

# Naïve Bayes Classifier – Concept of Laplace Smoothing



- So far you know how to represent feature vector  $X_j$
- How to create Vocabulary.
- And how to develop a model using Naïve Bayes assumption.
- Next task is to get an e mail and predict using my model whether it is a spam or non spam mail.

$$P(X_j | Y) \text{ \& } P(Y)$$

When an unknown word appears, what will happen to my predictor?

$$P(Y=1|X) = \frac{\prod_{j=1}^n P(X_j | Y=1) P(Y=1)}{\prod_{j=1}^n P(X_j | Y=1) P(Y=1) + \prod_{j=1}^n P(X_j | Y=0) P(Y=0)}$$

$$= \frac{0}{0}$$

$$X = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \begin{matrix} \text{er} \\ \text{absu} \\ \text{abra} \\ \vdots \\ \text{lottery} \\ \text{rev} \end{matrix}$$

10000x1



# Laplace Smoothing

It is a bad idea to put  $P(X_{i,00}/Y=1)=0$

How to fix it?



Take the example of Cricket tournament with following teams: India, Pakistan, Sri Lanka, Bangladesh, Australia, New Zealand, West Indies, England, Netherland. Following is the winning record of Netherland:

Date	Afghanistan's winning record	Win
15/09	Played with India	0
17/09	Played with Australia	0
19/09	Played with England	0
20/09	Played with Pakistan	0
22/09	Played with Sri Lanka	0
23/09	Going to play with Bangladesh	?

The maximum likelihood estimate of  $\phi_j / Y=1$  is

$$= \frac{\#1 + 1}{\#0 + 1 + \#1 + 1}$$
$$= \frac{0 + 1}{5 + 1 + 0 + 1} = \frac{1}{7}$$



# Laplace Smoothing



So In general the Laplace smoothing for Multivariate Bernoulli event model :

$$\phi_{j|Y=1} = \frac{\sum_{i=1}^m 1\{X_j^{(i)}=1 \wedge Y^{(i)}=1\} + 1}{\sum_{i=1}^m 1\{Y^{(i)}=1\} + 2}$$

$$\phi_{j|Y=0} = \frac{\sum_{i=1}^m 1\{X_j^{(i)}=0 \wedge Y^{(i)}=0\} + 1}{\sum_{i=1}^m 1\{Y^{(i)}=0\} + 2}$$

And for multinomial event model

$$\phi_j = \frac{\sum_{i=1}^m 1\{Z^{(i)}=j\} + 1}{m + K}, \quad \left\{ \begin{array}{l} \text{where } Z \text{ is a multinomial random variable -} \\ \text{parametrized by } \phi_j = P(Z=j) . \\ \text{Given a set of } m \text{ independent observations } [Z^{(1)}, \dots, Z^{(m)}], \\ \text{the maximum likelihood estimate are given by :} \\ \phi_j = \frac{\sum_{i=1}^m 1\{Z^{(i)}=j\}}{m} . \end{array} \right.$$