

Assignment - 1

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Q1: We know that joint distribution is

$$P(A, B) = P(\text{event A happening} \& \text{event B happening})$$

$$P(A, B) = P(B|A) P(A)$$

if B is conditionally dependent on A

Joint distribution
for bivariate random variables.

→ Linear Regression

our hypothesis

$$H(X) = F(x_1, x_2, x_3, x_4)$$

This is learnt by working on minimizing the cost function (Mean Squared error), and doesn't deal with probabilities.

Thus, we can't calculate $F(x_1, x_2, x_3, x_4, y)$ on this information

→ Logistic Regression

our hypothesis

$H(x)$ is function.

$$H(x) = P(y | x_1, x_2, x_3, x_4) \cdot P(x_1, x_2, x_3, x_4)$$

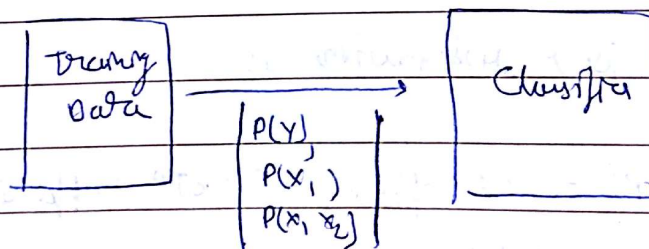
This is just $P(x=2)$ given $\{x_1, x_2, x_3, x_4\}$

But H doesn't contain $P(x_1, x_2, x_3, x_4)$ from that classifier.

$$\frac{1}{1 + e^{-\theta x}}$$

Probability that
 $y=1$

→ Naive Bayes Classifier ^{Bayes}
(when developing Naive Classifier)



our classifier is already built by calculating joint probabilities. Thus it contains information by which we can compute

$$P(Y, x_1, x_2, x_3, x_4)$$