

CS6462

*Probabilistic and Explainable AI*

# Lesson 1

## *Probability of Events*

# Probability



## *Definition*

- mathematics concerned with numerical descriptions of how likely an event is to occur
- degree of belief of an experiment  $X$  (or phenomena) where all possible outcomes are known (denoted as sample space  $S$ )
- any subset of  $S$  is called an event  $E$
- event  $E$  occurs if the outcome of experiment  $X$  is contained in  $E$

## *Example*

$X$  = tossing a die

$S = \{1, 2, 3, 4, 5, 6\}$

$E = \{2, 4, 6\}$  - an event "the number is even"

# Probability (cont.)



## *Example 2*

$X$  = tossing a coin twice

$S = \{hh, ht, th, tt\}$

$E = \{hh, ht\}$  event “the first toss results in a Head”

## *Example 3*

$X$  = tossing a die twice

$S = \{f(i, j) : i, j = 1, 2, \dots, 6\}$  contains 36 elements

$E = \{f(i, j) : f + j = 10\}$  event “the sum of the results of the two tosses is equal to 10”

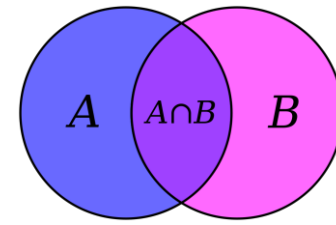
## *Example 4*

$X$  = choosing a point from the interval  $[0; 1]$

$S \subseteq \mathbb{Q}$  rational numbers,  $s \in S$  and  $0 \leq s \leq 1$

$E = \{1/3\}$

# Events and Set Theory

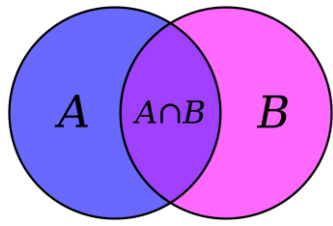


**Set Theory** provides the *notation* to describe and manipulate events:

- $E \subseteq S, F \subseteq S$  events  $E$  and  $F$  are *subsets* of the sample space  $S$
- $E^c$  complement of  $E$  - the set of all outcomes not in  $E$
- $E \cap F$  intersection of  $E$  and  $F$  - the set of all outcomes in both  $E$  and  $F$
- $E \cup F$  union of  $E$  and  $F$  - the set of all outcomes in  $E$  or in  $F$  or in both  $E$  and  $F$
- $E \subseteq F$  event  $E$  is subset of  $F$
- $E \cap F = \emptyset$  events  $E$  and  $F$  are *mutually exclusive* (disjoint )
- union and intersection of more than two events:

$$\bigcup_{i=1}^n E_i, \bigcup_{i=1}^{\infty} E_i, \bigcap_{i=1}^n E_i, \bigcap_{i=1}^{\infty} E_i$$

# Events and Set Theory (cont.)



## Commutativity:

- $E \cup F = F \cup E, E \cap F = F \cap E$

## Associativity:

- $(E \cup F) \cup G = E \cup (F \cup G), (E \cap F) \cap G = E \cap (F \cap G)$

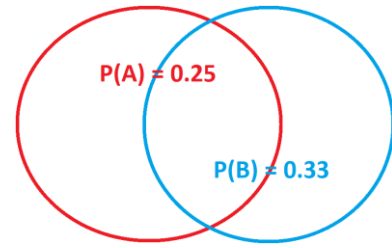
## Distributivity:

- $(E \cup F) \cap G = (E \cap G) \cup (F \cap G), (E \cap F) \cup G = (E \cup G) \cap (F \cup G)$

## Morgan's Laws:

- $(E \cup F)^c = E^c \cap F^c$
- $(E \cap F)^c = E^c \cup F^c$

# Probability of Events – Properties



Properties of probability  **$P$**  with respect to sample space  **$S$** :

- *notion*: probability  **$P$**  of an event  **$E$**

$$P(E)$$

- *scaling*: measures uncertainty on a scale from 0 to 1, with 1 representing certainty

$$P(S) = 1, 0 \leq P(E) \forall E \in S$$

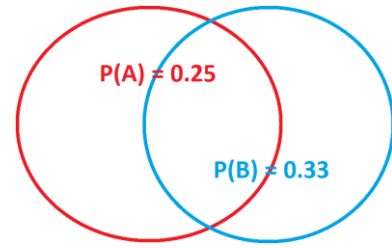
- *additivity*: imposes order on the assignment of probabilities

$$P(E \cup F) = P(E) + P(F), \text{ iff } E \cap F = \emptyset$$

for any sequence of events  $E_1, E_2, \dots, E_n$  which are mutually exclusive

$$P(\cup_{i=1}^n E_i) = \sum_{i=1}^n P(E_i)$$

# Probability of Events – Properties (cont.)



- *complementarity*:

$$P(E) + P(E^c) = 1 \quad \forall E \in S$$

- *general additivity*: additivity for non-mutually exclusive events

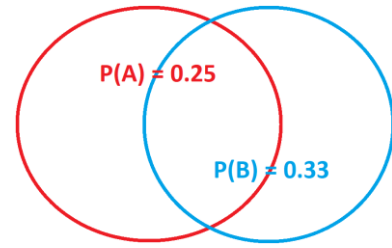
$$P(E \cup F) = P(E) + P(F) - P(E \cap F) \quad \forall E, F \in S$$

- *log odds*: measures the *odds of success* - probability of success/probability of failure

$$\text{LogOdds}(E) = \ln(P(E)/P(E^c))$$

$$\text{LogOdds}(E) = \ln(P(E)/(1 - P(E)))$$

# Probability of Events – Examples



*Example 1:  $X$  = Tossing a fair coin*

$$S = \{h, t\}$$

$$P(S) = 1, P(h) = P(t) = \frac{1}{2}$$

*Example 2:  $X$  = Tossing a fair die*

$$S = \{1, 2, 3, 4, 5, 6\}$$

$$P(S) = 1, P(1) = P(2) = P(3) = P(4) = P(5) = P(6) = \frac{1}{6}$$

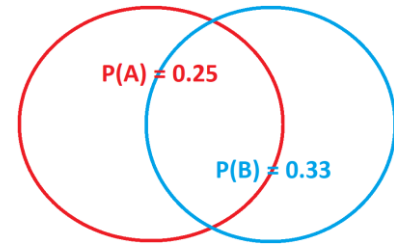
*Example 3:  $X$  = Tossing a fair coin twice.*

$$S = \{hh, ht, th, tt\}$$

$$P(S) = 1, P(hh) = P(ht) = P(th) = P(tt) = \frac{1}{4}$$



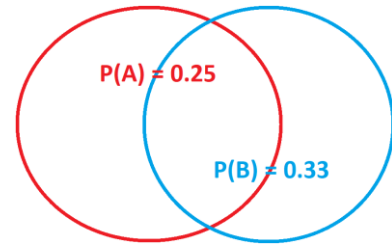
# Probability of Events – Examples (cont.)



*Example 4:  $X$  = Tossing a die -  $P(E \cup F)$  ?*

- $S = \{1, 2, 3, 4, 5, 6\}$
- $P(S) = 1, P(1) = P(2) = P(3) = P(4) = P(5) = P(6) = 1/6$
- $E = \{1, 3, 6\}, F = \{3, 4, 6\}$
- $E = \{1\} \cup \{3\} \cup \{6\}, F = \{3\} \cup \{4\} \cup \{6\}$
- $E \cap F = \{3, 6\} = \{3\} \cup \{6\}$
- recall *additivity rule*:  $P(\cup_{i=1}^n E_i) = \sum_{i=1}^n P(E_i)$   
 $P(E) = P(1) + P(3) + P(6) = 3/6,$   
 $P(F) = P(3) + P(4) + P(6) = 3/6$   
 $P(E \cap F) = P(3) + P(6) = 2/6$
- recall *general additivity rule*:  $P(E \cup F) = P(E) + P(F) - P(E \cap F)$   
 $P(E \cup F) = 3/6 + 3/6 - 2/6 = 4/6$

# Probability of Events – Examples (cont.)



*Example 5:  $X$  = 80% chance of rain*

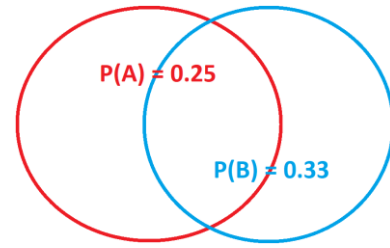
- $S = [0,1]$
- $E$  – chance of rain
- $P(E) = 0.8$  odds of success 80%
- $P(E^c) = 0.2$  odds of failure 20%
- recall log odds rule:  $\text{LogOdds}(E) = \ln(P(E)/P(E^c))$

$$\text{LogOdds}(E) = \ln(P(E)/P(E^c)) = \ln(0.8/0.2) = 1.38629436$$

the odds ratio (the probability of success/probability of failure) = 4

the log odds = 1.38629436

# Summary



## *Probability of Events*

- *Probability* – mathematics concerned with numerical descriptions of how likely an event is to occur
- *Events* – outcomes ***E*** of experiment ***X*** with known set of possible outcomes ***S***  
event ***E*** occurs if the outcome of experiment ***X*** is contained in ***E***
- *Set Theory* provides the *notation* to describe and manipulate events:
  - notion ***P(E)***
  - properties (rules) – scaling, additivity, complementarity, general additivity
  - log odds – another way to express probability
- *Next Lesson* – Conditional Probability

# Thank You!

Questions?