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## Knowledge Discovery in Databases II SS 2019

# **Exercise 4: High Dimensional Data Clustering**

#### Exercise 4-1 Recap: Frequent Itemset Mining

Subspace clustering algorithms usually utilize bottom-up subspace search, which is the same as the frequent itemset mining introduced in KDD1. Two bottom-up subspace search techniques: Apriori and Divide-and-Conquer (Database-projection), were discussed. Given the following data set, please find out all frequent itemsets with support not less than 2:

- (a) using Apriori approach
- (b) using Divide-and-Conquer approach

tid	a	b	c	d	e
1	1	1	0	1	1
2	1	1	0	0	0
3	0	1	1	1	0
4	1	0	1	0	1
5	0	1	0	1	1

Tabelle 1: Transaction dataset.

#### Exercise 4-2 Grid-based Subspace-Clustering (CLIQUE)

How many subspace clusters will be found by CLIQUE algorithm in the following dataset? ( $\tau = 3, \xi = 4$ ) NOTE: to keep consistency, a dense unit contains more than **or equal to**  $\tau$  objects



## Exercise 4-3 Density-based Subspace-Clustering (SubClu)

Show that the following statement (monotonicity of the core point property) holds:

Let D be a set of d-dimensional feature vectors, A the set of all attributes (dimensions/features). Further let  $p \in D$  and  $S \subseteq A$  be a subspace (attribute subset).

Then the following holds for arbitrary  $\epsilon \in \mathbb{R}^+$  and  $minPts \in \mathbb{N}$ :

$$\forall T \subseteq S : |\mathcal{N}^{S}_{\epsilon}(p)| \geq minPts \Rightarrow |\mathcal{N}^{T}_{\epsilon}(p)| \geq minPts$$

with  $|\mathcal{N}_{\epsilon}^{S}(p)| := \{q \in D \mid L_{P}(\pi_{S}(p), \pi_{S}(q)) \leq \epsilon\}.$