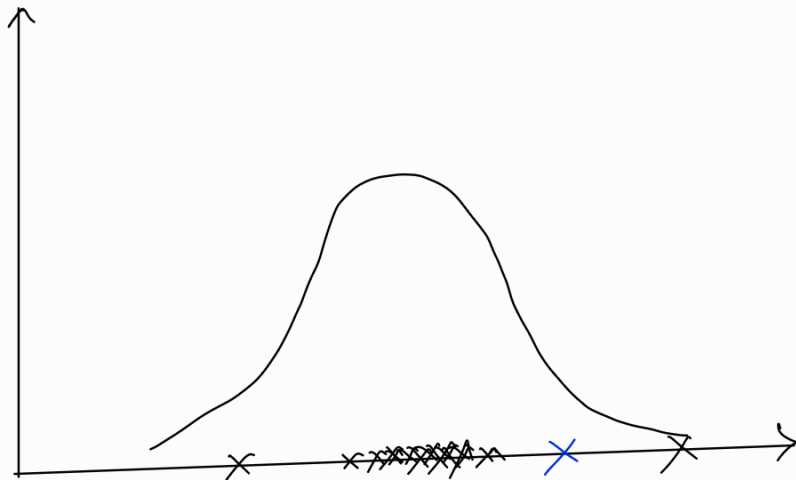


06/02/24

$$E(X) = \sum x \cdot p(x)$$

When we need to sample, we use CDF.

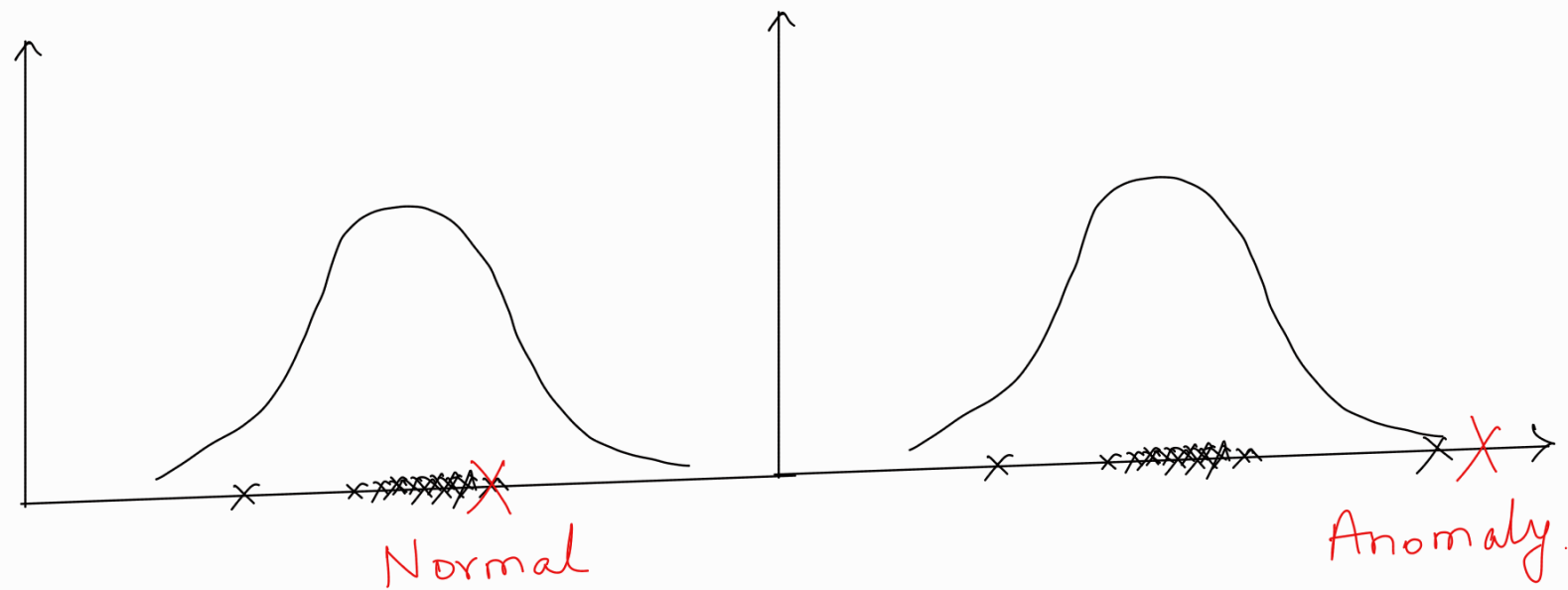
- 1] Draw CDF $CDF = \int f(x)$ → distribution
- 2] Select random number
- 3] Distribution is largely Gaussian



Prob of blue point belonging to PDF.

Put value in Gaussian equation and find likelihood.

→ Anomaly detection is possible using density estimation. If the data point does not lie in the normal range, it can be flagged as anomaly.



$$P(x|y_1, y_2) = P(x|y_1) P(x|y_2)$$

Naive Bayes

→ Will not work if data is correlated

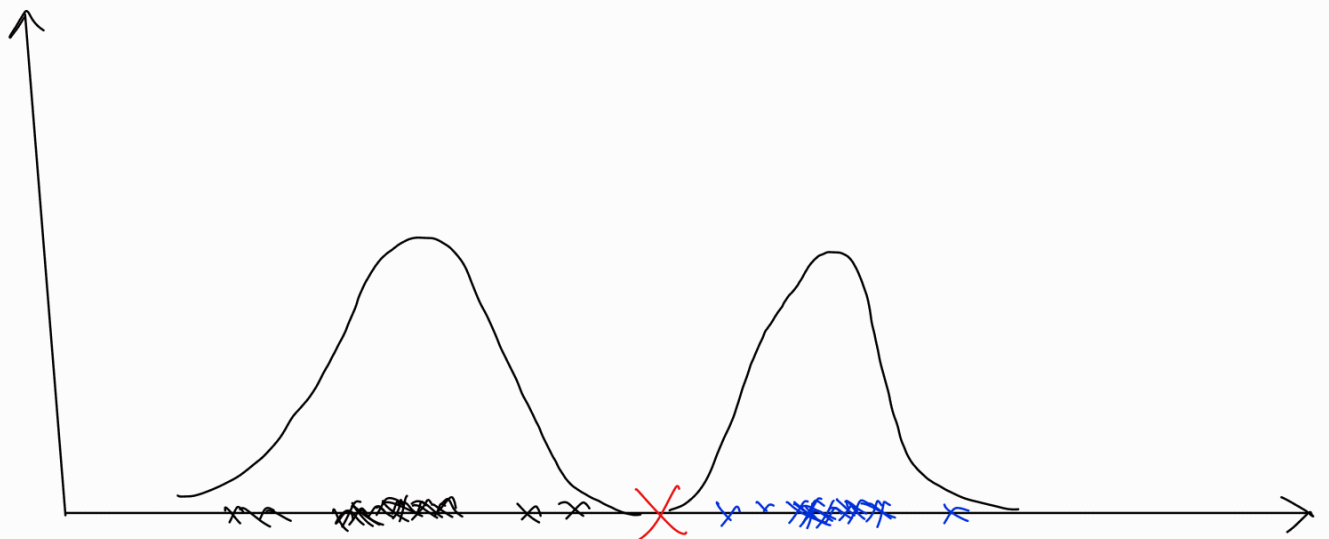
Multivariate Density Estimation

$$f(x) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma|^{\frac{1}{2}}} e^{\left[-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu) \right]}$$

Slide 19

If covariance was 0.4 in middle block,
tilt will be visible.

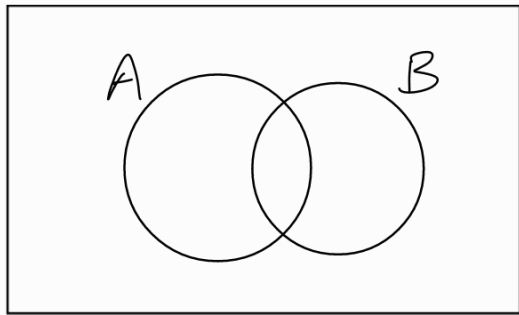
Generative Classification



Compute $P(y_1|x)$ and $P(y_2|x)$

$$\underbrace{P(y_1|x)}_{\text{posterior}} = \frac{\underbrace{P(x|y_1)}_{\text{likelihood}} \cdot \underbrace{P(y_1)}_{\text{prior}}}{P(x)}$$

$$P(x) = P(x|y_1) P(y_1) + P(x|y_2) P(y_2)$$



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

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

Use this to find $P(A)$ and $P(B)$

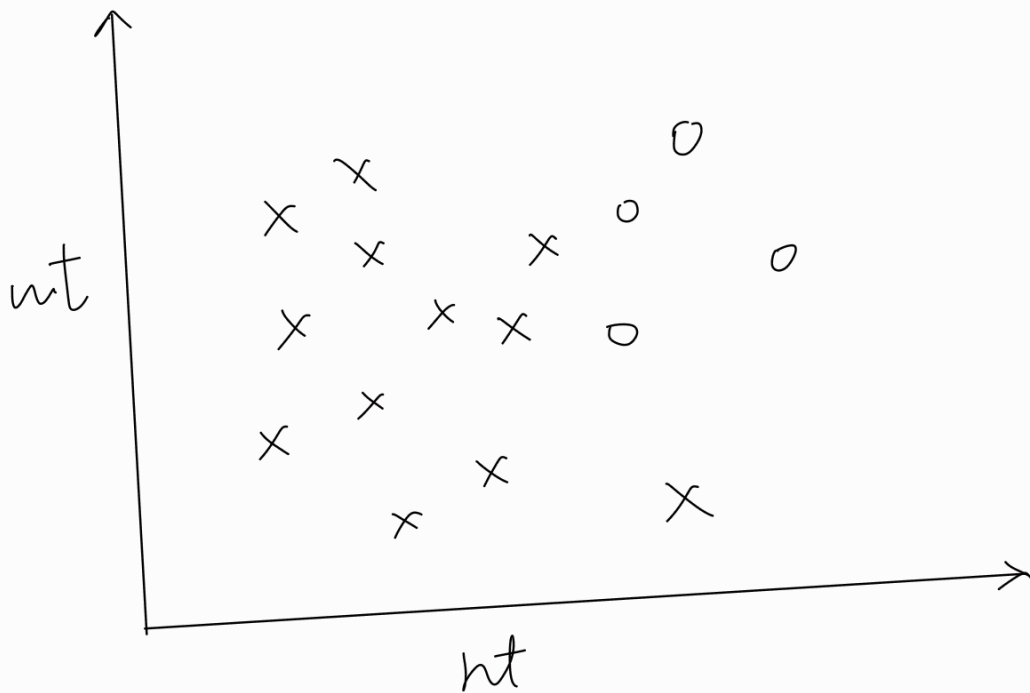
Find $P(y_1|x)$ and $P(y_2|x)$ and assign the class whose probability is higher.

→ Ignore denominator for classification tasks.

Naive Bayes

$$P(x_1, \dots, x_d | y) = \prod$$

eg]



$$p(a) = \frac{4}{16} \quad p(c) = \frac{12}{16}$$

$$p(x|a) = p(h_x|a) \cdot p(w_x|a)$$

$$p(x|c) = p(h_x|c) \cdot p(w_x|c)$$

→ This is possible as weight and height are not correlated. Take as independent events & multiply probs.

Spam Classification

→ Data follows multinomial distribution

Dictionary : Dear, Friend, Lunch, Money

$$P(N) = 0.67$$

$$P(\text{Dear} | N) = 0.471$$

$$P(\text{Friend} | N) = 0.291$$

$$P(\text{Lunch} | N) = 0.181$$

$$P(\text{Money} | N) = 0.061$$

$$P(S) = 0.33 \quad P(\text{Dear} | S) = 0.29$$

$$P(\text{Friend} | S) = 0.14$$

$$P(\text{Lunch} | S) = 0$$

$$P(\text{Money} | S) = 0.57$$

Email : Dear friend

$$\begin{aligned}
 P(N | \text{"Dear Friend"}) &= P(\text{"D.F"} | N) \cdot P(N) \\
 &= P(\text{Dear} | N) P(\text{Friend} | N) \cdot P(N) \\
 &= 0.471 \times 0.291 \times 0.67 \\
 &= 0.091
 \end{aligned}$$

$$\begin{aligned}
 P(S | \text{"DF"}) &= 0.29 \times 0.14 \times 0.33 \\
 &= 0.013
 \end{aligned}$$

\therefore The email classified as normal.

Email: Lunch Money Money Money

$$P(N | \text{Email}) = (0.181) \times (0.061)^3 \times 0.67 = 0.00002$$

$$P(S | \text{Email}) = \underbrace{0.001} \times (0.57)^3 \times 0.33 = 0.00006$$

$P(\text{Lunch} | S) = 0$ - But we add 0.001 to every probability.

\therefore Email classified as spam.

