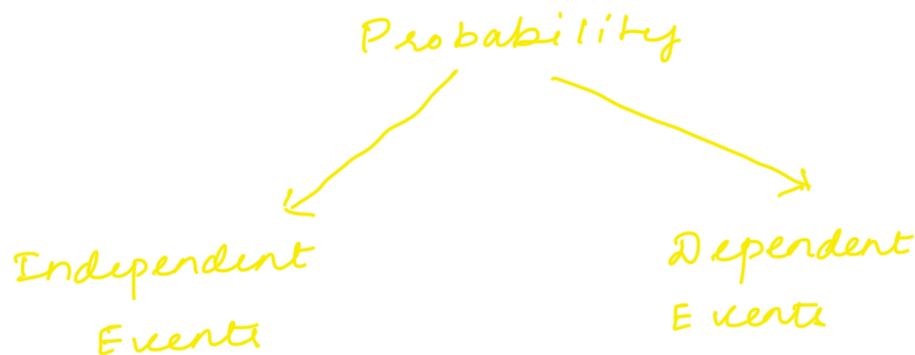


Naive Bayes Algorithm (Classification)

* It is specifically used to solve the classification problem - Both binary and multiclass classification.

① Probability

② Bayes' theorem.



Independent Events

Rolling a Die $\{1, 2, 3, 4, 5, 6\}$

$$Pr(1) = \frac{1}{6} \quad (\text{Chances of getting 1})$$

$$Pr(2) = \frac{1}{6} \quad (\text{Chances of getting 2})$$

* above is an example of chance of an single event.

why it is called as Independent event?

* The probability of one event
... the probability of

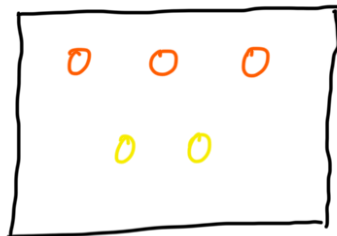
does not change the probability of the other event.

* If you consider the above example.

→ probability of occurring 1 when rolling a die does not change the probability of occurring 2.

Dependent Events

example, consider I have a bag of marbles



① what is the probability of removing an orange marble and then a yellow marble.

$$P(\text{orange}) = \frac{3}{5} \rightarrow \text{1st Event} \left(\frac{\text{total orange marbles}}{\text{total marbles}} \right)$$



| O O | removed.

$$P(\text{yellow}) = \frac{2}{4} \rightarrow \text{2nd Event}$$

therefore, this dependent probability of

this can be represented as

$$P(\text{orange and yellow}) = P(\text{orange}) * P(\text{yellow} \mid \text{orange})$$

1st event ← 2nd event ←



↓

this is nothing but,
probability of orange
has been already taken
place.

this equation is called as conditional probability.

therefore, this can be evaluated as

$$= \frac{3}{5} \times \frac{1}{2}$$

$$= \frac{3}{10} \rightarrow P(\text{orange and yellow})$$

In a generic way this can be represented as

$$P(A \text{ and } B) = P(A) * P(B \mid A)$$



Probability of
B given A (meaning
 $P(A)$ has been already
taken place)

Bayes Theorem

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

$$P(A) * P(B/A) = P(B) * P(A/B)$$

$$P(A/B) = \frac{P(A) P(B/A)}{P(B)}$$

||

Bayes theorem.

$P(A/B)$ = Probability of event A,
given B has occurred.

$P(A)$ = Probability of Event A

$P(B)$ = Probability of Event B

$P(B/A)$ = Probability of event B,
given A has occurred.

Independent Features

Dependent features.

x_1	x_2	x_3	y
—	—	—	yes

-	-	-	no
-	-	-	yes
-	-	-	no

we know that Bayes theorem

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

this can be represented as,

$$P(Y/(x_1, x_2, x_3)) = \frac{P(Y) * P(x_1, x_2, x_3 | Y)}{P(x_1, x_2, x_3)}$$

$$= \frac{P(Y) * P(x_1 | Y) * P(x_2 | Y) * P(x_3 | Y)}{P(x_1) * P(x_2) * P(x_3)}$$

Given our data set

Independent			Dependent
x_1	x_2	x_3	Y
-	-	-	yes
-	-	-	no
-	-	-	yes
-	-	-	no

the equation can be represented as,

$$Pr(Yes/(x_1, x_2, x_3)) =$$

$$\frac{P(yes) * P(x_1 / yes) * P(x_2 / yes) * P(x_3 / yes)}{P(x_1) * P(x_2) * P(x_3)}$$

$$Pr(No/(x_1, x_2, x_3)) =$$

$$P(no) * P(x_1 / no) * P(x_2 / no) * P(x_3 / no)$$

$$\frac{P(x_3 / \text{No})}{P(x_1) + P(x_2) + P(x_3)}$$

If you observe the above two equations the denominator is constant. So we will be removing the denominator.

therefore

$$Pr(\text{Yes} | (x_1, x_2, x_3)) = P(\text{Yes}) * P(x_1 / \text{Yes}) * P(x_2 / \text{Yes}) * P(x_3 / \text{Yes})$$

$$Pr(\text{No} | (x_1, x_2, x_3)) = P(\text{No}) * P(x_1 / \text{No}) * P(x_2 / \text{No}) * P(x_3 / \text{No})$$

Let's consider the Below problem

Day	Outlook	Temperature	Humidity	Wind	Play Tennis
1	Sunny	Hot	High	Weak	No
2	Sunny	Hot	High	Strong	No
3	Overcast	Hot	High	Weak	Yes
4	Rain	Mild	High	Weak	Yes
5	Rain	Cool	Normal	Weak	Yes
6	Rain	Cool	Normal	Strong	No
7	Overcast	Cool	Normal	Strong	Yes
8	Sunny	Mild	High	Weak	No
9	Sunny	Cool	Normal	Weak	Yes
10	Rain	Mild	Normal	Weak	Yes
11	Sunny	Mild	Normal	Strong	Yes
12	Overcast	Mild	High	Strong	Yes
13	Overcast	Hot	Normal	Weak	Yes
14	Rain	Mild	High	Strong	No

<u>Outlook</u>	Yes	No	$P(E / \text{Yes})$	$P(E / \text{No})$
Sunny	2	3	2/9	3/5
Overcast	4	0	4/9	0/5
Rain	3	2	3/9	2/5

run,

<u>Temperature</u>					
	yes	no	$P(E/\text{yes})$	$P(E/\text{no})$	
hot	2	2	$2/9$	$2/5$	
mild	4	2	$4/9$	$2/5$	
cold	3	1	$3/9$	$1/5$	

Play (yes / no)		$P(\text{yes})$	$P(\text{no})$
yes	9	$9/14$	$5/14$
no	5		

Test (sunny, hot) \rightarrow output?

$$P(\text{yes} / \text{sunny, hot}) = P(\text{yes}) * P(\text{sunny}) * P(\text{hot})$$
$$= \frac{9}{14} * \frac{2}{9} * \frac{2}{9}$$

$$= \frac{2}{63}$$

$$= 0.031$$

$$P(\text{no} / \text{sunny, hot}) = P(\text{no}) * P(\text{sunny}/\text{no}) * P(\text{hot}/\text{no})$$
$$= \frac{5}{14} * \frac{3}{5} * \frac{3}{5}$$

$$= \frac{3}{35}$$

$$= 0.085$$

$$P(\text{Yes} | (\text{sunny}, \text{hot})) = \frac{0.031}{(0.031 + 0.085)}$$

$$= 0.27$$

$$= 27\%$$

$$P(\text{No} | (\text{sunny}, \text{hot})) = \frac{0.085}{(0.031 + 0.085)}$$

$$= 0.73$$

$$= 73\%$$

Now with the new test data

<u>Outlook</u>	<u>Temperature</u>	<u>O/P</u>
Sunny	Hot	73%

therefore, there are 73% chances of not (no) playing tennis and 27% chances of playing tennis.

↓↓

... given test

The final outcome for this game, if the data will be no (0) — is not going to play tennis.