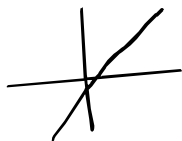


Today: Linear Functions + Matrices

linear: if  
additivity:  $F(x+y) = F(x) + F(y)$   
homogeneity:  $F(ax) = aF(x)$

exs:

- line:



- expectation functions are linear

non ex:  parabola

↳ called higher order polynomials,

where degree  $> 1$

$x^2$ ,  $x^3$  etc

we will approximate w/ linear equations

linear equations:

$$b = a_1 x_1 + a_2 x_2 + \dots + a_n x_n$$

( $a_1, \dots, a_n$  are #s - parameters)

( $x_1, \dots, x_n$  variables)

• group of linear equations = linear system

ex: tanks purchased from NATO allies, by  
Ukraine  $1 \dots i \dots n$

• sps  $y_i$  constant

$$y_i = 20$$

• sps it depends on # of senior leadership  
visits  $x_i$

$$y_i = 20 + 2 x_i$$

$$y_i = \beta_0 + \beta_1 x_i$$

→ we could graph  
this as a line!

tells us that each additional unit of  $x$   
(leadership visits) will produce a 2 unit  
increase in  $y$

in real life, don't observe what  $\beta_0$  and  $\beta_1$  are -  
we observe  $x$  and  $y$ . task: figure out  $\beta_0$  &  $\beta_1$ !

sps: france has 1 visit. france sells 1 tank  
germany has 2. germany sells 2 tanks

$$\begin{aligned} 1 &= \beta_0 + \beta_1 \cdot 2 \\ 3 &= \beta_0 + \beta_1 \cdot 1 \end{aligned}$$

solving for  $\beta_1, \beta_2$ :

$$4 - 3 = \beta_0 - \beta_0 + 2\beta_1 - 1\beta_1$$

$$1 = \beta_1$$

$$2 = \beta_0$$

## matrix

when we input data into excel or R,  
stored as a matrix.

looks like this:

final grade	hw #1	hw #2	hw #3	final ex
95	5	10	100	45
30	0	10	13	40
100	100	0	100	100

in matrix form, actually...

$$\begin{bmatrix} 95 \\ 30 \\ 100 \end{bmatrix} = \begin{bmatrix} 5 & 10 & 100 \\ 0 & 10 & 13 \\ 100 & 0 & 100 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

from our germany example:

$$\begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$



# matrix dimensionality

$m \times n$  matrices have  $m$  rows  
 $n$  columns

Rows x Columns  
(Roman catholic, Rum and coke)

Matrices are column of row vectors or row  
of column vectors

$$\begin{matrix} \text{rows:} & 1 & 2 & \dots & n \\ \left[ \begin{array}{cccc} a_{11} & \dots & & a_{1n} \\ \vdots & & & \\ a_{m1} & \dots & & a_{mn} \end{array} \right] & \begin{matrix} 1 \\ \vdots \\ m \end{matrix} \end{matrix}$$

$$= \left[ \begin{array}{c|c|c|c} 1 & & & \\ \hline \vec{C}_1 & \vec{C}_2 & \dots & \vec{C}_n \\ \hline 1 & 1 & & 1 \end{array} \right]$$

where  $\vec{C}_i = \begin{bmatrix} a_{1i} \\ \vdots \\ a_{mi} \end{bmatrix}$

OR

$$= \left[ \begin{array}{c} \vec{R}_1^T \\ \vec{R}_2^T \\ \vdots \\ \vec{R}_m^T \end{array} \right]$$

where  $\vec{R}_i = [a_{i1} \dots a_{in}]$

we've already seen this . . . diff in means estimator

$$\sum \mathbb{I} [X] = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix}$$

matrix transpose:

$$X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix},$$

$$X^T \text{ or } X' = [x_1 \ x_2 \ \dots \ x_n]$$

(more on this later!)

matrices & how computers store large datasets. this is essentially what is going on when we run regressions (more on that later)

$$\begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$

$$= \begin{bmatrix} 1\beta_0 + 2\beta_1 \\ 1\beta_0 + 1\beta_1 \end{bmatrix}$$

called matrix multiplication!

examples:

$$X^T = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 4 & 6 \\ 7 & 10 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \cdot 1 + 1 \cdot 3 & 1 \cdot 2 + 1 \cdot 4 \\ 1 \cdot 1 + 2 \cdot 3 & 1 \cdot 2 + 2 \cdot 4 \end{bmatrix}$$

more generally,

$$AB = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$a_{11} = [\text{1st row of } A][\text{1st column of } B]$$

$$a_{21} = [\text{2nd row of } A][\text{1st column of } B]$$

$$a_{12} = [\text{1st row of } A][\text{2nd column of } B]$$

$$a_{22} = [\text{2nd row of } A][\text{2nd column of } B]$$

# Laws:

$$A + B = B + A$$

$$\text{BUT... } AB \neq BA$$

associative laws:

$$(A + B) + C = A + (B + C)$$

$$(AB)C = A(BC)$$

distributive laws:

$$A(B + C) = AB + AC$$

$$(A + B)C = AC + BC$$

identity matrix:  $I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$AI = A$$

$$IA = A$$



$AB \neq BA$  example

$$A = \begin{bmatrix} 3 & 4 \\ 5 & 6 \end{bmatrix} \quad B = \begin{bmatrix} 7 & 8 \\ 9 & 0 \end{bmatrix}$$

$$AB = \begin{bmatrix} 57 & 24 \\ 89 & 40 \end{bmatrix}$$

$$BA = \begin{bmatrix} 61 & 76 \\ 27 & 36 \end{bmatrix}$$

(can also click "show solution" on matrix calculator for steps)

Can't multiply everything!

$$A = \begin{matrix} \text{rows: } & 1 & 2 & \dots & n \\ \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & & \end{bmatrix} & & & & \end{matrix} \begin{matrix} 1 \\ \vdots \\ m \end{matrix}$$

Call this an  $m \times n$  matrix  
if  $A$  is  $m \times n$  and  $B = r \times p$

$$\begin{bmatrix} m \text{ rows} \\ n \text{ columns} \\ \mathbb{R} \subset \text{Roman catholic} \end{bmatrix}$$

$AB$  exists iff  $n = r$

why?

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ q \end{bmatrix} = B$$

when we multiply this out,  $q$  doesn't have any parameters.

ex: in survey experiments,  $q$  may be some 4<sup>th</sup> variable that we think affects the outcome (say, voting choice). But if we didn't ask about  $q$ , we don't know what it is for each person. Thus we can't say anything about the relationship between  $q$  and our outcome.

# Practice

$$a) \begin{bmatrix} 1 & 3 \end{bmatrix} \times \begin{bmatrix} 2 & 0 \\ 1 & 3 \end{bmatrix} = \begin{bmatrix} 5 & 9 \end{bmatrix}$$

$$b) \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 2 & 6 & 1 \\ 1 & 3 & 2 \end{bmatrix} = \begin{bmatrix} 5 & 9 & 7 \\ 8 & 12 & 10 \end{bmatrix}$$

$$c) \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \\ 7 & 8 \end{bmatrix} \times \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 2 & 7 \\ 6 & 15 \\ 10 & 23 \\ 14 & 31 \end{bmatrix}$$

$$d) \begin{bmatrix} 1 \\ 3 \\ 5 \\ 7 \end{bmatrix} \times \begin{bmatrix} 2 & 1 \\ 0 & 3 \end{bmatrix} \therefore \text{A}$$

how do we use matrices to  
solve systems of linear equations?

- some systems have infinite solutions -
  - some systems have no solutions
- both mean that given the data we have, we cannot make <sup>good!</sup> inferences  
(more on that later)

\* maybe show in R what's going on here?

here, we have written matrices as follows:

$$[Y] = [X][\beta]$$

where  $Y$  is our outcome.

Sometimes, we write matrices like this:

$$\begin{bmatrix} \vec{Y} & \vec{X}_1 & \vec{X}_2 \end{bmatrix}$$

called an augmented matrix, solved by a process called gauss-jordan elimination

notice that  $\vec{\beta}$  is always an  $n \times 1$  vector! - in the augmented ex, we will solve for  $\beta_s$  - see Bretscher pg 12