

Learning from Data, Tutorial Sheet for Week 3: Answers

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1. Principal Components Analysis in *MATLAB*

See *w3ans.m* for the code

2. Principal Components Analysis

$$\vec{x}^\mu \approx \vec{m} + \sum_{i=1}^M a_i \vec{e}_i$$

a) Optimal coefficients a_i

$$\frac{d}{da_j} (\vec{x}^\mu - \vec{m} - \sum_{i=1}^M a_i \vec{e}_i)^2 = 0 \Leftrightarrow$$

$$\Leftrightarrow 2(\vec{x}^\mu - \vec{m} - \sum_{i=1}^M a_i \vec{e}_i) \frac{d}{da_j} (-\sum_{i=1}^M a_i \vec{e}_i) = 2(\vec{x}^\mu - \vec{m} - \sum_{i=1}^M a_i \vec{e}_i)(-\vec{e}_j) = 0 \Leftrightarrow$$

$$\Leftrightarrow (\vec{x}^\mu - \vec{m}) \cdot \vec{e}_j = \sum_{i=1}^M a_i \vec{e}_i \cdot \vec{e}_j$$

Since the basis set \vec{e}_i is orthonormal (orthogonal¹ and normalised²), we have³:
 $\vec{e}_i \cdot \vec{e}_j = 1, i = j$ and $\vec{e}_i \cdot \vec{e}_j = 0, i \neq j$.

Taking this into account, we get for our optimisation expression:

$$a_j = (\vec{x}^\mu - \vec{m}) \cdot \vec{e}_j$$

b) Calculating distances

$$(\vec{x}^a - \vec{x}^b)^2 \approx (\vec{m} + \sum_{i=1}^M a_i \vec{e}_i - \vec{m} - \sum_{i=1}^M b_i \vec{e}_i)^2$$

$$\begin{aligned} (\sum_{i=1}^M a_i \vec{e}_i - \sum_{i=1}^M b_i \vec{e}_i)^2 &= \sum_{i=1}^M a_i \vec{e}_i \sum_{j=1}^M a_j \vec{e}_j + \sum_{i=1}^M b_i \vec{e}_i \sum_{j=1}^M b_j \vec{e}_j - 2 \sum_{i=1}^M a_i \vec{e}_i \sum_{j=1}^M b_j \vec{e}_j = \\ &= \sum_{i=1}^M a_i \sum_{j=1}^M a_j \vec{e}_i \cdot \vec{e}_j + \sum_{i=1}^M b_i \sum_{j=1}^M b_j \vec{e}_i \cdot \vec{e}_j - 2 \sum_{i=1}^M a_i \sum_{j=1}^M b_j \vec{e}_i \cdot \vec{e}_j \end{aligned}$$

Again, since $\vec{e}_i \cdot \vec{e}_j = 1, i = j$ and $\vec{e}_i \cdot \vec{e}_j = 0, i \neq j$,

$$(\sum_{i=1}^M a_i \vec{e}_i - \sum_{i=1}^M b_i \vec{e}_i)^2 = \sum_{i=1}^M a_i^2 + \sum_{i=1}^M b_i^2 - 2 \sum_{i=1}^M a_i b_i = \vec{a} \cdot \vec{a} + \vec{b} \cdot \vec{b} - 2\vec{a} \cdot \vec{b}$$

$$(\vec{x}^a - \vec{x}^b)^2 \approx (\vec{a} - \vec{b})^2$$

3. Calculating distances

Instead of calculating the distances using the original high-dimensional representation of the data points (considerable computational effort), we can calculate the approximated distances using the lower-dimensional representation.

¹The dot product of any two different vectors of the set is 0

²Length, or norm, equal to 1

³Simply $\vec{e}_i \cdot \vec{e}_j = \delta_{ij}$