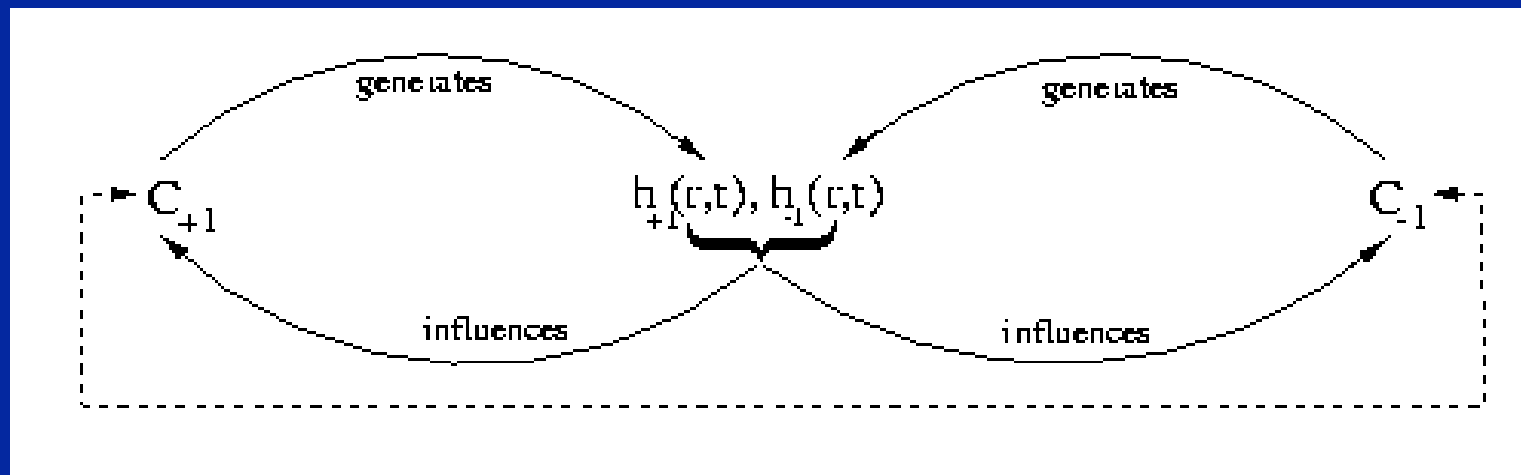
Communication Field

$$\frac{\partial}{\partial t} h_{\theta}(r, t) = \sum_{i=1}^N q_i(\theta_i, t) \delta_{\theta, \theta_i} \delta(r - r_i) - k_{\theta} h_{\theta}(r, t) + D_{\theta} \Delta h_{\theta}(r, t)$$

multi-component spatio-temporal field:

- ◇ reflects *spatial* distances between agents (r_i)
- ◇ permanent *local* individual contribution: $q_i(\theta_i, t)$
- ◇ information generated has a certain life time, $1/k_{\theta}$
⇒ considers memory effects
- ◇ information can spread throughout the system, D_{θ}
⇒ determines how fast information is distributed

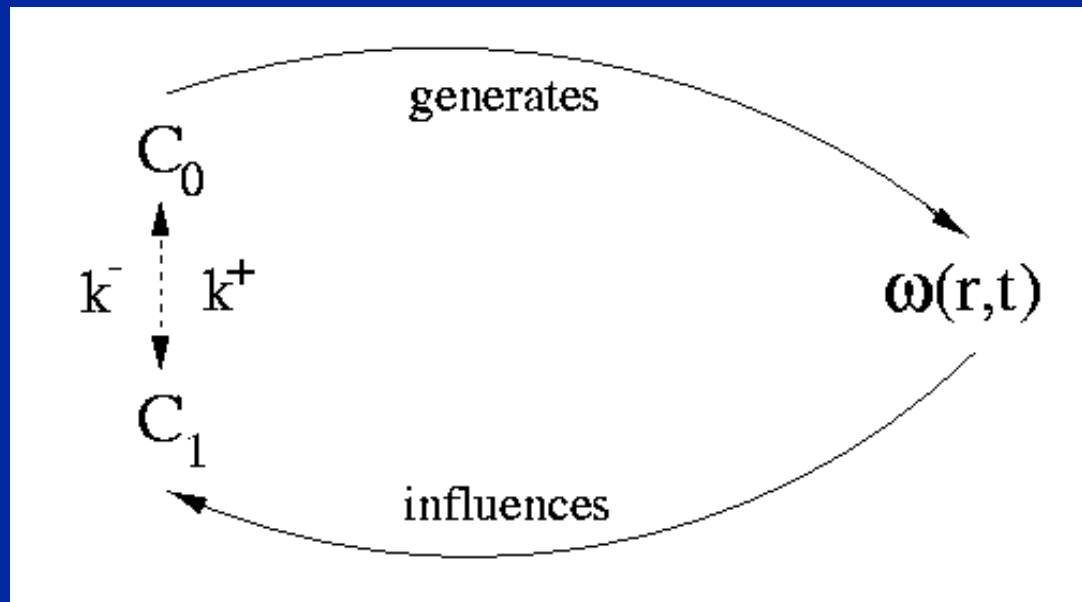
Example: Social Interaction



- ◇ agents with two different opinions: $\theta_i \in \{+1, -1\}$
- ◇ two types of strategies:
 - (1) migration: $h^e(r, t) \Rightarrow$ different types of response
 - (2) change of opinion

$$w(\theta_i|\theta_i) = \eta \exp\{[h_{\theta}(r_i, t) - h_{\theta}(r_i, t)]/\varepsilon_i\}$$

Example: Economic Agglomeration



- ◇ $\theta = 0$: *employed agents*, C_0 (immobile) $\Rightarrow l(\mathbf{r}, t)$
- ◇ $\theta = 1$: *unemployed agents*, C_1 (mobile) $\Rightarrow n(\mathbf{r}, t)$
- ◇ migration due to spatial wage differences: $\omega(r)$
- ◇ “hiring” and “firing”: k^+ , k^-

Economic Assumptions

- ◇ wage: marginal product of labor:

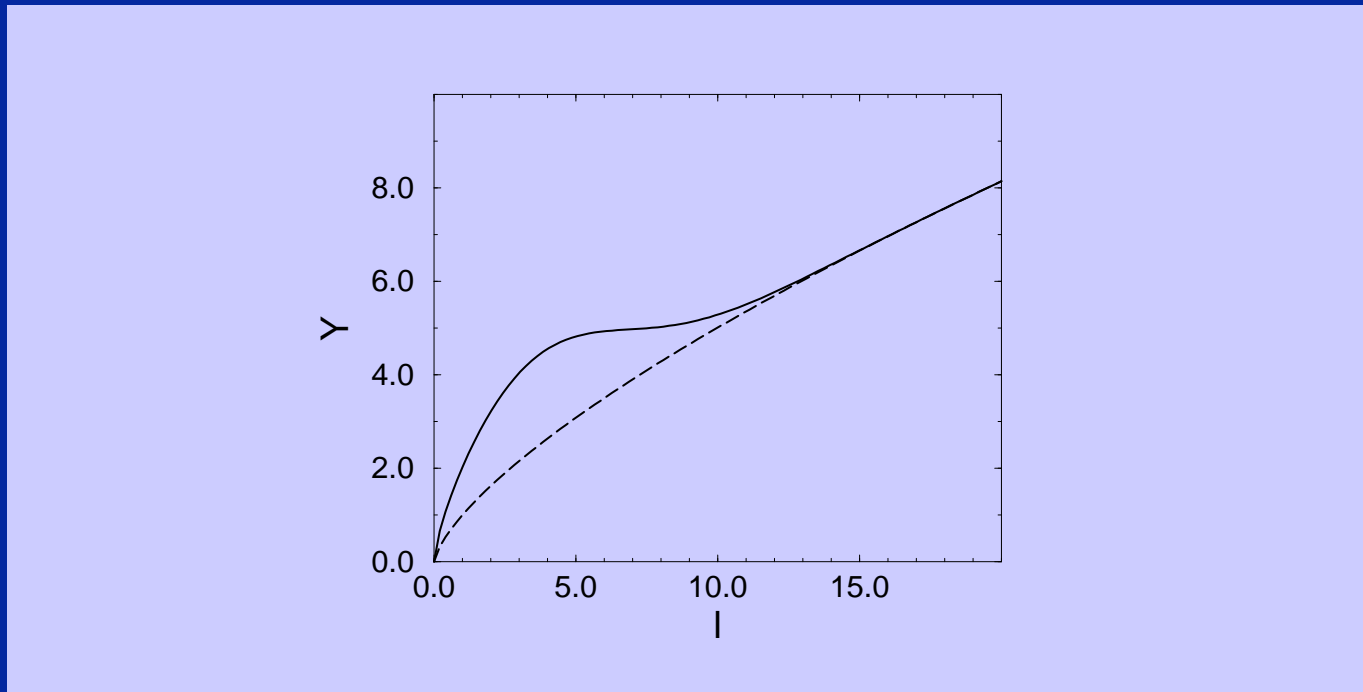
$$w\{l(\mathbf{r}, t)\} = \frac{\delta Y\{l(\mathbf{r}, t)\}}{\delta l}$$

- ◇ Cobb-Douglas production function

$$Y\{l(r, t)\} = A l^\beta(r, t), \quad \beta < 1$$

- ◇ A : represents level of productivity
considers *cooperative effects* resulting from interactions
among the workers \Rightarrow non-linear function

$$Y(l) = \frac{\bar{A}}{2} [1 + \exp(a_1 l + a_2 l^2)] l^\beta$$



- ◇ $a_2 < 0$: saturation effects \Rightarrow advantages of cooperative effects compensated by disadvantages of crowding

“hiring” and “firing” rates: ω^* : minimum wage

hiring rate k^+ : firms hire workers as long as $\frac{\delta Y}{\delta l} > \omega^*$
 (maximum profit condition)

$$k^+ = k^+ \{l(\mathbf{r}, t)\} = \eta \exp \left\{ \frac{\delta Y \{l(\mathbf{r}, t)\}}{\delta l} - \omega^* \right\}$$

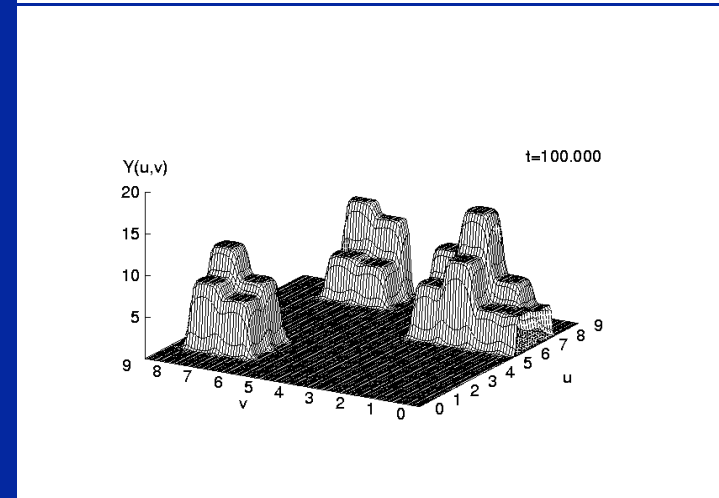
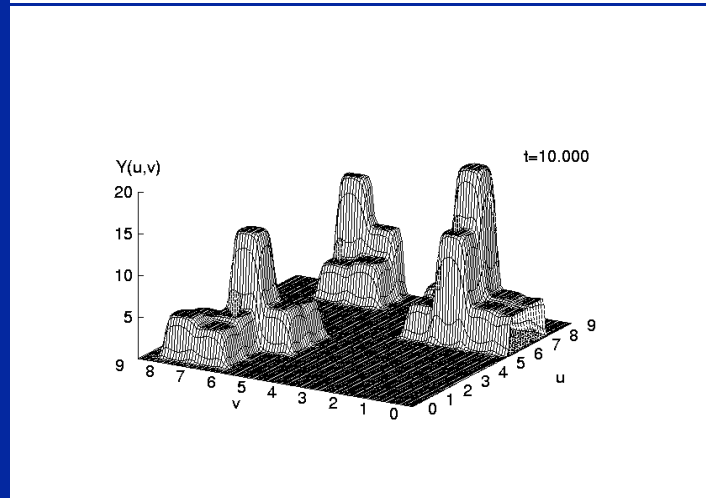
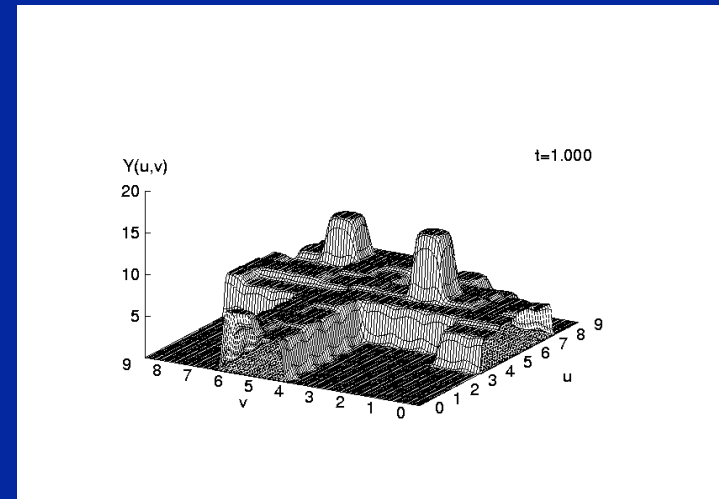
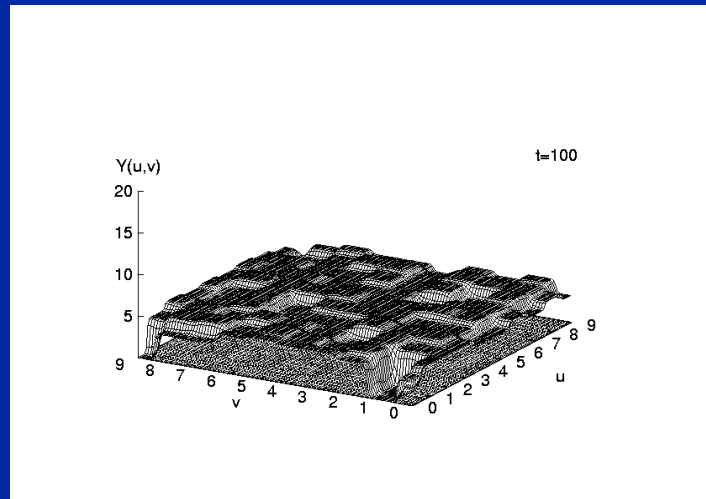
firing rate k^- : external and internal reasons

(i) workers are fired if $\frac{\delta Y}{\delta l} < \omega^*$

(ii) workers can quit their job for better opportunities

$$k^- = k^- \{l(\mathbf{r}, t)\} = \eta \exp \left\{ - \left[\frac{\delta Y \{l(\mathbf{r}, t)\}}{\delta l} - \omega^* \right] + c \frac{\partial \omega(\mathbf{r})}{\partial r} \right\}$$

Spatial distribution of production



Results of Computer Simulations

$t = 0$: random initial distribution

$t = 100.000$: distinct *extended major economic regions*

(i) $t < 1.000$:

- ◇ coexistence of numerous small economic centers
basis: cooperative effects, mutual stimulations

(ii) $t > 1.000$:

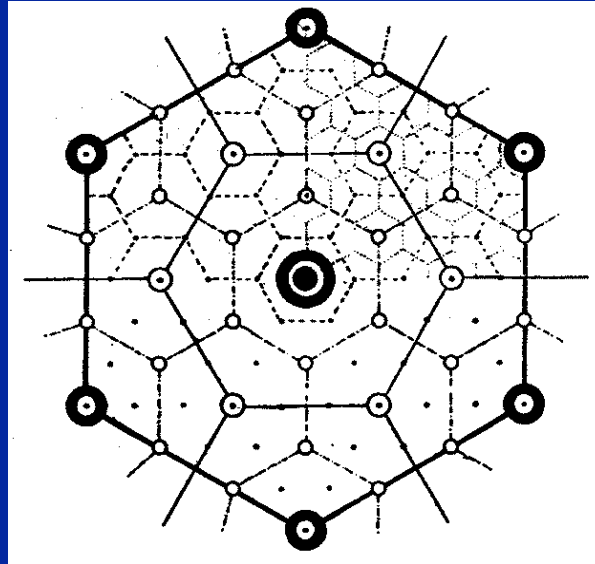
- ◇ some small centers overcome economic bottleneck
⇒ increase of marginal output
- ◇ competition: *local* attraction of labor force at the expense of the former small economic centers

Final stage: distinct extended economic regions

- (i) *stable coexistence* of the major economic regions
reason: *critical distance*
 - ◇ each economic center has its own attraction/supply area
⇒ prediction of the *central place theory*

- (ii) *quasi-stationary non-equilibrium* within the major economic regions
 - ◇ each economic region consists of some *subregions*
centers do not have same number of employed agents
⇒ still follow a stochastic eigendynamics

Central Place Theory



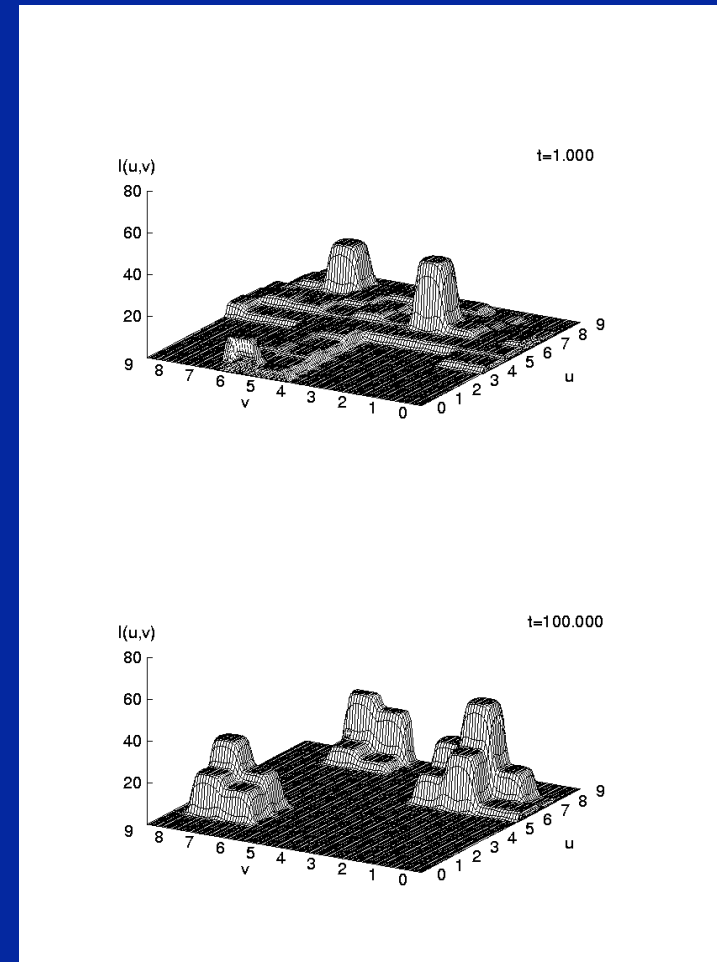
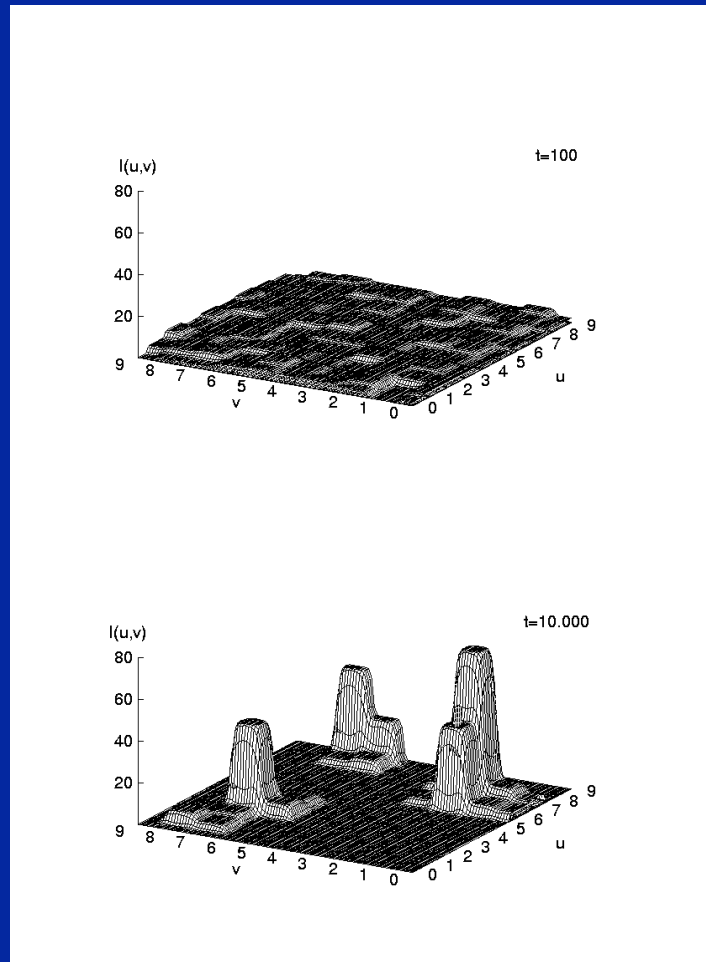
Walter Christaller: *Die zentralen Orte in Süddeutschland.*

Eine ökonomisch-geographische Untersuchung über die Gesetzmäßigkeit der Verbreitung und Entwicklung der Siedlungen mit städtischen Funktionen, Jena: Fischer, 1933
(Reprint: Darmstadt: Wissenschaftliche Buchgesellschaft, 1980)

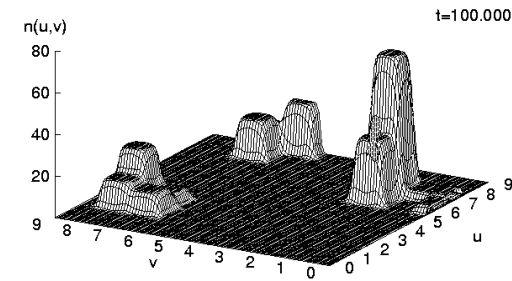
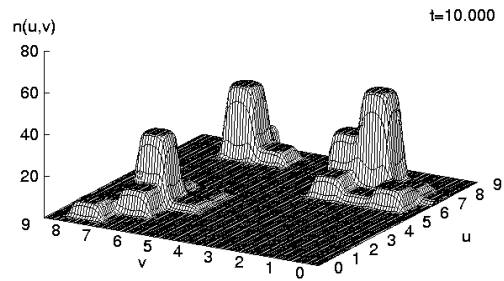
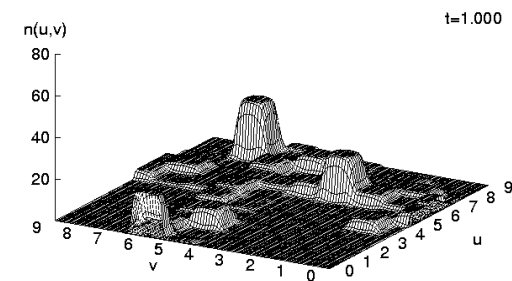
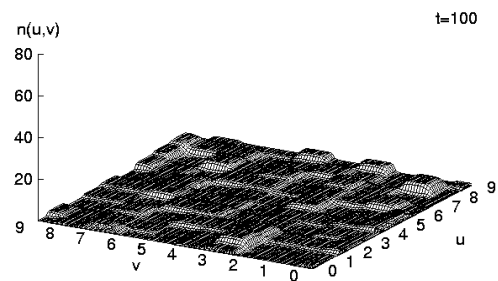
English translation by C.W. Baskin:

Central Places in Southern Germany, London: Prentice Hall, 1966

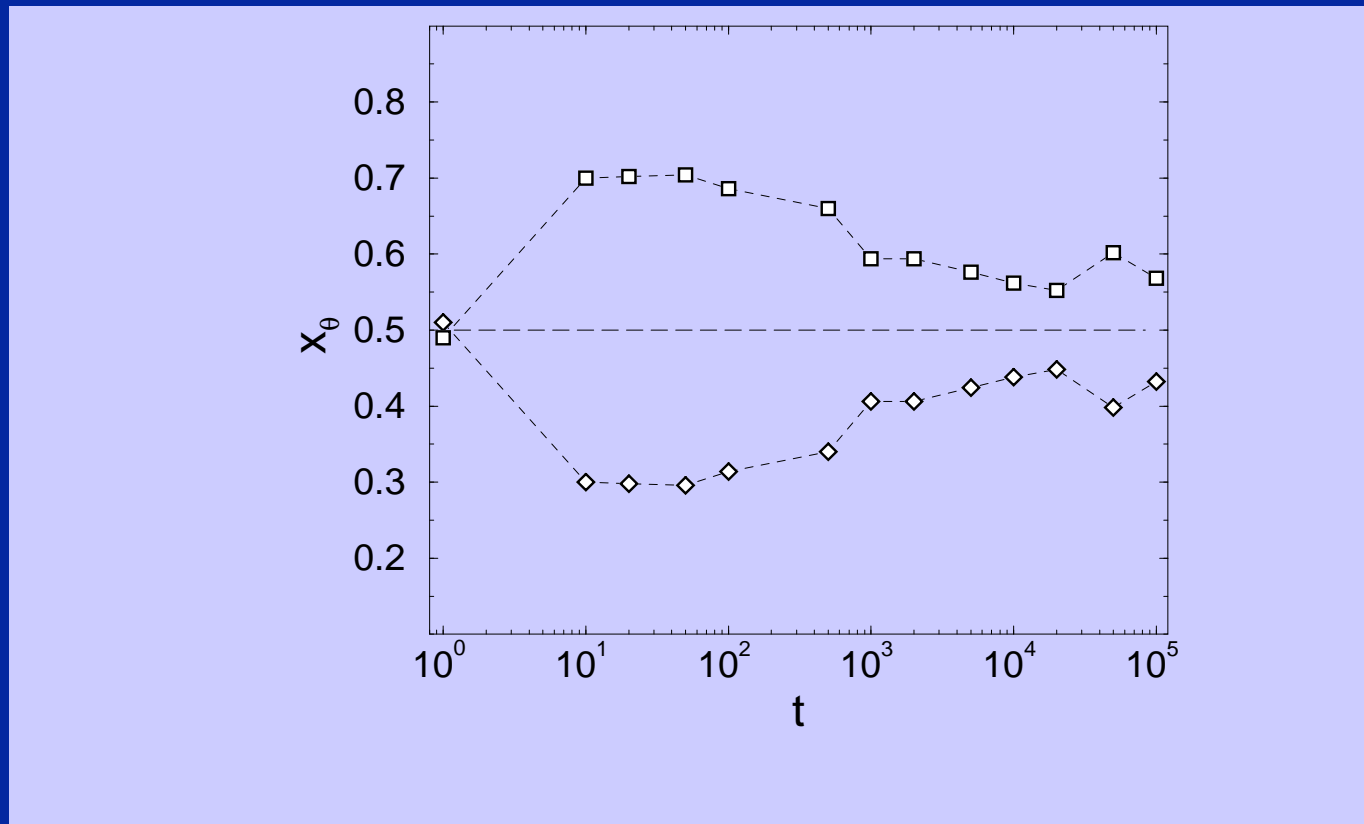
Spatial density of employed agents



Spatial density of unemployed agents



Total share $x_\theta \equiv N_\theta/N$



employed agents: (\square)

unemployed agents (\diamond)

Spatio-Temporal Evolution of Unemployment

(i) small scale production

- ◇ significant higher share of employed agents (~ 70 percent)
broadly distributed

(ii) large scale production

- ◇ increase of wage \Rightarrow affects migration of unemployed agents
concentrate in the productive regions
- ◇ important for the further growth: agents to hire
- ◇ concentration of employed *and* unemployed agents in the
same regions
- ◇ new (larger) centers employ 60 percent of the agents
 \Rightarrow *increase of unemployment*

Conclusions

Minimalistic multi-agent system: two levels of description

Agent's level:

- ◇ simplex - not complex \Rightarrow “intermediate” complexity
- ◇ *local* response/changes of the environment
- ◇ internal degrees of freedom \Rightarrow variety of responses
- ◇ parallel *independent* actions *coupled* by information flow

System's level:

- ◇ emergence of complexity
- ◇ adaption to changes in the environment
- ◇ failure tolerant (single agents, external perturbations)

⇒ Reason: *Self-Organization*

solutions result from the non-linear interaction

boundary conditions (semi-structured environment)

Characteristics of the approach:

- ◇ *non-deterministic*
random events (fluctuations) play a considerable role
- ◇ *non-finalistic*
final (global) solutions cannot be predicted from local interactions \Rightarrow solutions *emerge* \Rightarrow *path dependent*
- ◇ *bottom-up approach*: create a solution \Rightarrow self-organization
top-down approach: design a solution \Rightarrow planning

Self-Organization

Self-Organization is the process by which individual subunits achieve, through their cooperative interactions, states characterized by new, emergent properties transcending the properties of their constitutive parts.

Biebricher, C. K.; Nicolis, G.; Schuster, P.: *Self-Organization in the Physico-Chemical and Life Sciences*, EU Report 16546 (1995)

Self-organization is defined as spontaneous formation, evolution and differentiation of complex order structures forming in non-linear dynamic systems by way of feedback mechanisms involving the elements of the systems, when these systems have passed a critical distance from the statical equilibrium as a result of the influx of unspecific energy, matter or information.

SFB 230 “Natural Constructions”, Stuttgart, 1984 - 1995