

# Probability Review

Note Title

9/12/2006

Event - set of possible outcomes of an experiment  
 $A, B, C$

dice roll = 6

flipped coin lands on heads

Discrete Probability ( $Pr[\ ]$ )

$[ ]$   $0 \leq Pr[A] \leq 1$

- likelihood
- Percent outcome

Side Note:

assume representative

Determine if a light will be red at 11:34 am

only happens ONCE!

Ergodic  $\rightarrow$  doing a test multiple times will yield same results as sampling over time

# Important Sets

$S$  - Universal set

$\emptyset$  - NULL set

$$\Pr[S] = 1$$

$$\Pr[\emptyset] = 0$$

Events are treated as sets

- all sets are included in the Universal set

$$\Pr[\overline{A}] = 1 - \Pr[A]$$

↖ compliment

$A$  = flip of a coin is Heads

More interesting sets:

### Mutually Exclusive (ME)

- at most, one outcome can occur

Eg:

- a coin cannot land on both HEADs and Tails
- a die cannot land on both side 1 and 6

### Collectively Exhaustive (CE)

- at least one of the events must occur

dice 1, 2, 3, 4, 5, 6

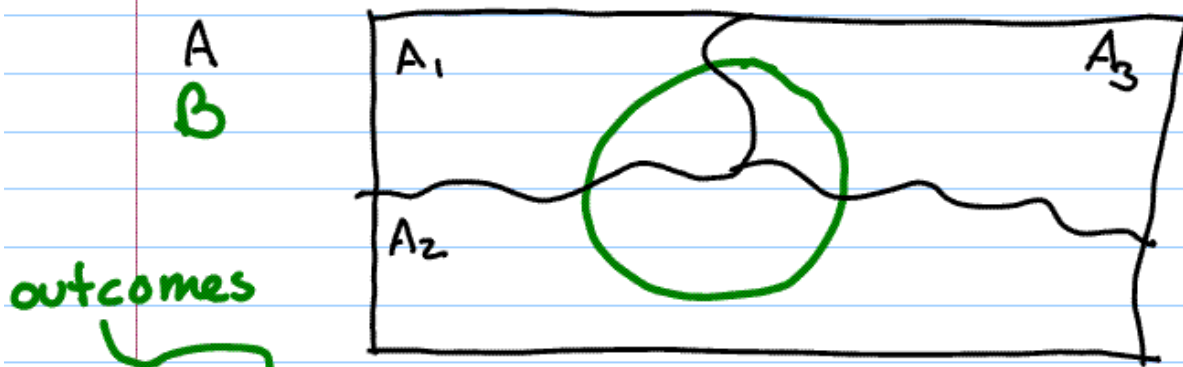
Six sided die roll is both ME + CE

$$A = [1, 2, 3, \dots, m]$$

A is both ME + CE iff

$$\sum_{i=1}^m \Pr[A_i] = 1 \quad \text{and}$$

$$\Pr[B] = \sum_{i=1}^m \Pr[A_i \cap B]$$



if  $A_i$  are ME then

$$\Pr[A_j \cup A_i] = \Pr[A_j] + \Pr[A_i]$$

# Conditional Probability

What is the Probability of A given that we know B

if B is in the set of possible outcomes is A also in this set?

$$P[A|B] = \frac{P[A \cap B]}{P[B]}$$

$$P[A \cap B] = P[A|B] \cdot P[B]$$

What if A and B are Independent

$$P[A \cap B] = P[A] \cdot P[B]$$

$$P[A] = P[A|B]$$

telling me about B tells me nothing about A

An Example?

$\Pr[\text{Graduate GT} \mid \text{We work REALLY Hard}]$

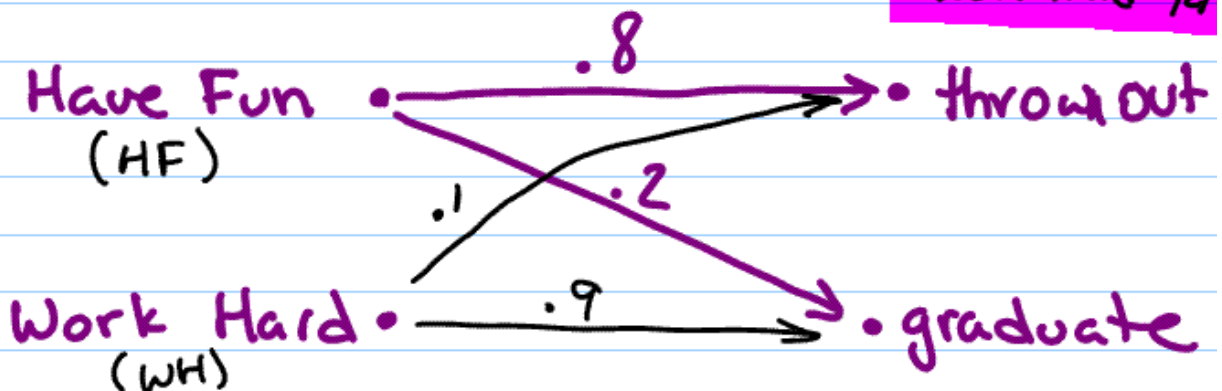
2 ME class

B { Have Fun?  
Work Hard

Prior  
 $\frac{1}{4}$   
 $\frac{3}{4}$

A { Thrown out of Tech  
Graduate

Prior have fun  $\frac{1}{4}$   
work hard  $\frac{3}{4}$



$$\Pr[\text{Error}] = \Pr[\text{HF} \cap \text{Error}] + \Pr[\text{WH} \cap \text{Error}]$$

$$\begin{aligned} &= \Pr[E | HF] \cdot \Pr[HF] + \\ &\quad \Pr[E | WH] \cdot \Pr[WH] \\ &= (.2 \cdot .25) + (.1 \cdot .75) \\ &= .05 + .075 = .125 \\ &= 1/8 \end{aligned}$$

Priors impact everything

- where do they come from?

- how do we pick them

BAD

→ counting

→ fiddle with them to get results you want

$$P(A \cap B) = P(B \cap A)$$

Bayes Rule

$$P[A|B] = \frac{P[A \cap B]}{P[B]}$$

$$= \frac{P[B|A] \cdot P[A]}{P[B]}$$

How do we use this for  
Pattern Rec?

$A \approx$  class  $\omega_i \in \Omega$

$B \approx$  data  $d$

$w_i = \text{Salmon}$

$$Pr[w_i | d]$$

What is the probability of class  $w_i$  given that we see data  $d$ ?

Bayes Rule

$$Pr[w_i | d] = \underbrace{Pr[d | w_i]}_{\text{likelihood}} \cdot \underbrace{Pr[w_i]}_{\text{Prior}}$$

$$\underbrace{Pr[d \wedge w_i] + Pr[d \wedge \bar{w}_i]}_{\text{evidence}} \rightarrow Pr[d]$$

who cares what the probability  $Pr[d]$  is? it is your data and you are stuck with it

you can compute  $Pr[w_i | d]$  without  $Pr[d]$

- Normalize!  $\nabla$

Normalize by  $\sum P_r^*[w_i | d]$

$$P_r^*[w_i | d] = P_r[w_i | d] P_r[d]$$

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Next class

- Continuous Probability
- Gaussians
- Distance Metrics
- Bayesian Classifiers