# Bayes' Theorem

- Bayes' Theorem is a way of finding a probability when we know certain other probabilities.
- The formula is:

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

how often A happens given that B happens, written P(A|B), how often B happens given that A happens, written P(B|A) and how likely A is on its own, written P(A) and how likely B is on its own, written P(B)

- Let us say P(Fire) means how often there is fire, and P(Smoke) means how often we see smoke, then:
- P(Fire | Smoke) means how often there is fire when we can see smoke P(Smoke | Fire) means how often we can see smoke when there is fire
- "Forwards" P(Fire | Smoke) when we know "Backwards" P(Smoke | Fire)

### **Example:**

- dangerous fires are rare (1%)
- but smoke is fairly common (10%) due to barbecues,
- and 90% of dangerous fires make smoke

#### Solution

We can then discover the probability of dangerous Fire when there is Smoke:

```
P(Fire|Smoke) =P(Fire) P(Smoke|Fire)/P(Smoke)
=1% x 90%/10%
=9%
So it is still worth checking out any smoke to be sure.
```

## Remembering

• First think "AB AB AB" then remember to group it like: "AB = A BA / B"

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

- The Bayes Rule provides the formula for the probability of Y given X. But, in real-world problems, you typically have multiple X variables.
- When the features are independent, we can extend the Bayes Rule to what is called Naive Bayes.
- It is called 'Naive' because of the naive assumption that the X's are independent of each other. Regardless of its name, it's a powerful formula.

When there are multiple X variables, we simplify it by assuming the X's are independent, so the **Bayes** rule

```
P (Y=k | X) = P(X | Y=k) * P (Y=k)
P (X)
```

where, k is a class of Y

## becomes, Naive Bayes

```
P (Y=k | X1..Xn) = P (X1 | Y=k) * P (X2 | Y=k) ... * P (Xn | Y=k) * P (Y=k)

P (X1 | Y=k) * P (X2 | Y=k) ... * P (Xn | Y=k) * P (Y=k)

P (X1) * P (X2) ... * P (Xn)
```