

# LECTURE 8: CLUSTERING AGGREGATION

## 1) Clustering Aggregation

### a) Terminology:

- i) **Clustering**: group of clusters output by clustering algorithm
- ii) **Cluster**: group of points

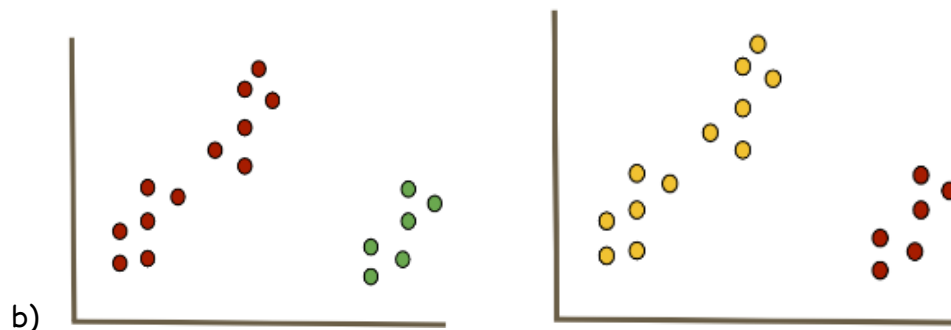
### b) Goals:

- i) Compare clusterings
- ii) Combine information from multiple clusterings to create a new clustering

## 2) Comparing Clusterings

### a) Need to compare clustering by looking at assignment of points in clusters

- i) Many points assigned to same cluster in both clustering C and P, then they should have a small distance
- ii) Identifying which cluster in P and C are not easy



- i) Clusterings are the same, but assignments/labels not consistent
- ii) Asking "is x in red cluster" in left clustering = "is x in yellow cluster" in right clustering
  - 1) However, won't know conversion unless we know set of conventions

### 3) Disagreement Distance

a) Given 2 clusterings P and C

$$D(P, C) = \sum_{x,y} \mathbb{I}_{P,C}(x, y)$$

i) Where

$$\mathbb{I}_{P,C}(x, y) = \begin{cases} 1 & \text{if P \& C disagree on which clusters x \& y belong to} \\ 0 & \end{cases}$$

	P	C
$x_1$	1	1
$x_2$	1	2
$x_3$	2	1
$x_4$	3	3
$x_5$	3	4

b) Ex:

i) Disagreement distance for P and C

$x_2$	$x_1$	1
$x_3$	$x_1$	1
$x_4$	$x_1$	0
$x_5$	$x_1$	0
$x_3$	$x_2$	0
$x_4$	$x_2$	0
$x_5$	$x_2$	0
$x_4$	$x_3$	0
$x_5$	$x_3$	0
$x_4$	$x_5$	1

c)

1.  $D(C, P) = 0$  iff  $C = P$
2.  $D(C, P) = D(P, C)$
3. Triangle Inequality:

$$\mathbb{I}_{C_1, C_3}(x, y) \leq \mathbb{I}_{C_1, C_2}(x, y) + \mathbb{I}_{C_2, C_3}(x, y)$$

i)  $I_{C,P}$  can only be 0 or 1 and the above is violated iff

$$I_{x,y}(C_1, C_3) = 1, I_{x,y}(C_1, C_2) = 0, I_{x,y}(C_2, C_3) = 0$$

#### 4) Aggregate Clustering

- a) **Goal:** From set of clusterings  $C_1, \dots, C_m$  generate a clustering  $C^*$  that minimizes

$$\sum_{i=1}^m D(C^*, C_i)$$

- i) Problem equivalent to clustering categorical data

- b) **Benefits:**

- i) Identify best number of clusters

- 1) Optimization function not make assumptions on number of clusters

- ii) Handle/detect outliers

- iii) Improve robustness of clustering algorithms → combining clusters can produce better results

- iv) Privacy preserving clustering: aggregate clustering without sharing data

- c) NP-Hard problem

- i) Often solve with approximations

- ii) **Rule:** only works if it produces clustering

	City	Profession	Nationality
$x_1$	NY	Doctor	US
$x_2$	NY	Teacher	French
$x_3$	Boston	Lawyer	Canada
$x_4$	Boston	Doctor	US
$x_5$	LA	Lawyer	Canda
$x_6$	LA	Actor	French

- d)

- i) Majority saying

1.  $x_1$  &  $x_2$  together
2.  $x_2$  &  $x_3$  together
3.  $x_1$  &  $x_3$  separate

