

Probabilistic Machine Learning:

1. Probabilistic refresher

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1/19



HR EXCELLENCE IN RESEARCH



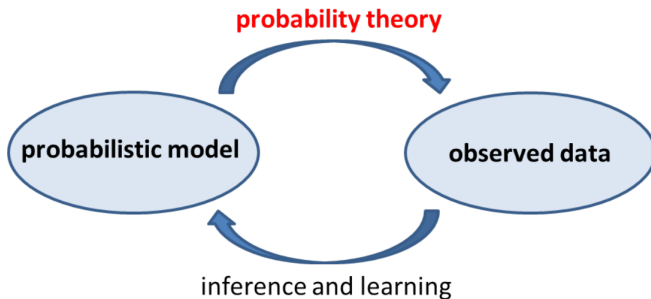
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The presentation has been inspired and in some parts totally based on Prof. Mario A. T. Figueiredo presentation at LxMLS'2017, Instituto Superior Tecnico & Instituto de Telecomunicacoes, Lisboa, Portugal.

Appropriate agreements to propagate his ideas has been acquired.



Probability theory



- ▶ has its origins in gambling
- ▶ great names: Fermat, Pascal, Bernoulli, Huygens, Laplace, Kolmogorov, Poisson, Cauchy, Boltzman, Bayes, Cardano, ...
- ▶ tool to handle uncertainty, information, knowledge, observations, ...
- ▶ ...thus also learning, decision making, inference, science, data science ...

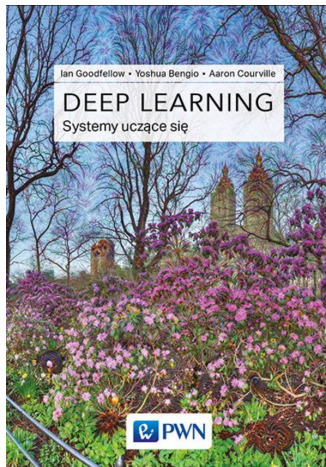
Do we still need to know probability theory?

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What book is this from?



Do we still need to know probability theory?



What is probability?

Example

$\mathbb{P}(\text{randomly drawn card is } \heartsuit) = 13/52$

$\mathbb{P}(\text{getting 1 in throwing a fair die}) = 1/6$

- ▶ **Classical** definition: $\mathbb{P}(A) = \frac{N_A}{N}$
...with N mutually exclusive equally likely outcomes, N_A of which result in the occurrence of A .
- ▶ **Frequentist** definition: $\mathbb{P}(A) = \lim_{N \rightarrow \infty} \frac{N_A}{N}$
...relative frequency of occurrence of A in infinite number of trials
- ▶ **Subjective probability**:
...gives meaning to $\mathbb{P}(\text{"it will rain today"})$, or $\mathbb{P}(\text{"I'll have passed the PUMa's exam next winter"})$

Key concepts: Sample space and events

- ▶ **Sample space** \mathcal{X} = set of possible outcomes of a random experiment.

Example

- ▶ Tossing two coins: $\mathcal{X} = \{HH, TH, HT, TT\}$
 - ▶ Roulette: $\mathcal{X} = \{1, 2, \dots, 36\}$
 - ▶ Draw a card from a shuffled deck $\mathcal{X} = \{A\heartsuit, 2\heartsuit, \dots, Q\spadesuit, K\spadesuit\}$
- ▶ An **event** A is a subset of $\mathcal{X} : A \subseteq \mathcal{X}$ (also written $A \in 2^{\mathcal{X}}$)

Example

- ▶ exactly one H in 2-coin toss: $A = \{TH, HT\}$
- ▶ odd number in the roulette: $A = \{1, 3, \dots, 35\}$
- ▶ drawn a \heartsuit card: $A = \{A\heartsuit, 2\heartsuit, \dots, K\heartsuit\}$

Key concepts: Sample space and events

- ▶ **Sample space** \mathcal{X} = set of possible outcomes of a random experiment.

Example

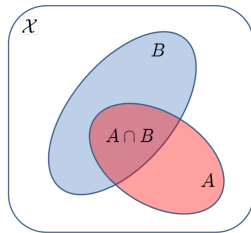
- ▶ Distance travelled by tossed die: $\mathcal{X} = \mathbb{R}_+$
- ▶ Location of the next rain drop on a given square tile: $\mathcal{X} = \mathbb{R}^2$
- ▶ Properly handling the continuous case requires deeper concepts:
 - ▶ Sigma algebras
 - ▶ Measurable functions
 - ▶ ... and other heavier stuff, not covered here

Kolmogorov's Axioms for Probability

- ▶ Probability is a function that maps events A into the interval $[0, 1]$.
Kolmogorov's axioms (1933) for probability
 - ▶ For any A , $\mathbb{P}(A) \geq 0$
 - ▶ $\mathbb{P}(\mathcal{X}) = 1$
 - ▶ If $A_1, A_2, \dots \subseteq \mathcal{X}$ are disjoint events, then $\mathbb{P}(\bigcup_i A_i) = \sum \mathbb{P}(A_i)$
- ▶ From these axioms, many results can be derived

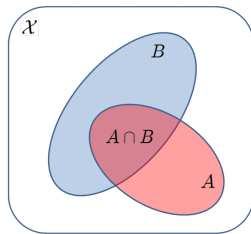
Example

- ▶ $\mathbb{P}(\emptyset) = 0$
- ▶ $C \subset D \implies \mathbb{P}(C) \leq \mathbb{P}(D)$
- ▶ $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B)$
- ▶ $\mathbb{P}(A \cup B) \leq \mathbb{P}(A) + \mathbb{P}(B)$



Conditional Probability and Independence

- ▶ If $\mathbb{P}(B) > 0$, $\mathbb{P}(A|B) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}$
(conditional probability of A, given B)
- ▶ ...satisfies all of Kolmogorov's axioms:
 - ▶ For any $A \subseteq \mathcal{X}$, $\mathbb{P}(A|B) \geq 0$
 - ▶ $\mathbb{P}(\mathcal{X}|B) = 1$
 - ▶ If $A_1, A_2, \dots \subseteq \mathcal{X}$ are disjoint,
 $\mathbb{P}(\bigcup_i A_i | B) = \sum_i \mathbb{P}(A_i | B)$
- ▶ **Independence:** A, B are independent ($A \perp B$):
 $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$



Conditional Probability and Independence

- ▶ If $\mathbb{P}(B) > 0$, $\mathbb{P}(A|B) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}$
- ▶ Events A, B are independent ($A \perp B$) $\iff \mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$
- ▶ Relationship with conditional probabilities:
 $(A \perp B) \iff \mathbb{P}(A|B) = \mathbb{P}(A)$

Example

\mathcal{X} = "52 cards", $A = \{4\heartsuit, 4\diamondsuit, 4\clubsuit, 4\spadesuit\}$, and $B = \{A\heartsuit, 2\heartsuit, \dots, K\heartsuit\}$;
then

$$\mathbb{P}(A) = 1/13, \mathbb{P}(B) = 1/4$$

$$\mathbb{P}(A \cap B) = \mathbb{P}(4\heartsuit) = \frac{1}{52}$$

$$\mathbb{P}(A)\mathbb{P}(B) = \frac{1}{13} \frac{1}{4} = \frac{1}{52}$$

$$\mathbb{P}(A|B) = \mathbb{P}(\text{"4" | "\heartsuit"}) = \frac{1}{13} = \mathbb{P}(A)$$

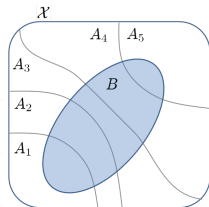
Bayes Theorem

- Law of total probability: if A_1, \dots, A_n are a partition of \mathcal{X}

$$\mathbb{P}(B) = \sum_i \mathbb{P}(B|A_i)\mathbb{P}(A_i) = \sum_i \mathbb{P}(B \cap A_i)$$

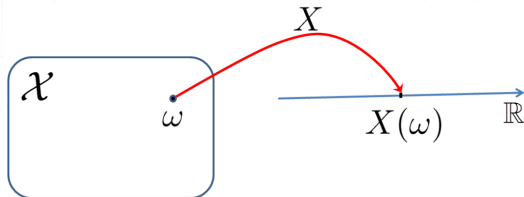
- Bayes' theorem: if $\{A_1, \dots, A_n\}$ is a partition of \mathcal{X}

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



Random Variables

- ▶ A (real) **random variable** (RV) is a function: $X : \mathcal{X} \rightarrow \mathbb{R}$



- ▶ **Discrete RV**: range of X is countable (e.g., \mathbb{N} or $\{0, 1\}$)
- ▶ **Continuous RV**: range of X is uncountable (e.g., \mathbb{R} or $[0, 1]$)

Example

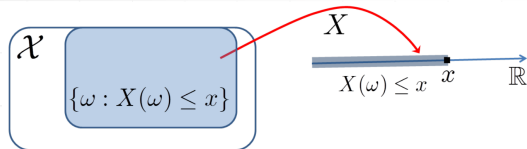
number of heads in tossing two coins, $\mathcal{X} = \{HH, HT, TH, TT\}$,
 $X(HH) = 2, X(HT) = X(TH) = 1, X(TT) = 0$, range of $X = \{0, 1, 2\}$

Example

distance traveled by a tossed coin; range of $X = \mathbb{R}_+$

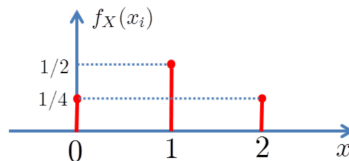
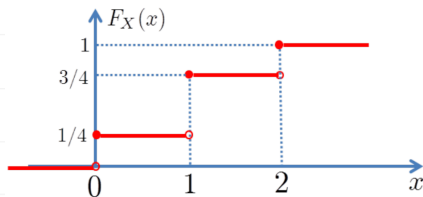
Random Variables: Distribution Function

- **Distribution function:** $F_X(x) = \mathbb{P}(\{\omega \in \mathcal{X} : X(\omega) \leq x\})$



Examples

number of heads in tossing 2 coins; $\text{range}(X) = \{0, 1, 2\}$



- **Probability mass function** (discrete RV): $f_X(x) = \mathbb{P}(X = x)$, $F_X(x) = \sum_{x_i \leq x} f_X(x_i)$

Important Discrete Random Variables

- **Uniform:** $X \in \{x_1, \dots, x_K\}$, pmf $f_X(x_i) = 1/K$

Examples

a fair roulette $X \in \{1, \dots, 36\}$, with $f_X(x) = 1/36$

a fair die $X \in \{1, \dots, 6\}$, with $f_X(x) = 1/6$

- **Bernoulli RV:** $X \in \{0, 1\}$, pmf $f_X(x) = \begin{cases} p & \text{if } x = 1 \\ 1 - p & \text{if } x = 0 \end{cases}$

Compact form: $f_X(x) = p^x(1 - p)^{1-x}$

Examples

an unfair coin (heads = 0, tails = 1), with $p \neq 1/2$.

Important Discrete Random Variables

- **Binomial RV:** $X \in \{0, 1, \dots, n\}$ (sum of n Bernoulli RVs)

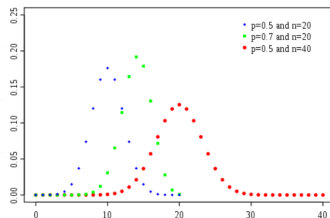
$$f_X(x) = \text{Binomial}(x; n, p) = \binom{n}{x} p^x (1-p)^{n-x}$$

Binomial coefficients ("n choose x"):

$$\binom{n}{x} = \frac{n!}{(n-x)!x!}$$

Example

number of heads in n coin tosses.



Other Important Discrete Random Variables

- **Geometric(p)**: $X \in \mathbb{N}$, pmf
 $f_X(x) = p(1 - p)^{x-1}$

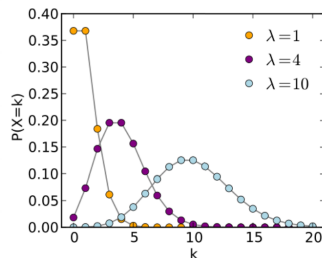
Example

number of coin tosses until first heads

- **Poisson(λ)**:
 $X \in \mathbb{N} \cup \{0\}$
pmf $f_X(x) = \frac{e^{-\lambda} \lambda^x}{x!}$

Example

“...probability of the number of independent occurrences in a fixed (time/space) interval, if these occurrences have known average rate”

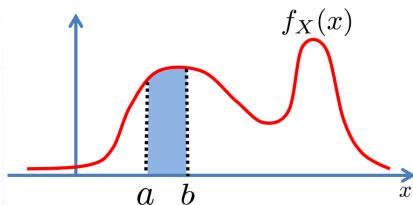


Continuous Random Variables

- **Probability density function (pdf, continuous RV):** $f_X(x)$

$$\int_{-\infty}^{\infty} f_X(x) = 1$$

$$\mathbb{P}(X \in [a, b]) = \int_a^b f_X(x) dx$$

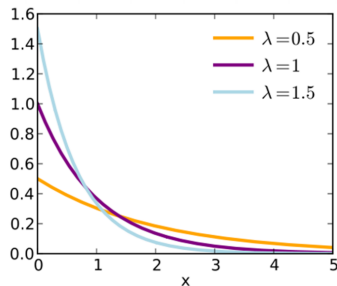
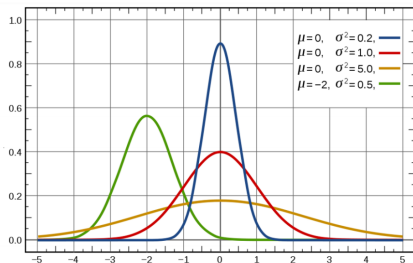


- Notice: $\mathbb{P}(X = c) = 0$

Important Continuous Random Variables

► **Uniform:** $f_X(x) = \text{Uniform}(x; a, b) = \begin{cases} \frac{1}{b-a} & \text{if } x \in [a, b] \\ 0 & \text{if } x \notin [a, b] \end{cases}$

► **Gaussian:** $f_X(x) = \mathcal{N}(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$



► **Exponential:** $f_X(x) = \text{Exp}(x; \lambda) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$