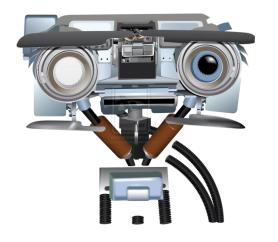


## **Preliminaries**

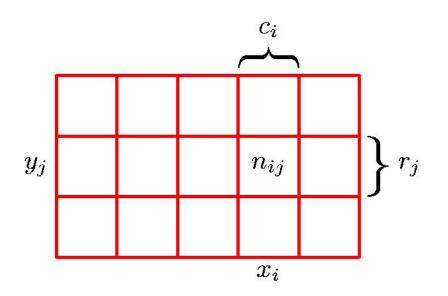
Probability Theory and Linear Algebra



# Probability Theory Review

Rules of Probability

# **Probability Theory**



#### **Marginal Probability**

$$p(X = x_i) = \frac{c_i}{N}.$$

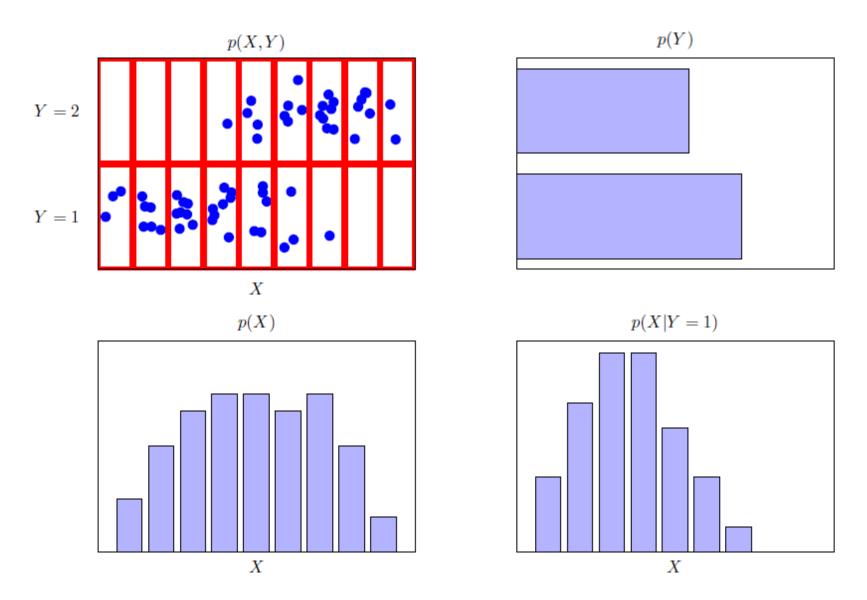
#### **Joint Probability**

$$p(X = x_i, Y = y_j) = \frac{n_{ij}}{N}$$

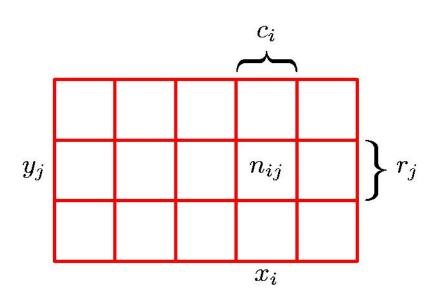
#### **Conditional Probability**

$$p(Y = y_j | X = x_i) = \frac{n_{ij}}{c_i}$$

# Example



## **Probability Theory**



#### Sum Rule

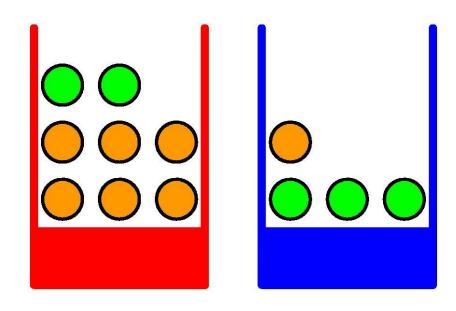
$$r_j$$
  $p(X = x_i) = \frac{c_i}{N} = \frac{1}{N} \sum_{j=1}^{L} n_{ij}$   
=  $\sum_{j=1}^{L} p(X = x_i, Y = y_j)$ 

#### **Product Rule**

$$p(X = x_i, Y = y_j) = \frac{n_{ij}}{N} = \frac{n_{ij}}{c_i} \cdot \frac{c_i}{N}$$
$$= p(Y = y_j | X = x_i) p(X = x_i)$$

# **Probability Theory**

see Bishop Chapter 1.2



- Pick a random box
- Pick a random fruit
- Observe the fruit type (orange or apple)
- Put it back in the box
- Repeat trial many times

What is the probability of picking an apple?

$$Pr(B = r) = 0.4, Pr(B = b) = 0.6$$
  
 $Pr(f = a|B = r) = 0.25, Pr(f = o|B = r) = 0.75$   
 $Pr(f = a|B = b) = 0.75, Pr(f = o|B = b) = 0.25$ 

# The Rules of Probability

Sum Rule

$$p(X) = \sum_{Y} p(X, Y)$$

**Product Rule** 

$$p(X,Y) = p(Y|X)p(X)$$

# Bayes' Theorem

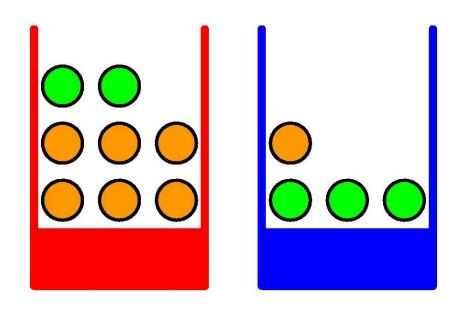
$$p(Y|X) = \frac{p(X|Y)p(Y)}{p(X)}$$

$$p(X) = \sum_{Y} p(X|Y)p(Y)$$

posterior ∝ likelihood × prior

# **Probability Theory**

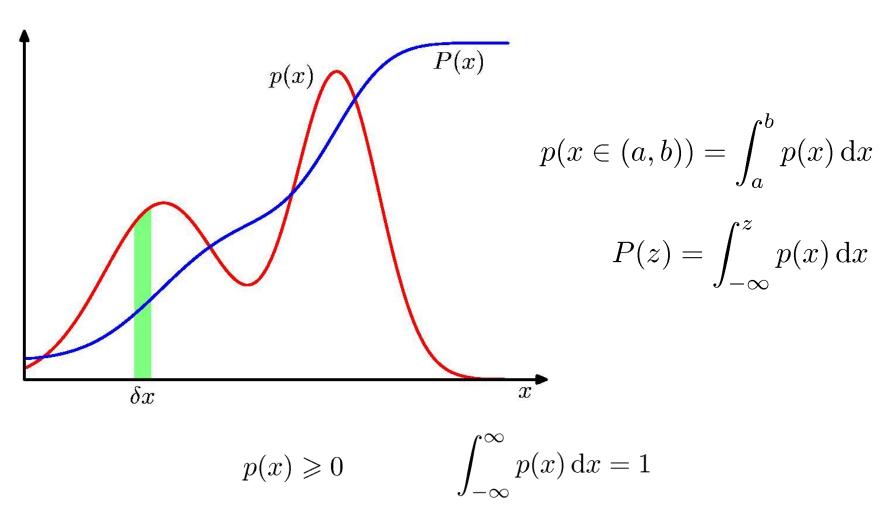
see Bishop Chapter 1.2



- Suppose we picked an orange
- What is the probability it came from the red box?

## **Probability Densities**

for continuous variables



# Expectations

$$\mathbb{E}[f] = \sum_{x} p(x) f(x)$$

$$\mathbb{E}[f] = \int p(x)f(x) \, \mathrm{d}x$$

$$\mathbb{E}_{x}[f|y] = \sum_{x} p(x|y)f(x)$$

$$\mathbb{E}[f] \simeq \frac{1}{N} \sum_{n=1}^{N} f(x_n)$$

Approximate Expectation (discrete and continuous)

## Variances and Co-variances

$$\operatorname{var}[f] = \mathbb{E}\left[ (f(x) - \mathbb{E}[f(x)])^{2} \right] = \mathbb{E}[f(x)^{2}] - \mathbb{E}[f(x)]^{2}$$

$$\operatorname{cov}[x, y] = \mathbb{E}_{x, y} \left[ \{x - \mathbb{E}[x]\} \{y - \mathbb{E}[y]\} \right]$$

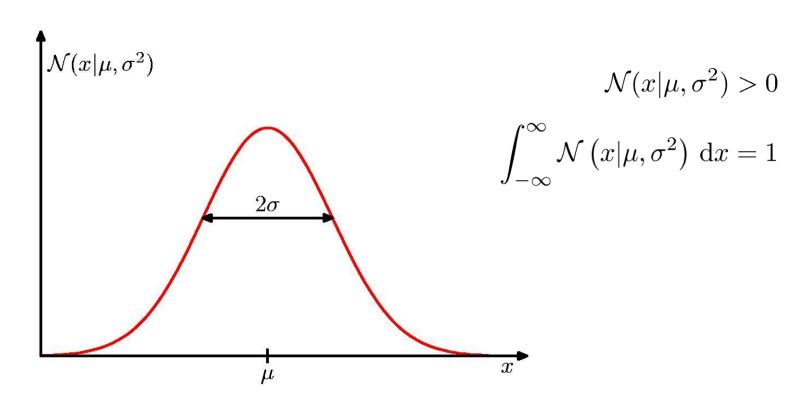
$$= \mathbb{E}_{x, y} [xy] - \mathbb{E}[x]\mathbb{E}[y]$$

$$\operatorname{cov}[\mathbf{x}, \mathbf{y}] = \mathbb{E}_{\mathbf{x}, \mathbf{y}} \left[ \{\mathbf{x} - \mathbb{E}[\mathbf{x}]\} \{\mathbf{y}^{\mathrm{T}} - \mathbb{E}[\mathbf{y}^{\mathrm{T}}]\} \right]$$

$$= \mathbb{E}_{\mathbf{x}, \mathbf{y}} [\mathbf{x}\mathbf{y}^{\mathrm{T}}] - \mathbb{E}[\mathbf{x}]\mathbb{E}[\mathbf{y}^{\mathrm{T}}]$$

## The Gaussian Distribution

$$\mathcal{N}\left(x|\mu,\sigma^2\right) = \frac{1}{(2\pi\sigma^2)^{1/2}} \exp\left\{-\frac{1}{2\sigma^2}(x-\mu)^2\right\}$$



## Gaussian Mean and Variance

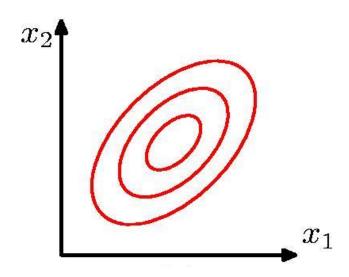
$$\mathbb{E}[x] = \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) x \, \mathrm{d}x = \mu$$

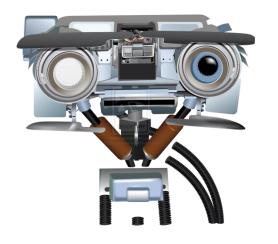
$$\mathbb{E}[x^2] = \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) x^2 dx = \mu^2 + \sigma^2$$

$$var[x] = \mathbb{E}[x^2] - \mathbb{E}[x]^2 = \sigma^2$$

## The Multivariate Gaussian

$$\mathcal{N}(\mathbf{x}|\boldsymbol{\mu}, \boldsymbol{\Sigma}) = \frac{1}{(2\pi)^{D/2}} \frac{1}{|\boldsymbol{\Sigma}|^{1/2}} \exp\left\{-\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu})^{\mathrm{T}} \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu})\right\}$$





# Linear Algebra review

Matrices and vectors

### Matrix Elements (entries of matrix)

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

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

**Vector:** An n x 1 matrix.

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

$$y_i=i^{th}$$
 element

#### 1-indexed vs 0-indexed:

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} \qquad y = \begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

#### **Matrix Addition**

$$\begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} + \begin{bmatrix} 4 & 0.5 \\ 2 & 5 \\ 0 & 1 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & 0 \\ 2 & 5 \\ 3 & 1 \end{bmatrix} + \begin{bmatrix} 4 & 0.5 \\ 2 & 5 \end{bmatrix} =$$

## **Scalar Multiplication**

$$\begin{vmatrix}
1 & 0 \\
2 & 5 \\
3 & 1
\end{vmatrix} =$$

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

## **Combination of Operands**

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



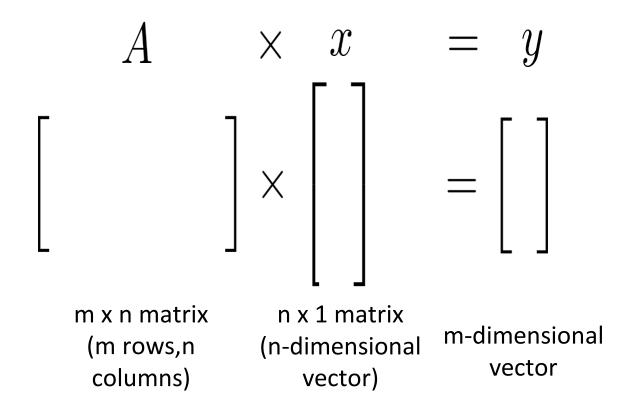
# Linear Algebra review

Matrix-vector multiplication

## **Example**

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

#### **Details:**



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

### **Example**

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

#### House sizes:

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

1416

1534

852

How do we get predicted price as matrix-vector product?



# Linear Algebra review

Matrix-matrix multiplication

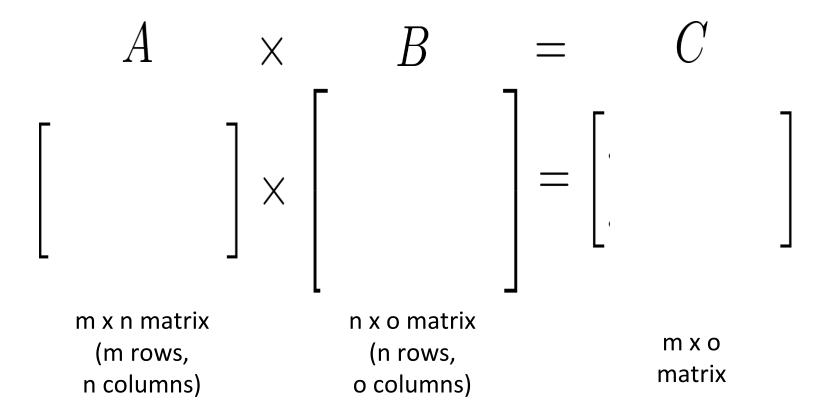
### **Example**

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

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

$$\begin{bmatrix} 1 & 3 & 2 \\ 4 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} 3 \\ 1 \\ 2 \end{bmatrix} =$$

#### **Details:**



The  $i^{th}$  column of the matrix C is obtained by multiplying A with the  $i^{th}$  column of B. (for i = 1,2,...,0)

## Given house

sizes:

2104 1416 1534

852

What is the price of each house?

# Have 3 competing linear functions:

1. 
$$h_{\theta}(x) = -40 + 0.25x$$

2. 
$$h_{\theta}(x) = 200 + 0.1x$$

3. 
$$h_{\theta}(x) = -150 + 0.4x$$

#### Matrix

1 2104

852

#### Matrix

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



# Linear Algebra Review

Matrix multiplication properties

Let A and B be matrices. Then in general,  $A\times B\neq B\times A.$  (not commutative.)

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

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

#### Associative

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

Let  $D = B \times C$ . Compute  $A \times D$ .

Let  $E = A \times B$ . Compute  $E \times C$ .

### **Identity Matrix**

Denoted I (or  $I_{n \times n}$ ). Examples of identity matrices:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ \mathbf{2} \times \mathbf{2} \\ \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ \mathbf{3} \times \mathbf{3} \\ \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

For any matrix A,

$$A \cdot I = I \cdot A = A$$

In general, is AB = BA?



# Linear Algebra review

Inverse and transpose

## Not all numbers have an inverse

#### **Matrix inverse:**

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

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

For a 2 x 2 matrix, what is a sufficient condition for it to have an inverse?

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

### **Matrix Transpose**

 $A = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 5 & 9 \end{bmatrix} \qquad A^T = \begin{bmatrix} 1 & 3 \\ 2 & 5 \\ 0 & 9 \end{bmatrix}$  Let A be an m x n matrix, and let  $B = A^T$ . Then B is an n x m matrix, and

$$B_{ij} = A_{ji}$$
.