

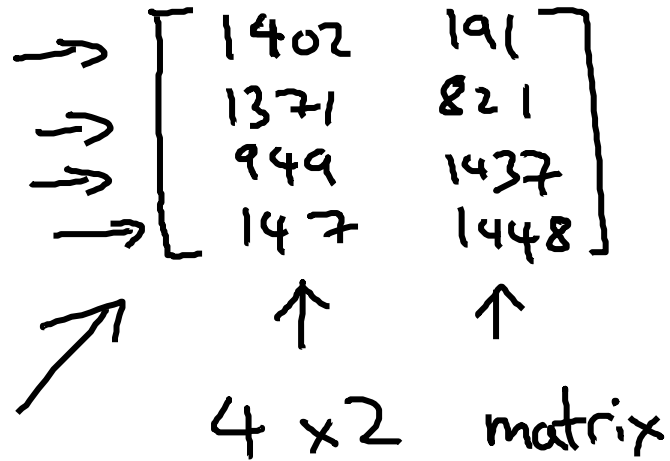
Machine Learning

Linear Algebra  
review (optional)

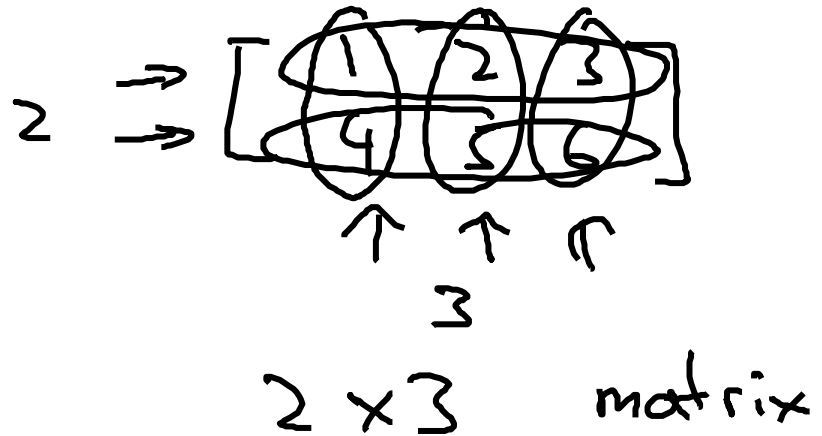
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Matrices and  
vectors

**Matrix:** Rectangular array of numbers:



→  $\mathbb{R}^{4 \times 2}$



$\mathbb{R}^{2 \times 3}$

Dimension of matrix: number of rows x number of columns

# Matrix Elements (entries of matrix)

$$A = \begin{bmatrix} 1402 & 191 \\ 1371 & 821 \\ 949 & 1437 \\ 147 & 1448 \end{bmatrix}$$

$A_{ij}$  = " $i, j$  entry" in the  $i^{\text{th}}$  row,  $j^{\text{th}}$  column.

$$A_{11} = 1402$$

$$A_{12} = 191$$

$$A_{32} = 1437$$

$$A_{41} = 147$$

~~$A_{43}$~~  = Undefined (error)

Vector: An n x 1 matrix.

$$y = \begin{bmatrix} 460 \\ 232 \\ 315 \\ 178 \end{bmatrix}$$

$n = 4$

← 4-dimensional vector.

~~$\mathbb{R}^{3 \times 2}$~~

$\mathbb{R}^4$

$y_i = i^{th}$  element

$$y_1 = 460$$

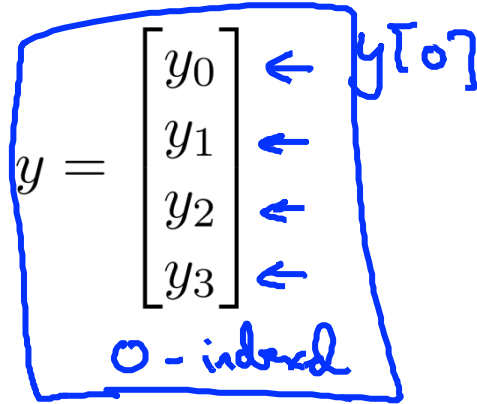
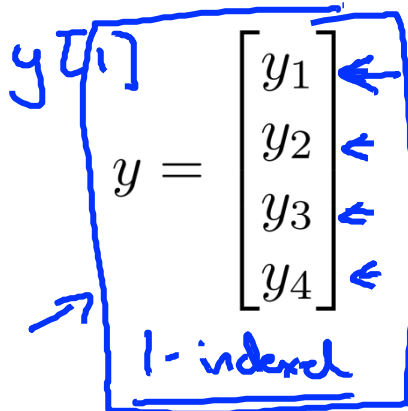
$$y_2 = 232$$

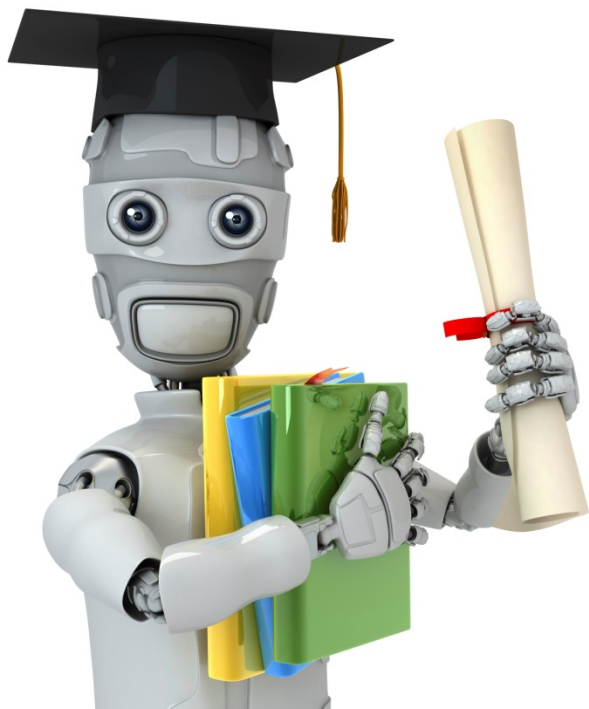
$$y_3 = 315$$

→ A, B, C, X

a, b, x, y

1-indexed vs 0-indexed:





Machine Learning

# Linear Algebra review (optional)

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Addition and scalar  
multiplication

# Matrix Addition

$$\begin{array}{c} \downarrow \quad \downarrow \\ \rightarrow \begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} + \begin{bmatrix} 4 & 0.5 \\ 2 & 5 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 5 & 0.5 \\ 4 & 10 \\ 3 & 2 \end{bmatrix} \\ \text{3x2 matrix} \quad \text{3x2} \quad \text{3x2} \end{array}$$

$$\begin{array}{c} \rightarrow \begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} + \begin{bmatrix} 4 & 0.5 \\ 2 & 5 \end{bmatrix} = \text{error} \\ \text{3x2} \quad \text{2x2} \end{array}$$

# Scalar Multiplication

real number

$$3 \times \begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 6 & 5 \\ 9 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} \times 3$$

3x2                      3x2

$$\begin{bmatrix} 4 & 0 \\ 6 & 3 \end{bmatrix} / 4 = \frac{1}{4} \begin{bmatrix} 4 & 0 \\ 6 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \frac{3}{2} & \frac{3}{4} \end{bmatrix}$$

# Combination of Operands

$$3 \times \begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix} - \begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix} / 3$$

Scalar multiplication

Scalar division

$$= \begin{bmatrix} 3 \\ 12 \\ 6 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 2/3 \end{bmatrix}$$

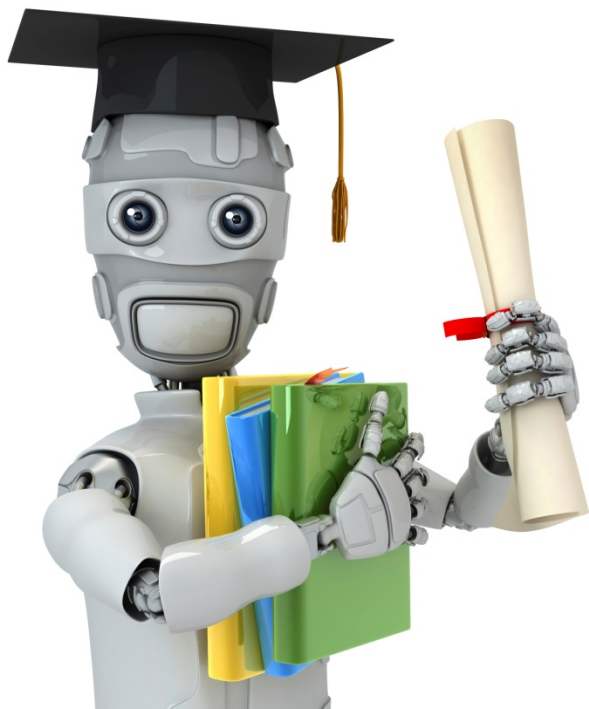
matrix subtraction / vector subtraction

matrix addition / vector addition

$$= \begin{bmatrix} 2 \\ 12 \\ 10 \frac{1}{3} \end{bmatrix}$$

3x1 matrix  
3-dimensional vector





Machine Learning

Linear Algebra  
review (optional)

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Matrix-vector  
multiplication

# Example

$$\begin{bmatrix} 1 & 3 \\ 4 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 5 \end{bmatrix} = \begin{bmatrix} 16 \\ 4 \\ 7 \end{bmatrix}$$

$3 \times 2$        $2 \times 1$        $3 \times 1$  matrix

$$1 \times 1 + 3 \times 5 = 16$$

$$4 \times 1 + 0 \times 5 = 4$$

$$2 \times 1 + 1 \times 5 = 7$$

## Details:

$$\underline{A} \times \underline{x} = \underline{y}$$

$m \times n$  matrix  
( $m$  rows,  
 $n$  columns)

$n \times 1$  matrix  
( $n$ -dimensional  
vector)

$m$ -dimensional  
vector

→ To get  $y_i$ , multiply  $A$ 's  $i^{\text{th}}$  row with elements of vector  $x$ , and add them up.

# Example

$$\begin{bmatrix} 1 & 2 & 1 & 5 \\ 0 & 3 & 0 & 4 \\ -1 & -2 & 0 & 0 \end{bmatrix}_{3 \times 4} \begin{bmatrix} 1 \\ 3 \\ 2 \\ 1 \end{bmatrix}_{4 \times 1} = \begin{bmatrix} 14 \\ 13 \\ -7 \end{bmatrix}_{3 \times 1} = \begin{bmatrix} 14 \\ 13 \\ -7 \end{bmatrix}$$

$$\left. \begin{array}{l} 1 \times 1 + 2 \times 3 + 1 \times 2 + 5 \times 1 = 14 \\ 0 \times 1 + 3 \times 3 + 0 \times 2 + 4 \times 1 = 13 \\ -1 \times 1 + (-2) \times 3 + 0 \times 2 + 0 \times 1 = -7 \end{array} \right\}$$

House sizes:

- 2104
- 1416
- 1534
- 852

Matrix

4x2

1	2104
1	1416
1	1534
1	852

$$h_{\theta}(x) = -40 + 0.25x$$

$h_{\theta}(x)$

2x1

Vector

-40
0.25

X

=

4x1 matrix

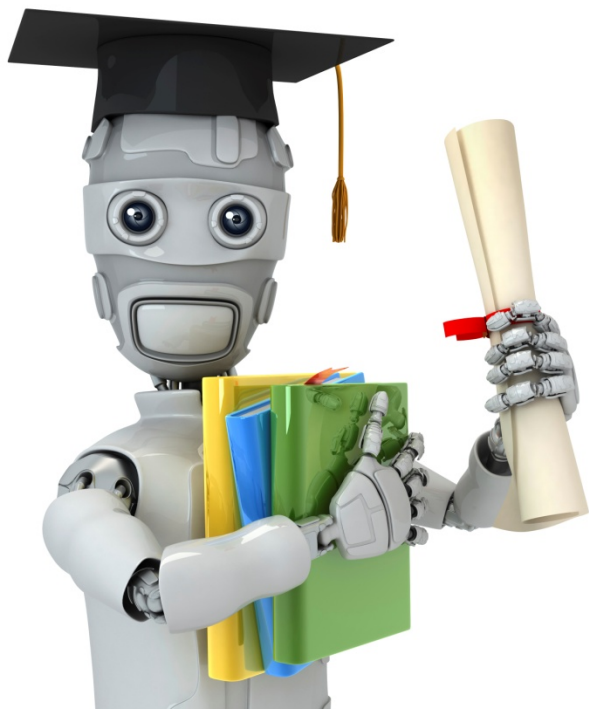
$-40 \times 1 + 0.25 \times 2104$
$-40 \times 1 + 0.25 \times 1416$

$h_{\theta}(1416)$

Prediction = Data Matrix \* Parameters

4x1

for  $i = 1:1000$ ,  
prediction (i) = ...



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Linear Algebra  
review (optional)

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Matrix-matrix  
multiplication

# Example

$$\begin{array}{l} \begin{bmatrix} 1 & 3 & 2 \\ 4 & 0 & 1 \end{bmatrix} \begin{array}{|c|} \hline 1 \\ \hline 3 \\ \hline \end{array} \begin{array}{|c|} \hline 0 \\ \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline 5 \\ \hline 2 \\ \hline \end{array} \\ \hline \textcircled{2 \times 3} \quad \textcircled{3 \times 2} \\ \begin{bmatrix} 1 & 3 & 2 \\ 4 & 0 & 1 \end{bmatrix} \times \begin{array}{|c|} \hline 1 \\ \hline 0 \\ \hline 5 \\ \hline \end{array} \\ \hline \begin{bmatrix} 1 & 3 & 2 \\ 4 & 0 & 1 \end{bmatrix} \times \begin{array}{|c|} \hline 3 \\ \hline 1 \\ \hline 2 \\ \hline \end{array} \end{array} = \begin{array}{l} \begin{bmatrix} 11 & 10 \\ 9 & 14 \end{bmatrix} \\ \begin{bmatrix} 11 \\ 9 \end{bmatrix} \\ \begin{bmatrix} 10 \\ 14 \end{bmatrix} \end{array}$$

*Handwritten annotations: Green boxes around the second and third columns of the first matrix, and the second and third rows of the second and third matrices. Green arrows point from the second and third rows of the second matrix to the second and third columns of the first matrix. A green arrow points from the second and third rows of the third matrix to the second and third columns of the first matrix. A green arrow points from the second and third columns of the first matrix to the second and third rows of the first result matrix. A green arrow points from the second and third columns of the first matrix to the second and third rows of the second result matrix. A green arrow points from the second and third columns of the first matrix to the second and third rows of the third result matrix.*

## Details:

$$\underline{A} \times \underline{B} = \underline{C}$$

$m \times n$  matrix  
( $m$  rows,  
 $n$  columns)

$n \times o$  matrix  
( $n$  rows,  
 $o$  columns)

$m \times o$   
matrix

The  $i^{th}$  column of the matrix  $C$  is obtained by multiplying  $A$  with the  $i^{th}$  column of  $B$ . (for  $i = 1, 2, \dots, o$ )



# Example

$$\begin{matrix} 2 \times 2 \\ \left[ \begin{array}{cc} 1 & 3 \\ 2 & 5 \end{array} \right] \end{matrix} \begin{matrix} 2 \times 2 \\ \left[ \begin{array}{cc} 0 & 1 \\ 3 & 2 \end{array} \right] \end{matrix} =$$

$$\begin{matrix} 2 \times 2 \\ \left[ \begin{array}{cc} 9 & 7 \\ 15 & 12 \end{array} \right] \end{matrix}$$

$$\begin{matrix} \left[ \begin{array}{cc} 1 & 3 \\ 2 & 5 \end{array} \right] \end{matrix} \begin{matrix} \left[ \begin{array}{c} 0 \\ 3 \end{array} \right] \end{matrix} =$$

$$\begin{matrix} \left[ \begin{array}{cc} 1 \times 0 + 3 \times 3 \\ 2 \times 0 + 5 \times 3 \end{array} \right] = \begin{matrix} \left[ \begin{array}{c} 9 \\ 15 \end{array} \right] \end{matrix} \end{matrix}$$

$$\begin{matrix} \left[ \begin{array}{cc} 1 & 3 \\ 2 & 5 \end{array} \right] \end{matrix} \begin{matrix} \left[ \begin{array}{c} 1 \\ 2 \end{array} \right] \end{matrix} =$$

$$\begin{matrix} \left[ \begin{array}{cc} 1 \times 1 + 3 \times 2 \\ 2 \times 1 + 5 \times 2 \end{array} \right] = \begin{matrix} \left[ \begin{array}{c} 7 \\ 12 \end{array} \right] \end{matrix} \end{matrix}$$

House sizes:

$$\begin{pmatrix} 2104 \\ 1416 \\ 1534 \\ 852 \end{pmatrix}$$

Have 3 competing hypotheses:

1.  $h_{\theta}(x) = -40 + 0.25x$
2.  $h_{\theta}(x) = 200 + 0.1x$
3.  $h_{\theta}(x) = -150 + 0.4x$

Matrix

$$\begin{bmatrix} 1 & 2104 \\ 1 & 1416 \\ 1 & 1534 \\ 1 & 852 \end{bmatrix} \times$$

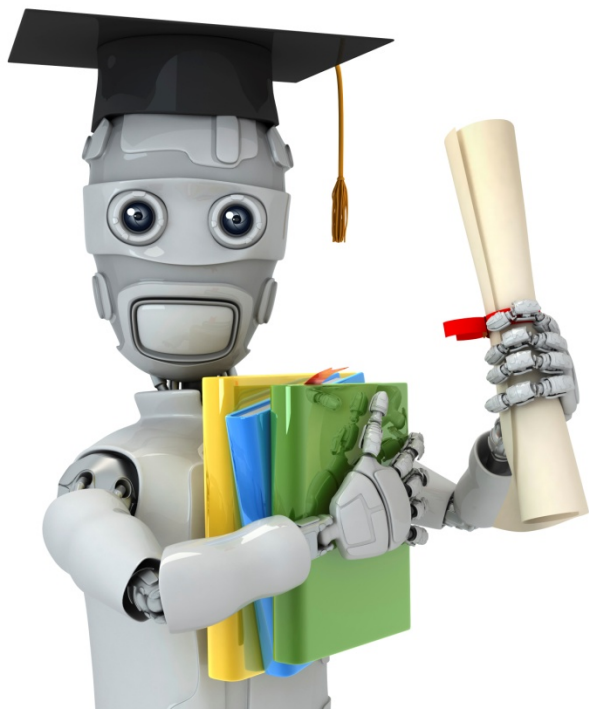
Matrix

$$\begin{bmatrix} -40 \\ 0.25 \end{bmatrix} \begin{bmatrix} 200 \\ 0.1 \end{bmatrix} \begin{bmatrix} -150 \\ 0.4 \end{bmatrix} =$$

$$\begin{bmatrix} 486 \\ 314 \\ 344 \\ 173 \end{bmatrix} \begin{bmatrix} 410 \\ 342 \\ 353 \\ 285 \end{bmatrix} \begin{bmatrix} 692 \\ 416 \\ 464 \\ 191 \end{bmatrix}$$

Prediction  
of first  
 $h_{\theta}$

Predictions  
of 2<sup>nd</sup>  
 $h_{\theta}$



Machine Learning

# Linear Algebra review (optional)

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## Matrix multiplication properties

$$3 \times 5 = 5 \times 3$$

"Commutative"

Let A and B be matrices. Then in general,  
A × B ≠ B × A. (not commutative.)

E.g.

$$\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 2 & 2 \end{bmatrix}$$

A × B  
 $m \times n$       $n \times m$

A × B is  $m \times m$

B × A is  $n \times n$

$$\underline{3 \times 5 \times 2}$$

$$3 \times 10 = 30 = 15 \times 2$$

$$3 \times (5+2) = (3 \times 5) \times 2$$

"Associative"

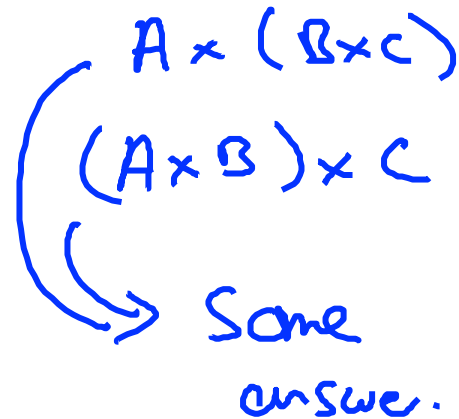
$$\begin{array}{l} A \times (B \times C) \leftarrow \\ \underline{(A \times B)} \times C \leftarrow \end{array}$$



$$A \times B \times C.$$

Let  $\underline{D = B \times C}$ . Compute  $A \times D$ .

Let  $\underline{E = A \times B}$ . Compute  $E \times C$ .



1 is identity

$1 \times z = z \times 1 = z$   
for any  $z$

# Identity Matrix

Denoted  $I$  (or  $I_{n \times n}$ ).

Examples of identity matrices:

$[1]$   
 $1 \times 1$   
 $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$   
 $2 \times 2$

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$   
 $3 \times 3$

$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$   
 $4 \times 4$

Informally:

$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}$

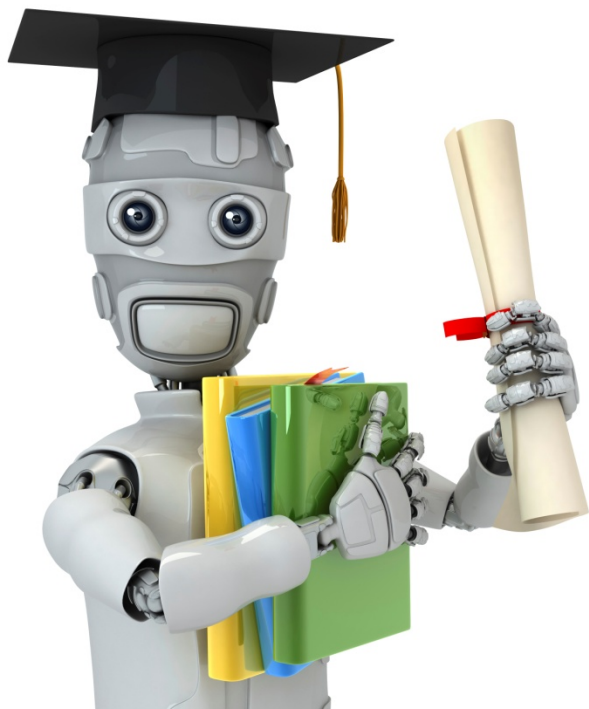
For any matrix  $A$ ,

$A \cdot I = I \cdot A = A$   
 $I_{m \times n}$     $n \times n$     $m \times m$     $m \times n$     $m \times n$

Note:

$AB \neq BA$  in general

$AI = IA$  ✓



Machine Learning

Linear Algebra  
review (optional)

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Inverse and  
transpose

I = "identity"

$$3 \underbrace{\left( \frac{1}{3} \right)}_{\frac{1}{3}} = 1$$

$$12 \times \underbrace{\left( \frac{1}{12} \right)}_{\frac{1}{12}} = 1$$

0 (0<sup>-1</sup>) undefined

Not all numbers have an inverse.

**Matrix inverse:**

square matrix  
(#rows = #columns)

$A^{-1}$

$$A = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

If A is an m x m matrix, and if it has an inverse,

$$\rightarrow \underline{A(A^{-1})} = \underline{A^{-1}A} = \underline{I}$$

E.g.

$$\underbrace{\begin{bmatrix} 3 & 4 \\ 2 & 16 \end{bmatrix}}_A$$

$$\underbrace{\begin{bmatrix} 0.4 & -0.1 \\ -0.05 & 0.075 \end{bmatrix}}_{A^{-1}}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_{2 \times 2}$$

Matrices that don't have an inverse are "singular" or "degenerate"



# Matrix Transpose

Example:

$$\underline{A} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 5 & 9 \end{bmatrix}_{2 \times 3}$$
$$\underline{B} = \underline{A}^T = \begin{bmatrix} 1 & 3 \\ 2 & 5 \\ 0 & 9 \end{bmatrix}_{3 \times 2}$$

Let  $A$  be an  $m \times n$  matrix, and let  $B = A^T$ .

Then  $B$  is an  $n \times m$  matrix, and

$$\underline{B}_{ij} = \underline{A}_{ji}.$$

$$B_{12} = A_{21} = 2$$

$$B_{32} = 9 \quad A_{23} = 9.$$