# **Matrices in Julia**

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# Outline

### Matrices

Matrix operations

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## Matrices

matrices in Julia are repersented by 2D arrays

• to create the  $2 \times 3$  matrix

$$A = \left[ \begin{array}{rrr} 2 & -4 & 8.2 \\ -5.5 & 3.5 & 63 \end{array} \right]$$

use

A = [2 -4 8.2; -5.5 3.5 63]

semicolons delimit rows; spaces delimit entries in a row

```
size(A) returns the size of A as a pair, i.e.,
A_rows, A_cols = size(A) # or
A_size = size(A)
# A_rows is A_size[1], A_cols is A_size[2]
row vectors are 1 × n matrices, e.g., [4 8.7 -9]
```

#### Matrices

# Indexing and slicing

- $A_{13}$  is found with A[1,3]
- ranges can also be used: A[2,1:2:end]
- selects all elements along that dimension
  - A[:,3] selects the third column
  - A[2,:] selects the second row
  - A[:,end:-1:1] reverses the order of columns
- ▶ A[:] returns the columns of A stacked as a vector, *i.e.*, if
  - A = [2 7; 8 1]

then A[:] returns

[2, 8, 7, 1]

#### Matrices

## **Block matrices**

the block matrix

$$X = \left[ \begin{array}{cc} A & B \\ C & D \end{array} \right]$$

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

X = [A B; C D]

- all matrices in a row must have the same height
- the total number of columns in each row be consistent (c.f. standard math notation, in which A and C must have the same number of columns)

## **Common matrices**

- ▶  $\mathbf{0}_{m \times n}$  is zeros(m,n)
- $m \times n$  matrix with all entries 1 is ones(m,n)
- ▶  $I_{n \times n}$  is eye(n)
- $\operatorname{diag}(x)$  is  $\operatorname{diagm}(x)$  (where x is a vector)
- random m × n matrix with entries from standard normal distribution: randn(m,n)
- ▶ random  $m \times n$  matrix with entries from uniform distribution on [0,1]: rand(m,n)

# Outline

Matrices

Matrix operations

## Transpose and matrix addition

- $A^T$  is written A' (single quote mark)
- ▶ +/- are overloaded for matrix addition/substraction
- for example,

$$\left[\begin{array}{rrr} 4.0 & 7 \\ -10.6 & 89.8 \end{array}\right] + \left[\begin{array}{rrr} 19 & -34.7 \\ 20 & 1 \end{array}\right]$$

is written

[4.0 7; -10.6 89.8] + [19 -34.7; 20 1] matrices must have the same size (unless one is a scalar)

### Matrix-scalar operations

- ▶ all matrix-scalar operations (+,-,\*) apply elementwise
- for example, matrix-scalar addition:

```
[1 2; 3 4] + 10
```

gives

$$\left[\begin{array}{rrr}1&2\\3&4\end{array}\right]+10\left[\begin{array}{rrr}1&1\\1&1\end{array}\right]=\left[\begin{array}{rrr}11&12\\13&14\end{array}\right]$$

scalar-multiplication:

gives

$$10\left[\begin{array}{rrr}1&2\\3&4\end{array}\right] = \left[\begin{array}{rrr}10&20\\30&40\end{array}\right]$$

## Matrix-vector multiplication

- the \* operator is used for matrix-vector multiplication
- for example,

Γ	1	2 ]	[5]
L	3	4	6

is written

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

- ▶ for vectors x and y, x'\*y finds their inner product
  - unlike dot(x,y), x'\*y returns a  $1 \times 1$  array, not a scalar

## **Matrix multiplication**

\* is overloaded for matrix-matrix multiplication:

$$\left[\begin{array}{rrrr} 2 & 4 & 3 \\ 3 & 1 & 5 \end{array}\right] \left[\begin{array}{rrrr} 3 & 10 \\ 4 & 2 \\ 1 & 7 \end{array}\right]$$

is written

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

 $\blacktriangleright A^k$  is A^k for square matrix A and nonnegative integer k

## **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
- maximum(A) and minimum(A) return the largest and smallest entries of A
- 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  $\hat{y} = X^T \beta + v$  ( $\beta$  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))