

# Extracting information from regional data using Self-Organizing Maps

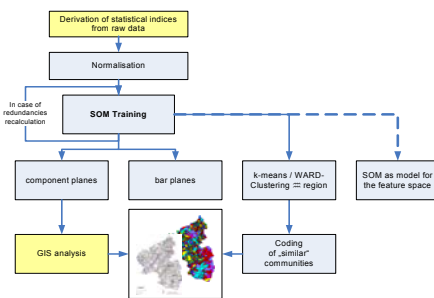
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## Problem

The extraction of information from regional statistical data involves processing high-dimensional data sets (i.e. with many variables) in such a way that similarities between the individual spatial units and relationships in their characteristics become accessible to our cognitive abilities.

## Data and processing approach

- birth rate per 1000
- population density
- age ratio ( $\geq 60 / \leq 18$ )
- population dynamics
- commuter balance
- per capita income
- per capita expenses
- hotels per 1000
- agricultural holdings
- fraction of agricultural areas
- business enterprises per 1000

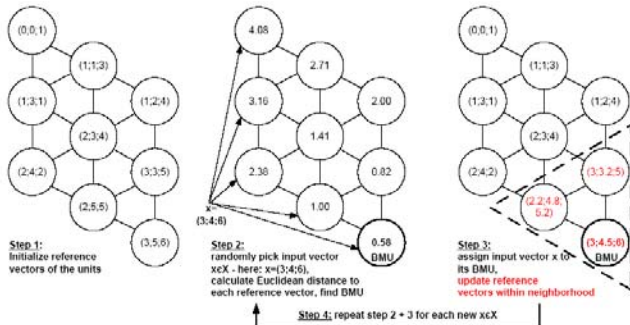


## Training of a Self-Organizing Map

A Self-Organizing Map (SOM), consisting of nodes  $i=1...k$  which are characterized by a vector

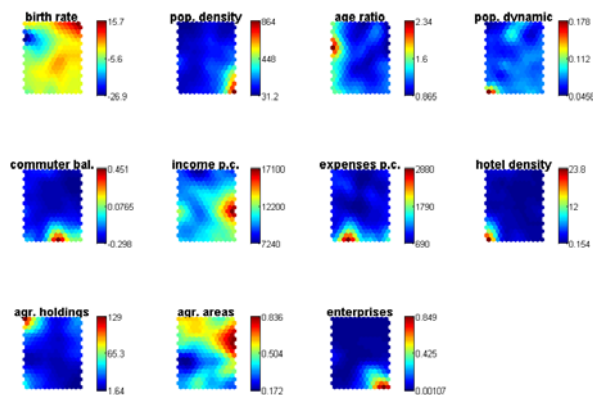
$$m_i = [\mu_{i1}, \mu_{i2}, \dots, \mu_{in}]^T \in \mathcal{R}^n$$

with the same dimensionality as the data sets  $x \in X$  is trained:

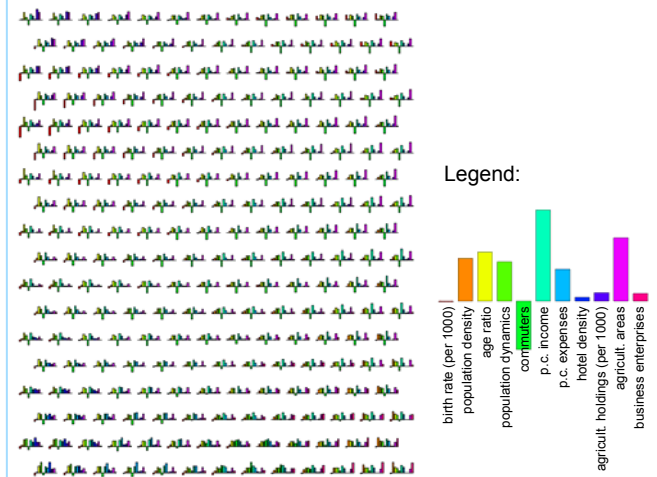


## Results

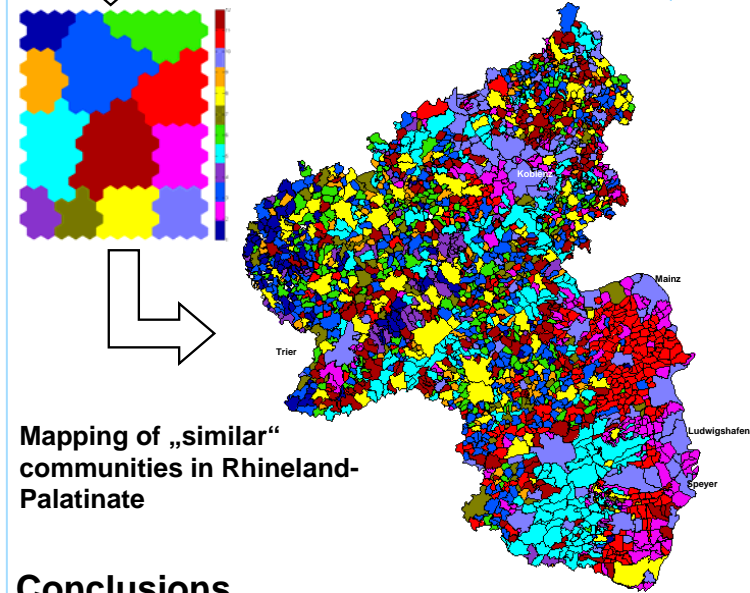
„Component planes“ (i.e. distribution for individual variables)



## „Bar plane“



## K-means clustering of SOM mapping (12 clusters)



## Mapping of „similar“ communities in Rhineland-Palatinate

## Conclusions

- SOM allows topologically ordered projection of high-dimensional data by similarity (i.e. similarity in data characteristics translates into distance on the SOM)
- this allows an intuitive exploration of regional statistical data, e.g. „correlation hunting“
- k-means clustering helps delineating „similar“ regions on the SOM for subsequent mapping in space
- the clusters can be interpreted as urbanized regions, forested regions, rural areas, peripheral communities etc.
- it can be seen that spatial vicinity does not always imply similarity of socio-economic characteristics
- the SOM also allows extrapolating new data sets

## Further Information on Self-Organizing Maps and their application

This work was carried out using the SOM-Toolbox for Matlab by the "SOM Toolbox Team", Helsinki University of Technology (<http://www.cis.hut.fi/projects/somtoolbox>)

Kohonen, T.: Self-Organizing Maps, 3rd ed., Berlin, Heidelberg, New York, 501pp., 2001.

Herbst, M., Casper, M.C.: Towards model evaluation using Self-Organizing Maps, Hydrol. Earth Syst. Sci., 12, 657-667, 2008.

Herbst, M., Gupta, H.V. and Casper, M.C.: Mapping model behaviour using Self-Organizing Maps, Hydrol. Earth Syst. Sci., 13, 395-409, 2009.

Herbst, M., Casper, M.C., Grundmann, J. and Buchholz, O.: Comparative analysis of model behaviour for flood prediction purposes using Self-Organizing Maps, Nat. Hazards Earth Syst. Sci., 9, 373-392, 2009.

