

# Linear Algebra Background

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# Motivation

Multivariate linear regression:

$$y^{(1)} \approx \theta_0 + \theta_1 x_1^{(1)} + \theta_2 x_2^{(1)} + \dots + \theta_n x_n^{(1)}$$

$$y^{(2)} \approx \theta_0 + \theta_1 x_1^{(2)} + \theta_2 x_2^{(2)} + \dots + \theta_n x_n^{(2)}$$

...

$$y^{(m)} \approx \theta_0 + \theta_1 x_1^{(m)} + \theta_2 x_2^{(m)} + \dots + \theta_n x_n^{(m)}$$

# Motivation

Multivariate linear regression:

$$y^{(1)} \approx \theta_0 + \theta_1 x_1^{(1)} + \theta_2 x_2^{(1)} + \dots + \theta_n x_n^{(1)}$$

$$y^{(2)} \approx \theta_0 + \theta_1 x_1^{(2)} + \theta_2 x_2^{(2)} + \dots + \theta_n x_n^{(2)}$$

...

$$y^{(m)} \approx \theta_0 + \theta_1 x_1^{(m)} + \theta_2 x_2^{(m)} + \dots + \theta_n x_n^{(m)}$$

After linear algebra

$$\mathbf{y} \approx \mathbf{X}\boldsymbol{\theta}$$

# Linear Algebra in ML

## Linear Algebra

- ▶ Succinct notation for models and algorithms
- ▶ Numerical tools (save coding!)

$$\boldsymbol{\theta} = (X^T X)^{-1} X^T \mathbf{y}$$

- ▶ Inspiration for new models and problems: Netflix

# Netflix Movie Recommendations

|       | Gladiator | Silence of the Lambs | WALL-E | Toy Story |
|-------|-----------|----------------------|--------|-----------|
| Alice | 5         | 4                    | 1      |           |
| Bob   |           | 5                    |        | 2         |
| Carol |           |                      |        | 5         |
| David |           |                      | 5      | 5         |
| Eve   | 5         | 4                    |        |           |

Matrix completion problem, matrix factorization

# Topics

- ▶ Matrices
- ▶ Vectors
- ▶ Matrix-Matrix multiplication (and special cases)
- ▶ Tranpose
- ▶ Inverse

# Matrices

- ▶ A matrix is an rectangular array of numbers

$$A = \begin{bmatrix} 101 & 10 \\ 54 & 13 \\ 10 & 47 \end{bmatrix}$$

- ▶ When  $A$  has  $m$  rows and  $n$  columns, we say that:
  - ▶  $A$  is an  $m \times n$  matrix
  - ▶  $A \in \mathbb{R}^{m \times n}$
- ▶ The entry in row  $i$  and column  $j$  is denoted  $A_{ij}$ 
  - ▶ sometimes  $a_{ij}$  or  $(A)_{ij}$

# Matrices

## Example

$$A = \begin{bmatrix} 101 & 10 \\ 54 & 13 \\ 10 & 47 \end{bmatrix}$$

- ▶  $A \in \mathbb{R}^{3 \times 2}$
- ▶  $A_{11} = 101$
- ▶  $A_{32} =$
- ▶  $A_{22} =$
- ▶  $A_{23} =$



# Vectors

- ▶ A vector is an  $n \times 1$  matrix:

$$\mathbf{x} = \begin{bmatrix} 8 \\ 2.4 \\ 1 \\ -10 \end{bmatrix}$$

- ▶ We write  $\mathbf{x} \in \mathbb{R}^n$  (instead of  $\mathbf{x} \in \mathbb{R}^{n \times 1}$ )
- ▶ The  $i$ th entry is  $x_i$

# Vectors

## Example

$$\mathbf{x} = \begin{bmatrix} 8 \\ 2.4 \\ 1 \\ -10 \end{bmatrix}$$

- ▶  $\mathbf{x} \in \mathbb{R}^4$
- ▶  $x_1 =$
- ▶  $x_4 =$

# Addition

- ▶ If two matrices have the same size, we can add them by adding corresponding elements

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

- ▶ Subtraction is similar
- ▶ Matrices of different sizes *cannot be added or subtracted*

# Scalar Multiplication

- ▶ A *scalar*  $x \in \mathbb{R}$  is a real number (i.e., not a vector)

e.g., 2, 3,  $\pi$ ,  $\sqrt{2}$ , 1.843, ...

- ▶ Scalar times matrix:

$$2 \cdot \begin{bmatrix} 1 & 3 \\ -2 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 6 \\ -4 & 0 \end{bmatrix}$$

(multiply each entry by the scalar)

# Matrix-Matrix Multiplication

- ▶ Can multiply two matrices *if their inner dimensions match*

$$A \in \mathbb{R}^{m \times n}, B \in \mathbb{R}^{n \times p}$$

$$C = AB \in \mathbb{R}^{m \times p}$$

- ▶ The product has entries

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

# Matrix-Matrix Multiplication

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

Move along  $i$ th row of  $A$  and  $j$ th row of  $B$ . Multiply corresponding entries, then add.

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \\ c_{41} & c_{42} & c_{43} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{bmatrix}$$

$$c_{32} = a_{31}b_{12} + a_{32}b_{22}$$

# Matrix-Matrix Multiplication

## Example

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 2 \\ -1 & 0 \end{bmatrix}$$

$$AB = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 3 & 2 \\ -1 & 0 \end{bmatrix} =$$

# Matrix-Matrix Multiplication

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$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 2 \\ -1 & 0 \end{bmatrix}$$

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# Multiplication Properties

- ▶ Associative

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

- ▶ Distributive

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

$$(B + C)D = BD + CD$$

- ▶ Not commutative

$$AB \neq BA$$

# Matrix-Vector Multiplication

A (worthy) special case of matrix-matrix multiplication:

$$A \in \mathbb{R}^{m \times n}, \quad \mathbf{x} \in \mathbb{R}^n$$

$$\mathbf{y} = A\mathbf{x} \in \mathbb{R}^m$$

Definition

$$y_i = \sum_{j=1}^n A_{ij}x_j$$

# Matrix-Vector Multiplication

$$y_i = \sum_{j=1}^n A_{ij}x_j$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$y_3 = a_{31}x_1 + a_{32}x_2$$

# Matrix-Vector Multiplication

## Example

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \mathbf{z} = \begin{bmatrix} 8 \\ 1.5 \end{bmatrix}$$

►  $A\mathbf{x} = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} =$

# Matrix-Vector Multiplication

## Example

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \mathbf{z} = \begin{bmatrix} 8 \\ 1.5 \end{bmatrix}$$

►  $A\mathbf{x} = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ -3 \end{bmatrix}$

# Matrix-Vector Multiplication

## Example

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- ▶  $A\mathbf{x} = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ -3 \end{bmatrix}$
- ▶  $A\mathbf{z} =$

# Matrix-Vector Multiplication

## Example

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \mathbf{z} = \begin{bmatrix} 8 \\ 1.5 \end{bmatrix}$$

►  $A\mathbf{x} = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ -3 \end{bmatrix}$

►  $A\mathbf{z} = \begin{bmatrix} 6.5 \\ 4.5 \end{bmatrix}$

# Transpose

Transposition of a matrix swaps the rows and columns

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix}, \quad A^T = \begin{bmatrix} 1 & 0 \\ -1 & 3 \end{bmatrix}.$$

Definition:

- ▶ Let  $A \in \mathbb{R}^{m \times n}$
- ▶ The *transpose*  $A^T \in \mathbb{R}^{n \times m}$  has entries

$$(A^T)_{ij} = A_{ji}.$$



# Transpose

## Example

$$A = \begin{bmatrix} 3 & 2 \\ -1 & 0 \\ 1 & 4 \end{bmatrix} \quad A^T =$$

# Transpose

## Example

$$A = \begin{bmatrix} 3 & 2 \\ -1 & 0 \\ 1 & 4 \end{bmatrix} \quad A^T = \begin{bmatrix} 3 & -1 & 1 \\ 2 & 0 & 4 \end{bmatrix}$$

# Transpose

## Example

$$A = \begin{bmatrix} 3 & 2 \\ -1 & 0 \\ 1 & 4 \end{bmatrix} \quad A^T = \begin{bmatrix} 3 & -1 & 1 \\ 2 & 0 & 4 \end{bmatrix}$$

## Example

$$\mathbf{x} = \begin{bmatrix} 1 \\ -3 \\ 2 \end{bmatrix} \quad \mathbf{x}^T =$$

# Transpose

## Example

$$A = \begin{bmatrix} 3 & 2 \\ -1 & 0 \\ 1 & 4 \end{bmatrix} \quad A^T = \begin{bmatrix} 3 & -1 & 1 \\ 2 & 0 & 4 \end{bmatrix}$$

## Example

$$\mathbf{x} = \begin{bmatrix} 1 \\ -3 \\ 2 \end{bmatrix} \quad \mathbf{x}^T = [1 \quad -3 \quad 2]$$

# Dot product

- ▶ A *special* special-case of matrix-matrix multiplication
- ▶ Let  $\mathbf{x}, \mathbf{y}$  be vectors of same size ( $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ ).
- ▶ Their dot product is

$$\begin{aligned}\mathbf{x}^T \mathbf{y} &= \sum_{i=1}^n x_i y_i \\ &= [x_1 \quad x_2 \quad \dots \quad x_n] \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}\end{aligned}$$

# Vector Norm

- ▶ The *norm* of a vector

$$\begin{aligned}\|\mathbf{x}\| &= \sqrt{x_1^2 + x_2^2 + \dots + x_n^2} \\ &= \sqrt{\mathbf{x}^T \mathbf{x}}\end{aligned}$$

- ▶ Geometric interpretation: length of the vector

# Transpose Properties

- ▶ Transpose of transpose

$$(A^T)^T = A$$

- ▶ Transpose of sum

$$(A + B)^T = A^T + B^T$$

- ▶ Transpose of product

$$(AB)^T = B^T A^T$$

# Identity

- ▶ The identity matrix  $I \in \mathbb{R}^{n \times n}$  has entries

$$I_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases},$$

$$I_{1 \times 1} = [1], \quad I_{2 \times 2} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad I_{3 \times 3} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

- ▶ For any  $A, B$  of appropriate dimensions

$$IA = A$$

$$BI = B$$



# Inverse

- ▶ The inverse  $A^{-1} \in \mathbb{R}^{n \times n}$  of a *square* matrix  $A \in \mathbb{R}^{n \times n}$  satisfies

$$AA^{-1} = I = A^{-1}A$$

- ▶ Compare to division of scalars

$$xx^{-1} = 1 = x^{-1}x$$

- ▶ **Not all matrices are invertible**

- ▶ E.g.,  $A$  not square,  $A = [0]$ ,  $A = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ , many more

# Inverse

## Example

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$$

Is  $B$  the inverse of  $A$ ?

# Inverse

## Example

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$$

Is  $B$  the inverse of  $A$ ?

## Example

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 3 \end{bmatrix} \quad A^{-1} = \begin{bmatrix} 1 & \frac{1}{3} \\ 0 & \frac{1}{3} \end{bmatrix}$$

Verify on your own.

# Inverse Properties

- ▶ Inverse of inverse

$$(A^{-1})^{-1} = A$$

- ▶ Inverse of product

$$(AB)^{-1} = B^{-1}A^{-1}$$

- ▶ Inverse of transpose

$$(A^{-1})^T = (A^T)^{-1} := A^{-T}$$

In MATLAB

See demo.m

# What You Should Know

- ▶ Definitions of matrices and vectors
- ▶ Meaning of matrix multiplication
  - ▶ Systems of equations  $\longrightarrow$  matrix-vector equations
- ▶ Properties of multiplication
- ▶ Properties of inverse, transpose
  - ▶ Get familiar with these as course goes on