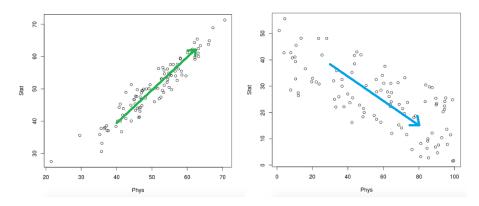
DM561 — Linear Algebra with Applications

Sheet 8, Fall 2020

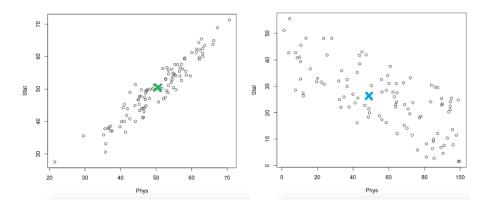
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Exercise 1

1. The direction the data varies the most are marked below



2. The point representing the mean value is marked below (roughly):



Exercise 2

1. The matrix is

$$X' = \begin{bmatrix} 90 & 60 & 90 \\ 90 & 90 & 30 \\ 60 & 60 & 60 \\ 60 & 60 & 90 \\ 30 & 30 & 30 \end{bmatrix}$$

The i'th row represents student i (1-indexed), ie. the feature values for that student. The first column holds math grades, the second english grades, and the third art grades.

2. The averages for the features are

- Math: $\frac{1}{5} \cdot (90 + 90 + 60 + 60 + 30) = 66$
- English: $\frac{1}{5} \cdot (90 + 90 + 60 + 60 + 30) = 60$
- Art: $\frac{1}{5} \cdot (90 + 90 + 60 + 60 + 30) = 60$

For each entry we subtract the corresponding mean for the column. We obtain

$$X = \begin{bmatrix} 66 - 90 & 60 - 60 & 60 - 90 \\ 66 - 90 & 60 - 90 & 60 - 30 \\ 66 - 60 & 60 - 60 & 60 - 60 \\ 66 - 60 & 60 - 60 & 60 - 90 \\ 66 - 30 & 60 - 30 & 60 - 30 \end{bmatrix} = \begin{bmatrix} -24 & 0 & -30 \\ -24 & -30 & 30 \\ 6 & 0 & 0 \\ 6 & 0 & -30 \\ 36 & 30 & 30 \end{bmatrix}$$

It trivial to observe the mean of each column (each feature) is 0.

- 3. The covariance is probably
 - a) positive are relatively high. It seems that there are correlation between the grade for Math and English; when one is high the other is also high.
 - b) close to 0. It seems that there are no tendencies indicating a correlation between grades for English and Art.
- 4. We should use the formulae $C_X = \frac{1}{n}X^TX$. Each row of X correspond to all m=3 measurements (a Mathgrade, an English grade, and an Art grade) from one particular trial (a student). Each column corresponds to all n=5 measurements of a particular type (either Math grades, English grades, or Art grades).

Remark: It important that X is mean centered.

- 5. Let's compute the covariance without Bessel's correction. Let $M = \{90, 90, 60, 60, 30\}$ be the Math grades, $E = \{60, 90, 60, 60, 30\}$ the English grades, and $A = \{90, 30, 60, 90, 30\}$ the Art grades. The mean of
 - $M \text{ is } \bar{m} = 66$
 - E is $\bar{e} = 60$
 - $A \text{ is } \bar{a} = 60$

cf. (2).

Now we can compute

•
$$\sigma_{MM}^2 = \frac{1}{5} \sum_{i=1}^5 (m_i - \bar{m})^2 = 504$$

•
$$\sigma_{EE}^2 = \frac{1}{5} \sum_{i=1}^5 (e_i - \bar{e})^2 = 360$$

•
$$\sigma_{AA}^2 = \frac{1}{5} \sum_{i=1}^5 (a_i - \bar{a})^2 = 720$$

•
$$\sigma_{ME}^2 = \frac{1}{5} \sum_{i=1}^{5} (m_i - \bar{m})(e_i - \bar{e}) = 360$$

•
$$\sigma_{MA}^2 = \frac{1}{5} \sum_{i=1}^5 (m_i - \bar{m})(a_i - \bar{a}) = 180$$

•
$$\sigma_{EA}^2 = \frac{1}{5} \sum_{i=1}^5 (e_i - \bar{e})(a_i - \bar{a}) = 0$$

It would of course have been easier just to use the formulae from (4). Let's do that in (6). Here we also use Bessel's correction.

- 6. We have that
 - a) the covariance matrix with Bessel's correction is

$$C_X = \frac{1}{5-1} X^T X = \begin{bmatrix} 630 & 450 & 225 \\ 450 & 450 & 0 \\ 225 & 0 & 900 \end{bmatrix}$$

b) the covariance matrix without Bessel's correction is

$$C_X = \frac{1}{5}X^T X = \begin{bmatrix} 504 & 360 & 180 \\ 360 & 360 & 0 \\ 180 & 0 & 720 \end{bmatrix}$$

For X we use the mean centered matrix from (2).

Exercise 3

1) We construct a transformation matrix W of the form

$$W = [w_1 \mid w_2]$$

Afterwards, we make the projection onto the new feature space by computing $Y = X \cdot W$. This is not quite the formulae from [1, p. 7] but it's close. By transposing each side we get

$$Y = X \cdot W \Leftrightarrow Y^T = W^T \cdot X^T$$

The matrix W^T corresponds to the matrix P from [1, p. 7] as the row vectors of P should be the new basis vectors for expressing the columns of X^T . The new feature space's basis vectors are the vectors w_1 and w_2 . These are the column vectors of W; thus, they are the row vectors of W^T . The matrix X^T has features as rows and samples as columns (since X had samples as rows and different features as columns). The matrix Y^T is the new representation of the data with principle components as rows (in a sense "new features") and samples as columns.

2) The size of W is 4×2 . The size of X is 150×4 . Since Y is computed as $X \cdot W$ its size becomes 150×2 (in a sense we have the same samples but now the features have changed to the principle components).

Exercise 4

- 1) Just do it!
- 2) Just do it!
- 3) Just do it!
- 4) Add this after cell 8:

```
# First image of test dataset X_test. Note: A row of X_test is an image. v = X_{test}[0]
```

References

[1] Daniel Merkle. Principal component analysis and eigenfaces. URL https://dm561.github.io/assets/DM561-DM562-PCA-Eigenfaces.pdf, 2020.