

Problem Set 1: Matrix Algebra Review

Stat 154, Fall 2017, Prof. Sanchez

Due date: Tu Sep-12 (before midnight)

The purpose of this assignment is to apply in R some of the introductory material, giving you the opportunity to do some work with matrices and vectors.

Use an R markdown (`.Rmd`) file to write your code and answers. You can *knit* the `Rmd` file as html or pdf. Please submit both your `Rmd` and knitted file to bCourses. Make sure to include your name, and your lab section. No late assignments will be accepted.

Problem 1

Create the following matrices in R and compute the operations in parts (a) to (e)

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & -3 \\ 4 & 0 & 1 \end{bmatrix}; \quad \mathbf{B} = \begin{bmatrix} 2 & 3 & 4 \\ -1 & 2 & 0 \end{bmatrix}; \quad \mathbf{C} = \begin{bmatrix} 0 & 1 & 0 \\ 4 & -1 & -2 \end{bmatrix}$$

- a. $\mathbf{A} + \mathbf{B}$
- b. $(\mathbf{A} + \mathbf{C}) + \mathbf{B}$
- c. $\mathbf{A} - (\mathbf{C} + \mathbf{B})$
- d. $-(\mathbf{A} + \mathbf{B})$
- e. $(\mathbf{A} - \mathbf{B}) + \mathbf{C}$

Problem 2

Assume the following matrix \mathbf{X}

$$\mathbf{X} = \begin{array}{ccccc} & Y & X_1 & X_2 & X_3 \\ a & 2 & 1 & 0 & 9 \\ b & 4 & 2 & 3 & 8 \\ c & 3 & 5 & 2 & 4 \\ d & 7 & 3 & 4 & 5 \\ e & 8 & 7 & 7 & 2 \\ f & 9 & 8 & 7 & 1 \end{array}$$

Note that \mathbf{X} has row-names and column-names.

Create the matrix \mathbf{X} in R (with row and column names), and find, via vector-matrix operations:

- a. $\sum Y$
- b. \bar{X}_1
- c. $\sum YX_2$
- d. $\sum X_3^2 - (\sum X_3)^2/6$
- e. the mean-centered matrix
- f. the (sample) covariance matrix \mathbf{S}

You are not allowed to use `sum()`, `apply()`, `sweep()`, nor loops. You can only use these operators: `+`, `-`, `*`, `/` and `%*%`.

Problem 3

Let \mathbf{a} and \mathbf{b} be vectors with given lengths and angle θ . Use R to compute their scalar product under the conditions:

- a. $\|\mathbf{a}\| = 0.5$; $\|\mathbf{b}\| = 4$; $\theta = 45^\circ$
- b. $\|\mathbf{a}\| = 4$; $\|\mathbf{b}\| = 1$; $\theta = 90^\circ$
- c. $\|\mathbf{a}\| = 1$; $\|\mathbf{b}\| = 1$; $\theta = 120^\circ$

Problem 4

Refer to the Gram-Schmidt orthonormalization process described in the following wikipedia entry:

https://en.wikipedia.org/wiki/Gram%E2%80%93Schmidt_process

This procedure is a method for orthonormalizing a set of vectors in an inner product space. In other words, it allows you to find an orthogonal basis for a set of vectors.

Write an R function `proj()` for the *projection operator* given by:

$$proj_{\mathbf{u}}(\mathbf{v}) = \frac{\langle \mathbf{u}, \mathbf{v} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle} \mathbf{u}$$

This projector operator projects the vector \mathbf{v} orthogonally onto the line spanned by vector \mathbf{u} . Given two vectors \mathbf{u} and \mathbf{v} , you should be able to call your function like:

```
proj(v, u)
```

Test `proj(u, v)` with $\mathbf{u} = (1, 3, 5)$ and $\mathbf{v} = (2, 4, 6)$

Problem 5

Once you have your function `proj()`, write R code to apply the Gram-Schmidt orthonormalization procedure to the following sets of vectors:

- a. $\mathbf{x} = (1, 2, 3)$; $\mathbf{y} = (3, 0, 2)$; $\mathbf{z} = (3, 1, 1)$
- b. $\mathbf{x} = (2, 1)$; $\mathbf{y} = (1, 2)$; $\mathbf{z} = (1, 1)$

Start by setting $\mathbf{u}_1 = \mathbf{x}$, and report the set of vectors \mathbf{u}_k and the orthonormalized vectors \mathbf{e}_k , for $k = 1, 2, 3$.

Problem 6

The length of a vector $\mathbf{x} = (x_1, x_2, \dots, x_n)$ in the n -dimensional real vector space \mathbb{R}^n is usually given by the Euclidean norm:

$$\|\mathbf{x}\| = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2}$$

In many situations, the Euclidean distance is insufficient for capturing the actual distances in a given space. The class of p -norms generalizes the notion of *length* of a vector and it is defined by:

$$\|\mathbf{x}\|_p = (|x_1|^p + |x_2|^p + \dots + |x_n|^p)^{\frac{1}{p}}$$

where p is a real number ≥ 1 .

Write a function `lp_norm()` that computes the L_p -norm of a vector. This function should take two arguments:

- \mathbf{x} the input vector
- p the value for p
- Give p a default value of 1
- Allow the user to specify $p = \text{"max"}$ to compute the L_∞ -norm

You should be able to call `lp_norm()` like this:

```
lp_norm(x)           # default p = 1
lp_norm(x, p = 2)
lp_norm(x, p = "max") # L-max norm
```

Test your function `lp_norm()` with the following vectors and values for p :

- a. `zero <- rep(0, 10)` and $p = 1$
- b. `ones <- rep(1, 5)` and $p = 3$
- c. `u <- rep(0.4472136, 5)` and $p = 2$

- d. `u <- -40:0` and `p = 100`
- e. `u <- 1:1000` and `p = "max"`

Problem 7

Show that the set $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ is orthonormal, with:

- $\mathbf{u}_1 = \frac{1}{\sqrt{11}}(3, 1, 1)$
- $\mathbf{u}_2 = \frac{1}{\sqrt{6}}(-1, 2, 1)$
- $\mathbf{u}_3 = \frac{1}{\sqrt{66}}(-1, -4, 7)$

Hint: You need to check that they are orthogonal, and of unit norm.

Problem 8

For this problem, use the data set `USArrests` that comes in R.

- a. Convert `USArrests` as a matrix; this will be the data matrix \mathbf{X} . Confirm that \mathbf{X} is an object of class `"matrix"`
- b. Let n be the number of rows of \mathbf{X} , and p the number of columns of \mathbf{X}
- c. Create a diagonal matrix $\mathbf{D} = \frac{1}{n}\mathbf{I}$ where \mathbf{I} is the $n \times n$ identity matrix. Display the output of `sum(diag(D))`.
- d. Compute $\mathbf{g} = \mathbf{X}^T \mathbf{D} \mathbf{1}$ where $\mathbf{1}$ is a vector of 1's of length n . Display \mathbf{g} .
- e. Calculate the mean-centered matrix $\mathbf{X}_c = \mathbf{X} - \mathbf{1}\mathbf{g}^T$. Display the output of `colMeans(Xc)`.
- f. Compute the (population) variance-covariance matrix $\mathbf{V} = \mathbf{X}^T \mathbf{D} \mathbf{X} - \mathbf{g}\mathbf{g}^T$. Display the output of \mathbf{V} .
- g. Let $\mathbf{D}_{1/S}$ be a $p \times p$ diagonal matrix with elements on the diagonal equal to $1/S_j$, where S_j is the standard deviation for the j -th variable. Display only the elements in the diagonal of $\mathbf{D}_{1/S}$
- h. Compute the matrix of standardized data $\mathbf{Z} = \mathbf{X}_c \mathbf{D}_{1/S}$. Display the output of `colMeans(Z)` and `apply(Z, 2, sd)`
- i. Compute the (population) correlation matrix $\mathbf{R} = \mathbf{D}_{1/S} \mathbf{V} \mathbf{D}_{1/S}$. Display the matrix \mathbf{R}
- j. Confirm that \mathbf{R} can also be obtained as $\mathbf{R} = \mathbf{Z}^T \mathbf{D} \mathbf{Z}$