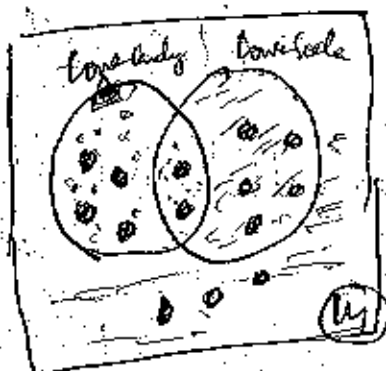


# Conditional Probability

(1)

		Candy		
		loves Candy	Doesn't love Candy	
Soda	loves Soda	2 $\frac{2}{14}$	5 $\frac{5}{14}$	$2+5 = \frac{7}{14}$
	doesn't love Soda	4 $\frac{4}{14}$	3 $\frac{3}{14}$	$4+3 = \frac{7}{14}$
		$2+4 = \frac{6}{14}$	$5+3 = \frac{8}{14}$	



Contingency Table

$$P(\text{LC and S}) = \frac{2}{14} = 0.14$$

$$P(\text{LC and NS}) = \frac{4}{14} = 0.29$$

(LC only)

$$P(\text{NC and LS}) = \frac{5}{14} = 0.36$$

only LS

$$P(\text{NC and LS}) = \frac{3}{14} = 0.21$$

We can also determine the probability that someone loves Soda, regardless of how they feel about Candy.

$$2+5 = \frac{7}{14}$$

~~that~~ Knowing that the next person we meet will love Soda. What is the probability that the next person also loves Candy?

In other words, what is the probability that someone loves Candy and Soda given that we know they love Soda?

$$P(\text{LC and LS} | \text{loves soda}) =$$

This is called the conditional probability.

$$P(\text{LC and LS}) = \frac{2}{14} = 0.14 \quad \leftarrow \text{different}$$

$$P(\text{LC and LS} | \text{LS}) = \frac{2}{2+5} = 0.29$$

In this case, knowing that they loved soda increased the probability that they would love candy.

Now, we have probability that someone does not love candy and loves soda.

$$P(\text{NC and LS} | \text{LS}) = \frac{5}{2+5} = \frac{5}{7} = 0.71$$

$$= \frac{\frac{5}{14}}{\frac{2+5}{14}} = 0.71$$

We got the same probability that we got before.

No need divide with the total population.

$$\frac{\left(\frac{5}{14}\right) P(\text{NC and LS})}{\frac{2+5}{14} P(\text{LS})}$$

$$P(\text{NC and LS} | \text{LS}) = \frac{P(\text{NC and LS})}{P(\text{LS})} \quad (3)$$

Conditional Probability is the probability that something will happen,  $\propto$  scaled by whatever knowledge we already have about the event.

Bayes' Theorem

$$P(\text{NC} \& \text{LS} | \text{LS}) = \frac{P(\text{NC} \& \text{LS})}{P(\text{LS})} = 0.71$$

Standard notation

Redundant

$$P(\text{NC} | \text{LS}) = \frac{P(\text{NC} \& \text{LS})}{P(\text{LS})}$$

$$P(\text{NC} \& \text{LS} | \text{LS}) = \frac{P(\text{NC} \& \text{LS})}{P(\text{LS})} = 0.71$$

$$P(\text{NC} \& \text{LS} | \text{NC}) = \frac{P(\text{NC} \& \text{LS})}{P(\text{NC})} = 0.63$$

$$P(\text{NC} \& \text{LS} | \text{LS}) \times P(\text{LS}) = \frac{P(\text{NC} \& \text{LS}) \times P(\text{LS})}{P(\text{LS})}$$

$$P(\text{NC} \& \text{LS} | \text{NC}) \times P(\text{NC}) = \frac{P(\text{NC} \& \text{LS}) \times P(\text{NC})}{P(\text{NC})}$$

$$P(NC \& LS | LS) \times P(LS)$$

$$= P(NC \& LS | NC) \times P(NC)$$

$$P(NC \& LS | LS) = \frac{P(NC \& LS | NC) \times P(NC)}{P(LS)}$$

$$P(NC \& LS | NC) = \frac{P(NC \& LS | LS) \times P(LS)}{P(NC)}$$

~~Derive~~ derived Bayes' Theorem.

A = doesn't have candy (NC)

B = looks sick (LS)

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

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