

MUNICIPAL DEBT RESTRUCTURING: DESIGN A TOOL FOR THE EVALUATION OF THE LONG RUN SUSTAINABILITY OF LG'S FINANCIAL STRUCTURE



LITHUANIAN MUNICIPALITIES CLUSTERING



CLUSTERING

CLUSTER ANALYSIS OR CLUSTERING IS THE TASK OF GROUPING A SET OF OBJECTS IN SUCH A WAY THAT OBJECTS IN THE SAME GROUP (CLUSTER) ARE MORE SIMILAR TO EACH OTHER THAN TO THOSE IN OTHER GROUPS



MUNICIPAL DEBT RESTRUCTURING

Statistical classification
technique

Data are sub-divided into
groups (clusters)

Items in a cluster are very
similar (but not identical)
to one another and very
different from the items
in other clusters

It is a main task of
exploratory data mining
and a common technique
for statistical data analysis
used in many fields

It is a useful tool that
reveals associations,
patterns, relationships,
and structures in masses
of data

MUNICIPAL DEBT RESTRUCTURING

Lithuanian Municipalities

Cluster analysis or clustering:

identification of groups of Municipalities more similar to each other than to those in other groups





PANEL DATA

Strongly balanced dataset
of **CONTEXT** information
from 2013 to 2017



ANALYSIS

CLUSTERING ANALYSIS:

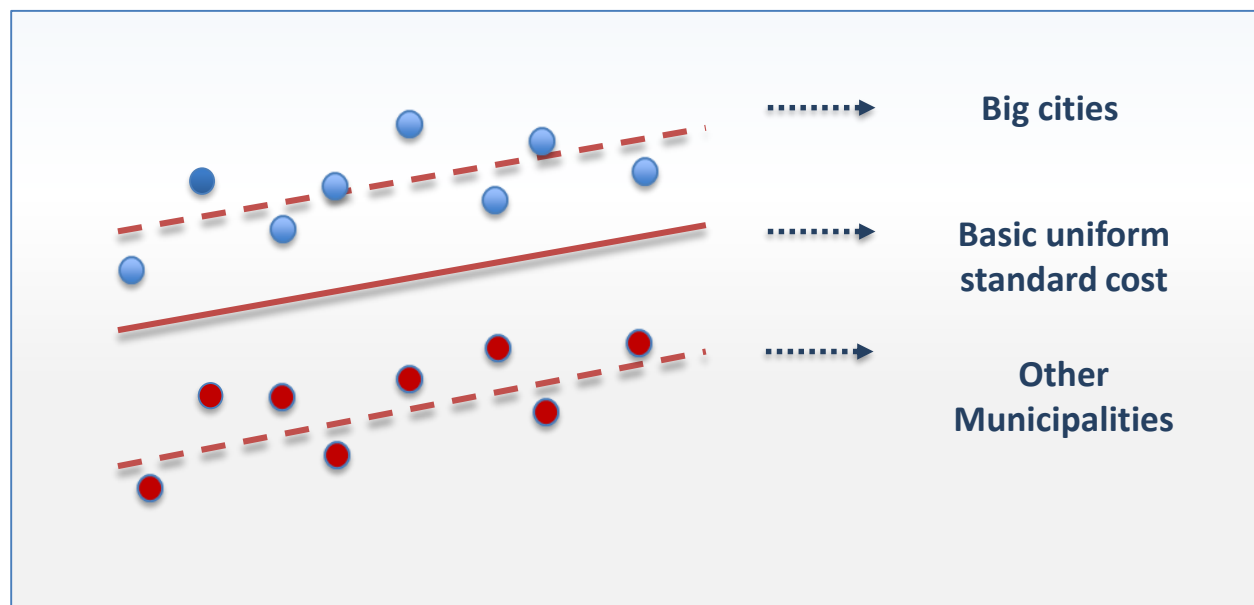
- ✓ ESTIMATION ON 2017
- ✓ CLUSTERS ARE TIME INVARIANT





SHIFTING EFFECT

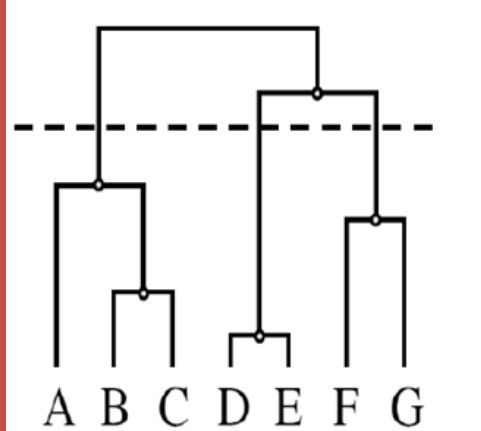
Different Cluster features can be estimated



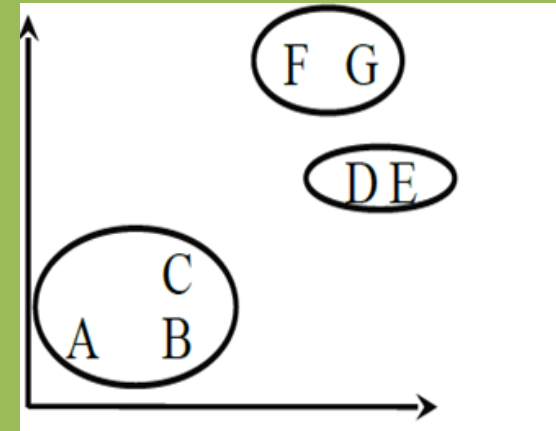
WHY CLUSTERS?

COST SHIFT ESTIMATION FOR MUNICIPALITIES WITH DIFFERENT CONTEXT FEATURES

Cluster TAXONOMY



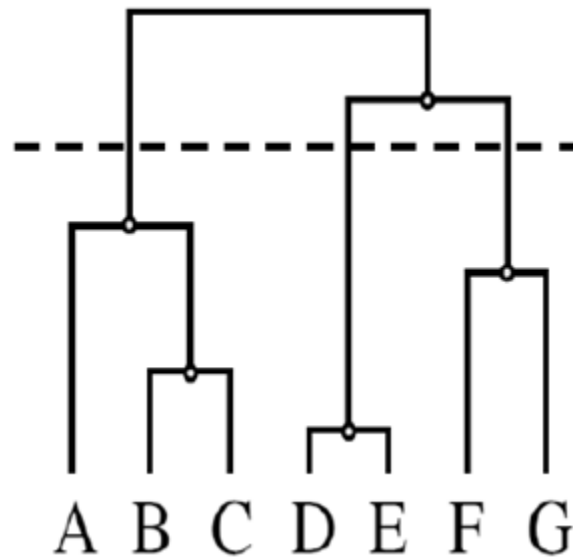
**Hierarchical Cluster
analysis**



**Partitional
algorithm**

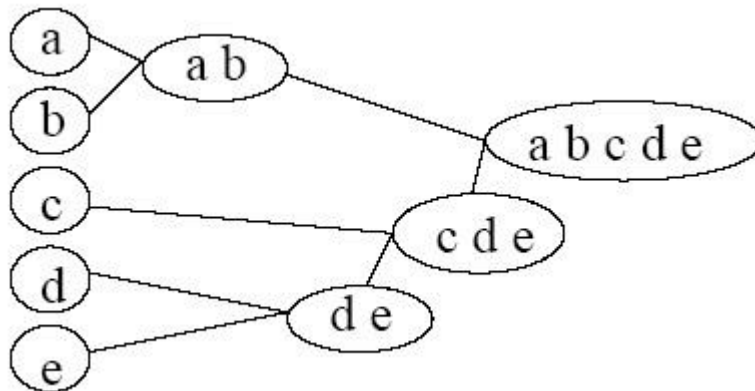
Hierarchical Cluster analysis

- Sequence of hierarchies
- Focused on the relations between dataset patterns
- Requires a matrix of proximity
- Results are presented with DENDOGRAMS



Hierarchical Cluster analysis

- **Strategies:**
 - **AGGLOMERATIVE:** all units disjoint (bottom up)
 - **DISJUNCTIVE:** all units grouped together (top- own)

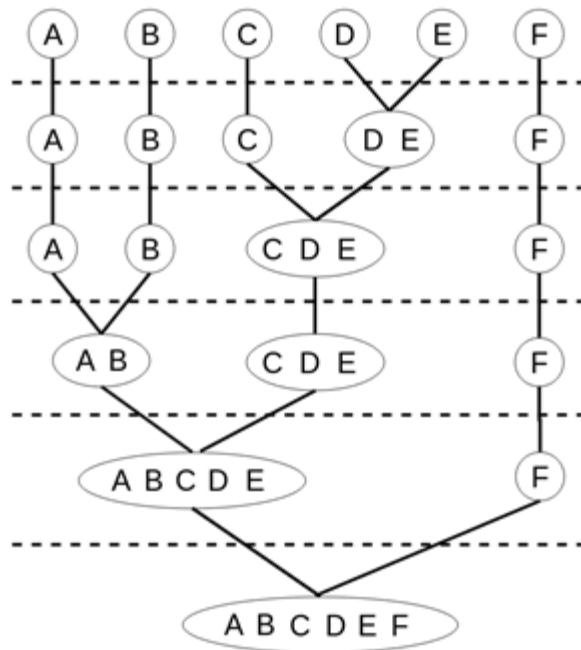


Bottom Up
strategy

Dendrogram

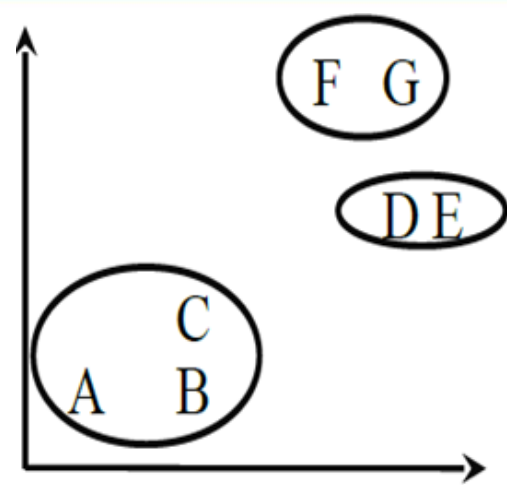
Hierarchical Cluster analysis

- **Strategies:**
 - **AGGLOMERATIVE:** all units disjoint (bottom up)
 - **DISJUNCTIVE:** all units grouped together (top-own)



Top Down
strategy

Dendogram



Partitional algorithm

- The result is a single partition of data
- Identify natural groups present in the data
- Identify a group of disjoint cluster
- The union of disjoint clusters gives back the initial data set



Mixture Models

Model-Based Clustering

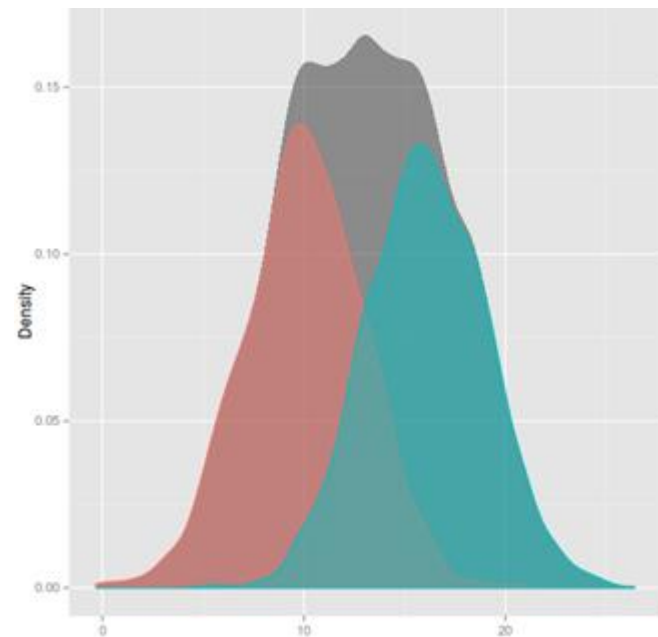
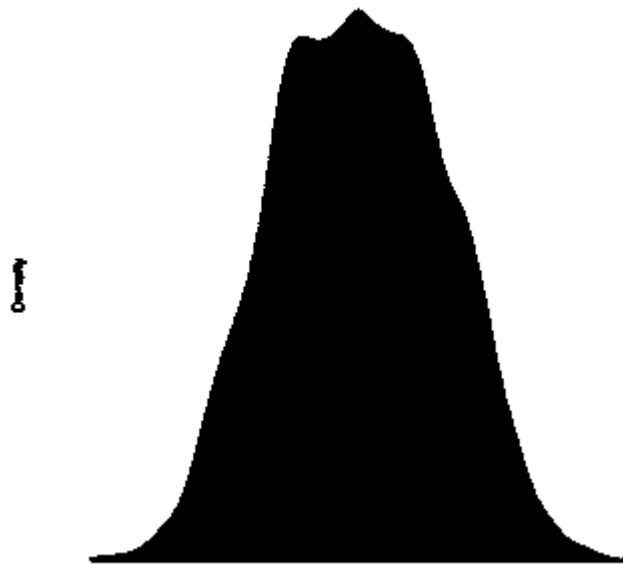
**Partitional
algorithm**

Mixture Models (Model-Based Clustering)

Use of probabilistic models

ASSUMPTION: data are generated by a MIXTURE of probabilistic distributions and each distribution identifies a CLUSTER

TARGET: maximize the fit between data and model

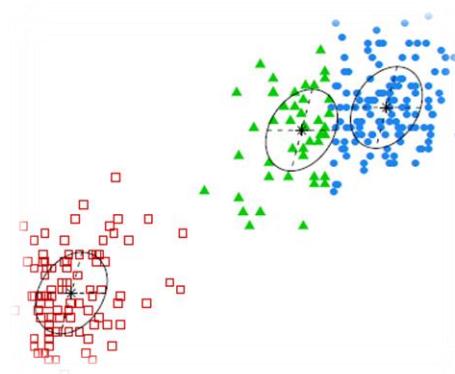


Mixture Models (Model-Based Clustering)

Mixture of distributions

$$\sum_{k=1}^G \tau_k \phi_k(\mathbf{x}_i \mid \mu_k, \Sigma_k)$$

Model based
Clustering



More frequent Technique: Gaussian Mixture Models

The assumption is that every component of the MIXTURE
(cluster) is **Gaussian**

Mistura di distribuzioni Gaussianne

$$\sum_{k=1}^G \tau_k \phi_k(\mathbf{x}_i \mid \mu_k, \Sigma_k)$$

dove:

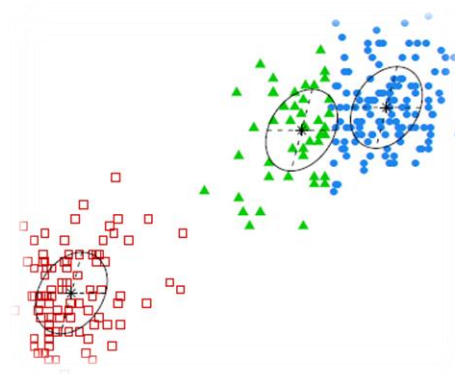
X data matrix

G number of components (**CLUSTER**)

τ_k probability of k-th **CLUSTER** $(\tau_k \geq 0; \sum_{k=1}^G \tau_k = 1)$

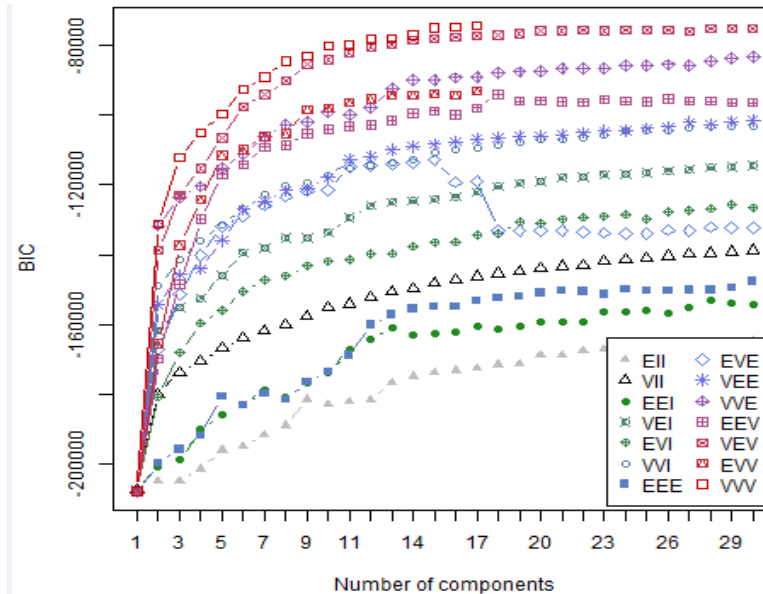
Σ_k covariance matrix **CLUSTER**

Model based Clustering



$$\phi_k(\mathbf{x} \mid \mu_k, \Sigma_k) = (2\pi)^{-\frac{p}{2}} |\Sigma_k|^{-\frac{1}{2}} \exp \left\{ -\frac{1}{2} (\mathbf{x}_i - \mu_k)^T \Sigma_k^{-1} (\mathbf{x}_i - \mu_k) \right\}$$

Mixture Models (Model-Based Clustering)



Criterion for
identify the
optimal number
of CLUSTER

- **BIC** (Bayesian Information Criterion)

Clustering Results



LITHUANIA

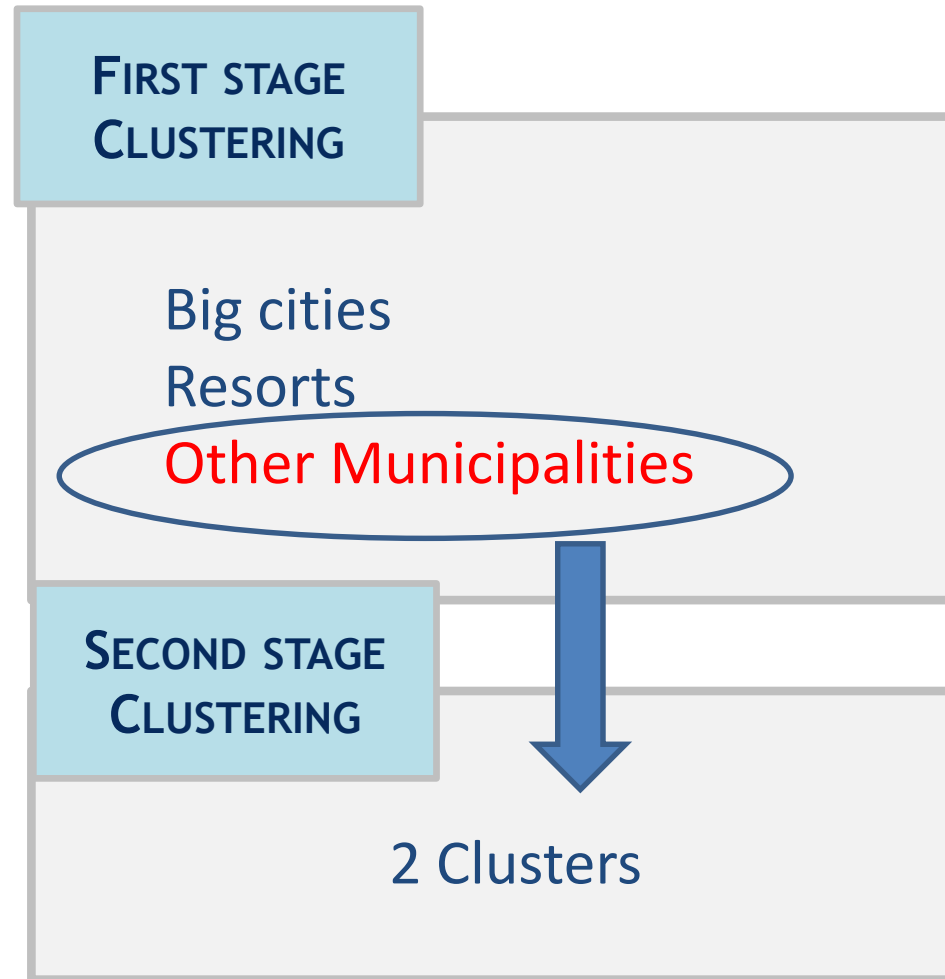
60 MUNICIPALITIES

**2 stage
processing**

4 CLUSTERS



Clustering Results



Clustering Results

FIRST STAGE



Big cities

6 Municipalities

Resorts

4 Municipalities

Other
Municipalities

50 Municipalities

Clustering Results



BIG CITIES

VILNIAUS M.
ALYTAUS M.
KAUNO M.
KLAIPĖDOS M.
PANEVĖŽIO M.
ŠIAULIŲ M.

Municipalities with a big amount of
population and a high population
density

Clustering Results

Municipalities strongly oriented to
touristic activity



Resorts

RESORTS

BIRŠTONO M.
DRUSKININKŲ M.
NERINGOS M.
PALANGOS M.

Clustering Results

Residual group of Municipalities



These Municipalities have been furtherly divided into 2 groups according to contextual features
(second stage Clustering)



Other
Municipalities

Clustering Results

Residual group of Municipalities



Other
predominantly urban

Other
predominantly rural



Other
Municipalities

Clustering Results

FIRST and SECOND STAGE



Big cities

6 Municipalities

Other
predominantly urban

12 Municipalities

Other
predominantly rural

38 Municipalities

Resorts

4 Municipalities

Clustering Results



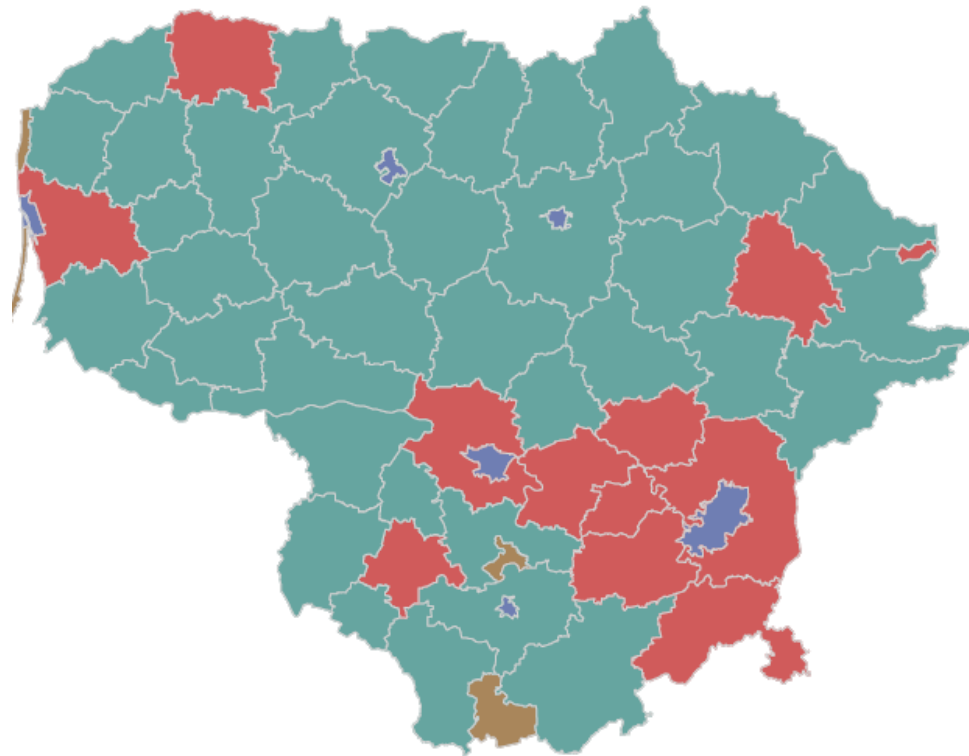
- Population
- Density of population
- Population 0-19
- Population over 65
- Working age population

DISCRIMINANT VARIABLES

- Area (squared km of land)
- Agricultural land
- Forest land
- Built-up area
- Waters

- Number of tourists accommodated in accommodation establishments
- Number of overnight stays in accommodation establishments

Clustering Results



Cluster  1  2  3  4

Soluzioni per il Sistema Economico

Via Mentore Maggini 48/C – 00143 Roma

 info@pec.sose.it **t.** +39 06 508311 **f.** +39 06 50831301