

## Chapter 9: Exploratory Data Analysis (EDA)

- dot product:  $\langle u, v \rangle = u_1 v_1 + u_2 v_2 + \dots$
- norm:  $\|v\| = \sqrt{\langle v, v \rangle}$
- unit vector:  $\|v\| = 1$
- orthogonal:  $\langle v, u \rangle = 0$
- projection =  $\frac{\langle x, u_1 \rangle}{\langle u_1, u_1 \rangle} \cdot u_1 + \frac{\langle x, u_2 \rangle}{\langle u_2, u_2 \rangle} \cdot u_2 + \dots$

First check that,  $u_1$  &  $u_2$  are orthogonal

- PCA
- Transpose:  $d_1 \times d_2$  and  $d_2 \times d_1$
- Square matrix:  $A' = A$
- $A \cdot u = \lambda \cdot u$ ,  $u$  = eigenvector and  $\lambda$  = eigenvalue

• Centroid  $(\bar{x}) = \frac{1}{n} (x_1 + \dots + x_n)$  } Centering

• Replace each  $x_i$  by  $x_i - \bar{x}$

• Normalization:  $\frac{\text{values}}{\text{standard deviation}}$

( $std = \sqrt{\text{Variance}}$ )

Variance =  $\frac{1}{n} (x_1^2 + x_2^2 + \dots)$

- subspace spanned by  $u$
- When the variance along  $u$  is large indicate that the projection along  $u$  are interesting.
- Maximizing variance.
- Let  $u$  a unit vector
  - The projection of  $x$  along vector  $u$  is  $\langle x, u \rangle \cdot u$  (coefficient)
  - The variance is  $\frac{1}{N} \sum_{i=1}^N \langle x_i, u \rangle^2$
  - $N$  times the variance is

$$\sum_{i=1}^N \langle x_i, u \rangle^2 = \sum_{i=1}^N u' x_i (x_i)' u = u' \left( \sum_{i=1}^N x_i (x_i)' \right) u$$

so,  $S = \sum_{i=1}^N x_i (x_i)'$  • the scatter matrix



Variance = eigenvalue.

- The unit vector  $u$  that maximizes  $u'Su$  is same as
  - the eigenvector with largest eigenvalue.
  - the eigenvector is called the first principal component

- Summary of PCA

- centering and normalization
- finding a unit vector  $u$ , such that the variance of the projections along direction  $u$  is maximized
- the variance is  $\frac{1}{n} \cdot u'Su$
- computing the unit vector  $u$  which maximizes  $u'Su$ 
  - eigenvector of  $S$  with the largest eigenvalue

1. centering and normalization

2. compute the scatter matrix

3. compute the eigenvectors & eigenvalues of the scatter matrix.

$$S = \sum_{i=1}^n x_i (x_i)^T$$