## Maximum Likelihood Estimation (MLE)

- Consider a sequence of coin flips, for example 🛮 😽 🖂
- HHTTTTHTTHTTHTTHTTHTT (5 times H, 15 times T)
  - Which Pr(H) and Pr(T) are the most likely?
  - Looks like  $Pr(H) = \frac{1}{4}$  and  $Pr(T) = \frac{3}{4}$  ...  $P(T) = \frac{1}{17} = \frac{1}{20}$   $= \frac{3}{4}$

## Conditional probabilities

- Let A and B be events in a probability space  $\Omega$
- Denote by  $Pr(A \mid B)$  the probability of  $A \cap B$  in the space B
- (1) Pr(A | B) := Pr(A ∩ B) / Pr (B)
- (2)  $Pr(A \mid B) \cdot Pr(B) = Pr(B \mid A) \cdot Pr(A)$

# Pr(A|B)=Pr(B|A).P(A)/P(B)

# Naive Bayes

$$P(H) = \frac{|H|}{|F|} = 8$$

- Ω = {1,2,3,4,5,6}
- $A = \{2, 4, 6\} \Rightarrow P(A) = \frac{|A|}{|\Omega|} = \frac{3}{6} = \frac{1}{2}$
- B= 21, 2,3]=) P(B)=131=3/6=1/2 P(A|B)= 1A(B) 1521

# $\frac{1}{3} = \frac{1}{3}$

# Probabilistic assumptions

- Underlying probability distributions:
  - A distribution  $p_c$  over the classes ... where  $\Sigma_c$   $p_c = 1$ For each c, a distr.  $p_{\underline{w}c}$  over the words ... where  $\Sigma_{\underline{w}}$   $p_{\underline{w}c}$  = 1
- Naïve Bayes assumes the following process for generating a document D with m words W<sub>1</sub>...W<sub>m</sub> and class label C:
  - Pick C=c with prob.  $p_c$  , then pick each word  $W_i$ =w with probability  $p_{wc}$  , independent of the other words

## Learning phase

– For a training set <u>T</u> of objects, let:

 $T_c$  = the set of documents from class c

 $n_{wc}$  = #occurrences of word w in documents from  $T_c$ 

 $n_c$  = #occurrences of all words in documents from  $T_c$ 

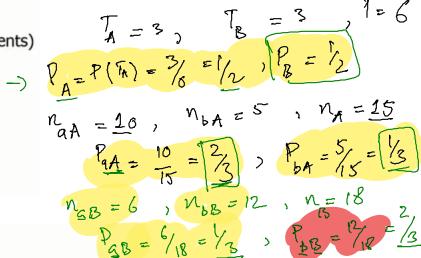
- We compute the following probabilities or likelihoods:

- $p_c := |T_c| / |T_c|$  global likelihood of a class  $p_{wc} := n_{wc} / n_c$  likelihood of a word for a class
- > Thorror = 10 | > Tooc = 25 | > Tomedy = 15 | P Jupyler, Doc = 20 | = 1/4 = 0.25

### Learning phase, example

Consider Example 2 (artificial documents)

,	
aba	Α
baabaaa	A
bbaabbab	В
abbaa	A
abbb	В
bbbaab	В



#### Prediction

 For a given document d we want to compute the probability of each class **c**, given document d:

- Using Bayes Theorem, we have:
$$Pr(C=c \mid D=d) \neq Pr(D=d \mid \underline{C=c}) \cdot Pr(C=c) / Pr(C=c)$$

- Using our (naive) probabilistic assumptions, we have:

$$Pr(D=d \mid C=c) = Pr(W_1=w_1 \cap ... \cap W_m=w_m \mid C=c)$$

$$= \prod_{i=1,...,m} Pr(W_i=w_i \mid C=e)$$

#### Prediction ... continued

We thus obtain that

$$Pr(C=c \mid D=d)$$

$$= \prod_{i=1,...,m} Pr(W_i=w_i \mid C=c) \cdot Pr(C=c) / Pr(D=d)$$

$$= \prod_{i=1,...,m} \mathbf{p}_{w_ic} \cdot \mathbf{p}_c / Pr(D=d)$$

- Note 1: for  $\Pi_{i=1,...,m}$   $p_{w_ic}$  just take the  $p_{wc}$  for all words w in the document and multiply them (if a word w occurs multiple times, also take the factor p<sub>wc</sub> multiple times)
- Note 2: the Pr(D=d) is the same for all classes c

### Prediction, example

Consider Example 2 (artificial documents)

– Let us predict the class for aab ... A or B?

$$P(A|aab) > P(B|aab) \rightarrow Reduct$$

P(A|B) = 
$$\frac{14 \text{ NB}}{18}$$

$$= \frac{14 \text{ NB}}{18} / \frac{121}{20} = \frac{14 \text{ NB}}{18}$$

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