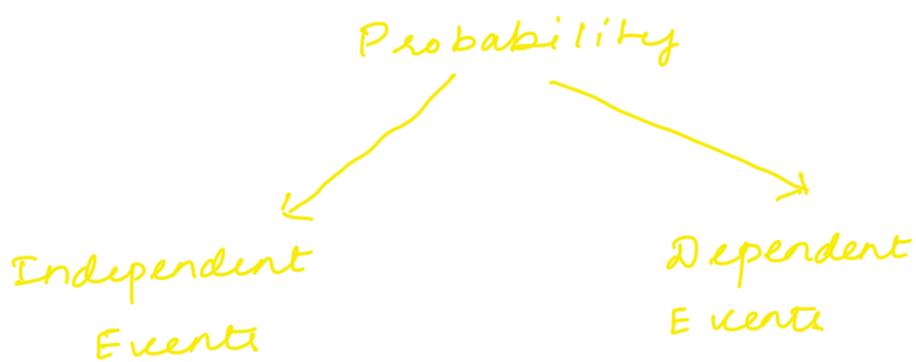


Naive Bayes Algorithm (Classification)

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

- ① Probability
- ② Bayes theorem.



Independent Events

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

$$P_r(1) = \frac{1}{6} \quad (\text{chance of getting 1})$$

$$P_r(2) = \frac{1}{6} \quad (\text{chance of getting 2})$$

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

why it is called as Independent event?

- * The probability of one event $\dots \dots \dots$ the probabilities of

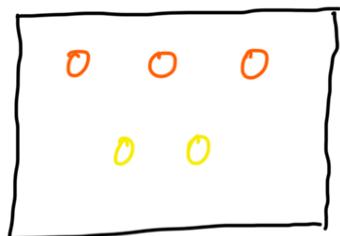
does not change the probability of the other event.

* If you consider the above example.

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

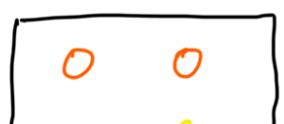
Dependent Events

example, consider I have a bag of marbles



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

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



→ orange marble has been

[O O] removed.

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

therefore, this dependent probability of-

this can be represented as

1st event \hookrightarrow 2nd event \hookleftarrow

$$P(\text{orange and yellow}) = P(\text{orange}) * P\left(\frac{\text{yellow}}{\text{orange}}\right)$$



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

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

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

In a generic way this can be
represented as

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



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

Bayes Theorem

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

$$P(A) \times P(B/A) = P(B) \times P(A/B)$$

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

↓
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

x_1 x_2 x_3

— — —

Dependent features.

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 | the equation can be represented as, |
|-----------------|-----------|---|
| x_1, x_2, x_3 | y | $P(y x_1, x_2, x_3) =$ |
| - - - | yes | $P(yes) * P(x_1 yes) * P(x_2 yes)$ |
| - - - | no | $* P(x_3 yes)$ |
| - - - | yes | $\frac{P(x_1) * P(x_2) * P(x_3)}{P(x_1) * P(x_2) * P(x_3)}$ |
| - - - | no | $P(y no x_1, x_2, x_3) =$ |
| | | $P(no) * P(x_1 no) * P(x_2 no) * P(x_3 no)$ |

$$\left| \frac{P(x_3 | \text{No})}{P(x_1) + P(x_2) + P(x_3)} \right|$$

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 |

now,

Temperature

| | Yes | No | $P(E/Yes)$ | $P(E/No)$ |
|------|-----|----|------------|-----------|
| Hot | 2 | 2 | 2/4 | 2/5 |
| Mild | 4 | 2 | 4/6 | 2/5 |
| Cold | 3 | 1 | 3/4 | 1/5 |

$$P(\text{Play} | \text{Yes/No}) \quad P(\text{Yes}) \quad P(\text{No})$$

$$\begin{array}{lll} \text{Yes} & 9 & 9/14 \\ \text{No} & 5 & \end{array} \quad \frac{5}{14}$$

Test (sunny, hot) \rightarrow output ?

$$P(\text{Yes} | \text{sunny, hot}) = P(\text{Yes}) \times P(\text{sunny}) \times P(\text{hot})$$

$$= \frac{9}{14} \times \frac{2}{4} \times \frac{2}{5}$$

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

$$= 0.031$$

$$P(\text{No} | \text{sunny, hot}) = P(\text{No}) \times P(\text{sunny/No}) \times P(\text{hot/No})$$

$$= \frac{5}{14} \times \frac{3}{5} \times \frac{3}{5}$$

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

$$= 0.085$$

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

$$= 0.27$$

$$= 27\%$$

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

$$= 0.073$$

$$= 73\%$$

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.

||,

... i.e. given test

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