Vector and matrix algebra

### Packages for this section

• This is (almost) all base R! We only need this for one thing later:

library(tidyverse)

#### Vector addition

Adds 2 to each element.

• Adding vectors:

```
u <- c(2, 3, 6, 5, 7)
v <- c(1, 8, 3, 2, 0)
u + v
```

```
[1] 3 11 9 7 7
```

• Elementwise addition. (Linear algebra: vector addition.)

## Adding a number to a vector

• Define a vector, then "add 2" to it:

u

• adds 2 to each element of u.

### Scalar multiplication

As per linear algebra:

k

[1] 2

u

[1] 2 3 6 5 7

k \* u

[1] 4 6 12 10 14

• Each element of vector multiplied by 2.

#### "Vector multiplication"

What about this?

u

[1] 2 3 6 5 7

V

[1] 1 8 3 2 0

u \* v

[1] 2 24 18 10 0

Each element of  $\mathfrak u$  multiplied by *corresponding* element of  $\mathfrak v$ . Could be called elementwise multiplication.

(Don't confuse with "outer" or "vector" product from linear algebra, or indeed "inner" or "scalar" multiplication, for which the answer is a number.)

### Combining different-length vectors

• No error here (you get a warning). What happens?

u

[1] 2 3 6 5 7

```
W \leftarrow c(1, 2)
```

[1] 3 5 7 7 8

- Add 1 to first element of u, add 2 to second.
- Go back to beginning of w to find something to add: add 1 to 3rd element of u, 2 to 4th element, 1 to 5th.

#### How R does this

- Keep re-using shorter vector until reach length of longer one.
- "Recycling".
- If the longer vector's length not a multiple of the shorter vector's length, get a warning (probably not what you want).
- Same idea is used when multiplying a vector by a number: the number keeps getting recycled.

#### **Matrices**

[2,] 2

• Create matrix like this:

```
(A <- matrix(1:4, nrow = 2, ncol = 2))

[,1] [,2]
[1,] 1 3
```

- First: stuff to make matrix from, then how many rows and columns.
- R goes down columns by default. To go along rows instead:

```
(B <- matrix(5:8, nrow = 2, ncol = 2, byrow = TRUE))

[,1] [,2]
[1,] 5 6
[2,] 7 8
```

• One of nrow and ncol enough, since R knows how many things in the matrix.

### Adding matrices

What happens if you add two matrices?

Α

В

$$A + B$$

# Adding matrices

• Nothing surprising here. This is matrix addition as we and linear algebra know it.

# Multiplying matrices

Now, what happens here?

Α

```
[,1] [,2]
[1,] 1 3
[2,] 2 4
```

В

A \* B

# Multiplying matrices?

- Not matrix multiplication (as per linear algebra).
- $\bullet$  Elementwise multiplication. Also called  ${\it Hadamard\ product}$  of A and B.

# Legit matrix multiplication

Like this:

Α

В

A %\*% B

### Reading matrix from file

The usual:

```
my url <- "http://ritsokiguess.site/datafiles/m.txt"
M <- read_delim(my_url, " ", col_names = FALSE )</pre>
M
# A tibble: 3 x 2
     X1 X2
  <dbl> <dbl>
    10 9
3
```

```
class(M)
```

```
[1] "spec tbl df" "tbl df" "tbl" "data.frame"
```

#### but...

• except that M is not an R matrix, and thus this doesn't work:

```
v <- c(1, 3)
M %*% v
```

Error in M %\*% v: requires numeric/complex matrix/vector arguments

# Making a genuine matrix

Do this first:

```
M <- as.matrix(M)
M</pre>
```

X1 X2 [1,] 10 9

[2,] 8 7

[3,] 6 5

V

[1] 1 3

and then all is good:

[,1]

### Linear algebra stuff

• To solve system of equations Ax = w for x:

Α

```
[,1] [,2]
[1,] 1 3
[2,] 2 4
```

W

[1] 1 0

#### Matrix inverse

To find the inverse of A:

#### Α

```
[,1] [,2]
[1,] 1 3
[2,] 2 4
```

#### solve(A)

 You can check that the matrix inverse and equation solution are correct.

#### Inner product

 Vectors in R are column vectors, so just do the matrix multiplication (t() is transpose):

```
a <- c(1, 2, 3)
b <- c(4, 5, 6)
t(a) %*% b
```

```
[,1]
[1,] 32
```

- ullet Note that the answer is actually a  $1 \times 1$  matrix.
- Or as the sum of the elementwise multiplication:

```
sum(a * b)
```

[1] 32

### Accessing parts of vector

• use square brackets and a number to get elements of a vector

b

[1] 4 5 6

b[2]

[1] 5

### Accessing parts of matrix

• use a row and column index to get an element of a matrix

```
Α
```

```
[,1] [,2]
[1,] 1 3
[2,] 2 4
```

#### A[2,1]

[1] 2

leave the row or column index empty to get whole row or column, eg.

[1] 1 3

# Eigenvalues and eigenvectors

ullet For a matrix A, these are scalars  $\lambda$  and vectors v that solve

$$Av = \lambda v$$

• In R, eigen gets these:

A

### Eigenvalues and eigenvectors

е

eigen() decomposition

# To check that the eigenvalues/vectors are correct

 $\bullet$   $\lambda_1 v_1$ : (scalar) multiply first eigenvalue by first eigenvector (in column)

```
e$values[1] * e$vectors[,1]
```

```
[1] -3.039462 -4.429794
```

ullet  $Av_1$ : (matrix) multiply matrix by first eigenvector (in column)

```
A %*% e$vectors[,1]
```

```
[,1]
[1,] -3.039462
[2,] -4.429794
```

- These are (correctly) equal.
- The second one goes the same way.

### A statistical application of eigenvalues

• A negative correlation:

```
d <- tribble(
    ~x,    ~y,
    10,    20,
    11,    18,
    12,    17,
    13,    14,
    14,    13
)
v <- cor(d)
v</pre>
```

```
x y
x 1.0000000 -0.9878783
y -0.9878783 1.0000000
```

• cor gives the correlation matrix between each pair of variables (correlation between x and y is -0.988)

# Eigenanalysis of correlation matrix

#### eigen(v)

eigen() decomposition

- first eigenvalue much bigger than second (second one near zero)
- two variables, but data nearly one-dimensional
- opposite signs in first eigenvector indicate that the one dimension is:
  - x small and y large at one end,
  - x large and y small at the other.