New York University Applied Statistics and Econometrics I Fall 2015

Problem Set 1

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- (I) The file Data.xls contains prices of 10 highly traded stocks. Write an R program to:
 - 1. Load the data on memory, and create a matrix containing daily returns.
 - 2. Calculate the average return, the standard deviation, and the correlation matrix of the returns. Check whether this matrix is singular or not.
 - 3. Get a histogram of the returns for each stock.
- (II) Write an R program to:
 - 1. Generate a sample of 200 replications of a 4-dimensional multivariate normal random vector with mean zero and variance equal to the identity matrix. Each of the rows of this 200×4 matrix are going to be artificial observations in this exercise.



- 2. Compute the residual maker matrix $M_0 := I i(i'i)^{-1}i'$, where i is a vector of ones. Check that this matrix is idempotent.
- 3. Compute the eigenvalues of M_0 . How many of them are 1. Compute now the rank of M_0 . Is this a coincidence? Prove your statement.
- 4. Compute the values of $a = x'M_0x$ for each row x. Plot a histogram with these values. What is the distribution of the values of a?
- (III) The trace operator $tr(\cdot)$ is defined on the family of squared matrices n-by-n as the sum of the elements of the main diagonal. That is, if $A := [a_{ij}]$, then:

$$tr(A) = \sum_{i=1}^{n} a_{ii}$$

Prove the following properties of the trace operator:

1. The trace is invariant under cyclical permutations (what is a cyclical permutation? An ordering that does not change the relative position of the elements is they are arranged in a ring):



$$tr(ABCD) = tr(BCDA) = tr(CDAB) = tr(DABC)$$

Provide an example of matrices such that the trace of the product of a non-cyclical permutation gives different results, i.e. matrices such that $tr(ABC) \neq tr(ACB)$.

2.
$$tr(X \otimes Y) = tr(X) \cdot tr(Y)$$

(IV) The projection matrix P_X is defined as:

$$P_X := X \left(X^\top X \right)^{-1} X^\top$$

Prove the following properties about the projection matrix:

- 1. P_X is symmetric and idempotent
- $2. P_X X = X$
- 3. The i-th diagonal element of $P_X = X (X^{\top} X)^{-1} X^{\top}$ is:

$$h_{ii} = x_i^\top \left(X^\top X \right)^{-1} x_i$$

which is called the **leverage** of each observation. Prove that:

$$\sum_{i=1}^{n} h_{ii} = tr(P) = k$$

where k is the number of columns in X.

(V) Show the following:

1. The inverse of a diagonal block matrix is given by:



$$\begin{pmatrix} A_{11} & & & & \\ & \cdot & & & \\ & & \cdot & & \\ & & & A_{NN} \end{pmatrix}^{-1} = \begin{pmatrix} A_{11}^{-1} & & & \\ & \cdot & & & \\ & & \cdot & & \\ & & & A_{NN}^{-1} \end{pmatrix}$$

2. The partitioned inverse 2x2 matrix has the inverse:

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}^{-1} = \begin{pmatrix} A_{11}^{-1} + A_{11}^{-1} A_{12} F_2 A_{21} A_{11}^{-1} & -A_{11}^{-1} A_{12} F_2 \\ -F_2 A_{21} A_{11}^{-1} & F_2 \end{pmatrix}$$

where $F_2 = (A_{11} - A_{12}A_{22}^{-1}A_{21})^{-1}$.

(VI) Consider the following matrix:

$$A = \left(\begin{array}{cc} 0.4 & 1 - c \\ 0.6 & c \end{array}\right)$$

- 1. Find all the eigenvalues of A (as a function of c).
- 2. Show that A has just one eigenvector dimension when c=1.6. Find a unitary vector with this direction.

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- 3. When c=0.4, this is a Markov matrix. What happens to A^n when $n\to\infty$? (Hint: use the spectral decomposition of a matrix. What is the *spectral decomposition*? Start here, section 2.3: http://web.stanford.edu/~damle/refresher/notes/matrixanalysis.pdf)
- (VII) A portfolio manager wants to know what is the tradeoff between returns and volatility in his portfolio. There are N assets in the economy. The return of each asset is a random variable whose mean and variance are known. These returns are not independent, so there is a $N \times N$ covariance matrix Ω including all covariances of returns between the assets. Suppose the manager is going to choose a portfolio w, which corresponds to a $N \times 1$ vector including the weights assigned to each asset in the portfolio. For example, if the portfolio has an equal weight for every asset, the vector has 1/N as entry in every position. A portfolio that allocates all the money in the first asset has a first entry 1 and the rest zero. The $N \times 1$ vector of expected returns is denoted by e. In order to study this mean-variance trade-off, the portfolio manager sets a reference expected return μ he wants to reach and calculates the minimum variance he can get by choosing the appropriate weights.
 - 1. Express the variance of the portfolio in terms of w and Ω .
 - 2. Write down the constraints expressing that the weights must be such that the expected return of the portfolio is equal to μ , and the one that expresses that all the weights sum up to one.
 - 3. Write down the variance minimization problem: minimize the portfolio variance subject to the two previous constraints. Write the Langrangian of the problem and get the first order conditions of the optimization problem.
 - 4. Express the optimal weight as a function of μ (without the Lagrange multipliers). You have just characterized the *portolio frontier* in the Markowitz problem. Nobel Prize insight!

(VIII) Give an non-trivial example of the following:



- 1. Two uncorrelated random variables X and Y that are not independent.
- 2. Two different discrete random variables X and Y with the same distribution.
- 3. Two normal random variables X and Y that are not jointly normal.
- (IX) The random variables X and Y are defined on \mathbb{N} .
 - 1. Suppose that know the joint probability distribution of X and Y:

$$\mathbb{P}(a \le x \le b, c \le y \le d) = \sum_{a \le x \le b} \sum_{c \le y \le d} f(x, y)$$

• Explain how to obtain the marginal distribution of X and Y from f(x,y)

 \bullet Define the covariance between X and Y. Show that:

$$Cov(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y)$$

- ullet State the condition we need to check that X and Y are independent
- 2. Let $\mathbb{P}_X(x=1) = 0.5$, $\mathbb{P}_X(x=2) = 0.3$, $\mathbb{P}_X(x=3) = 0.2$. Define $Y = X^2$. Derive the marginal distribution for both random variables, their covariance, and check if whether they are independent or not.