

★ Naive Bayes

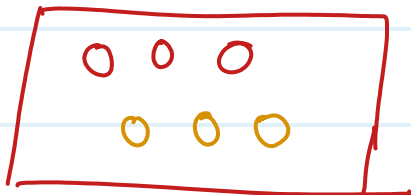
1

It works on probability based approach, and It is used to solve classification problem.

Independent event

$$\begin{array}{lcl} \text{coin toss (H)} & = & \frac{1}{2} \\ \text{Dice} & = & \frac{1}{6} \end{array}$$

Dependent probability



$$Pr(R) = \frac{3}{5}$$

After remove one red marble

$$Pr(Y) = \frac{2}{4}$$

$$\frac{3}{6} \times \frac{2}{5}$$

conditional probability

R = Red
G = Green

2

$$P_r(R \text{ and } G) = P(R) \times P(G|R)$$

★ formulas

$$P_r(A \text{ and } B) = P(A) \times P(B|A)$$

$$P_r(A \text{ and } B) = P_r(B \text{ and } A)$$

$$P(A) \times P(B|A) = P_r(B) \times P(A|B)$$

formulas

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

Baye's Theorem

example

A	B
$x_1 \ x_2 \ x_3 \ \dots \ x_n$	y

$$P(Y/(x_1 \ x_2 \ x_3 \ \dots \ x_n)) = \frac{P(Y) \times P[(x_1 \ x_2 \ x_3 \ \dots \ x_n)/Y]}{P(x_1 \ x_2 \ x_3 \ \dots \ x_n)}$$

$Y = \text{Yes/No}$
Binary class

yes

$$P[\text{yes}/(x_1, x_2, x_3)] = \frac{P(\text{yes}) \times P(x_1/\text{yes}) \times P(x_2/\text{yes}) \times P(x_3/\text{yes})}{\text{constant} - \cancel{P(x_1)} \cancel{P(x_2)} \cancel{P(x_3)}}$$

no

$$P[\text{no}/(x_1, x_2, x_3)] = \frac{P(\text{no}) \times P(x_1/\text{no}) \times P(x_2/\text{no}) \times P(x_3/\text{no})}{\text{constant} - \cancel{P(x_1)} \cancel{P(x_2)} \cancel{P(x_3)}}$$

Example - Dataset

outlook, Temp. humidity wind play tennis
 x_1, x_2, x_3, x_4, Y

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outlook	yes	no	$P(Y)$	$P(N)$
sunny	2	3	2/9	3/5
overcast	3	0	3/9	0/5
Rain	4	2	4/9	2/5
	$\frac{4}{9}$	$\frac{2}{5}$		
	$\frac{9}{14}$	$\frac{5}{14}$		
	Y	N		

* Temp

	yes	no	$p(y)$	$p(n)$
Hot	2	2	2/9	2/5
mild	4	2	4/9	2/5
cold	3	1	3/9	1/5
	$\frac{9}{9}$	$\frac{5}{5}$		

$$P(\text{yes}) = \frac{9}{14}$$

$$P(\text{no}) = \frac{5}{14}$$

For new dataset

Test (sunny, hot) = play ??

$$p(Y/\text{sunny, hot}) = p(Y) \times p(\text{sunny}/Y) \times p(\text{hot}/Y)$$

$$p(N/\text{sunny, hot}) = p(N) \times p(\text{sunny}/N) \times p(\text{hot}/N)$$

$$p(Y/\text{sunny, hot}) = \frac{9}{14} \times \frac{2}{9} \times \frac{2}{9} = 0.031$$

$$p(N/\text{sunny, hot}) = \frac{5}{14} \times \frac{3}{5} \times \frac{2}{5} = 0.085$$

$$p(Y/\text{sunny, hot}) = \frac{0.031}{0.031 + 0.085}$$

$$= 0.27 = 27\%$$

$$P(N / \text{Sunny, hot}) = \frac{0.085}{0.085 + 0.031}$$

$$= 0.73 \Rightarrow 73\%$$

// Naive bayes's classifier used for NLP
(natural language processing)

NLTK - Natural Language Toolkit

RE - Regular expression.

He is eating.

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He - - - -

He - - - -