

# Dimensionality reduction & clustering

## Lecture 18

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# Outline

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1. Revist the role of clustering.
2. Study the most popluar clustering method and its simple variant:
  - K-means algorithm
  - K-medoids algorithm
3. Explore another popluar method:
  - Hierarchical clustering (agglomerative clustering)

# Role of clustering

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Suppose: Data distribution is pretty wide

We need lots of data examples to ensure good generalization performance.

In case # of examples is not so big, one may want to classify the examples such that the distribution of the classified examples is concentrated.

Clustering is often employed for such classification.

# ***K*-means in words**

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*K* indicates the number of resultant clusters.

“mean” serves as a representative of each cluster.

# How $K$ -means works

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1. Choose  $K$  points randomly.
2. (*Assignment step*): Map each data point to either one of the  $K$  points depending on its distance.
3. (*Update step*): Compute the **means** of such  $K$  clusters.
4. Repeat 2 & 3 until assignment is not changed further.

# A variant of $K$ -means

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## $K$ -medoids

Very similar to  $K$ -means.

The only distinction is that we take “median (medoid)” instead of “mean”.

**Note:** Robust to **outliers**.

# Hierarchical clustering in words

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Do clustering in a **hierarchical** manner.

1. Start with the largest number of clusters (same as data points).
2. Merge clusters according a certain rule.
3. Repeat such merging until we reach down to  $K$  clusters.

# How hierarchical clustering works

Start with 6 clusters (same # of data points)



1. Choose a pair of cluster centroids with minimum distance
2. Compute the **centroid** of the pair.



# How hierarchical clustering works



3. Repeat Step 1 & 2: Choosing a pair of *updated* centroids with minimum distance, and then update centroids.
4. Repeat until  $K$  clusters are formed.

# Dimensionality reduction + clustering?

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How to combine?

Often: Do clustering **after** dimensionality reduction.

# What is next?

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From lots of project experiences with Hyundai Motor, found that many people are interested in:

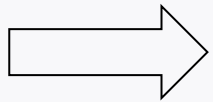
1. Anamoly detection

2. Fusion learning

# Important techniques for the problems

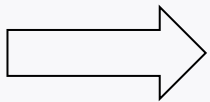
From lots of project experiences with Hyundai Motor, found that many people are interested in:

## 1. Anamoly detection



**autoencoder**

## 2. Fusion learning



**matrix completion**

# Look ahead

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Will study the two techniques:

1. **autoencoder**

2. **matrix completion**