

# 1 Matrix Algebra Review

Matrix

$$\mathbf{A}_{(n \times m)} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1m} \\ a_{21} & a_{22} & \cdots & a_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{bmatrix}$$

$n = \# \text{ of rows}, m = \# \text{ of columns}$

Square matrix :  $n = m$

Vector

$$\mathbf{x}_{(n \times 1)} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

## Remarks

- R is a matrix oriented programming language
- Excel can handle matrices and vectors in formulas and some functions
- Excel has special functions for working with matrices. There are called *array* functions. Must use

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to evaluate array function

## 1.1 Transpose of a Matrix

Interchange rows and columns of a matrix

$$\underset{(m \times n)}{\mathbf{A}}' = \text{transpose of } \underset{(n \times m)}{\mathbf{A}}$$

Example

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, \quad \mathbf{A}' = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

$$\mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \quad \mathbf{x}' = [1 \ 2 \ 3]$$

R function

`t(A)`

Excel function

`TRANSPOSE(matrix)`

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## 1.2 Symmetric Matrix

A square matrix  $\mathbf{A}$  is symmetric if

$$\mathbf{A} = \mathbf{A}'$$

Example

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}, \mathbf{A}' = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$$

Remark: Covariance and correlation matrices are symmetric

## 1.3 Basic Matrix Operations

### 1.3.1 Addition and Subtraction (element-by-element)

$$\begin{aligned} \begin{bmatrix} 4 & 9 \\ 2 & 1 \end{bmatrix} + \begin{bmatrix} 2 & 0 \\ 0 & 7 \end{bmatrix} &= \begin{bmatrix} 4 + 2 & 9 + 0 \\ 2 + 0 & 1 + 7 \end{bmatrix} \\ &= \begin{bmatrix} 6 & 9 \\ 2 & 8 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} 4 & 9 \\ 2 & 1 \end{bmatrix} - \begin{bmatrix} 2 & 0 \\ 0 & 7 \end{bmatrix} &= \begin{bmatrix} 4 - 2 & 9 - 0 \\ 2 - 0 & 1 - 7 \end{bmatrix} \\ &= \begin{bmatrix} 2 & 9 \\ 2 & -6 \end{bmatrix} \end{aligned}$$

### 1.3.2 Scalar Multiplication (element-by-element)

$$\begin{aligned}c &= 2 = \text{scalar} \\A &= \begin{bmatrix} 3 & -1 \\ 0 & 5 \end{bmatrix} \\2 \cdot A &= \begin{bmatrix} 2 \cdot 3 & 2 \cdot (-1) \\ 2 \cdot 0 & 2 \cdot 5 \end{bmatrix} = \begin{bmatrix} 6 & -2 \\ 0 & 10 \end{bmatrix}\end{aligned}$$

### 1.3.3 Matrix Multiplication (not element-by-element)

$$\begin{matrix} \mathbf{A} \\ (3 \times 2) \end{matrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}, \quad \begin{matrix} \mathbf{B} \\ (2 \times 3) \end{matrix} = \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{bmatrix}$$

Note:  $\mathbf{A}$  and  $\mathbf{B}$  are conformable matrices: # of columns in  $A = \#$  of rows in  $B$

$$\begin{matrix} \mathbf{A} \cdot \mathbf{B} \\ (3 \times 2) \quad (2 \times 3) \end{matrix} = \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} & a_{11}b_{13} + a_{12}b_{23} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} & a_{21}b_{13} + a_{22}b_{23} \\ a_{31}b_{11} + a_{32}b_{21} & a_{31}b_{12} + a_{32}b_{22} & a_{31}b_{13} + a_{32}b_{23} \end{bmatrix}$$

Remark: In general,

$$\mathbf{A} \cdot \mathbf{B} \neq \mathbf{B} \cdot \mathbf{A}$$

Example

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$
$$\mathbf{A} \cdot \mathbf{B} = \begin{bmatrix} 5 + 14 & 6 + 16 \\ 15 + 28 & 18 + 32 \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}$$

R operator

`A**B`

Excel function

`MMULT(matrix1, matrix2)`  
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### 1.3.4 Identity Matrix

The  $n$ -dimensional identity matrix has all diagonal elements equal to 1, and all off diagonal elements equal to 0.

Example

$$\mathbf{I}_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Remark: The identity matrix plays the roll of “1” in matrix algebra

$$\begin{aligned} \mathbf{I}_2 \cdot \mathbf{A} &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \\ &= \begin{bmatrix} a_{11} + 0 & a_{12} + 0 \\ 0 + a_{21} & 0 + a_{22} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \\ &= \mathbf{A} \end{aligned}$$

Similarly

$$\begin{aligned} \mathbf{A} \cdot \mathbf{I}_2 &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \mathbf{A} \end{aligned}$$

R function

`diag(n)`

creates  $n$ -dimensional identity matrix

## 1.3.5 Matrix Inverse

Let  $\mathbf{A}$   $\underset{(n \times n)}{=}$  square matrix.  $\mathbf{A}^{-1}$  = “inverse of  $\mathbf{A}$ ” satisfies

$$\mathbf{A}^{-1}\mathbf{A} = \mathbf{I}_n$$

$$\mathbf{A}\mathbf{A}^{-1} = \mathbf{I}_n$$

Remark:  $\mathbf{A}^{-1}$  is similar to the inverse of a number:

$$a = 2, a^{-1} = \frac{1}{2}$$

$$a \cdot a^{-1} = 2 \cdot \frac{1}{2} = 1$$

$$a^{-1} \cdot a = \frac{1}{2} \cdot 2 = 1$$

R function

```
solve(A)
```

Excel function

```
MINVERSE(matrix)
```

```
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```

## Representing Systems of Linear Equations Using Matrix Algebra

Consider the system of two linear equations

$$\begin{aligned}x + y &= 1 \\2x - y &= 1\end{aligned}$$

The equations represent two straight lines which intersect at the point

$$x = \frac{2}{3}, y = \frac{1}{3}$$

Matrix algebra representation:

$$\begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

or

$$\mathbf{A} \cdot \mathbf{z} = \mathbf{b}$$

where

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix}, \mathbf{z} = \begin{bmatrix} x \\ y \end{bmatrix} \text{ and } \mathbf{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

We can solve for  $z$  by multiplying both sides by  $\mathbf{A}^{-1}$

$$\begin{aligned}\mathbf{A}^{-1} \cdot \mathbf{A} \cdot \mathbf{z} &= \mathbf{A}^{-1} \cdot \mathbf{b} \\ \implies \mathbf{I} \cdot \mathbf{z} &= \mathbf{A}^{-1} \cdot \mathbf{b} \\ \implies \mathbf{z} &= \mathbf{A}^{-1} \cdot \mathbf{b}\end{aligned}$$

or

$$\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Remark: As long as we can determine the elements in  $\mathbf{A}^{-1}$ , we can solve for the values of  $x$  and  $y$  in the vector  $\mathbf{z}$ . Since the system of linear equations has a solution as long as the two lines intersect, we can determine the elements in  $\mathbf{A}^{-1}$  provided the two lines are not parallel.

There are general numerical algorithms for finding the elements of  $\mathbf{A}^{-1}$  and typical spreadsheet programs like Excel have these algorithms available. However, if  $\mathbf{A}$  is a  $(2 \times 2)$  matrix then there is a simple formula for  $\mathbf{A}^{-1}$ . Let

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}.$$

Then

$$\mathbf{A}^{-1} = \frac{1}{\det(\mathbf{A})} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}.$$

where

$$\det(\mathbf{A}) = a_{11}a_{22} - a_{21}a_{12}$$

Let's apply the above rule to find the inverse of  $\mathbf{A}$  in our example:

$$\mathbf{A}^{-1} = \frac{1}{-1 - 2} \begin{bmatrix} -1 & -1 \\ -2 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & \frac{-1}{3} \end{bmatrix}.$$

Notice that

$$\mathbf{A}^{-1}\mathbf{A} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & \frac{-1}{3} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

Our solution for  $\mathbf{z}$  is then

$$\begin{aligned} \mathbf{z} &= \mathbf{A}^{-1}\mathbf{b} \\ &= \begin{bmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & \frac{-1}{3} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} \end{aligned}$$

so that  $x = \frac{2}{3}$  and  $y = \frac{1}{3}$ .

In general, if we have  $n$  linear equations in  $n$  unknown variables we may write the system of equations as

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n &= b_2 \\ &\vdots = \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n &= b_n \end{aligned}$$

which we may then express in matrix form as

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

or

$$\underset{(n \times n)}{\mathbf{A}} \cdot \underset{(n \times 1)}{\mathbf{x}} = \underset{(n \times 1)}{\mathbf{b}}.$$

The solution to the system of equations is given by

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$$

where  $\mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$  and  $\mathbf{I}$  is the  $(n \times n)$  identity matrix. If the number of equations is greater than two, then we generally use numerical algorithms to find the elements in  $\mathbf{A}^{-1}$ .

## 1.4 Representing Summation Using Matrix Notation

$$\sum_{i=1}^n x_i = x_1 + x_2 + \cdots + x_n$$

$$\underset{(n \times 1)}{\mathbf{x}} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad \underset{(n \times 1)}{\mathbf{1}} = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}$$

Then

$$\begin{aligned}\mathbf{x}'\mathbf{1} &= \begin{pmatrix} x_1 & x_2 & \cdots & x_n \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix} \\ &= x_1 + x_2 + \cdots + x_n = \sum_{i=1}^n x_i\end{aligned}$$

Equivalently

$$\begin{aligned}\mathbf{1}'\mathbf{x} &= \begin{pmatrix} 1 & 1 & \cdots & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \\ &= x_1 + x_2 + \cdots + x_n = \sum_{i=1}^n x_i\end{aligned}$$

## Sum of Squares

$$\sum_{i=1}^n x_i^2 = x_1^2 + x_2^2 + \cdots + x_n^2$$

$$\begin{aligned} \mathbf{x}'\mathbf{x} &= \begin{pmatrix} x_1 & x_2 & \cdots & x_n \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \\ &= x_1^2 + x_2^2 + \cdots + x_n^2 = \sum_{i=1}^n x_i^2 \end{aligned}$$

## Sums of cross products

$$\sum_{i=1}^n x_i y_i = x_1 y_1 + x_2 y_2 + \cdots + x_n y_n$$

$$\mathbf{x}'\mathbf{y} = \begin{pmatrix} x_1 & x_2 & \cdots & x_n \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}$$

$$= x_1 y_1 + x_2 y_2 + \cdots + x_n y_n = \sum_{i=1}^n x_i y_i$$

$$= \mathbf{y}'\mathbf{x}$$

R function

```
crossprod(x,y)
```

Excel function

```
MMULT(TRANSPOSE(x),y)
```

```
MMULT(TRANSPOSE(y),x)
```

```
<ctrl>-<shift>-<enter>
```

## Portfolio Math with Matrix Algebra

### Three Risky Asset Example

Let  $R_i$  ( $i = A, B, C$ ) denote the return on asset  $i$  and assume that  $R_i$  follows CER model:

$$R_i \sim iid N(\mu_i, \sigma_i^2)$$
$$\text{cov}(R_i, R_j) = \sigma_{ij}$$

Portfolio “ $\mathbf{x}$ ”

$x_i$  = share of wealth in asset  $i$

$$x_1 + x_2 + x_3 = 1$$

Portfolio return

$$R_{p,x} = x_A R_A + x_B R_B + x_C R_C.$$

Portfolio expected return

$$\mu_{p,x} = E[R_{p,x}] = x_A\mu_A + x_B\mu_B + x_C\mu_C$$

Portfolio variance

$$\begin{aligned}\sigma_{p,x}^2 = \text{var}(R_{p,x}) &= x_A^2\sigma_A^2 + x_B^2\sigma_B^2 + x_C^2\sigma_C^2 \\ &+ 2x_Ax_B\sigma_{AB} + 2x_Ax_C\sigma_{AC} + 2x_Bx_C\sigma_{BC}\end{aligned}$$

Portfolio distribution

$$R_{p,x} \sim N(\mu_{p,x}, \sigma_{p,x}^2)$$

## Matrix Algebra Representation

$$\mathbf{R} = \begin{pmatrix} R_A \\ R_B \\ R_C \end{pmatrix}, \quad \boldsymbol{\mu} = \begin{pmatrix} \mu_A \\ \mu_B \\ \mu_C \end{pmatrix}, \quad \mathbf{1} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$
$$\mathbf{x} = \begin{pmatrix} x_A \\ x_B \\ x_C \end{pmatrix}, \quad \boldsymbol{\Sigma} = \begin{pmatrix} \sigma_A^2 & \sigma_{AB} & \sigma_{AC} \\ \sigma_{AB} & \sigma_B^2 & \sigma_{BC} \\ \sigma_{AC} & \sigma_{BC} & \sigma_C^2 \end{pmatrix}$$

Portfolio weights sum to 1

$$\begin{aligned} \mathbf{x}'\mathbf{1} &= (x_A \ x_B \ x_C) \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= x_1 + x_2 + x_3 = 1 \end{aligned}$$

Portfolio return

$$\begin{aligned}R_{p,x} &= \mathbf{x}'\mathbf{R} = (x_A \ x_B \ x_C) \begin{pmatrix} R_A \\ R_B \\ R_C \end{pmatrix} \\ &= x_A R_A + x_B R_B + x_C R_C \\ &= \mathbf{R}'\mathbf{x}\end{aligned}$$

Portfolio expected return

$$\begin{aligned}\mu_{p,x} &= \mathbf{x}'\boldsymbol{\mu} = (x_A \ x_B \ x_C) \begin{pmatrix} \mu_A \\ \mu_B \\ \mu_C \end{pmatrix} \\ &= x_A \mu_A + x_B \mu_B + x_C \mu_C \\ &= \boldsymbol{\mu}'\mathbf{x}\end{aligned}$$

Excel formula

```
MMULT(transpose(xvec),muvec)
```

```
<ctrl>-<shift>-<enter>
```

R formula

```
crossprod(x,mu)
```

```
t(x)%*%mu
```

Portfolio variance

$$\begin{aligned}\sigma_{p,x}^2 &= \mathbf{x}'\Sigma\mathbf{x} \\ &= (x_A \ x_B \ x_C) \begin{pmatrix} \sigma_A^2 & \sigma_{AB} & \sigma_{AC} \\ \sigma_{AB} & \sigma_B^2 & \sigma_{BC} \\ \sigma_{AC} & \sigma_{BC} & \sigma_C^2 \end{pmatrix} \begin{pmatrix} x_A \\ x_B \\ x_C \end{pmatrix} \\ &= x_A^2\sigma_A^2 + x_B^2\sigma_B^2 + x_C^2\sigma_C^2 \\ &\quad + 2x_Ax_B\sigma_{AB} + 2x_Ax_C\sigma_{AC} + 2x_Bx_C\sigma_{BC}\end{aligned}$$

## Excel formulas

```
MMULT(TRANSPOSE(xvec),MMULT(sigma,xvec))
```

```
MMULT(MMULT(TRANSPOSE(xvec),sigma),xvec)
```

```
<ctrl>-<shift>-<enter>
```

## R formulas

```
t(x)%*%sigma%*%x
```

## Portfolio distribution

$$R_{p,x} \sim N(\mu_{p,x}, \sigma_{p,x}^2)$$

## Covariance Between 2 Portfolio Returns

2 portfolios

$$\mathbf{x} = \begin{pmatrix} x_A \\ x_B \\ x_C \end{pmatrix}, \mathbf{y} = \begin{pmatrix} y_A \\ y_B \\ y_C \end{pmatrix}$$
$$\mathbf{x}'\mathbf{1} = 1, \mathbf{y}'\mathbf{1} = 1$$

Portfolio returns

$$R_{p,x} = \mathbf{x}'\mathbf{R}$$
$$R_{p,y} = \mathbf{y}'\mathbf{R}$$

Covariance

$$\text{cov}(R_{p,x}, R_{p,y}) = \mathbf{x}'\Sigma\mathbf{y}$$
$$= \mathbf{y}'\Sigma\mathbf{x}$$

Excel formula

```
MMULT(TRANSPOSE(xvec),MMULT(sigma,yvec))
```

```
MMULT(TRANSPOSE(yvec),MMULT(sigma,xvec))
```

```
<ctrl>-<shift>-<enter>
```

R formula

```
t(x)%*%sigma%*%y
```