



# Introduction to theory of probability

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# Outline

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- Important concepts in probability theory
- Bayes' rule
- Random variables and distributions



# Definition of Probability

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- **Experiment:** toss a coin twice
- **Sample space:** possible outcomes of an experiment
  - $S = \{HH, HT, TH, TT\}$
- **Event:** a subset of possible outcomes
  - $A = \{HH\}$ ,  $B = \{HT, TH\}$
- **Probability of an event :**
  - a number assigned to an event  $\Pr(A)$ 
    - Axiom 1:  $\Pr(A) \geq 0$
    - Axiom 2:  $\Pr(S) = 1$
    - Axiom 3: For every sequence of disjoint events

$$\Pr\left(\bigcup_i A_i\right) = \sum_i \Pr(A_i)$$

- Example:  $\Pr(A) = n(A)/N$ : frequentist statistics



# Conditional Probability

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- For events A and B, **conditional probability**  $\Pr(A|B)$  stands for the probability that A happens if B occurs.
  - Prob. of A given B.
- Example:  $A=\{HH\}$ ,  $B=\{HT, TH\}$ , what is the conditional probability  $\Pr(A|B)$ ?
  - 0
- Example:  $A=\{HH\}$ ,  $B=\{HH, TH\}$ , what is the conditional probability  $\Pr(B|A)$ ?
  - 1



# Joint Probability

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- For events A and B, **joint probability**  $\Pr(AB)$  stands for the probability that both events happen.
- Example:  $A=\{HH\}$ ,  $B=\{HT, TH\}$ , what is the joint probability  $\Pr(A,B)$ ?
  - 0
- $\Pr(AB)=\Pr(A)\Pr(B|A)$ 
  - $=\Pr(B)\Pr(A|B)$
- $A=\{HH\}$ ,  $B=\{HH, TH\}$ , the joint probability  $\Pr(A,B)$ ?
  - $=\Pr(A)\Pr(B|A)$
  - $=1/4 \times 1=1/4$



# Independence

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- Two events ***A and B are independent*** in case

$$\Pr(A|B) = \Pr(A)$$

OR

$$\Pr(B|A) = \Pr(B)$$

OR

$$\Pr(AB) = \Pr(A)\Pr(B)$$

- A set of events  $\{A_i\}$  is independent in case

$$\Pr(\bigcap_i A_i) = \prod_i \Pr(A_i)$$



# Independence

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$$\Pr(AB) = \Pr(A)\Pr(B)$$

- A set of events  $\{A_i\}$  is independent in case

$$\Pr(\bigcap_i A_i) = \prod_i \Pr(A_i)$$

- Example: Drug test

$A = \{\text{A patient is a Woman}\}$

$B = \{\text{Drug fails}\}$

Will event A be independent  
from event B ?

|         | Woman | Man  |
|---------|-------|------|
| Success | 200   | 1800 |
| Failure | 1800  | 200  |



# Independence

- Consider the experiment of tossing a coin twice
- Example I: HT or HH
  - $A = \{HT, HH\}$ ,  $B = \{HT\}$
  - Will event A independent from event B?
$$\begin{aligned} P(B|A) &= 1 \\ P(A|B) &= 0 \\ P(A) &= 0.5 \\ P(B) &= 0.25 \end{aligned}$$
- Example II:
  - $A = \{HT\}$ ,  $B = \{TH\}$
  - Will event A independent from event B?
$$\begin{aligned} P(B|A) &= 0 \\ P(A|B) &= 0 \\ P(A) &= 0.25 \\ P(B) &= 0.25 \end{aligned}$$
- Disjoint  $\neq$  Independence
- If A is independent from B, B is independent from C, will A be independent from C?

Independence mutual, but not transitive!

$$(P(AB)=P(A)P(B)) \ \& \ (P(BC)=P(B)P(C)) \ \nRightarrow \ (P(AC)=P(A)P(C))$$





# Conditioning

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- If A and B are events with  $\Pr(A) > 0$ , the ***conditional probability of B given A*** is

$$\Pr(B \mid A) = \frac{\Pr(AB)}{\Pr(A)}$$



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$A = \{\text{Patient is a Women}\}$

$B = \{\text{Drug fails}\}$

$\Pr(B|A) = ?$

$\Pr(A|B) = ?$



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A = {Patient is a Women}

B = {Drug fails}

|         | Women | Men  |
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| Success | 200   | 1800 |
| Failure | 1800  | 200  |

$\Pr(B|A) = ?$

$\Pr(A|B) = ?$

- Given A is independent from B, what is the relationship between  $\Pr(A|B)$  and  $\Pr(A)$ ?



# Bayes' Rule

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- Given two events A and B and suppose that  $\Pr(A) > 0$ . Then

$$\Pr(B \mid A) = \frac{\Pr(AB)}{\Pr(A)} = \frac{\Pr(A \mid B) \Pr(B)}{\Pr(A)}$$

- Example:

$$\Pr(R) = 0.8$$

| $\Pr(W R)$ | R   | $\neg R$ |
|------------|-----|----------|
| W          | 0.7 | 0.4      |
| $\neg W$   | 0.3 | 0.6      |

R: It is a rainy day

W: The grass is wet

$$\Pr(R|W) = ?$$

# Bayes' Rule

|          | R   | $\neg R$ |
|----------|-----|----------|
| W        | 0.7 | 0.4      |
| $\neg W$ | 0.3 | 0.6      |

R: It rains

W: The grass is wet

**Information**



**Inference**

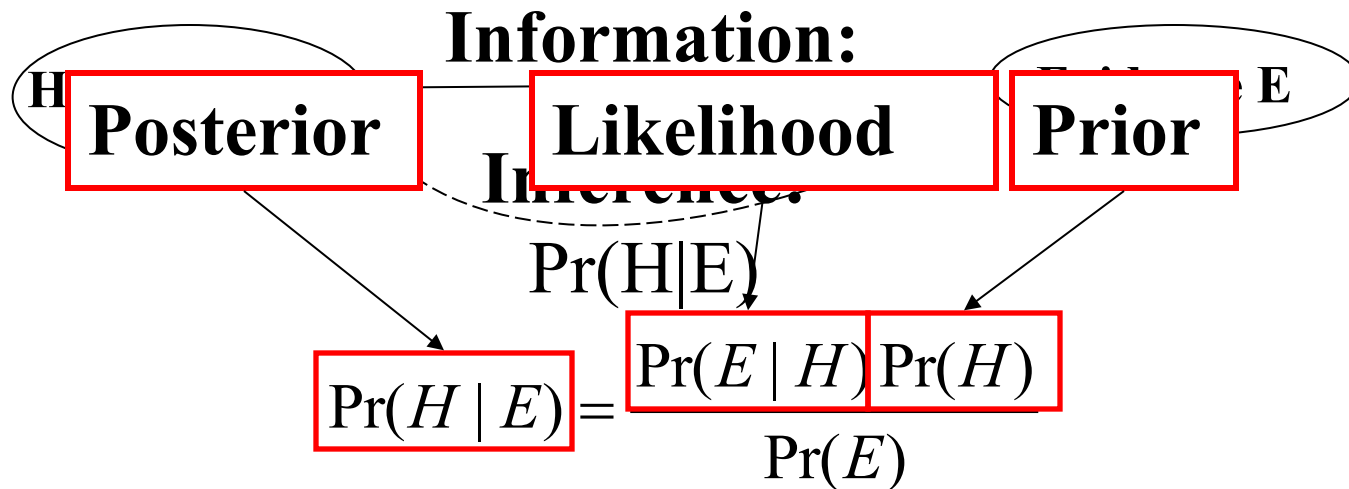
$\Pr(R|W)$

# Bayes' Rule

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# Bayes' Rule: Expanded

---

- Suppose that  $B_1, B_2, \dots, B_k$  form a partition of  $S$ :

$$B_i \cap B_j = \emptyset; \quad \bigcup_i B_i = S$$

Suppose that  $\Pr(B_i) > 0$  and  $\Pr(A) > 0$ . Then

$$\Pr(B_i | A) = \frac{\Pr(A | B_i) \Pr(B_i)}{\Pr(A)}$$



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$$\begin{aligned} \Pr(B_i | A) &= \frac{\Pr(A | B_i) \Pr(B_i)}{\Pr(A)} \\ &= \frac{\Pr(A | B_i) \Pr(B_i)}{\sum_{j=1}^k \Pr(AB_j)} \end{aligned}$$





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# Random Variable and Distribution

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- A ***random variable***  $X$  is a numerical outcome of a random experiment
- The ***distribution*** of a random variable is the collection of possible outcomes along with their probabilities:
  - Discrete case:  $\Pr(X = x) = p_{\theta}(x)$
  - Continuous case:  $\Pr(a \leq X \leq b) = \int_a^b p_{\theta}(x)dx$
  - Probability density function
  - Probability mass function



# Random Variable: Example

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- Let  $S$  be the set of all sequences of three rolls of a die. Let  $X$  be the sum of the number of dots on the three rolls.
- What are the possible values for  $X$ ?
- $\Pr(X = 5) = ?$ ,  $\Pr(X = 10) = ?$



# Expectation

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- A random variable  $X \sim \Pr(X=x)$ . Then, its expectation is

$$E[X] = \sum_x x \Pr(X = x)$$

- In an empirical sample,  $x_1, x_2, \dots, x_N$

$$E[X] = \frac{1}{N} \sum_{i=1}^N x_i$$

- Continuous case:  $E[X] = \int_{-\infty}^{\infty} x p_{\theta}(x) dx$

- Expectation of sum of random variables

$$E[X_1 + X_2] = E[X_1] + E[X_2]$$



# Expectation: Example

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- Let  $S$  be the set of all sequence of three rolls of a die. Let  $X$  be the sum of the number of dots on the three rolls.
- What is  $E(X)$ ?
- Let  $S$  be the set of all sequence of three rolls of a die. Let  $X$  be the product of the number of dots on the three rolls.
- What is  $E(X)$ ?



# Variance

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- The variance of a random variable  $X$  is the expectation of  $(X - E[X])^2$  :

$$\begin{aligned} \text{Var}(X) &= E((X - E[X])^2) \\ &= E(X^2 + E[X]^2 - 2XE[X]) \\ &= E(X^2 - E[X]^2) \\ &= E[X^2] - E[X]^2 \end{aligned}$$



# Bernoulli Distribution

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- The outcome of an experiment can either be success (i.e., 1) and failure (i.e., 0).
- $\Pr(X=1) = p, \Pr(X=0) = 1-p$
- $E[X] = p, \text{Var}(X) = p(1-p)$



# Binomial Distribution

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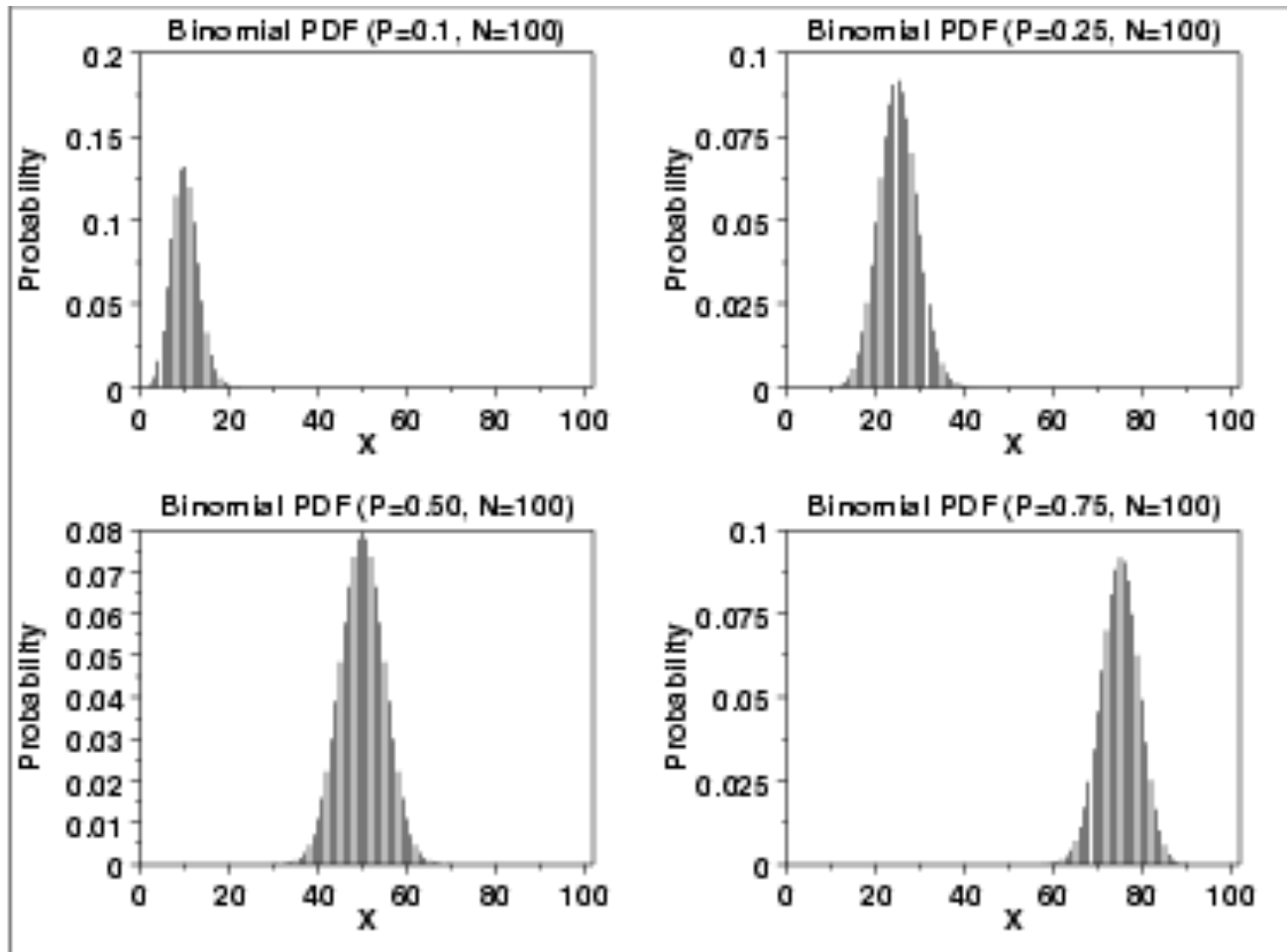
- $n$  draws of a Bernoulli distribution
  - $X_i \sim \text{Bernoulli}(p)$ ,  $X = \sum_{i=1}^n X_i$ ,  $X \sim \text{Bin}(p, n)$
- Random variable  $X$  stands for the number of times that experiments are successful.

$$\Pr(X = x) = p_{\theta}(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & x = 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

- $E[X] = np$ ,  $\text{Var}(X) = np(1-p)$



# Plots of Binomial Distribution





# Poisson Distribution

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- Coming from Binomial distribution

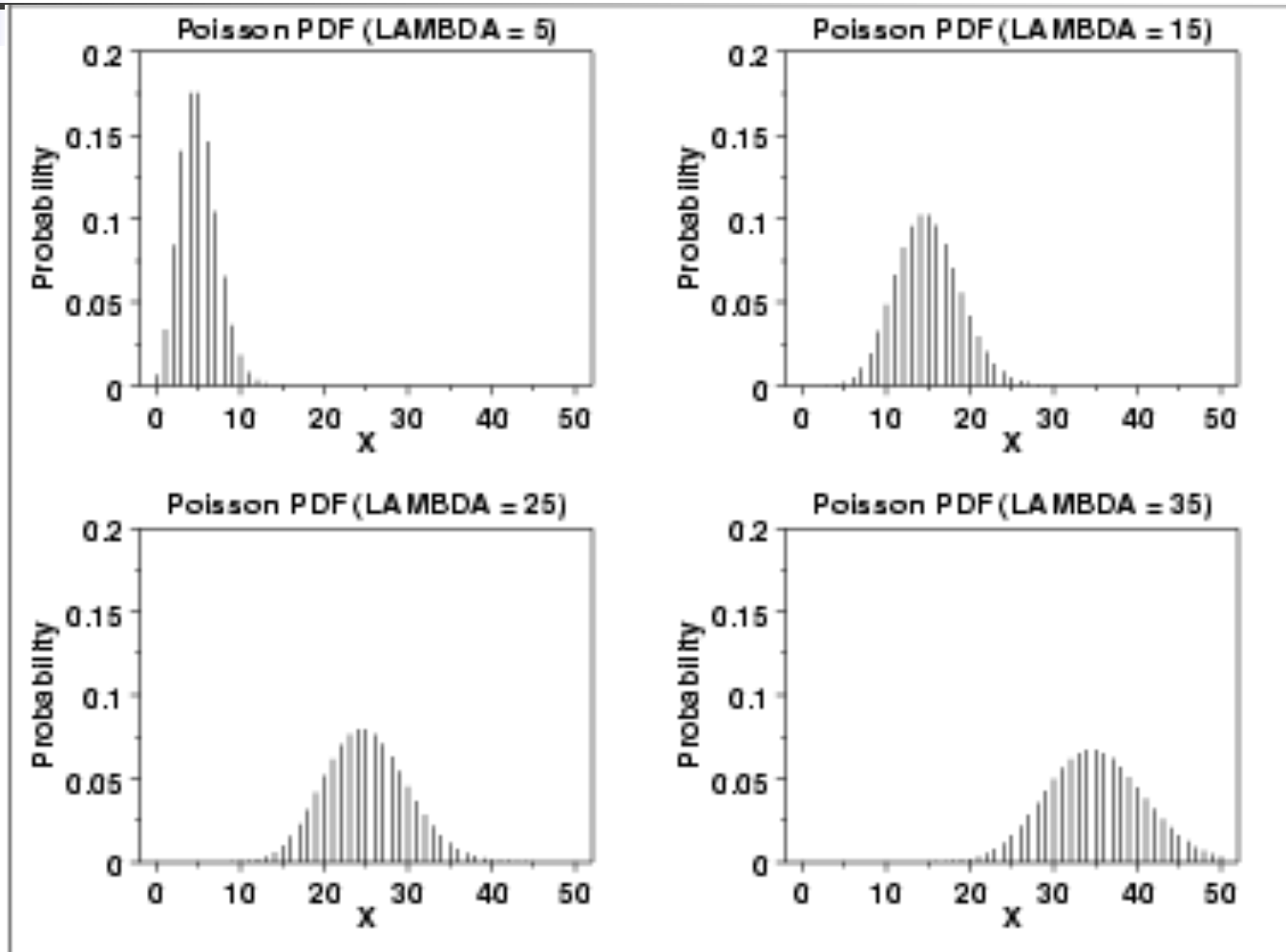
- Fix the expectation  $\lambda=np$
- Let the number of trials  $n\rightarrow\infty$

A Binomial distribution will become a Poisson distribution

$$\Pr(X = x) = p_{\theta}(x) = \begin{cases} \frac{\lambda^x}{x!} e^{-\lambda} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

- $E[X] = \lambda, \text{Var}(X) = \lambda$

# Plots of Poisson Distribution





# Normal (Gaussian) Distribution

- $X \sim N(\mu, \sigma^2)$

$$p_{\theta}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\}$$

$$\Pr(a \leq X \leq b) = \int_a^b p_{\theta}(x) dx = \int_a^b \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\} dx$$

- $E[X] = \mu, \text{Var}(X) = \sigma^2$
- If  $X_1 \sim N(\mu_1, \sigma_1^2)$  and  $X_2 \sim N(\mu_2, \sigma_2^2)$ ,  $X = X_1 + X_2$ ?

Not sum of Gaussian distributions.

Sum of normally distributed random variables.

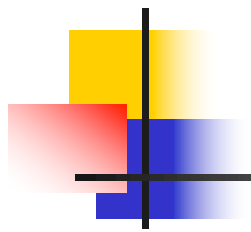
If independent, also Gaussian.  $N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$



# Summary

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- Basic concepts in probability theory
- Bayes' rule
- Random variable and probability distributions



Thank you!