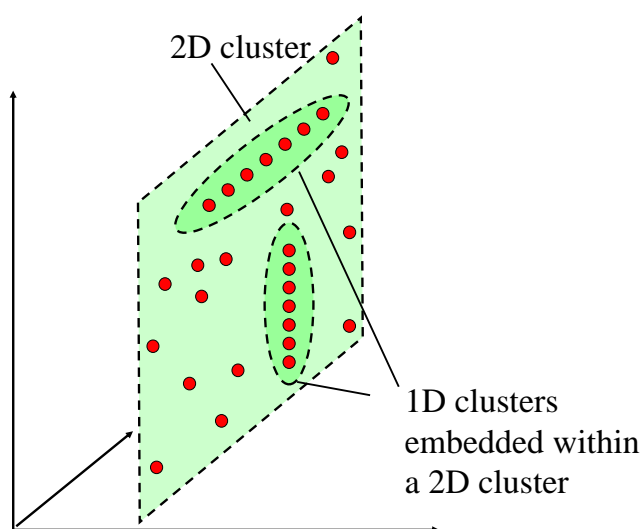


Objective and Definitions

Hierarchically Nested Subspace Clusters



Subspace Dimensionality of a Point p

- Variance along attribute A_i

$$VAR_{A_i}(NN_k(p)) = \frac{\sum_{q \in NN_k(p)} (\pi_{A_i}(q) - \pi_{A_i}(p))^2}{|NN_k(p)|}$$

$\pi_S(o)$: projection of point o into subspace S

- Subspace preference vector

$$w_i(p) = \begin{cases} 0, & \text{if } VAR_{A_i}(NN_k(p)) > \alpha \\ 1, & \text{if } VAR_{A_i}(NN_k(p)) \leq \alpha \end{cases}$$

- Subspace dimensionality

$$\lambda(p) = \sum_{i=1}^d \begin{cases} 1, & \text{if } w_i(p) = 0 \\ 0, & \text{if } w_i(p) = 1 \end{cases}$$

Subspace Preference Vector / Dimensionality of a Pair of Points (p, q)

- Subspace preference vector

$$w(p, q) = w_p \wedge w_q$$

- Subspace dimensionality

$$\lambda(p, q) = \sum_{i=1}^d \begin{cases} 1, & \text{if } w_i(p, q) = 0 \\ 0, & \text{if } w_i(p, q) = 1 \end{cases}$$

Subspace Distance between two Points p and q

$SDIST(p, q) = (d_1, d_2)$, where

$$d_1 = \lambda(p, q) + \begin{cases} 1, & \text{if } \max\{d_{w_p}(p, q), d_{w_q}(p, q)\} > \alpha \\ 0, & \text{else} \end{cases}$$

$$d_2 = d_{\bar{w}(p, q)}(p, q)$$

$$d_{w_p}(p, q) = \sum_{i=1}^d w_i(p) (\pi_{A_i}(p) - \pi_{A_i}(q))^2 :$$

weighted Euclidean distance w.r.t. w between two points p and q

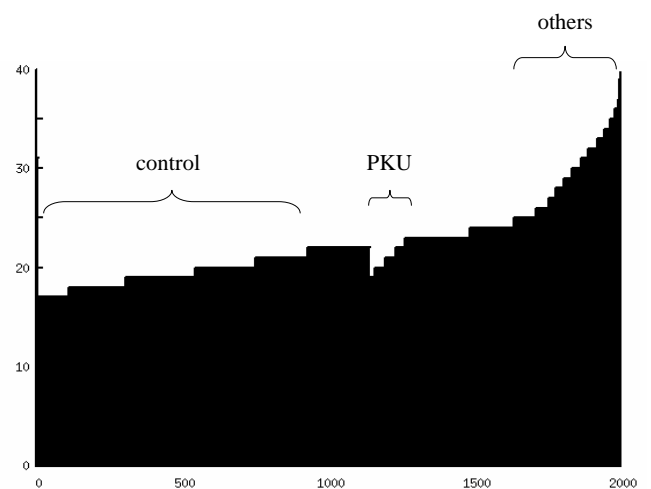
Algorithm and Results

Algorithm HiSC

```

algorithm HiSC(DB, k,  $\alpha$ )
  pq := empty priority queue;
  for each p  $\in$  DB do
    compute  $w_p$ ;
    p.SDIST =  $\infty$ ;
    insert p into pq;
  while pq  $\neq \emptyset$  do
    o := pq.next();
    for each p  $\in$  pq do
      pq.update(p, SDIST(o,p));
    append o to cluster order;
  return cluster order;
  
```

Result on Metabolome Dataset



Result on Synthetic Dataset

