

# *Intro to Unsupervised ML*

**TM Quest**

# Overview

## What Will we Learn in This Module?

- What does it mean for a model to be **unsupervised**?
  - What is the different between **supervised** and **unsupervised**?
  - **Why** use unsupervised learning?
  - **When** use unsupervised learning?
- What is the **Kmeans clustering** model?
  - How does the Kmeans clustering work?

# *What is Unsupervised Learning?*

# *Supervised VS. Unsupervised*

## Supervised Learning

We have features and **targets**.

## Unsupervised Learning

We have features.

### Example (Clustering your customer group)

- Want to understand your customers better.
- Divide them into groups based on behavior.
- Can use your better understanding of the different groups to tailor your marketing.

# *Unsupervised Learning*

## When to use Unsupervised Learning?

- When the labels are unavailable.
  - Impossible/illegal/hard/expensive to get.
  - Too slow to get for the task.
  - Unknown what the labels should be.

## Unsupervised Tasks

- Clustering
- Outlier/Anomaly Detection
- Recommender Systems

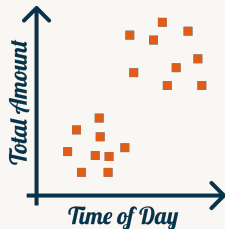
# Clustering

## Clustering

- Clustering models can both be supervised (e.g., KNN) and unsupervised.
- In unsupervised learning we need to do the clustering without the labels.
- We will learn more about the k-means model later.

## Example (Clustering)

- A retailer knows the following:
  - email-address,
  - the time of day,
  - total amount.
- Want to make custom promotions based on this information.



# *Outlier/Anomaly Detection*

## Outlier/Anomaly Detection

- **Outlier detection** is finding outliers in the system.
- Assumes that there are more normal data points than outliers.
- Can benefit from some supervised data.

## Example

- Spam filters: Outliers—Spam mails
- Fraud detection: Outliers—Fraudulent transactions
- Find mistakes in the system: Outliers—Mistakes
- Detect cyber attacks in your system: Outliers—Attacks

# Recommender System

## Recommender System

- **Recommender systems** are systems that give the user recommendations on what to do next.
- On smaller systems, it is often based on rules rather than machine learning.

## Example

- Recommending the next thing to read/watch (YouTube/Netflix/TikTok),
- Recommending additional wares in an online store (Amazon).
- Recommending further information (your bank/state/forum).



# *K-Means Clustering*

# K-Means Clustering

## Output Clusters

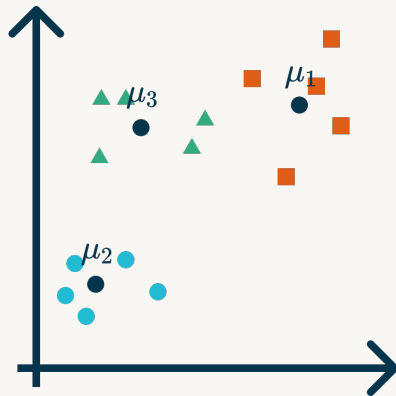
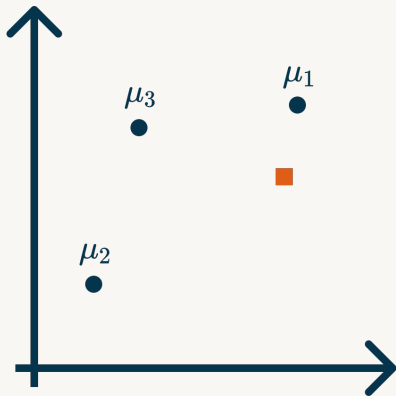
- The algorithm finds the **mean points**  $\mu_1, \dots, \mu_k$ .
- Each mean  $\mu_i$  gives us a **corresponding cluster**  $S_i$  of data points.
  - $S_i$  consists of all data points that are closer to  $\mu_i$  than any other mean point.
- Additionally, the mean points satisfy

$$\mu_i = \frac{1}{n_i} \sum_{x \in S_i} x,$$

where  $n_i$  is the number of points in  $S_i$ .

# 3-Means Clustering

## Example



## How to Find the Mean Points?

Given a cluster  $S_i$  with  $n_i$  points, define its **mean** by

$$\mu_i = \frac{1}{n_i} \sum_{x \in S_i} x.$$

Of all the ways to divide the points into  $k$ -clusters  $S_1, \dots, S_k$  the  $k$ -means algorithm tries to **minimize the quantity**

$$\sum_{i=1}^k \frac{1}{n_i} \sum_{x \in S_i} d(x, \mu_i)^2 = \sum_{i=1}^k \text{Var}(S_i)$$

where  $d(x, \mu_i)$  is the distance between  $x$  and  $\mu_i$ .

*The  $k$ -means algorithm tries to simultaneously minimize how much the clusters spread out.*