

Probability Theory

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Agenda

Introduction

Primary goals

Areas of growth

Timeline

Summary



Definitions

Sample Space : $\{1,2,3,4,5,6\}$

Countable sample space : $\{1,2,3,4,5,6\}$
uncountable : $0 < S < 5$

Sample Space

$A = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots} \}, B = \{ \text{die with 1 dot}, \text{die with 3 dots}, \text{die with 4 dots}, \text{die with 5 dots} \}$

Unions : $A \cup B = \{1,3,4,5,6\}$


Intersection (joint) : $A \cap B = \{3,5\}$

Complements : $A^c = \{2,4,6\}$





Definitions


$$A = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots} \}, B = \{ \text{die with 2 dots}, \text{die with 3 dots}, \text{die with 4 dots}, \text{die with 5 dots} \}$$

Event : *set of outcomes of the experiment (subset of sample space).*

Marginal probability: *the probability of a single event occurring, independent of other events.*

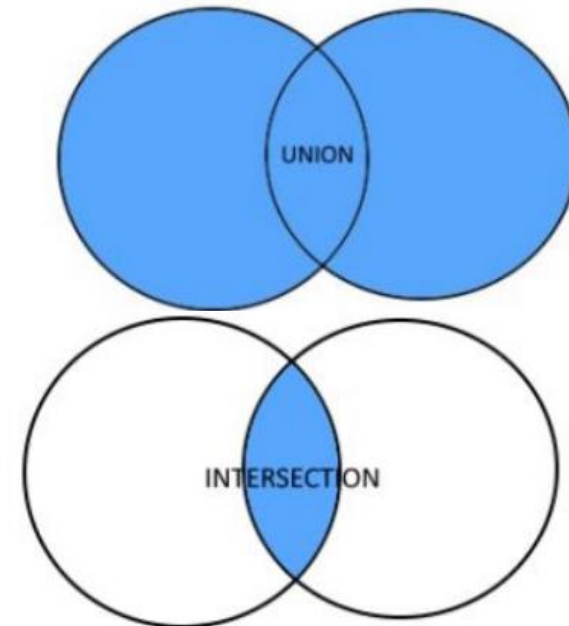
$$P(A) = \frac{3}{6} = 0.5$$

Union probability: *the probability of two events.*

$$P(A \cup B) = \frac{5}{6}$$


Intersection (joint) probability: *the probability of two events.*

$$P(A \cap B) = \frac{2}{6}$$





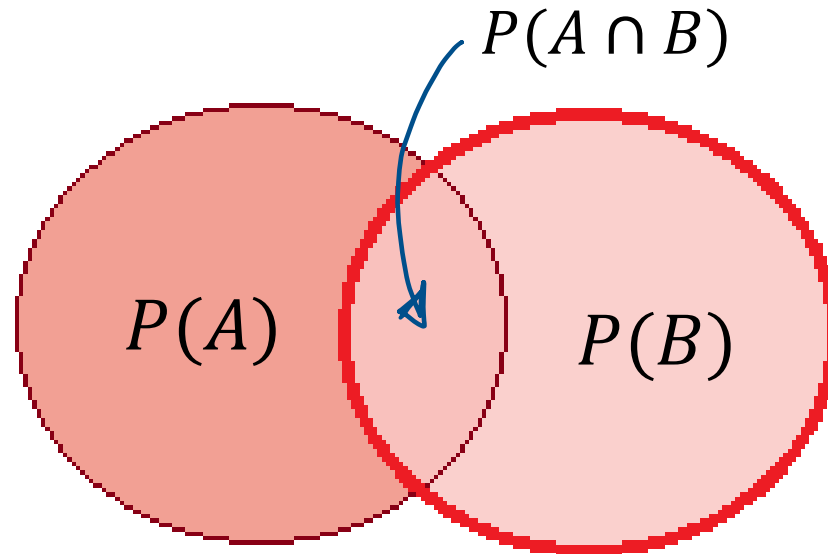
Definitions


$$A = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots} \}, B = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots}, \text{die with 4 dots} \}$$

Conditional probability:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{2}{6} \times \frac{6}{4} = 0.5$$

Given event





Definitions



Example : calculate the probability of tossing two fair dice were the first one get (2) and the sum of two dice less than 5

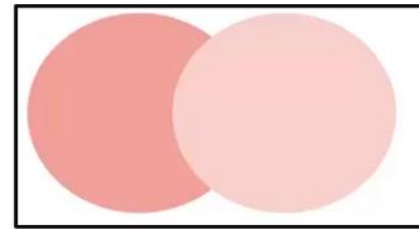
Event 1 = $P(D_2 = 2)$

Event 2 = $P(D_1 + D_2 \leq 5)$

$$P(D_2 = 2 | D_1 + D_2 \leq 5)$$

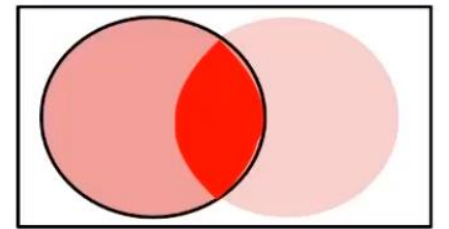
$$P(E_1 | E_2) = \frac{P(E_1 \cap E_2)}{P(E_1)}$$

$$P(D_1 + D_2 \leq 5) =$$



$P(E_1)$

$P(E_2)$



$P(E_1 \cap E_2)$

D1 →

D2 ↓

	1	2	3	4	5	6	
	1	2	3	4	5	6	7
	2	3	4	5	6	7	8
	3	4	5	6	7	8	9
	4	5	6	7	8	9	10
	5	6	7	8	9	10	11
	6	7	8	9	10	11	12





Definitions

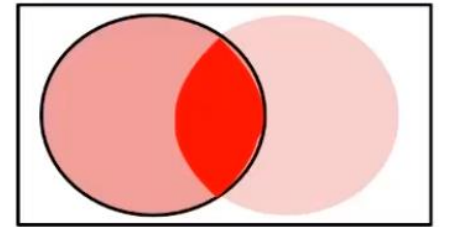
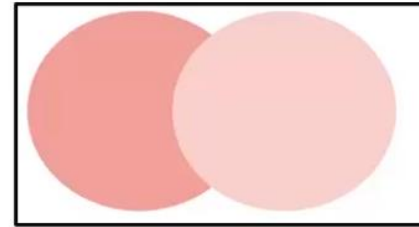


Example : calculate the probability of tossing two fair dice were the first one get (2) and the sum of two dice less than 5

$$P(D_2 = 2 | D_1 + D_2 \leq 5)$$

$$\text{Event 1} = P(D_2 = 2)$$

$$\text{Event 2} = P(D_1 + D_2 \leq 5)$$



$P(E_1)$

$P(E_2)$

$P(E_1 \cap E_2)$

$$P(E_1 | E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)}$$

$$P(E_2 = D_1 + D_2 \leq 5) = \frac{10}{36}$$

$$P(E_1 \cap E_2) = \frac{3}{36}$$

	1	2	3	4	5	6		1	2	3	4	5	6		
	1	2	3	4	5	6	7		1	2	3	4	5	6	7
	2	3	4	5	6	7	8		2	3	4	5	6	7	8
	3	4	5	6	7	8	9		3	4	5	6	7	8	9
	4	5	6	7	8	9	10		4	5	6	7	8	9	10
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Definitions

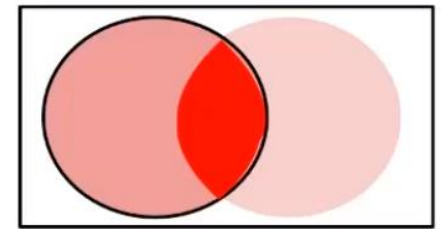
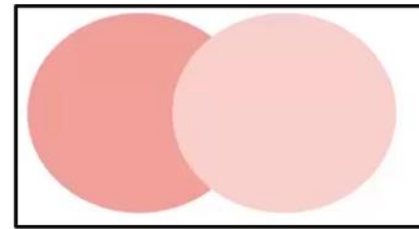


Example : calculate the probability of tossing two fair dice were the first one get (2) and the sum of two dice less than 5

$$P(D_2 = 2 | D_1 + D_2 \leq 5)$$

$$\text{Event 1} = P(D_2 = 2)$$

$$\text{Event 2} = P(D_1 + D_2 \leq 5)$$



$P(E_1)$

$P(E_2)$

$P(E_1 \cap E_2)$

$$P(E_1 | E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)}$$

$$P(E_2 = D_1 + D_2 \leq 5) = \frac{10}{36}$$

$$P(E_1 \cap E_2) = \frac{3}{36}$$

$$P(E_1 | E_2) = \frac{3/36}{10/36} = 0.3$$

		1	2	3	4	5	6			1	2	3	4	5	6
	1	2	3	4	5	6	7		1	2	3	4	5	6	7
	2	3	4	5	6	7	8		2	3	4	5	6	7	8
	3	4	5	6	7	8	9		3	4	5	6	7	8	9
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$$A = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots} \}, B = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots}, \text{die with 4 dots} \}$$

Conditional probability:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{2}{6} \times \frac{6}{4} = 0.5$$

Given event

Multiplication Rule:

$$P(A \cap B) = P(B) P(A|B)$$



Intersection
(joint)

Conditional

select someone from the entire group who has two characteristics.

pull out a subgroup that has one of the characteristics already, and you want the probability that someone from that subgroup has a second characteristic.

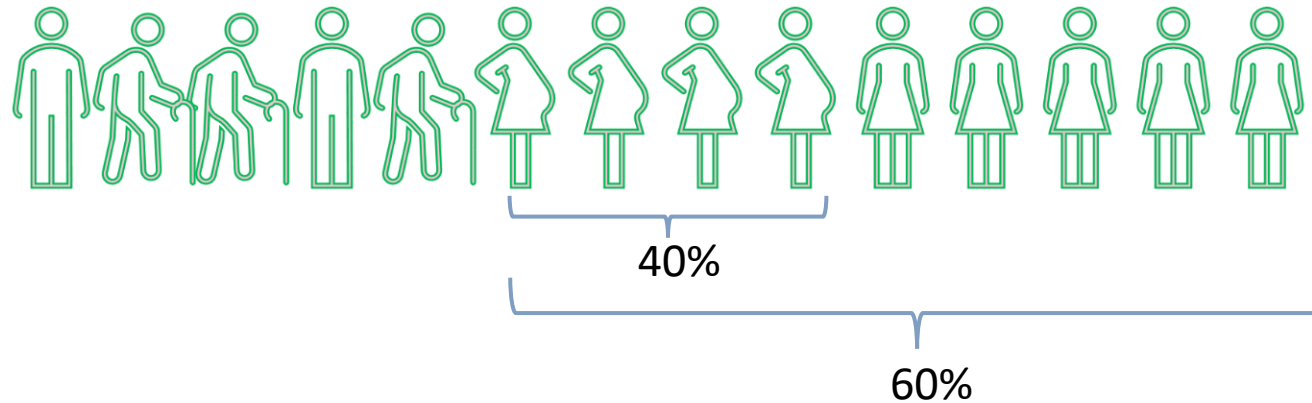




Exercise



Suppose there is an event 60% of attendance are woman (W) and 40% of them are pregnant (Pr). What the chance of a person you select from this event is a woman **AND** pregnant




$$\begin{aligned} P(W \cap Pr) &= P(W) P(W|Pr) \\ &= 0.6 \times 0.4 = 0.24 \end{aligned}$$





Definitions


$$A = \{ \text{die with 1 dot}, \text{die with 2 dots}, \text{die with 3 dots} \}, B = \{ \text{die with 2 dots}, \text{die with 3 dots}, \text{die with 4 dots}, \text{die with 5 dots} \}$$

Additional Rule:

$$P(A \cup B) = P(B) + P(A) - P(A \cap B)$$



union

Intersection
(joint)

select someone from the entire group who has two characteristics.

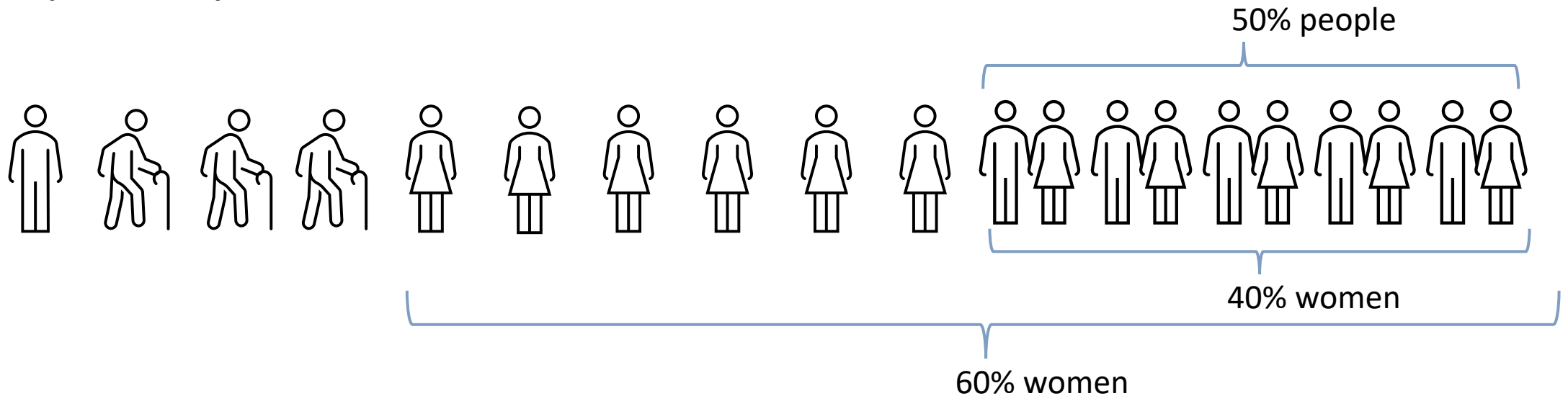




Exercise



Suppose there is an event 60% of attendance are woman (W) and 50% of class are married (M). What the chance of a person you select from this event is a woman **OR** married



$$\begin{aligned} P(W \cap MW) &= P(W) \times P(MW) \\ &= 0.6 \times 0.4 = 0.24 \end{aligned}$$

$$\begin{aligned} P(W \cup M) &= P(W) + P(MP) - P(W \cap MW) \\ &= 0.6 + 0.5 - 0.24 = 0.86 \end{aligned}$$





Bayes Theorem



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

$$P(A \cap B) = P(B) P(A|B) = P(B \cap A)$$

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





Bayes Theorem



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

Likelihood (arrow pointing to $P(A|B)$)
prior (arrow pointing to $P(B)$)
evidence (arrow pointing to $P(A)$)

$$P(C_1 | X_1 \& X_2 \& X_3 \& X_4) = \frac{P(X_1 | C_1) * P(X_2 | C_1) * P(X_3 | C_1) * P(X_4 | C_1) * P(C_1)}{P(X_1) * P(X_2) * P(X_3) * P(X_4)}$$





Gaussian Naïve Bayes Classifier:



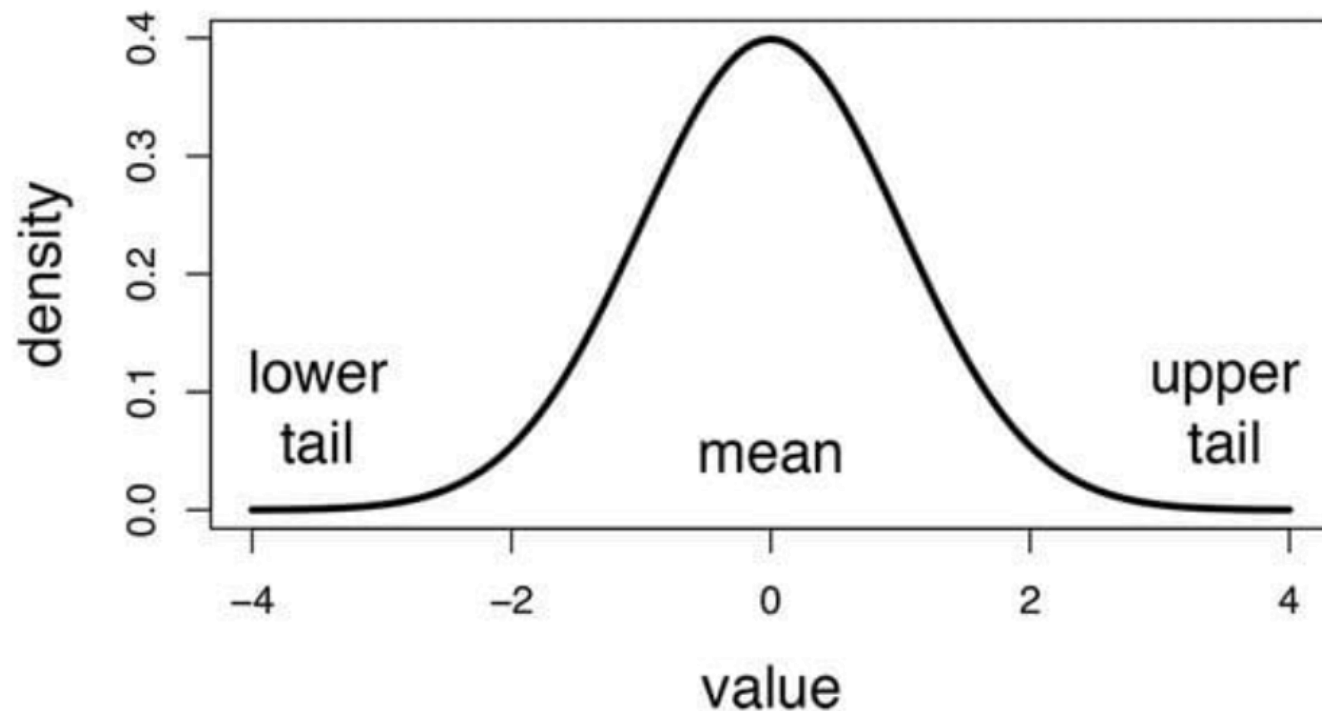
continuous values associated with each feature are assumed to be distributed according to a **Gaussian distribution**

Gaussian distribution

$$P(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\mu = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\sigma = \left[\frac{1}{n-1} \sum_{i=1}^n (x_i - \mu)^2 \right]^{0.5}$$





Multinomial Naïve Bayes Classifier



Feature vectors represent the **frequencies** with which certain events have been generated by a **multinomial distribution**.

This is the event model typically used for document classification.

Bernoulli Naïve Bayes Classifier:

In the multivariate Bernoulli event model, features are independent booleans (binary variables) describing inputs.

binary term occurrence (i.e. a **word occurs** in a document or not)





Lab time

Annual revenue growth



Thank you

