

K-Medians: Handling Outliers by Computing Medians

- Medians are less sensitive to outliers than means
 - □ Think of the median salary vs. mean salary of a large firm when adding a few top executives!
- \square *K-Medians*: Instead of taking the **mean** value of the object in a cluster as a reference point, **medians** are used (L_1 -norm as the distance measure)
- ☐ The criterion function for the *K-Medians* algorithm: $S = \sum_{i=1}^{K} \sum_{j=1}^{K} |x_{ij} med_{kj}|$
- ☐ The *K-Medians* clustering algorithm:
 - □ Select *K* points as the initial representative objects (i.e., as initial *K medians*)
 - Repeat
 - Assign every point to its nearest median
 - □ Re-compute the median using the median of each individual feature
 - ☐ Until convergence criterion is satisfied

K-Modes: Clustering Categorical Data

- □ K-Means cannot handle non-numerical (categorical) data
 - Mapping categorical value to 1/0 cannot generate quality clusters for highdimensional data
- □ *K-Modes*: An extension to *K-Means* by replacing means of clusters with *modes*
- □ Dissimilarity measure between object X and the center of a cluster Z
 - $\Phi(x_j, z_j) = 1 n_j^r / n_j$ when $x_j = z_j$; 1 when $x_j \neq z_j$
 - \square where z_j is the categorical value of attribute j in Z_l , n_l is the number of objects in cluster l, and n_j^r is the number of objects whose attribute value is r
- ☐ This dissimilarity measure (distance function) is **frequency-based**
- □ Algorithm is still based on iterative *object cluster assignment* and *centroid update*
- □ A *fuzzy K-Modes* method is proposed to calculate a *fuzzy cluster membership* value for each object to each cluster
- ☐ A mixture of categorical and numerical data: Using a *K-Prototype* method