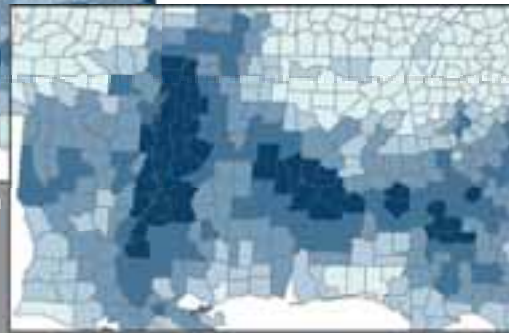


Jenks Classification



Autocorrelation based Regioclustering

Spatially aware Data Classification for Choropleth Maps

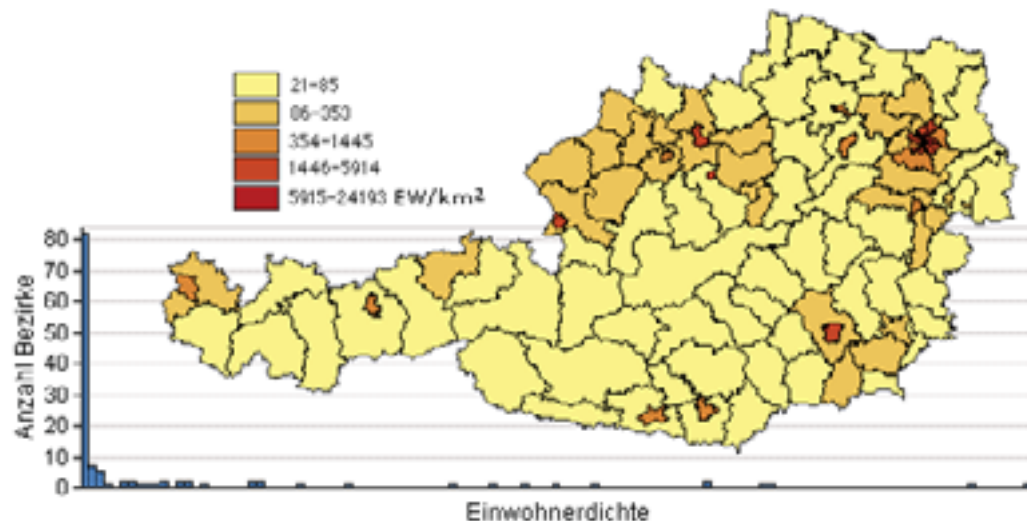
Esri International User Conference, July 15th 2014

Christoph Traun

What a choropleth map should do...

Choropleth maps should

- provide an overview of the statistical distribution of the data.
- allow to identify approximate attribute values for each location, and
- **provide an image of the spatial configuration of the mapped phenomenon by focusing on boundary lines between different classes/shadings.**

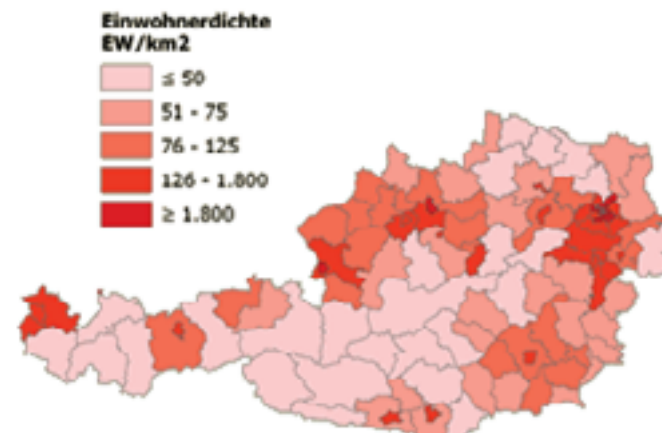
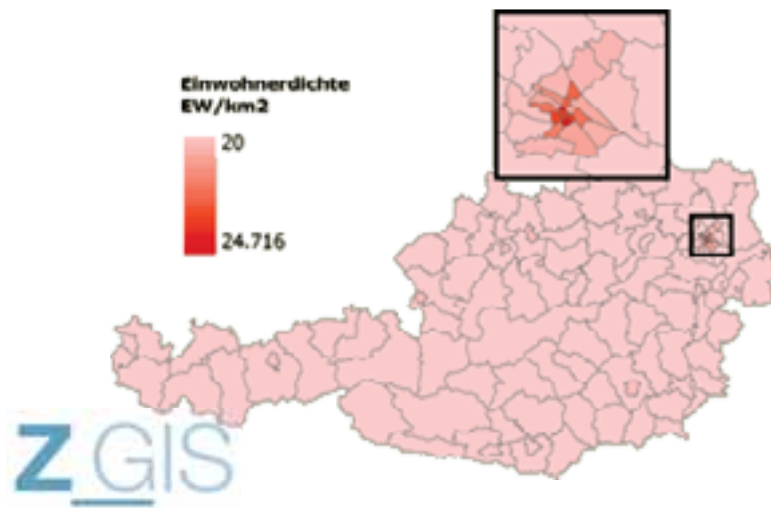


BEZIRK	Dichte
AMSTETTEN <305>	351
BADEN <306>	167
ELLIDENZ <801>	47
BRALNAJ AM INN <404>	91
DREGENZ <002>	140
BRUCK AN DER LEITHA <307>	80
BRUCK AN DER MUR <602>	49
DEUTSCHLANDSBERG <603>	71
DORNBIRN <803>	437
EFERDING <405>	118

What data-classification should do...

Data classification should

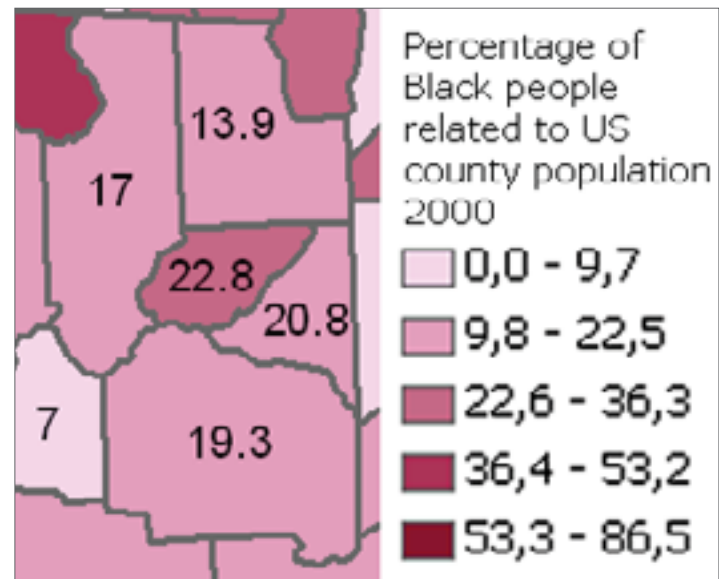
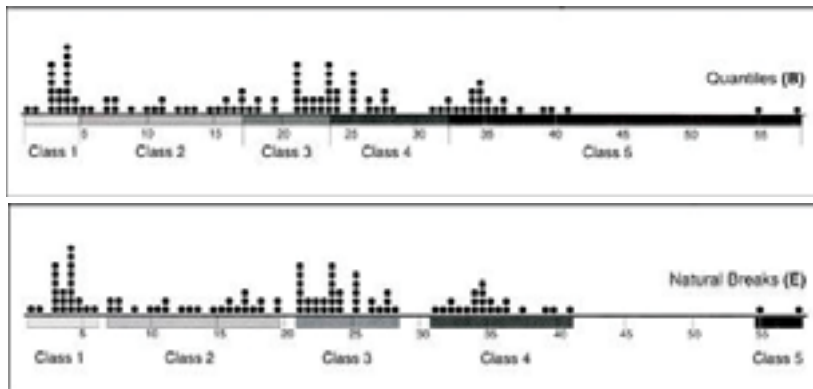
- reduce (map) complexity to enhance readability/perception
- enable the unambiguous visual relation between a map feature shading and the corresponding value (range).
- enable the comparison of multiple datasets (time series, ...), and
- help to properly reflect data inherent spatial patterns



What data classification does...

Classification routines implemented in GIS and mapping software

- group data into classes based on the distribution of attribute values along the number line
 - Equal Interval
 - Quantile
 - Optimal (natural breaks)
 - ...



- ...and completely ignore the spatial context of data.

There is a gap!

Autocorrelation Based Regioclassification

Choropleth maps should clearly communicate the main spatial patterns. They should be optimized in this respect.

Classification should help to properly reflect data inherent spatial patterns and reduce map complexity

classification
ignoring
spatial
context

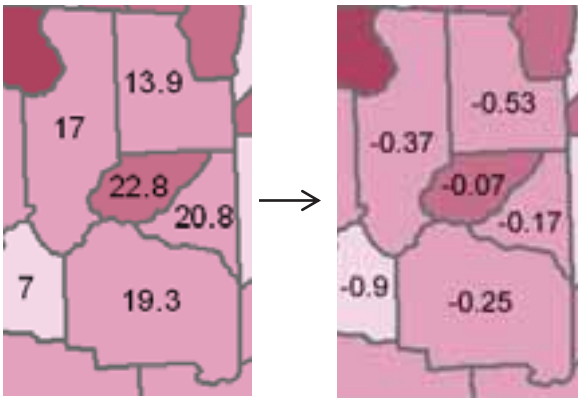
Autocorrelation based Regioclassification – Step 1

[Moran's I Scatterplot: see Luc Anselin 1995]

1. z-standardizing all values:
 $z_i = (x_i - \text{mean}) / \text{standard deviation}$

Overall Mean (n= 755): **24.1**

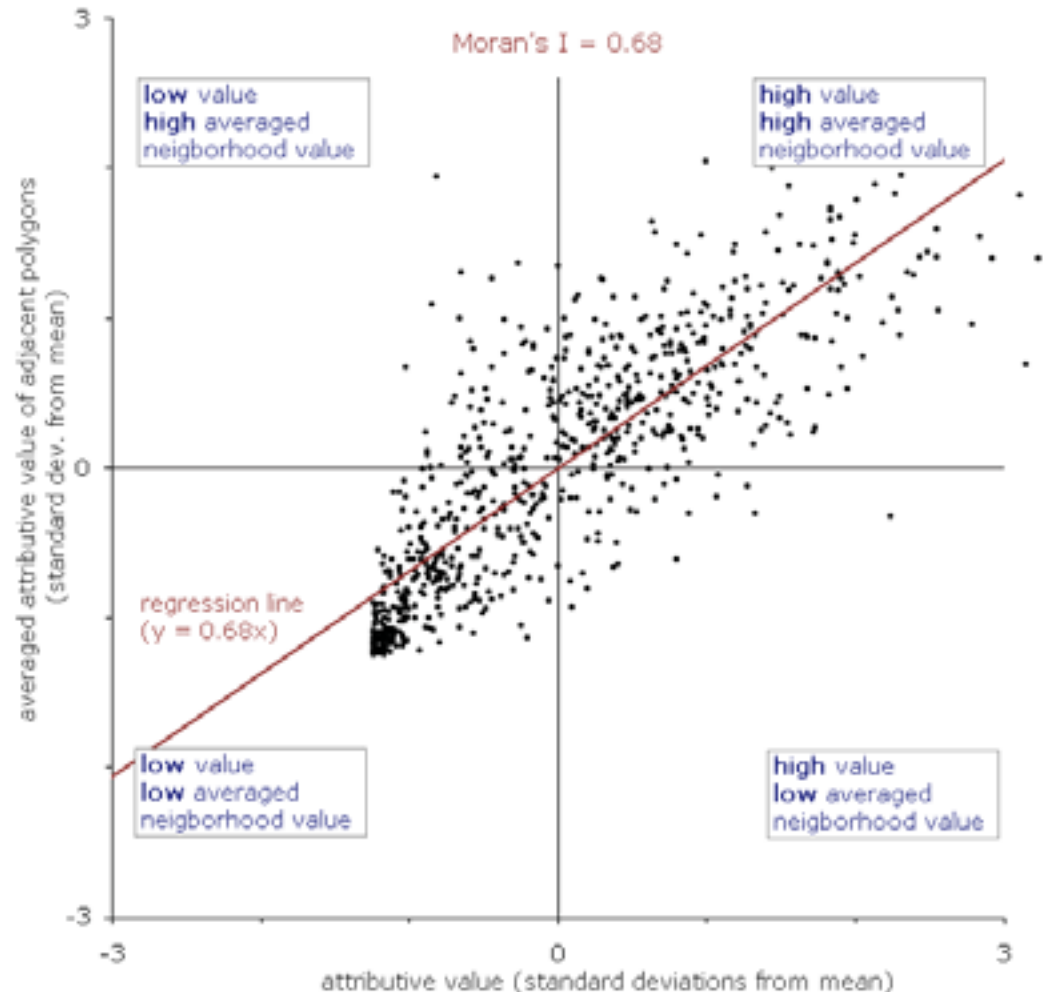
Overall Std.Dev.: **19.3**

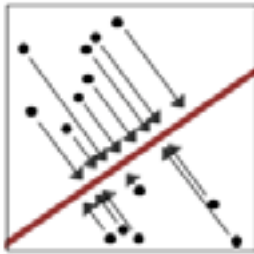


2. Calculate mean value of local neighborhood (adjacent polygons):

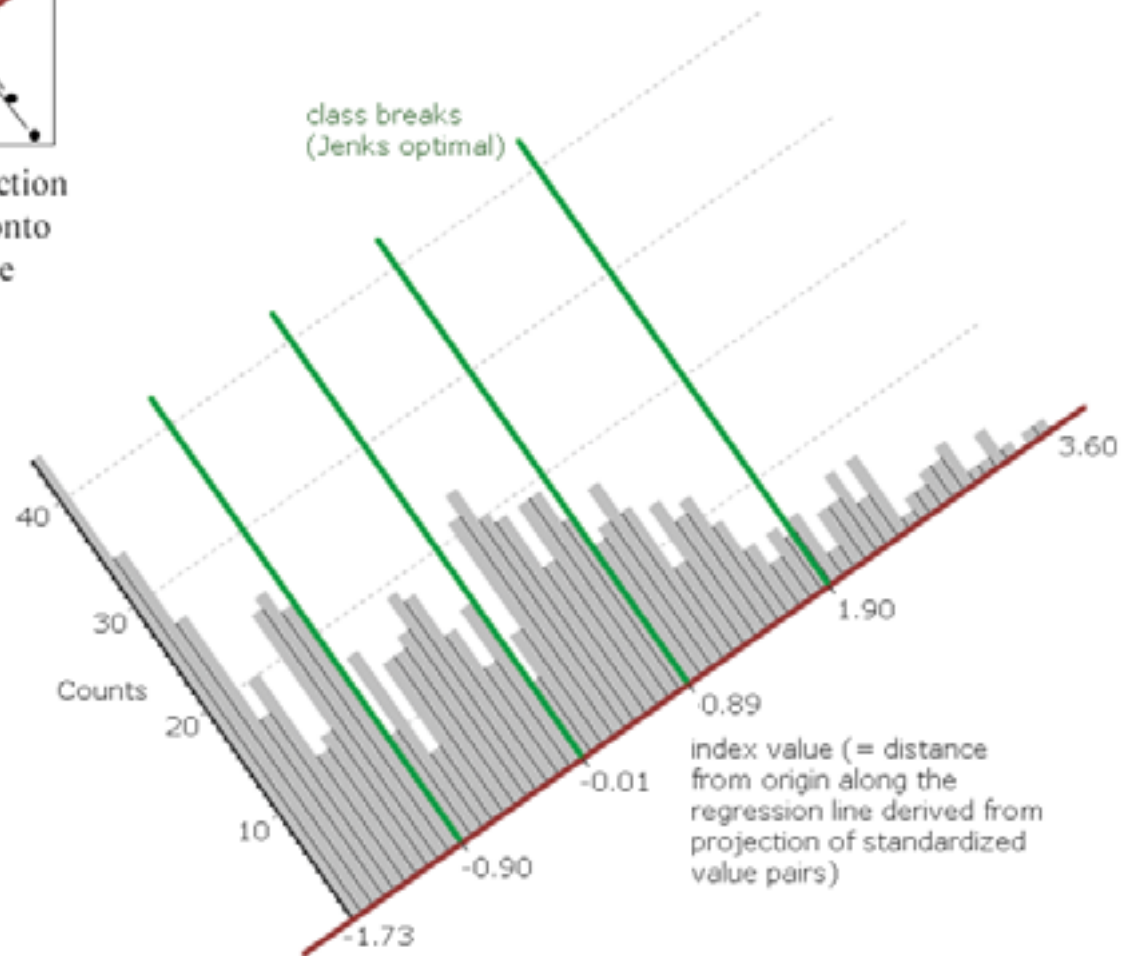
$$(-0.37 + -0.53 + -0.17 + -0.25) / 4 = \mathbf{-0.33}$$

(Example for polygon $z = -0.07$)



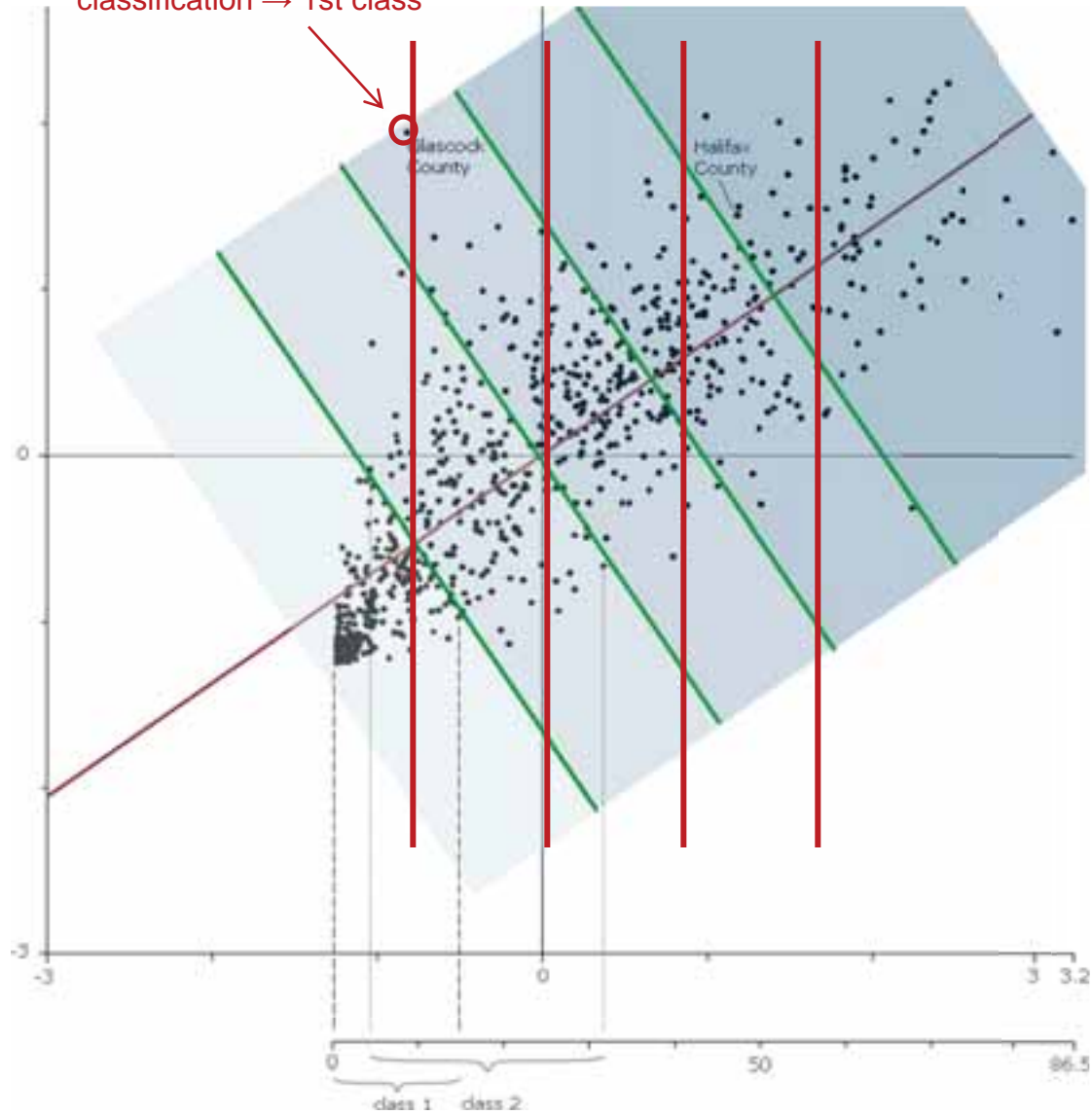


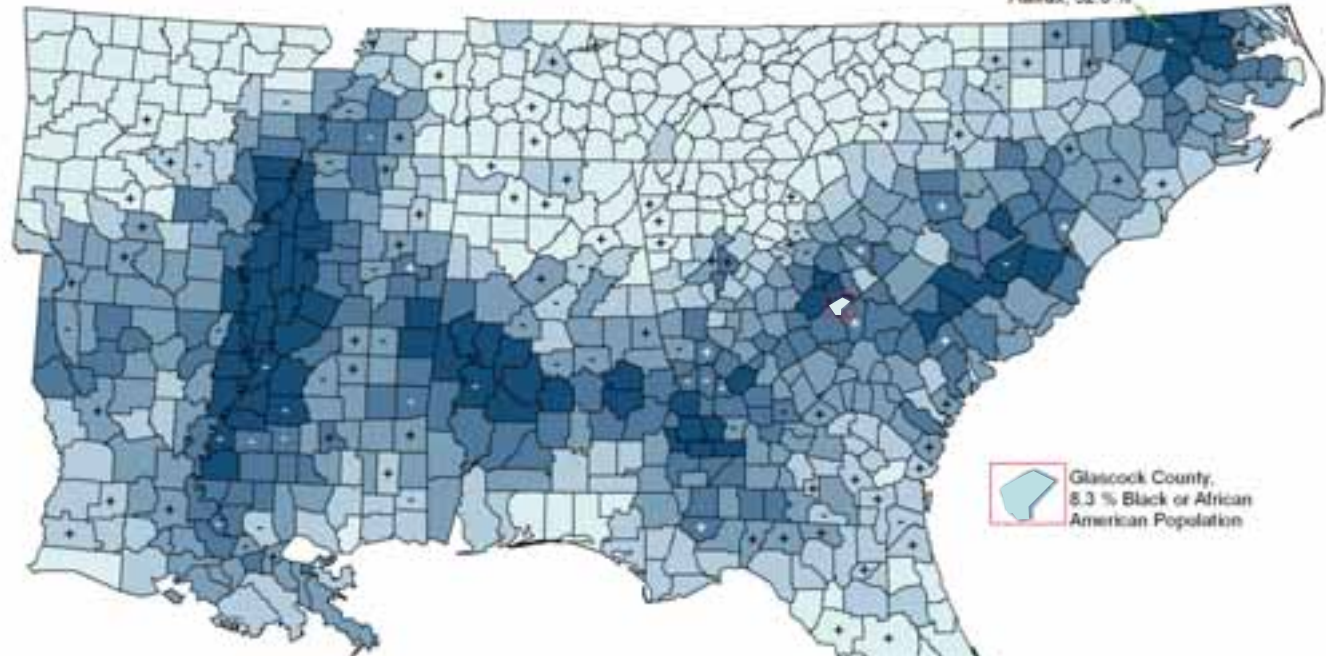
Step 2: Projection of all points onto regression line (detail view)



Step 3: Determination of class breaks based on the distribution of projected, standardized value pairs

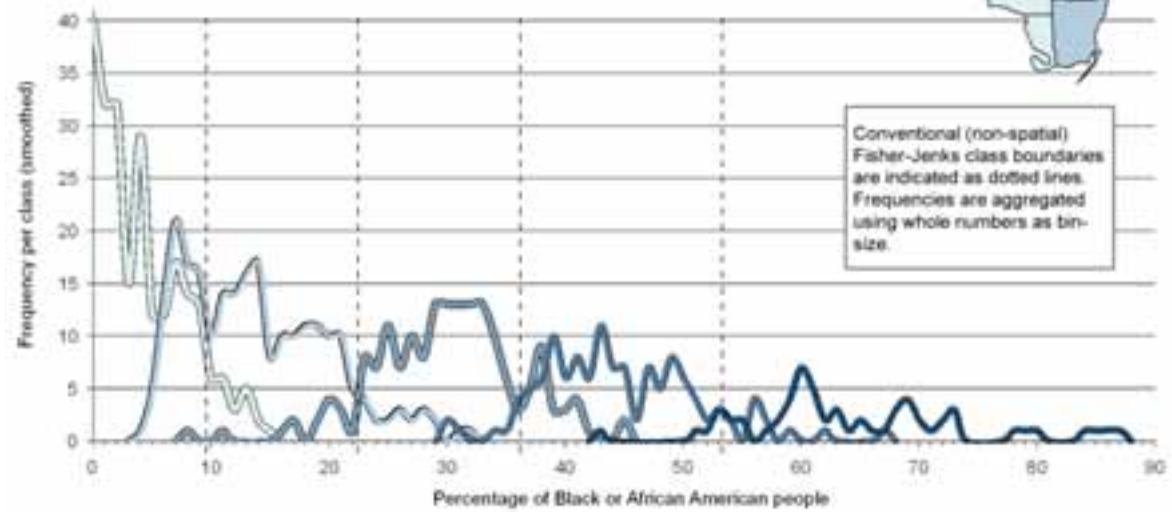
LISA outlier → conventional
classification → 1st class

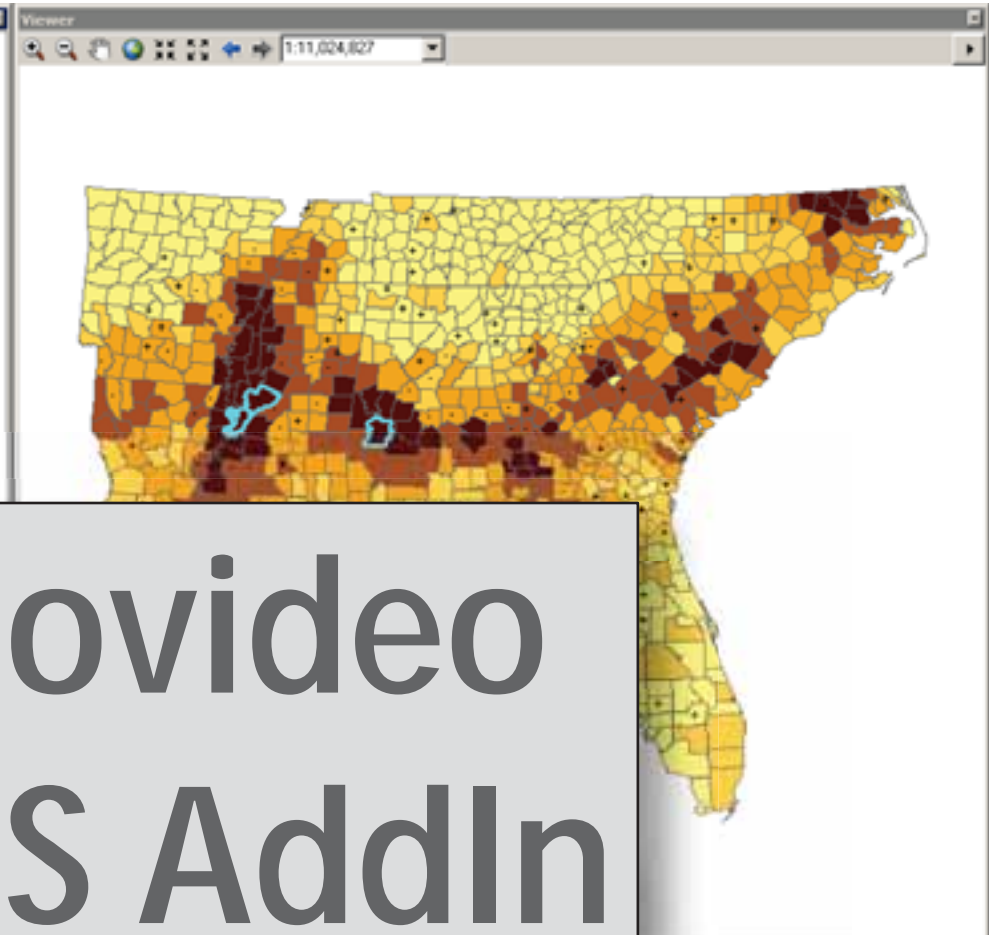
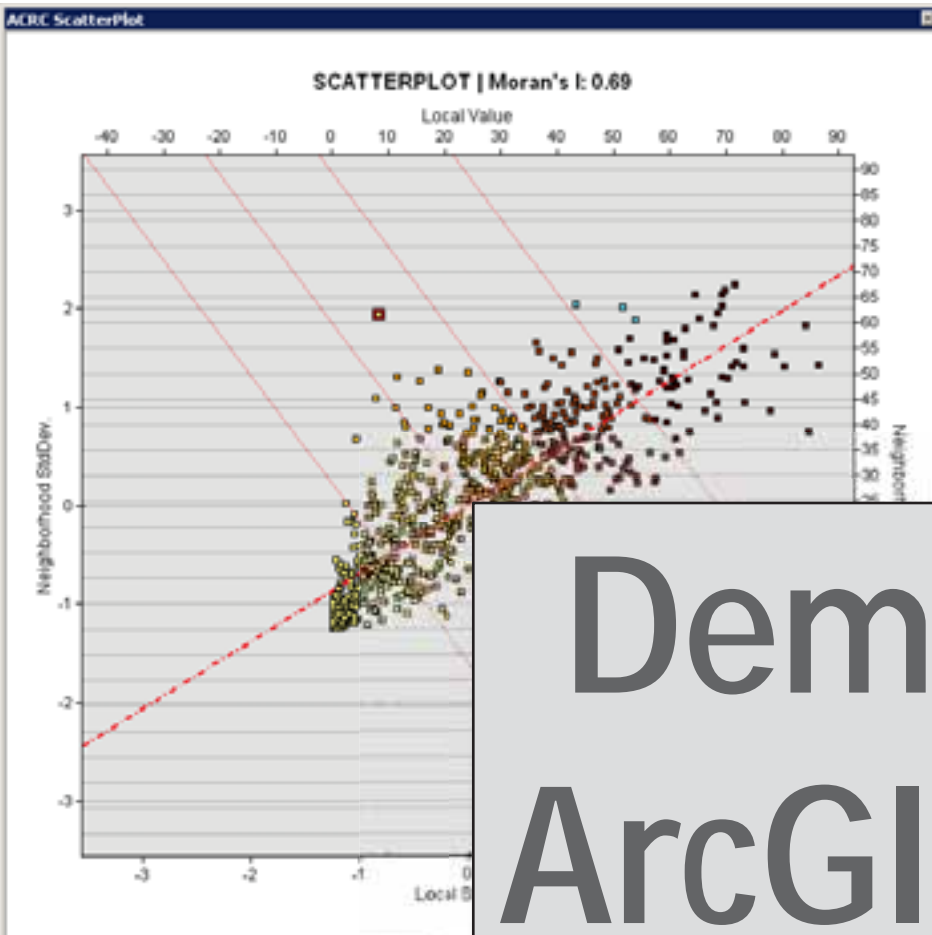




Glascock County,
8.3 % Black or African
American Population

Percentage of Black or African American people related to total county-population in 2000:





Demovideo ArcGIS AddIn

ACRC - ACRC_rel_black

Statistics Classification Graphs

Classification [ACRC_rel_black]

Method (Advanced - Read help before changing)

Natural Breaks (Jenks)

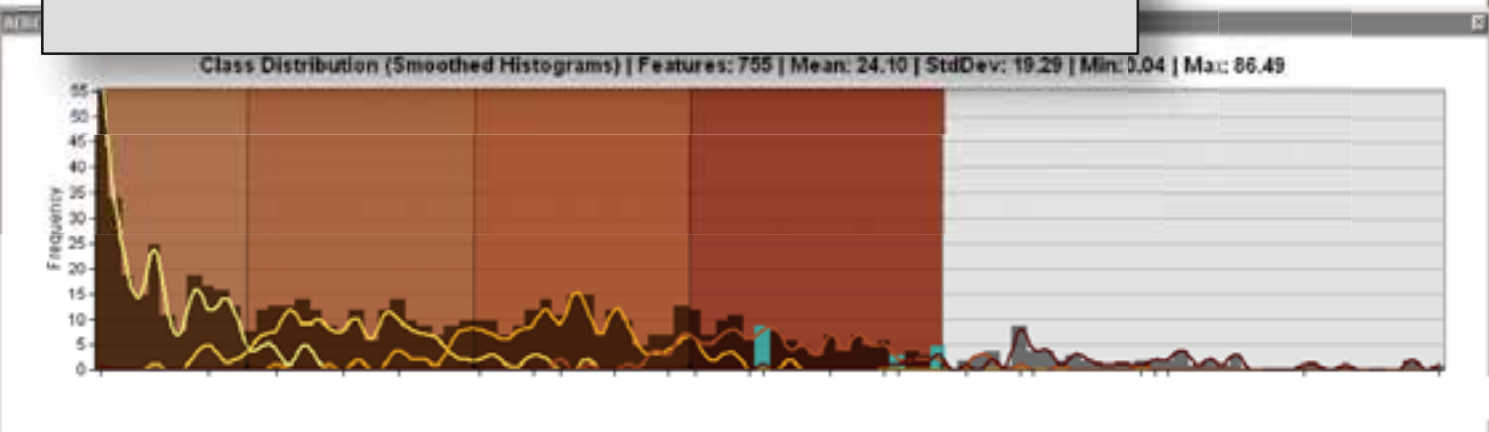
Classes

5

Exclude Outliers No Gaps in Class Labels

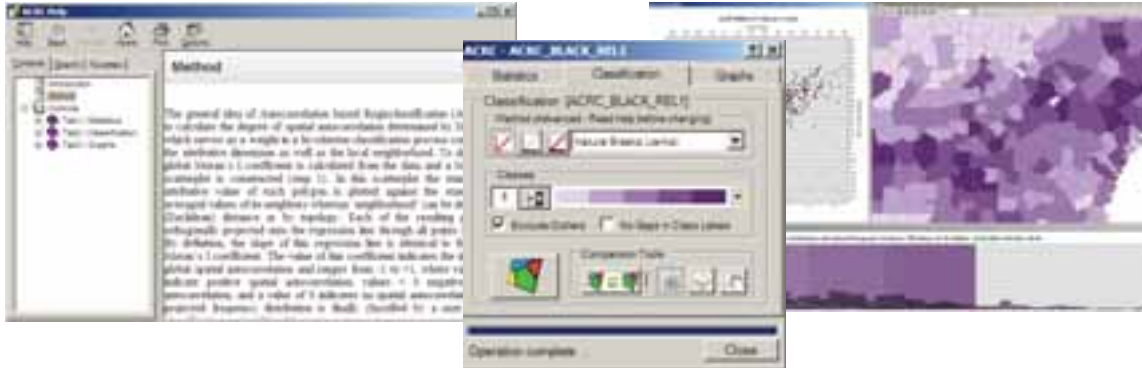
Comparison Tools

Operation complete ... Close



Ressources and further research

- [Download ArcGIS AddIn](#) (Author: Christoph Mayrhofer, University of Salzburg)



- Publications:
 - Traun, C. & Loidl, M. (2012): Autocorrelation-Based Regioclassification - A self-calibrating classification approach for choropleth maps explicitly considering spatial autocorrelation. In: International Journal of Geographical Information Science 26(5), p.923-939.
 - Loidl, M. & Traun, C. (2013): The Effect of ACRC on the Results of Cartographic Classification Depending on Spatial Autocorrelation. In: International Journal of Geoinformatics 9(2), p. 29-36.
- Current research (publication in finalization): Use of autocorrelation in space *and* time to reduce spatial and transitional complexity in visualisation of time series data (small multiples & map animation)