



# **California Dreamin'**

## **An analysis on housing in California**

**Gabriele Cialdea, Leonardo Filippone, Pietro Pianini**



# Introduction

## Research Question ?

How can we predict the **expected value of an house** basing on the **features of the neighbourhood?**



Analysis of the **structure** and the **variability of the sample**



Selecting a **prediction model** assessing several **supervised classification methodologies**



# Agenda

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## Sample and Data Selection

Clustering

Principal Components Analysis

Supervised Classification

Conclusions



# Data Description

① Sample and Data Selection

## Source and dimension of the dataset and attributes of the observations

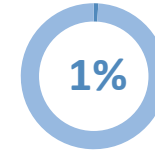
### Source

Kaggle dataset based on 1990 census on housing in California



### # observations

20.640<sup>1</sup>



**Missing values**  
(artificially added)

For each observation the following **attributes**:

- Longitude
- Latitude
- Housing median age
- Total rooms
- Total bedrooms
- Population
- Households median income
- Median house value
- Ocean proximity (artificially added)

## Selection and manipulation process

### Step 1

Cleaning the dataset from **missing values**

Final number of observations: **20.433**

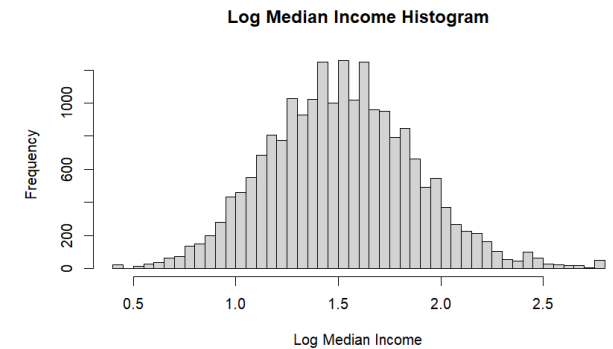
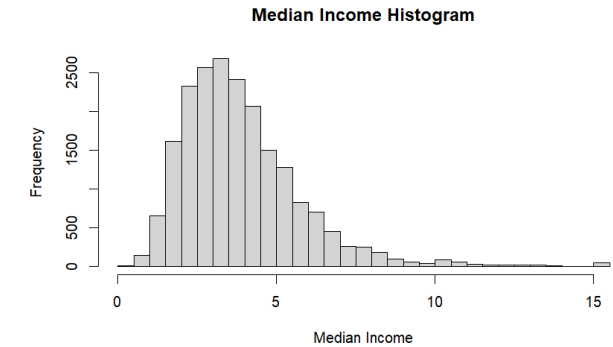
### Step 2

Log-transformed the values of the **most skewed attributes**

### Step 3

Scaled values for all the observations and categorized the **target attribute**

## 🔍 Focus on Step 2





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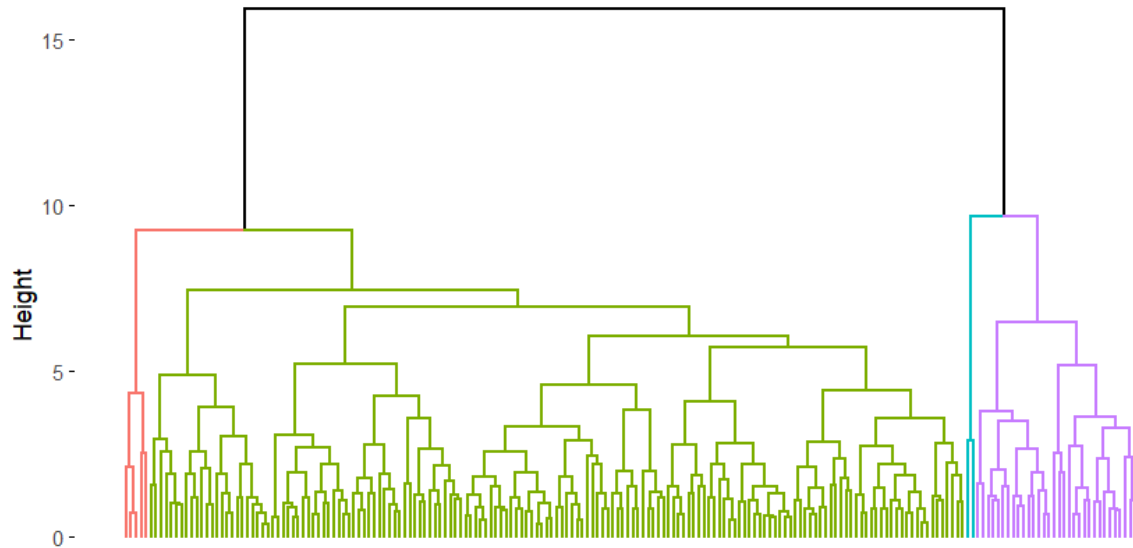


# Hierarchical Clustering

② Clustering

## Dendrogram representation

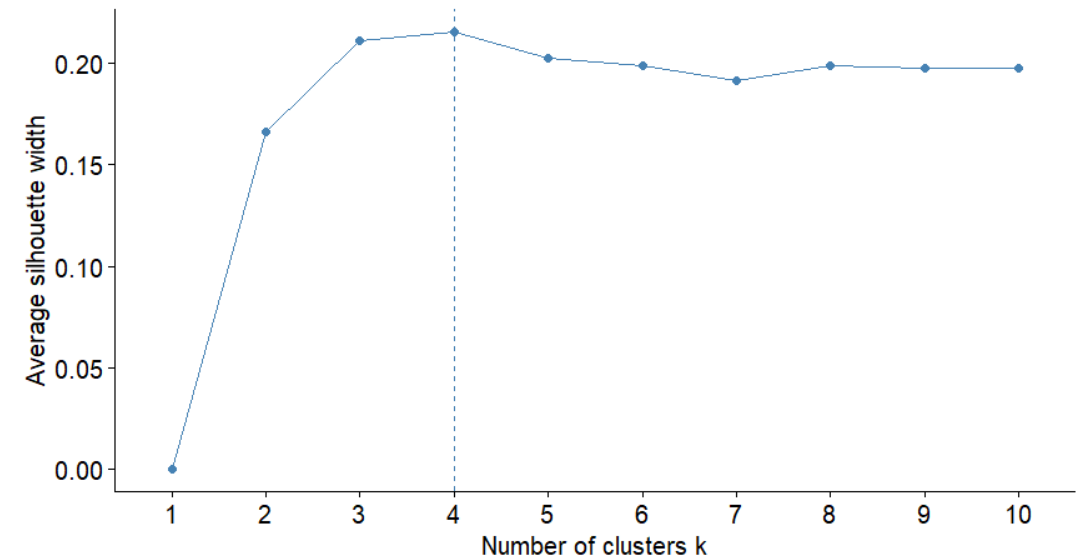
Euclidean-Complete with eclust



Number of clusters **4**

## Silhouette evaluation

Optimal number of clusters  
Silhouette method AHC



Average Silhouette Width **0,22**

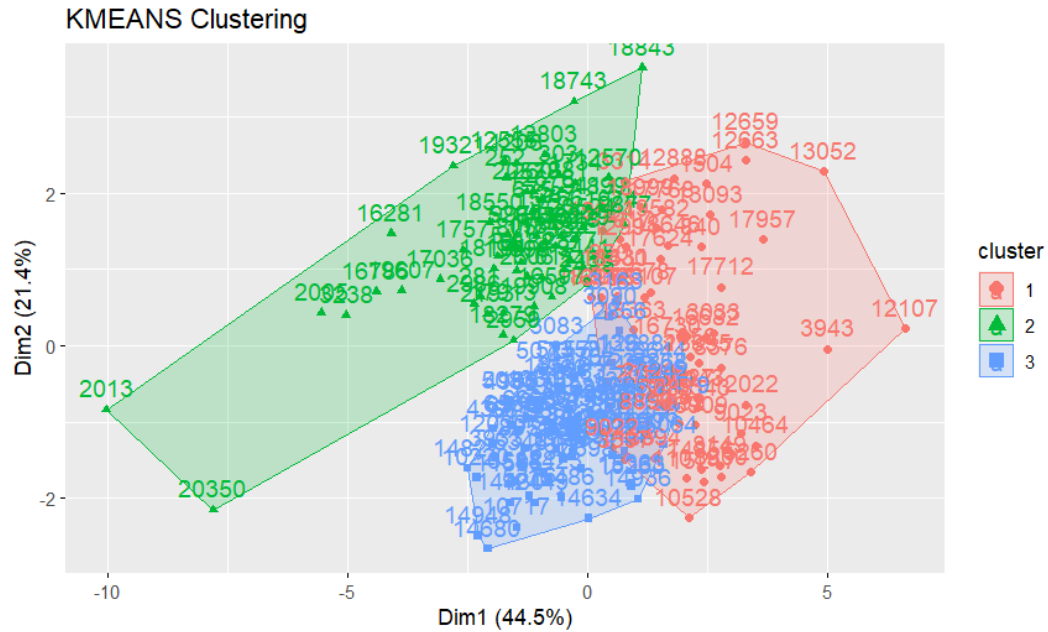
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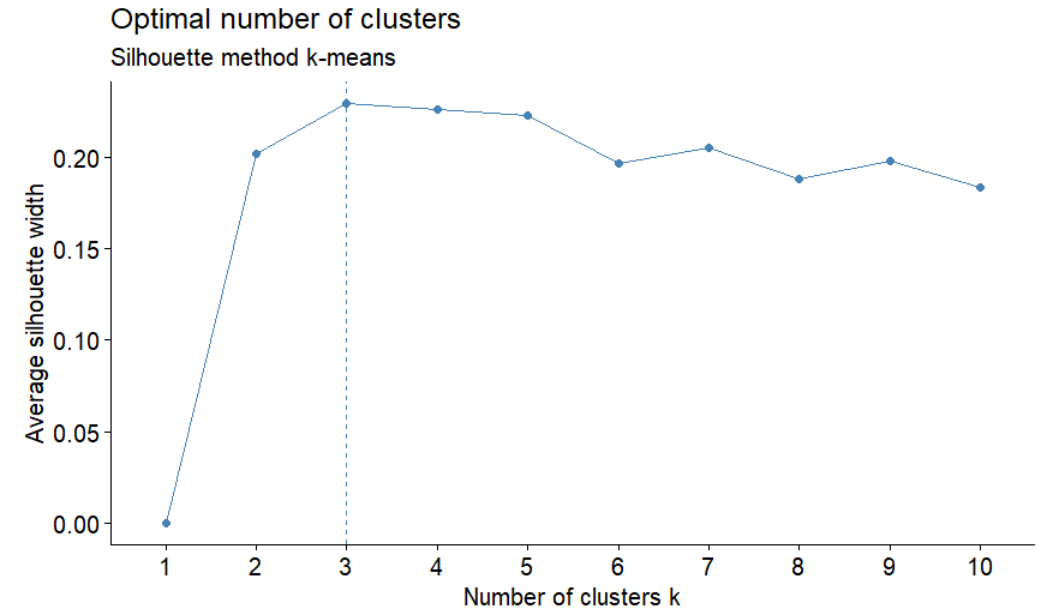
# K-Means Clustering

## K-Means cluster representation



Number of clusters **3**

## Silhouette evaluation



Average Silhouette Width **0,23**





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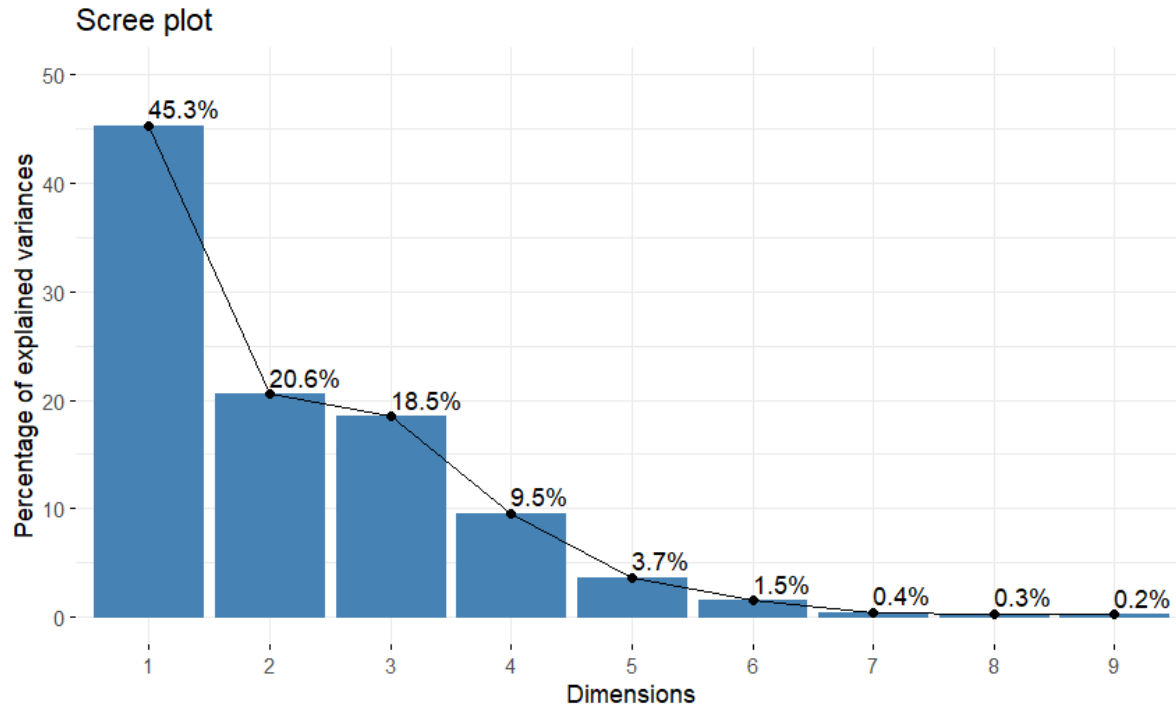


# Optimal number of components

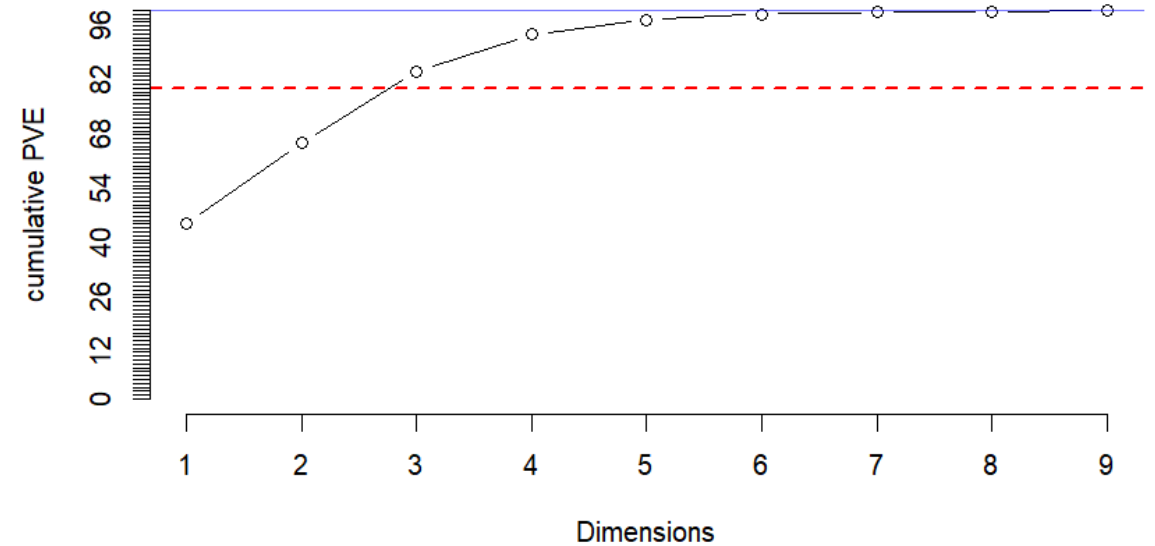
③

PCA

## Screeplot



## Cumulative PVE



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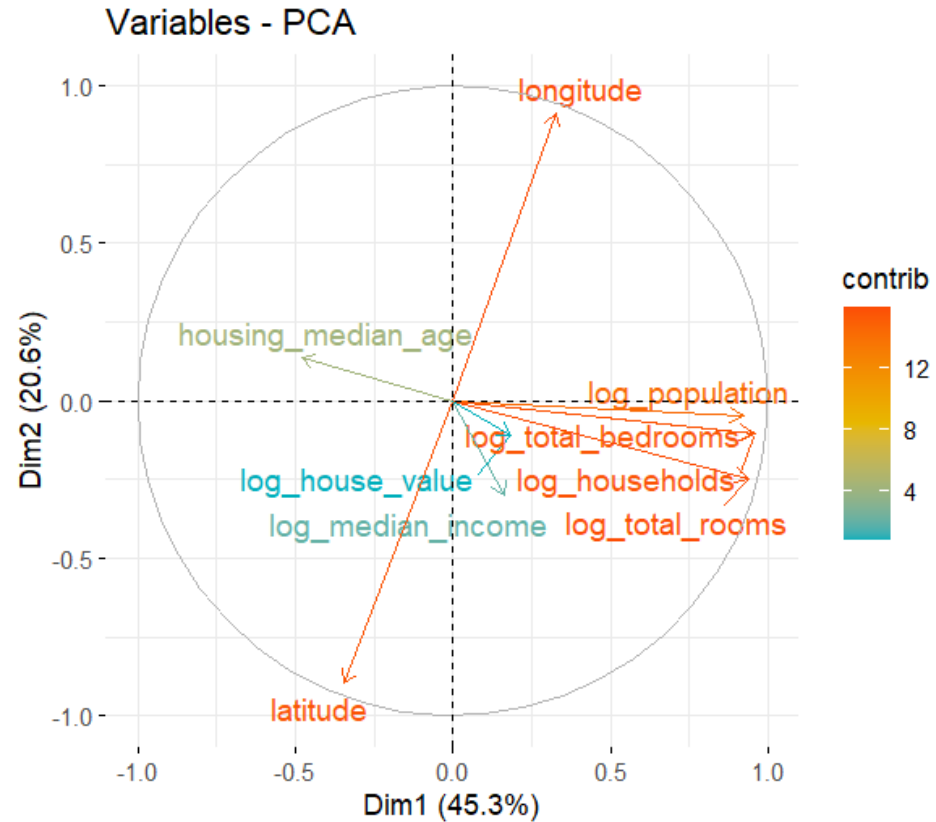


# Principal components representation

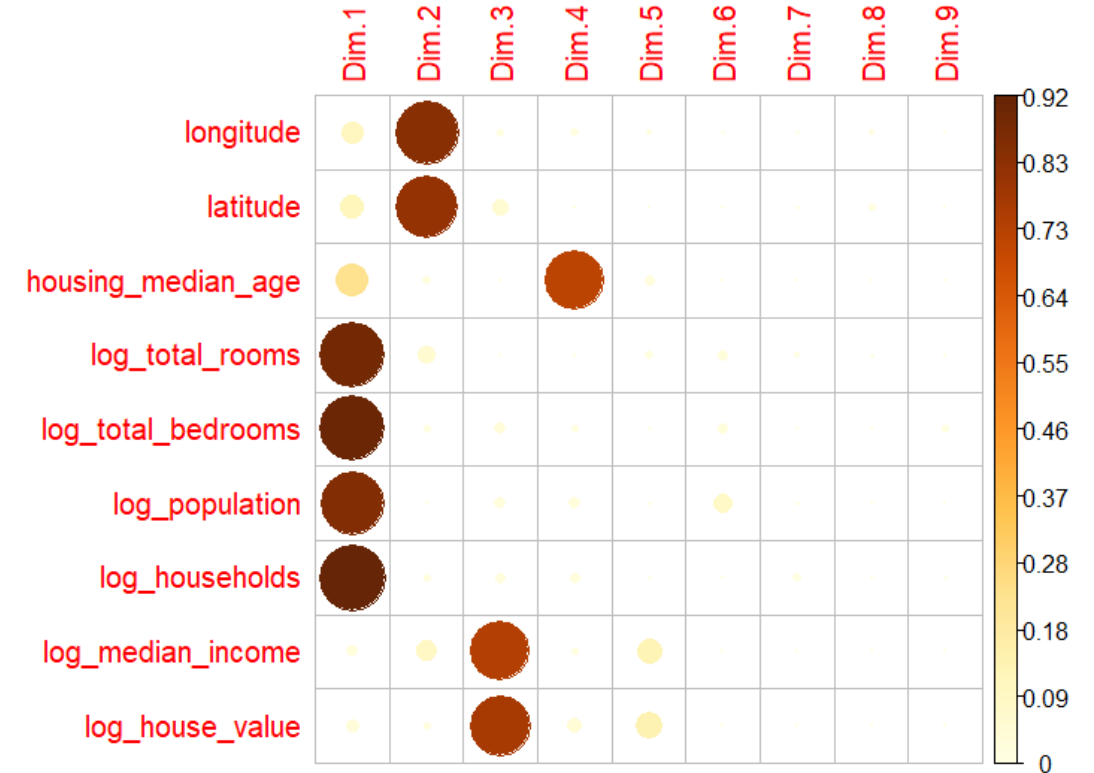
③

PCA

## Biplot



## Corrplot



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Sample and Data Selection

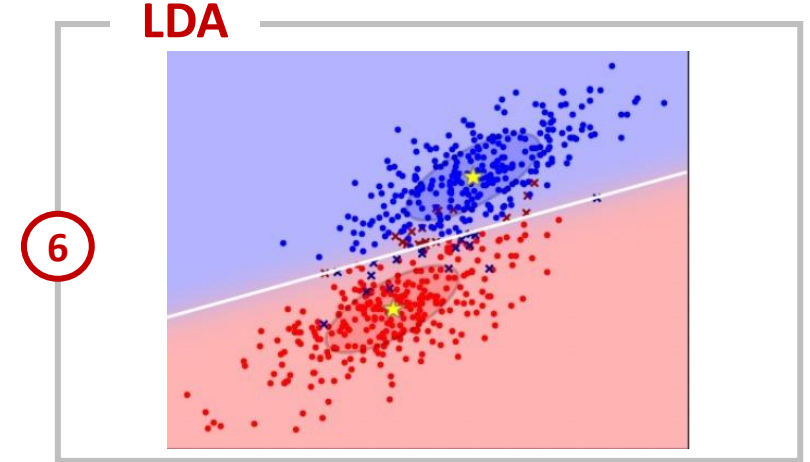
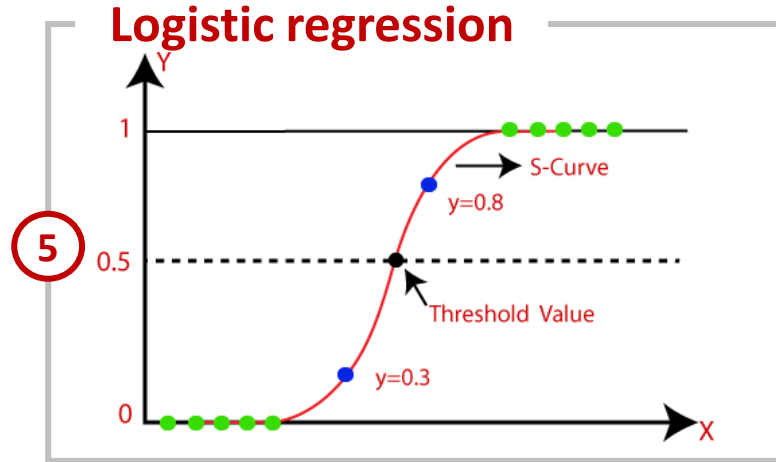
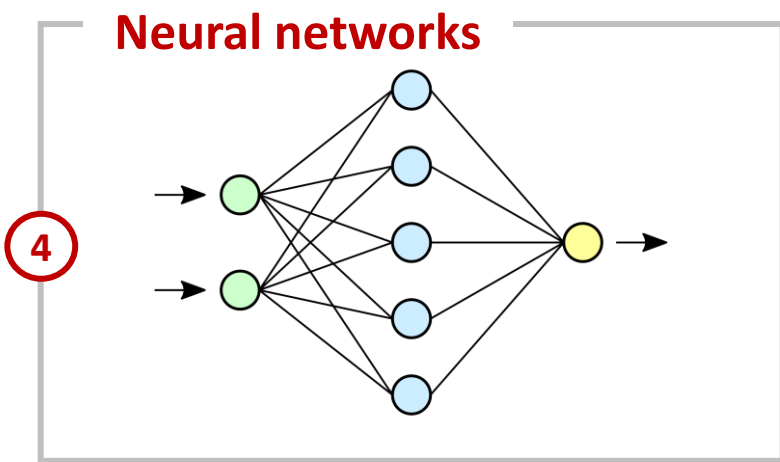
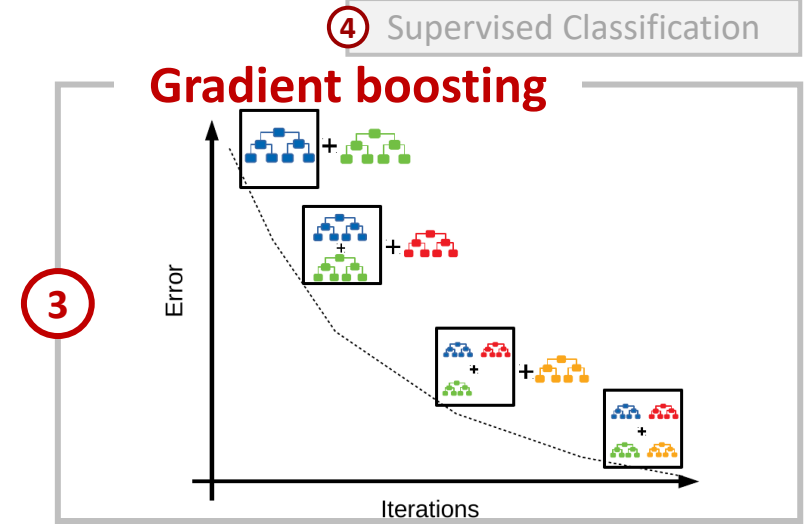
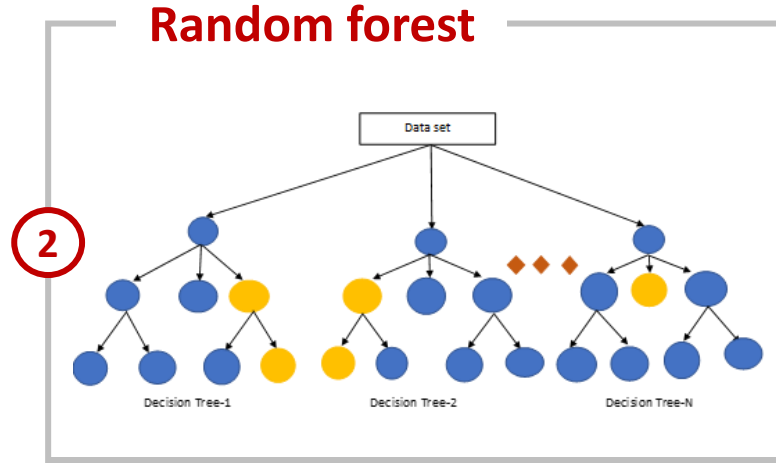
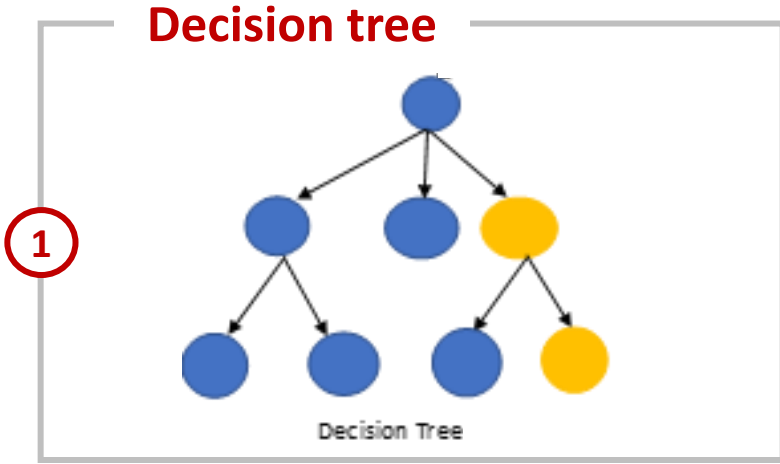
Clustering

Principal Components Analysis

**Supervised Classification**

Conclusions

# Main Classifiers Overview

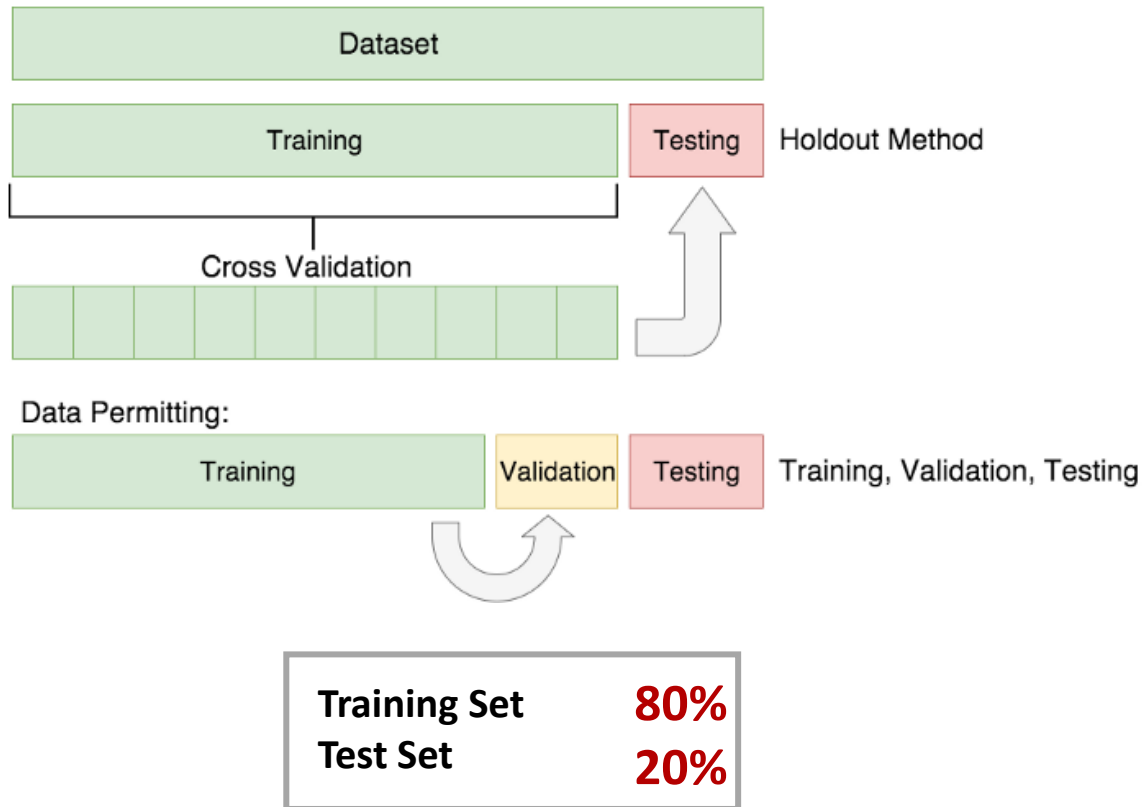




# Cross Validation Methodology and Evaluation Metric

④ Supervised Classification

## Cross validation



## Evaluation metric

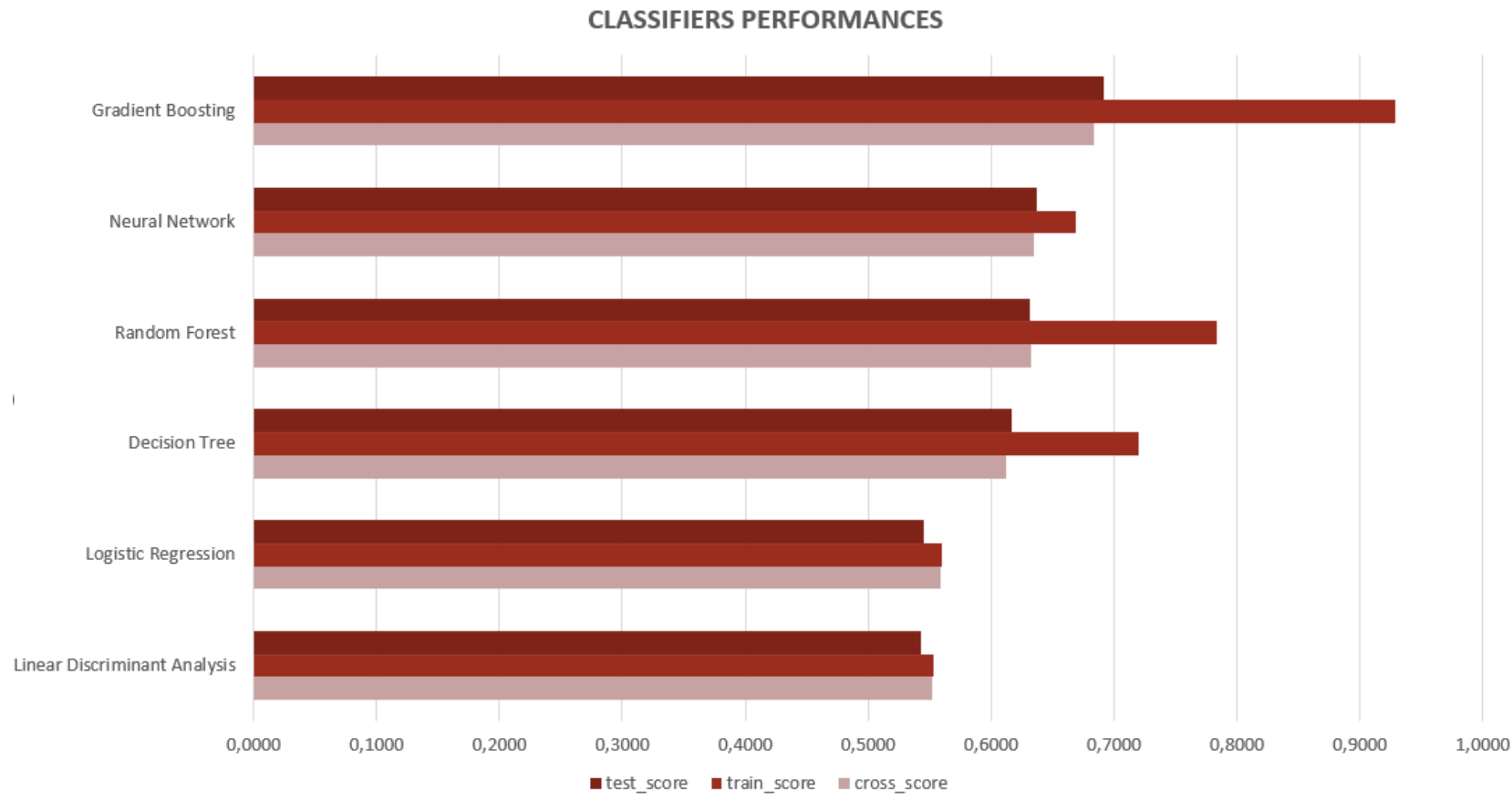
Metric	Formula
True positive rate, recall	$\frac{TP}{TP+FN}$
False positive rate	$\frac{FP}{FP+TN}$
Precision	$\frac{TP}{TP+FP}$
<b>Accuracy</b>	<b><math>\frac{TP+TN}{TP+TN+FP+FN}</math></b>
F-measure	$\frac{2 \cdot \text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}}$



# Results

## ④ Supervised Classification

### Performance scores and ranking of the main classifiers



- **Gradient Boosting** is recognized as the **best-fitting classifier**
- These the **optimal hyperparameters**
  - Learning rate: 0.01
  - Number of estimators: 300
  - Max depth: 5



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# Further research

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**More recent data and/or** extension of the analysis to **other geographical areas** (e.g. robustness in other States of the U.S.)



**Different and more complex methodologies** for the supervised classification



Employment of **more explanatory attributes** for the observations in the prediction model



# References

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- Boehmke, Bradley; Greenwell, Brandon (2019). Hands-On Machine Learning with R. Chapman & Hall.
- James, G., Witten, D., Hastie, T., Tibshirani, R. (2021). An Introduction to Statistical Learning. Springer Texts in Statistics. Springer, New York, NY
- Pace, R. Kelley, and Ronald Barry. "Sparse spatial autoregressions." Statistics & Probability Letters 33.3 (1997): 291-297.