Lecture 12: Dimensionality reduction

Introduction

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Methods for feature reduction

# Lecture 12: Dimensionality reduction Introduction to Machine Learning

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## **Definition**

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#### Dimensionality reduction

**Dimensionality reduction** is the transformation of the data from a **high**-dimensional space into a **lower**-dimension space, with as little information loss as possible.

## **Definition**

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#### Dimensionality reduction

**Dimensionality reduction** is the transformation of the data from a **high**-dimensional space into a **lower**-dimension space, with as little information loss as possible.

#### We want to:

- Reduce the number of features
- Retain as much information as possible

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#### Question

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#### Question

Why do you think we need to reduce dimensions?

Computation time

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#### Question

- Computation time
- Easier data visualization

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#### Question

- Computation time
- Easier data visualization
- Possible unrelated features acting as noise

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#### Question

- Computation time
- Easier data visualization
- Possible unrelated features acting as noise
- Possible correlated features that do not bring any new information to solve the task

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#### Question

- Computation time
- Easier data visualization
- Possible unrelated features acting as noise
- Possible correlated features that do not bring any new information to solve the task
- The curse of dimensionality

# Curse of dimensionality

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#### Curse of dimensionality

The **curse of dimensionality\*** refers to various phenomena that arise when analyzing data in high-dimensional space.

# Curse of dimensionality

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#### Curse of dimensionality

The **curse of dimensionality\*** refers to various phenomena that arise when analyzing data in high-dimensional space.

The main problem is that the **dataset becomes sparse**: we do not have enough combinations of values to properly learn from it!

## Methods for feature reductions

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#### Question

What is in your opinion possible approaches to reduce the number of features ?

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#### Question

What is in your opinion possible approaches to reduce the number of features ?

■ Removing some features

## Methods for feature reductions

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#### Question

What is in your opinion possible approaches to reduce the number of features ?

- **Removing** some features
- **Projecting the features** into a lower dimensional space.

#### Feature selection

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Feature selection

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**Feature selection** consists in selecting the subset of features that bear the most information and removing the features that bear little information from the dataset.

#### Feature selection

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#### Feature selection

**Feature selection** consists in selecting the subset of features that bear the most information and removing the features that bear little information from the dataset.

#### Question

Do you remember what features we removed from the Titanic dataset ? Why and how ?

#### Feature selection

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#### Feature selection

**Feature selection** consists in selecting the subset of features that bear the most information and removing the features that bear little information from the dataset.

#### Question

Do you remember what features we removed from the Titanic dataset ? Why and how ?

We need some objective rules to define what subset of variables to keep.

## Feature projection

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Feature projection\* (also called feature extraction) transforms the data from the high-dimensional space to a space of fewer dimensions.

## Feature projection

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#### Feature projection

Feature projection\* (also called feature extraction) transforms the data from the high-dimensional space to a space of fewer dimensions.

We need to project the data into another space which is a combination of the features with as little information loss as possible.

## Feature projection

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#### Feature projection

Feature projection\* (also called feature extraction) transforms the data from the high-dimensional space to a space of fewer dimensions.

We need to project the data into another space which is a combination of the features with as little information loss as possible.

The transformation can be:

- Linear (PCA, NMF . . . )
- Non-linear (t-SNE, ...)

# Questions

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Questions ?