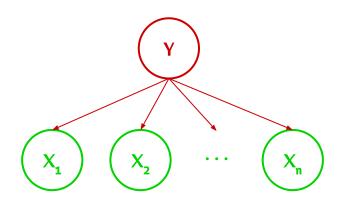
**Exam Prep Discussion 11** 

#### **Maximum Likelihood Estimation**

- Setup: n observations  $X_1$ , ...,  $X_n$  drawn from a distribution parameterized by  $\theta$ 
  - We know that  $\theta$  "determines" the  $X_i$ 's
  - O Now that we've seen the  $X_i$ 's, what is the most likely value of θ?

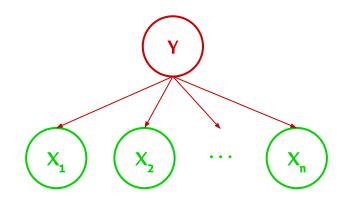
# Naive Bayes



## **Naive Bayes**

• Method to predict the probability of observing a label Y given features  $X_1, ..., X_n$ :

$$P(Y = y | X_1 = x_1, \dots, X_n = x_n)$$



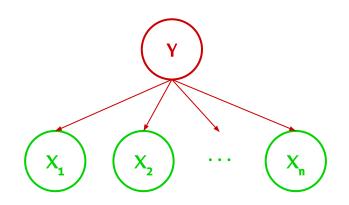
## **Naive Bayes**

• Method to predict the probability of observing a label Y given features  $X_1, ..., X_n$ :

$$P(Y = y | X_1 = x_1, \dots, X_n = x_n)$$

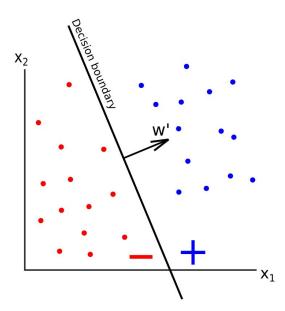
• Prediction formula:

$$Pred(X) = argmax_y P(Y = y | X_1 = x_1, ..., X_n = x_n)$$



## Perceptron

- Assume data  $\vec{x} = [x_1, ..., x_d]^T$  and has a label  $y^*$
- Machine learning algorithm that generates a linear decision boundary in d dimensions
  - Uses weight vector  $\vec{w} = [w_1, \dots, w_d]^T$
- Goal:  $\vec{w}^T \vec{x}$  correctly determines the sign of  $y^*$



### **Perceptron**

- Assume data  $\vec{x} = [x_1, ..., x_d]^T$  and has a label  $y^*$
- Machine learning algorithm that generates a linear decision boundary in d dimensions
  - Uses weight vector  $\vec{w} = [w_1, \dots, w_d]^T$
- Goal:  $\vec{w}^T \vec{x}$  correctly determines the sign of  $y^*$

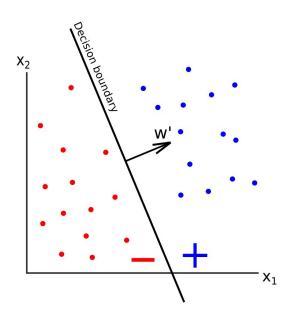
#### Algorithm.

- 1) Start with  $\vec{w} = 0$
- 2) While not all data points are correctly classified,
  - a) Predict label of point using current weight vector:

$$y_{\text{pred}} = \begin{cases} +1 & w^T x \ge 0 \\ -1 & w^T x < 0 \end{cases}$$

b) If prediction is *incorrect*, update weight vector:

$$\vec{w} \leftarrow \vec{w} + \alpha y^* \vec{x}$$



x: [2, 1], with label 
$$y^* = -1$$
.  $\alpha = 1$   
 $w = [1, 0]$   
 $w^T x = 2 + 0 = 2 \Rightarrow y_{pred} = +1$   
 $w' = [1, 0] + (-1)[2, 1] = [-1, -1]$   
Now,  $w'^T x = -2 - 1 = -3 \Rightarrow y_{pred} = -1 = y^*$