

1. Relationships between R.V.s
2. Infer info about 1 RV given another

Events  $A + B$

holds  $\left\{ \begin{array}{l} A > B \\ A \sim B \\ A < B \end{array} \right.$  more plausible  
as plausible

Universal Comparability

Transitivity

if  $A \succeq B$  and  $B \succeq C \Rightarrow A \succeq C$

$P(A) > P(B)$  iff  $A > B$

$P(A) = P(B)$  iff  $A \sim B$

Random Variable

Today only!

R.V. Capital

Happy Meal



Chipotle



Filet Minion



Variable

- take on finite set of values
- each has probability

$$P(X=1)=0.5$$

Variable

- Continuous
- Discrete
- Related to other RVs

$$P(X|Y)$$

↑

DMU

$(\Omega, \mathcal{F}, P)$

$$X: \Omega \rightarrow E$$

Vocab

support(X)

set of all values  
 $X$  can take

Coinflip

$\{h, t\}$   
 $\{1, 0\}$

$N(\mu, \sigma)$

$\mathbb{R}$

Distribution

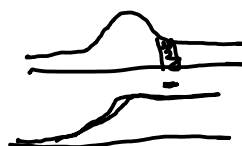
discrete: PMF  
continuous: PDF

function that matches  
each value in support  
to a real number

$$P(X=1)=0.5$$



$$p(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2}$$



CDF

$$\text{cdf}(x) = P(X \leq x)$$

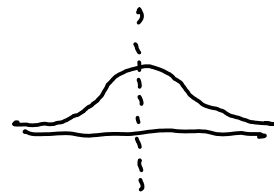
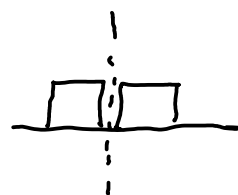
cdf



Expectation

$$E[X]$$

$$= \sum_{i \in X} i P(X=i) \quad \int_{-\infty}^{\infty} x p(x) dx$$



Multiple R.V.s

$X, Y$

Joint  
 $P(X, Y, Z)$

Conditional Dist  
 $P(X|Y, Z)$

Marginal Dist  
 $P(X), P(Y), P(Z)$

X	Y	Z	P(X, Y, Z)
0	0	0	0.08
0	0	1	0.31
0	1	0	0.09
0	1	1	0.37
1	0	0	0.01
1	0	1	0.05
1	1	0	0.02
1	1	1	0.07

Dist-valued function  
 $Y, Z \rightarrow \text{dist of } X$

X	$P(X Y=0, Z=0)$
0	0.888...
1	0.111...

$P(X|Y, Z)$

X	P(X)	Y	P(Y)
0	0.85	0	0.45
1	0.15	1	0.55

Z	P(Z)
0	0.20
1	0.80

Chipotle

Filet Mignon  
Axionization

3 Rules

- 1) a)  $0 \leq P(X|Y) \leq 1$   
b)  $\sum_{x \in X} P(x|Y) = 1$

- 2) "Law of total Probability"  
 $P(X) = \sum_{y \in Y} P(X, y)$

- 3) Def. of Conditional Prob

$$P(X|Y) = \frac{P(X, Y)}{P(Y)}$$

joint  $\rightarrow$  marginal

joint  $\rightarrow$  conditional

Marginal + Cond.  $\rightarrow$  Joint

$$P(X, Y) = P(X, Y) P(Y)$$

Often:

A: hidden

B: measure

know  $P(B|A)$

Bayes Rule:

$$P(A|B) = \frac{P(A,B)}{P(B)}$$

$$P(B|A) = \frac{P(B,A)}{P(A)}$$

$$P(A|B)P(B) = P(A,B) = P(B|A)P(A)$$

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

$$P(A|B,C) = \frac{P(B|A,C)P(A|C)}{P(B|C)}$$

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Breakout Rooms

Introduce Yourself:  
Favorite Prog. Lang.

$S$  = scam

$C_+ =$  clue

$$P(S=1) = 0.1$$

$$P(C_+=1 | S=1) = 0.9$$

$$P(C_+=1 | S=0) = 0.2$$

How many clues ( $C_1=1, C_2=1, C_3=1 \dots$ ) do you need to get in a row to be 90% sure that there is a scam?

$$P(S=1 | C_1=1 \dots C_n=1) = \frac{P(C_1=1 \dots C_n=1 | S=1) P(S=1)}{P(C_1=1 \dots C_n=1)}$$

$$P(C_{1:n}=1 | S=1) = 0.9^n$$

$$P(S=1) = 0.1$$

$$P(C_1=1 \dots C_n=1) = P(C_{1:n}=1 | S=0) P(S=0) + P(C_{1:n}=1 | S=1) P(S=1)$$

$$0.2^n \cdot 0.9 + 0.9^n \cdot 0.1$$

3 clues!

## Independence

Def  $X$  and  $Y$  are Independent iff  $P(X,Y) = P(X)P(Y)$   
 $X \perp Y$

$$P(X|Y) = P(X)$$

Def  $X$  and  $Y$  are conditionally Independent given  $Z$   
iff  $P(X,Y|Z) = P(X|Z)P(Y|Z)$

$$X \perp Y | Z$$

$$P(X|Y,Z) = P(X|Z)$$

$$P(C_1, C_2 | S) = P(C_1 | S)P(C_2 | S)$$