

# Repetition of basic statistics

MVE220

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Problems are translated from [1]

# Independent Events

Assume  $P(B) > 0$ . Then the **conditional probability** of A given B is

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

A and B are **independent** if

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

## Problem 1 - Independent Events (2.5.5. in [1])

Two dice are thrown. Assume their sum was 4. What is the conditional probability that

- a) The first dice showed 3
- b) The second dice showed 2 or less
- c) Both dice showed an odd number

# Distribution function

A random variable  $X$  can be discrete or continuous.

- If  $X$  discrete: **probability function**  $p_X(x) := P(X = x)$ ,  
 $x = x_1, x_2, \dots$
- If  $X$  continuous: **probability density function (pdf)**  $f_X$  s.t.  
 $P(X \in A) = \int_A f_X(t) dt$

In both cases: **(cumulative) distribution function (cdf)**

$$F_X(t) := P(X \leq t), \quad -\infty < t < \infty$$

**For the cdf we have**

- $F(t) \in [0, 1]$
- $F(t) \xrightarrow[t \rightarrow -\infty]{} 0$ , and  $F(t) \xrightarrow[t \rightarrow \infty]{} 1$
- $P(a < X \leq b) = F(b) - F(a)$
- $P(X > a) = 1 - F(a)$

## Problem 2 - Distribution function (3.3.1. in [1])

Let  $Y$  be a random variable with the cdf

$$F(t) = \begin{cases} 0 & t < 0 \\ t^2 & 0 \leq t \leq 1 \\ 1 & t > 1 \end{cases}$$

- Sketch  $F(t)$
- Calculate  $P(Y \leq 0.5)$
- Calculate  $P(0.5 < Y \leq 0.9)$

## Expected value etc.

### Expected value

$$E[X] = \sum_k kP(X = k) \text{ if } X \text{ discrete}$$

$$E[X] = \int_{-\infty}^{\infty} xf_X(x)dx \text{ if } X \text{ continuous}$$

### $E[g(X)]$

$$E[g(X)] = \sum_k g(k)P(X = k) \text{ if } X \text{ discrete}$$

$$E[g(X)] = \int_{-\infty}^{\infty} g(x)f_X(x)dx \text{ if } X \text{ continuous}$$

### k:th Moment

$$E[X^k]$$

### Variance

$$\sigma^2 = \text{Var}(X) := E[(X - E[X])^2] = E[X^2] - E[X]^2$$

### Standard deviation

$$\sigma = \sqrt{\text{Var}(X)}$$

### Problem 3 - Expected value etc. (3.5.1. in [1])

Calculate the expectation and standard deviation of  $X$ .

$k$	1	2	3	4	5
$p_X(k)$	0.2	0.1	0.3	0.1	0.3



## Covariance

$$\text{Cov}(X, Y) := E[(X - E[X])(Y - E[Y])] = E[XY] - E[X]E[Y]$$

## Correlation coefficient

$$\rho(X, Y) := \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

$X, Y$  are **uncorrelated** if  $\text{Cov}(X, Y) = 0$

## Problem 4 - Correlation (3.8.3. in [1])

Let  $(X, Y)$  be given with pdf  $f_{X,Y}(x, y) = x + y$ ,  $0 \leq x \leq 1$ ,  $0 \leq y \leq 1$ .  
Calculate  $E[X]$ ,  $E[Y]$ ,  $\text{Var}(X)$ ,  $\text{Var}(Y)$ ,  $\text{Cov}(X, Y)$  and  $\rho(X, Y)$ .

# Poisson processes

## Poisson distribution

$$X \sim Po(\lambda) \quad P(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad k = 0, 1, 2, \dots$$

$$E[X] = \lambda, \quad Var(X) = \lambda$$

## Stochastic process

$\{X(t), t \in I\}$  family of random variables with index  $t$  from the index set  $I$ .

## Poisson process

Let  $\{U_k, k = 1, 2, 3, \dots\}$  be a sequence of i.i.d.  $Exp(\lambda)$  distributed random variables.

$T_n := \sum_{k=1}^n U_k$  is the time of the  $n$ -th event.

Define  $N(t) :=$  events in  $(0, t]$ ,  $N(0) = 0$ .

Then  $\{N(t), t \geq 0\}$  is a **Poisson process** with intensity  $\lambda$ .

## Independent increments of Poisson process

$$N(s + t) - N(s) \sim Po(\lambda t)$$

$$N(t) \sim Po(\lambda t)$$

## Problem 5 - Poisson processes (4.3.1. in [1])

Let  $\{N(t), t \geq 0\}$  be a Poisson process with  $\lambda = 2$ . Calculate

- a)  $P(N(1) = 0)$
- b)  $P(N(3) = 4)$
- c)  $P(N(2) \leq 3)$
- d)  $P(N(0.5) > 1)$

# Point estimation

Let  $x_1, x_2, \dots, x_n$  be observations of  $X$  that follows a distribution with some unknown parameter  $\theta$ . Then we can **estimate**  $\theta$  with  $\theta^*$

## Maximum likelihood method

We want to find  $\theta^*$  that maximises the **Likelihood function**

$$L(\theta) = \begin{cases} \prod_{i=1}^n p(x_i; \theta) & X \text{ discrete} \\ \prod_{i=1}^n f(x_i; \theta) & X \text{ continuous} \end{cases}$$

Easier to handle the **log-likelihood function**

$$l(\theta) = \ln(L(\theta)) = \begin{cases} \sum_{i=1}^n \ln(p(x_i; \theta)) & X \text{ discrete} \\ \sum_{i=1}^n \ln(f(x_i; \theta)) & X \text{ continuous} \end{cases}$$

## Problem 6 - Point estimation (7.2.9. in [1])

Nine observations were obtained from a distribution with pdf

$$f(x) = \frac{x}{\theta^2} e^{x/\theta}$$

Find the ML-estimate of  $\theta$ .

$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$	$x_7$	$x_8$	$x_9$
2.2	5.5	1.7	1.3	3.5	3.2	0.6	3.8	1.9

Given observations  $\mathbf{x}$  we want to compute a  $(1 - \alpha)\%$  **confidence interval**  $I_\theta = (\theta_1(\mathbf{x}), \theta_2(\mathbf{x}))$ , where  $P(\theta_1(\mathbf{X}) < \theta < \theta_2(\mathbf{X})) = 1 - \alpha$



## Problem 7 - Confidence intervals (7.3.1. in [1])

The random variable  $X$  is Poisson distributed with expected value  $\mu$ . The 95% confidence interval for  $\mu$  is  $I_\mu = (0.8, 2.0)$ . Calculate the 95% confidence interval for  $p := P(X = 0)$ .

## References

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- [1] Sven Erick Alm and Tom Britton. *Stokastik: sannolikhets teori och statistikteori med tillämpningar*. Liber, 2008.