#### **Data Mining**

Data

**Data Preprocessing** 

**Dimensionality Reduction** 

**Fourier Transform** 

**Wavelet Transform** 

Haar Wavelet Transform

PCA - Principle component analysis

**Pattern Mining** 

**Basic Pattern Mining** 

Apriori

FP-Growth

**Advanced Pattern Mining** 

Sequential Pattern Mining

GSP - Generalized Sequential Pattern

Stream Data Pattern Mining

# **Data Mining**

### **Data**

## **Data Preprocessing**

### **Dimensionality Reduction**

**Fourier Transform** 

**Wavelet Transform** 

**Haar Wavelet Transform** 

$$\psi(t) = egin{cases} 1 & 0 \leq t < 1/2, \ -1 & 1/2 \leq t < 1, \ 0 & ext{otherwise.} \end{cases}$$

$$\phi(t) = \left\{ egin{array}{ll} 1 & & 0 \leq t < 1, \ 0 & & ext{otherwise.} \end{array} 
ight.$$

$$h[n] = egin{cases} rac{1}{\sqrt{2}} & ext{if n} = 0, 1 \ 0 & ext{otherwise} \end{cases}$$

#### **PCA - Principle component analysis**

PCA just project raw data to another space, using eigvectors as basis vectors. It reduces some dimensions of the data while remaining most of the information in the data.

1. Calculate covariance matrix *cov\_matrix* for each feature.

$$\sum_i \frac{(a_i - E_a)(b_i - E_b)}{n - 1}$$

- 2. Calculate Eigenvalues  $\lambda_1 \cdots \lambda_n$  and Eigenvectors  $\epsilon_1 \cdots \epsilon_n$  of  $cov\_matrix$
- 3. Order  $\lambda_1 \cdots \lambda_n$  and choose the top-k  $\lambda$ s and related  $\epsilon$ s
- 4. The size of origin data is  $m \times n$ , the transform matrix (size  $n \times k$ ) consists of k vectors  $[\epsilon_1 \cdots \epsilon_k]$ .
- 5. Transformed data is  $origin \times trans$ , and its size is  $m \times k$
- 6. To reconstruct the origin data, just use  $transformed \times trans^T$

## **Pattern Mining**

### **Basic Pattern Mining**

#### **Apriori**

Just one principle:

$$sup(S_1)>=sup(S_2)$$
 when  $S_1\subseteq S_2$ 

So we can do pruning using this principle.

#### **FP-Growth**

- 1. Order the supports of frequent items, and order items in transactions. Record the items (if on conditional FP-tree, also record the condition part).
- 2. Build FP-tree
- 3. Build conditional FP-tree, repeat this process on conditional FP-tree

### **Advanced Pattern Mining**

### **Sequential Pattern Mining**

#### **GSP - Generalized Sequential Pattern**

Just an algorithm to generate candidate (k+1)-sequences from k-sequences.

$$S_1+S_2
ightarrow S_3$$
 and  $S_3=S_1+S_2[-1]$  iff  $S_1[2:]=S_2[:-1]$ 

### **Stream Data Pattern Mining**

Total size N

Error rate  $\sigma$ 

Support rate s

Use buckets to process data, and after each bucket, decrease count by 1.

Pattern with support over  $(s-\sigma)N$  are found