

MacAnova Operations on Columns

Note: These differ from R and S-plus

```
Cmd> a # previously entered 4 by 3 matrix
(1,1)      1      2      -3
(2,1)      0     11      4
(3,1)      7     17     12
(4,1)     -3      4      5
```

Sum down columns

```
Cmd> sum(a)#sum down columns; row vector (1 by 4 matrix)
(1,1)      5      34     18
```

```
Cmd> sum(a,margin:2) # sum down columns; ordinary vector
(1)        5      34     18
```

(1,1) indicates result has 2 dimensions

(1) indicates result has 1 dimensions

Sum across rows

```
Cmd> sum(a') # sum across rows, row vector result
(1,1)      0     15     36      6
```

```
Cmd> sum(a')' # sum across rows, column vector sums
(1,1)      0
(2,1)     15
(3,1)     36
(4,1)      6
```

```
Cmd> sum(a,margin:1) # sum across rows, ordinary vector result
(1)        0     15     36      6
```

Products down columns

```
Cmd> prod(a) # multiply down columns
(1,1)      0     1496    -720
```

Extremes of columns

```
Cmd> min(a) # row vector result
(1,1)     -3      2      -3
```

```
Cmd> max(a) # row vector result
(1,1)      7     17     12
```

Displays for Statistics 5401

Lecture 3

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Class Web Page

<http://www.stat.umn.edu/~kb/classes/5401>

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Ordering columns by `sort()`

```
Cmd> sort(a) # Each column sorted in increasing order
(1,1)      -3      2      -3
(2,1)      0      4      4
(3,1)      1     11      5
(4,1)      7     17     12

Cmd> sort(a,down:T) # decreasing order
(1,1)      7     17     12
(2,1)      1     11      5
(3,1)      0      4      4
(4,1)     -3      2     -3
```

`grade()` is related to `sort()`:

```
Cmd> grade(a)
(1,1)      4      1      1 Row #s of col minima
(2,1)      2      4      2 Row #s of 2nd
(3,1)      1      2      4 Row #s of 3rd
(4,1)      3      3      3 Row #s of col maxima
```

The numbers in row i are the row numbers of the i^{th} smallest entries in each column. You can also use `down:T`.

```
Cmd> grade(a,down:T) # row numbers of elements in down order
(1,1)      3      3      3 Row #s of col maxima
(2,1)      1      2      4
(3,1)      2      4      2
(4,1)      4      1      1 Row #s of col minima
```

Reorder the rows of **a** so that column 1 is increasing.

```
Cmd> J <- grade(a[,1]) # grade() applied to just column 1 of a
Cmd> a[J,] # Rows now permuted so column 1 is sorted
(1,1)      -3      4      5
(2,1)      0     11      4
(3,1)      1      2     -3
(4,1)      7     17     12
```

Matrix Multiplication $A \times C = A C$

MacAnova: `a %**% b` *not* `a * b`

```
Cmd> c <- matrix(enter(1 3 5 2 4 6),3) # 3x2
Cmd> c
(1,1)      1      2
(2,1)      3      4
(3,1)      5      6
3 by 2

Cmd> b <- matrix(run(12),nrows(a)) # 4 by 3
Cmd> b
(1,1)      1      5      9
(2,1)      2      6     10
(3,1)      3      7     11
(4,1)      4      8     12
4 by 3
```

$A \times C = A C$ (A times C)

```
Cmd> a %**% c # (4 by 3) %**% (3 by 2)
(1,1)      -8      -8
(2,1)      53      68
(3,1)     118     154
(4,1)      34      40
4 by 2
```

$A \times B' = A B'$ (A times transpose of B)

```
Cmd> a %**% b' # 4 by 3 %**% 3 by 4
(1,1)     -16     -16     -16     -16
(2,1)      91     106     121     136
(3,1)     200     236     272     308
(4,1)      62      68      74      80
4 by 4
```

For **A B** to be defined, number of columns of **A** must = number of rows of **B**:

```
Cmd> a %**% b # 4 by 3 %**% 4 by 3 is an error
ERROR: Dimension mismatch: 4 by 3 %**% 4 by 3 near
a %**% b # 4 by 3 %**% 4 by 3
```

This produced an error because `ncols(a) ≠ nrows(b)`

Two *short cuts*:

`a %c% b` is the same as `a' %*% b`

`a %C% b` is the same as `a %*% b'`

```
Cmd> equal(a %c% b, a' %*% b)
(1) T
      T means True, a %c% b equals a' %*% b
```

```
Cmd> equal(a %C% b, a %*% b')
(1) T
```

`equal(x,y)` has value `T` (True) if and only if `x` and `y` are identical. Otherwise it has value `F` (False).

Help topic `arithmetic` summarizes the arithmetic operations and details of their use.

Help topic `matrices` has info on matrix multiplication and most functions that compute things from matrices such as eigenvalues and determinants.

To read these topics, you can

- use **Help** on the **Help** menu and then click on the topic, or
- type `help(matrices)` or `help(arithmetic)`.

Notational Conventions

You can view a matrix in several ways.

Suppose $A = [a_{ij}]_{1 \leq i \leq n, 1 \leq j \leq p}$ is an n by p matrix, that is, a rectangular table:

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1p} \\ a_{21} & a_{22} & \dots & a_{2p} \\ a_{31} & a_{32} & \dots & a_{3p} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{np} \end{bmatrix} \quad n \text{ rows, } p \text{ columns}$$

- You can view A as a set of n 1 by p row vectors, $\mathbf{a}_i' = [a_{i1}, a_{i2}, \dots, a_{ip}]$,

$$A = \begin{bmatrix} \mathbf{a}_1' \\ \mathbf{a}_2' \\ \dots \\ \mathbf{a}_n' \end{bmatrix} = [\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n]'$$

where $\mathbf{a}_i = \begin{bmatrix} a_{i1} \\ a_{i2} \\ \dots \\ a_{ip} \end{bmatrix}, i = 1, \dots, n$

\mathbf{a}_i' is 1 by p

- You can view **A** as a set of p $n \times 1$ *column* vectors

$$\mathbf{A} = [\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_p]$$

where $\mathbf{A}_j = \begin{bmatrix} a_{1j} \\ a_{2j} \\ a_{3j} \\ \vdots \\ a_{nj} \end{bmatrix} = [a_{1j}, a_{2j}, \dots, a_{nj}]'$

I often use the convention:

- Lower case letters such as **a** stand for the rows of a matrix
- Uppercase letters such as **A** stand for the columns.

$$\mathbf{A} = \begin{bmatrix} \mathbf{a}_1' \\ \mathbf{a}_2' \\ \dots \\ \mathbf{a}_n' \end{bmatrix} = [\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_p].$$

When **A** is a data matrix for n cases and p variables, \mathbf{a}_i has the data for *case* i and \mathbf{A}_j has the data for *variable* j .

Vocabulary: Inner product

When **a** and **b** are vectors with the *same length* p , their *inner product* is

$$\sum_{1 \leq i \leq p} a_i b_i$$

that is, an inner product is a *sum of products*. This is a useful operation because you use a lot of sums of products in statistics.

```
Cmd> x <- vector(3.32,3.00,1.61,2.53,3.61)
Cmd> y <- vector(11.21,10.64,8.16,9.47,10.41)
Cmd> innerxy <- sum(x*y); innerxy # x*y is elementwise product
(1)      143.81      Inner product of x and y
```

Inner product of **a** with itself is

$$\|\mathbf{a}\|^2 \equiv \sum_i a_i^2, \text{ a sum of } \textit{squares}.$$

You use a lot of sums of squares, too.

Vocabulary: $\|\mathbf{a}\| = \sqrt{\{\sum_i a_i^2\}}$ is the *norm* or *length* of **a**.

```
Cmd> normx <- sqrt(sum(x^2)); normx
(1)      6.4844
```

This is not the same as `length(a)` = number of elements of **a** in MacAnova.

Matrix product of row and column vectors

Suppose **a** and **b** are vectors of the same length. Then **a'** is a **row** vector and **b** is a **column** vector.

The matrix products $\mathbf{a}' \times \mathbf{b} = \mathbf{a}'\mathbf{b}$ and $\mathbf{b}' \times \mathbf{a} = \mathbf{b}'\mathbf{a}$ have the same value.

It is *defined to be* the inner product of **a** and **b**. That is

$$\mathbf{a}'\mathbf{b} = \mathbf{b}'\mathbf{a} = \sum_i a_i b_i$$

```
Cmd> x' %*% y
(1,1)      143.81
```

```
Cmd> y' %*% x
(1,1)      143.81
```

```
Cmd> innerxy # computed previously as sum(x*y)
(1)        143.81
```

A particular case is

$$\mathbf{a}'\mathbf{a} = \sum_i a_i^2 = \|\mathbf{a}\|^2$$

So you can represent both sums of products and sums of squares by matrix products.

```
Cmd> x' %*% x # same as sum(x^2)
(1,1)        42.047
```

Note that products $\mathbf{a}'\mathbf{b}$ and $\mathbf{b}'\mathbf{a}$ have the form

$$(1 \text{ by } p \text{ matrix}) \times (p \text{ by } 1 \text{ matrix})$$

The number of columns on the left = the number of rows on the right.

```
Cmd> a <- vector(78.2, 69.5, 32.4, 52.0, 66.2)
```

```
Cmd> b <- vector(56.5, 26.9, 54.5, 38.9, 67.3)
```

```
Cmd> sum(a*b) # inner product
```

```
(1)      14532
```

```
Cmd> a' %*% b # matrix product of a %c% b
```

```
(1,1)      14532
```

```
Cmd> b' %*% a # matrix product or b %c% a
```

```
(1,1)      14532
```

```
Cmd> a %*% b' # this is ab', not yet defined; differs from b'%*%a
```

```
(1,1)      4418.3      2103.6      4261.9      3042      5262.9
(2,1)      3926.8      1869.5      3787.8      2703.5      4677.3
(3,1)      1830.6      871.56      1765.8      1260.4      2180.5
(4,1)      2938      1398.8      2834      2022.8      3499.6
(5,1)      3740.3      1780.8      3607.9      2575.2      4455.3
```

This last illustrates that in general

$$\mathbf{AB} \neq \mathbf{BA}$$

Of course, I haven't yet even defined what \mathbf{ab}' is.

In fact, the i,j element of \mathbf{ab}' is $a_i b_j$.

Matrix multiplication in terms of inner products

You can define matrix multiplication for larger matrices in terms of inner products $\mathbf{a}_i' \mathbf{B}_j$.

Suppose \mathbf{A} is m by n and \mathbf{B} is n by q

$$\mathbf{A} = [\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_m]', \text{ with } \mathbf{a}_i \text{ } n \text{ by } 1$$

$$\mathbf{B} = [\mathbf{B}_1, \mathbf{B}_2, \dots, \mathbf{B}_q], \text{ with } \mathbf{B}_i \text{ } n \text{ by } 1$$

The matrix product $\mathbf{C} = [c_{ij}] = \mathbf{A} \mathbf{B}$ is *defined* by the inner products $c_{ij} \equiv \mathbf{a}_i' \mathbf{B}_j$

That is, $c_{ij} = (\text{inner product of } \mathbf{a}_i \text{ and } \mathbf{B}_j)$
 $= (\text{matrix product of row } i \text{ of } \mathbf{A} (\mathbf{a}_i')$ and column j of $\mathbf{B} (\mathbf{B}_j)$).

In more familiar terms

$$c_{ij} = \sum_{1 \leq k \leq n} a_{ik} b_{kj}, \quad i = 1, \dots, m, \quad j = 1, \dots, q$$

An important case

Suppose $n \times p$ $\mathbf{X} = [\mathbf{X}_1 \mathbf{X}_2 \dots \mathbf{X}_p]$ is a matrix of *data* on p variables for n cases.

Then the columns of \mathbf{X} and the rows of \mathbf{X}' are the vectors \mathbf{X}_j , $j = 1, \dots, p$, each of which have n elements. Therefore

$$\mathbf{X}'\mathbf{X} = [\mathbf{X}_j' \mathbf{X}_k]_{1 \leq j, k \leq p}$$

But

$\mathbf{X}_j' \mathbf{X}_k = \sum_{1 \leq i \leq n} x_{ij} x_{ik}$ and $\mathbf{X}_j' \mathbf{X}_j = \sum_{1 \leq i \leq n} x_{ij}^2$
 are **sums of products and squares**.

Similarly, if $\mathbf{Y} = [\mathbf{Y}_1 \mathbf{Y}_2 \dots \mathbf{Y}_q]$ is $n \times q$,

$$\mathbf{X}'\mathbf{Y} = [\mathbf{X}_j' \mathbf{Y}_k]_{1 \leq j \leq p, 1 \leq k \leq q}$$

where $\mathbf{X}_j' \mathbf{Y}_k = \sum_{1 \leq i \leq n} x_{ij} y_{ik}$ is a **sum of products**.

Matrix Multiplication in terms of outer products

Suppose

$x = [x_i]_{1 \leq i \leq m}$ a m -vector (m by 1 matrix)

$y = [y_i]_{1 \leq i \leq n}$ a n -vector (n by 1 matrix)

m and n may be different.

Vocabulary: The *outer product* $x \otimes y$ is defined to be the $m \times n$ matrix

$$x \otimes y = \begin{bmatrix} x_1 y_1 & x_1 y_2 & \dots & x_1 y_n \\ x_2 y_1 & x_2 y_2 & \dots & x_2 y_n \\ \dots & \dots & \dots & \dots \\ x_m y_1 & x_m y_2 & \dots & x_m y_n \end{bmatrix} = x y'$$

$m \times 1 \quad 1 \times n$

```
Cmd> x <- vector(29,4,-2,-58) # m = 4
Cmd> y <- vector(70,45,40,34,-147) # n = 5
Cmd> outer(x,y) # or x %**% y' or x %C% y
(1,1) 2030 1305 1160 986 -4263
(2,1) 280 180 160 136 -588
(3,1) -140 -90 -80 -68 294
(4,1) -4060 -2610 -2320 -1972 8526
Cmd> x %**% y'
(1,1) 2030 1305 1160 986 -4263
(2,1) 280 180 160 136 -588
(3,1) -140 -90 -80 -68 294
(4,1) -4060 -2610 -2320 -1972 8526
Cmd> x[4]*y[5] # check row 4 and column 5 of result
(1) 8526
```

Suppose A is m by n and B is n by q .

You can define the matrix product AB in terms of the *outer products* $A_j b_j'$ of the *columns* A_j of A and the *rows* b_j' of B .

$$AB = \sum_{1 \leq j \leq n} A_j b_j' = A_1 b_1' + A_2 b_2' + \dots + A_n b_n'$$

= sum of n outer products, all m by q

```
Cmd> a <- matrix(vector(1.04,0.696,-0.651,0.13,1.5,1.61), 2); a
(1,1) 1.04 -0.651 1.5 2 by 3
(2,1) 0.696 0.13 1.61
Cmd> b <- matrix(vector(0.367,0.658,0.796,\
0.024,-0.784,-0.851, 0.352,-0.082,1.143),3); b
(1,1) 0.367 0.024 0.352
(2,1) 0.658 -0.784 -0.082 3 by 3
(3,1) 0.796 -0.851 1.143
Cmd> a %**% b # matrix product
(1,1) 1.1473 -0.74116 2.134 2 by 3
(2,1) 1.6225 -1.4553 2.0746
Cmd> # Here is same thing as sum of outer prods
Cmd> a[,1] %**% b[1,] + a[,2] %**% b[2,] + a[,3] %**% b[3,]
(1,1) 1.1473 -0.74116 2.134
(2,1) 1.6225 -1.4553 2.0746
```

For example, $a[,2] \%**\% b[2,] =$ outer product of column 2 of a and row 2 of b

This might be a good time to show how “looping” can be helpful. I computed the same product by using a `for` loop.

I started by creating a 2 by 3 matrix of 0's to hold the sum of outer products

```
Cmd> s <- matrix(rep(0,6),2);s # set initial value of s to 0
(1,1)      0      0      0
(2,1)      0      0      0

Cmd> for(i,1,3){ # same thing using a loop
  s <- s + a[,i] %*% b[i,];
}

Cmd> s
(1,1)      1.1473     -0.74116      2.134
(2,1)      1.6225     -1.4553      2.0746
```

The “loop” was repeated 3 times, with $i = 1$, $i = 2$ and $i = 3$.

Note the “`;;`” after the last (and only) command in the loop. If that is omitted, output may be produced on every trip around a loop.

Note: Except in exceptional cases

$$AB \neq BA$$

That is, order of multiplication matters.

Notational conventions again

I will always use the conventions

- **Single** multivariate observation is column vector \mathbf{x}
- **Column** of data matrix \Leftrightarrow **Variable**
- **Row** of data matrix \Leftrightarrow **Case**
notated \mathbf{x}_i' , where \mathbf{x}_i is column vector

Linear Combinations of Variables

Often in statistics you are interested in a *linear combination* of variables in a multivariate vector $\mathbf{x} = [x_1, \dots, x_p]'$.

Vocabulary: A **linear combination** has the form

$$y = c_1x_1 + c_2x_2 + \dots + c_px_p = \sum_{1 \leq i \leq p} c_i x_i$$

Example:

$$\begin{aligned} y &= x_3 - x_1 \\ &= (-1)x_1 + (0)x_2 + (1)x_3 + \dots + (0)x_p, \end{aligned}$$

This is the inner product of

$$\mathbf{c} = [c_1, c_2, \dots, c_p]' = [-1, 0, 1, \dots, 0]'$$

and

$$\mathbf{x} = [x_1, x_2, \dots, x_p]'$$

You can write y as the matrix product

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

You should learn to recognize $\mathbf{c}'\mathbf{x}$ as representing a linear combination.

Symmetric Matrices

Vocabulary: **Square, symmetric**

If \mathbf{A} is a p by p matrix, then we say it is *square*.

A square matrix \mathbf{A} is *symmetric* when $a_{ji} = a_{ij}$ all i and j .

$$\mathbf{A} = \begin{bmatrix} 15.3 & 11.0 & 16.2 & 10.3 \\ 11.0 & 13.2 & 12.0 & 11.0 \\ 16.2 & 12.0 & 17.3 & 11.2 \\ 10.3 & 11.0 & 11.2 & 9.4 \end{bmatrix}$$

Equal values are connected by arrows.

When \mathbf{A} is square, then $.5*(\mathbf{A} + \mathbf{A}')$ is always symmetric.

Diagonal Matrices

When **B** is square and $b_{ij} = 0$ when $i \neq j$, then **B** is a **diagonal** matrix:

Example

$$B = \begin{bmatrix} 15.3 & 0.0 & 0.0 & 0.0 \\ 0.0 & 13.2 & 0.0 & 0.0 \\ 0.0 & 0.0 & 17.3 & 0.0 \\ 0.0 & 0.0 & 0.0 & 9.4 \end{bmatrix}$$

This is sometimes written

$$B = \text{diag}[b_{11}, b_{22}, \dots, b_{pp}].$$

Create **B** in MacAnova using `dmat()`:

```
Cmd> b <- dmat(vector(15.3,13.2,17.3,9.4)); b
(1,1) 15.3 0 0 0
(2,1) 0 13.2 0 0
(3,1) 0 0 17.3 0
(4,1) 0 0 0 9.4
```

You use `diag()` to extract the diagonal of a matrix:

```
Cmd> diag(b)
(1) 15.3 13.2 17.3 9.4
Cmd> a <- matrix(vector(15.3,11,16.2,10.3, 11,13.2,12,11,\
16.2,12,17.3,11.2, 10.3,11,11.2,9.4),4)
Cmd> a
(1,1) 15.3 11 16.2 10.3
(2,1) 11 13.2 12 11
(3,1) 16.2 12 17.3 11.2
(4,1) 10.3 11 11.2 9.4
Cmd> diag(a)
(1) 15.3 13.2 17.3 9.4
```

m by m identity matrix I_m

$$I_m = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

with 1's down the diagonal and 0's everywhere else. I_m is square and diagonal.

Create in MacAnova by, for example,

```
Cmd> dmat(rep(1,4)) # or dmat(vector(1,1,1,1))
(1,1)      1      0      0      0
(2,1)      0      1      0      0
(3,1)      0      0      1      0
(4,1)      0      0      0      1

Cmd> rep(1,4)
(1)      1      1      1      1
```

A shortcut form is

```
Cmd> dmat(4, 1)
(1,1)      1      0      0      0
(2,1)      0      1      0      0
(3,1)      0      0      1      0
(4,1)      0      0      0      1
```

$dmat(m, c)$ creates a m by m matrix with m values of c on the diagonal.

Multiplying a matrix by an identity matrix doesn't change it.

If A is m by n ,

- $I_m A = A$, that is $I_m A$ is *identical* to A
- $A I_n = A$, that is $A I_n$ is *identical* to A

MacAnova

```
Cmd> a # m = 2, n = 3
(1,1)      1.04      -0.651      1.5
(2,1)      0.696      0.13      1.61

Cmd> i2 <- dmat(2, 1) # 2 by 2 identity matrix

Cmd> i3 <- dmat(3, 1) # 3 by 3 identity matrix

Cmd> i2 %** a # identity matrix on left; same as a
(1,1)      1.04      -0.651      1.5
(2,1)      0.696      0.13      1.61

Cmd> a %** i3 # identity matrix on right; same as a
(1,1)      1.04      -0.651      1.5
(2,1)      0.696      0.13      1.61
```