

Matrices in Julia

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Matrices

- ▶ matrices in Julia are represented by 2D arrays
- ▶ `[2 -4 8.2; -5.5 3.5 63]` creates the 2×3 matrix

$$A = \begin{bmatrix} 2 & -4 & 8.2 \\ -5.5 & 3.5 & 63 \end{bmatrix}$$

- ▶ spaces separate entries in a row; semicolons separate rows
- ▶ `size(A)` returns the size of `A` as a pair, *i.e.*,
`A_rows, A_cols = size(A)` # or
`A_rows` is `size(A)[1]`, `A_cols` is `size(A)[2]`
- ▶ row vectors are $1 \times n$ matrices, *e.g.*, `[4 8.7 -9]`

Indexing and slicing

- ▶ A_{ij} is found with $A[i, j]$
- ▶ can use ranges: $A[1:2, 1:3]$ is 2×3 submatrix or slice $A_{1:2, 1:3}$
- ▶ $:$ selects all elements along that dimension
 - $A[:, 3]$ is third column
 - $A[2, :]$ is second row
- ▶ $A[:]$ stacks the columns of A as a vector (column-major order)
- ▶ $A'[:]$ stacks the rows of A as a vector (row-major order)

Block matrices

- ▶ block matrix

$$X = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

(with A , B , C , and D matrices) is formed with

$$X = [A \ B; \ C \ D]$$

- ▶ usual rules governing dimensions of A , B , C , and D apply

Useful matrices in Julia

- ▶ $\mathbf{0}_{m \times n}$ is `zeros(m,n)`
- ▶ $m \times n$ matrix with all entries 1 is `ones(m,n)`
- ▶ $I_{n \times n}$ is `eye(n)`
- ▶ `diag(x)` is `diagm(x)` (where x is a vector)

Transpose and matrix addition

- ▶ A^T is written A' (single quote mark)
- ▶ +/- are used for matrix addition/substraction (matrices must have the same size)
- ▶ for example,

$$\begin{bmatrix} 4.0 & 7 \\ -10.6 & 89.8 \end{bmatrix} + \begin{bmatrix} 19 & -34.7 \\ 20 & 1 \end{bmatrix}$$

is written

$$[4.0 \ 7; -10.6 \ 89.8] + [19 \ -34.7; 20 \ 1]$$

Matrix-scalar operations

- ▶ matrix-scalar operations (+, -, *, \) apply elementwise
- ▶ scalar-matrix multiplication:

$$10 * [1 \ 2; \ 3 \ 4]$$

gives

$$10 \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 10 & 20 \\ 30 & 40 \end{bmatrix}$$

(scalar can also appear on right of matrix)

- ▶ matrix-scalar addition:

$$[1 \ 2; \ 3 \ 4] + 10$$

gives

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + \begin{bmatrix} 10 & 10 \\ 10 & 10 \end{bmatrix} = \begin{bmatrix} 11 & 12 \\ 13 & 14 \end{bmatrix}$$

(which is not standard mathematical notation)

Matrix-vector multiplication

- ▶ * operator is used for matrix-vector multiplication
- ▶ for example,

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \end{bmatrix}$$

is written

$$[1 \ 2; 3 \ 4] * [5, 6]$$

Matrix multiplication

- ▶ * is also used for matrix-matrix multiplication:

$$\begin{bmatrix} 2 & 4 & 3 \\ 3 & 1 & 5 \end{bmatrix} \begin{bmatrix} 3 & 10 \\ 4 & 2 \\ 1 & 7 \end{bmatrix}$$

is written

$$[2 \ 4 \ 3; \ 3 \ 1 \ 5] * [3 \ 10; \ 4 \ 2; \ 1 \ 7]$$

- ▶ A^k is A^k (for square matrix A)

Other functions

- ▶ sum of entries of a matrix: `sum(A)`
- ▶ average of entries of a matrix: `mean(A)`
- ▶ `max(A,B)` and `min(A,B)` finds the element-wise max and min respectively
 - the arguments must have the same size unless one is a scalar
- ▶ `norm(A)` is not what you might think
 - to find $\left(\sum_{i,j} A_{ij}^2\right)^{1/2}$ use `norm(A[:])` or `vecnorm(A)`

Computing regression model RMS error

the math:

- ▶ X is an $n \times N$ matrix whose N columns are feature n -vectors
- ▶ y is the N -vector of associated outcomes
- ▶ regression model is $\hat{y} = X^T \beta + v$ (β is n -vector, v is scalar)
- ▶ RMS error is $\mathbf{rms}(\hat{y} - y)$

in Julia:

```
y_hat = X'*beta + v  
rms_error = norm(y_hat-y)/sqrt(length(y))
```