

## Expectation and variance

### Expected value of a discrete random variable

$$E[X] = \sum_{-\infty}^{\infty} x_k P(X = x_k) = \sum_{-\infty}^{\infty} x_k p(x_k) \quad (1)$$

- Expectation  $\Leftrightarrow$  expected value  $\Leftrightarrow$  average  $\Leftrightarrow$  mean
- $E[X]$  is a constant/non-random
- $E[X]$  may take a value different from  $X$

### Expectation of a continuous random variable

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx \quad (2)$$

### Linearity of expected value

$$E[aX + b] = aE[X] + b \quad (3)$$

Fra sidste forelæsning

$$f_Y(y) = \left| \frac{d}{dy} g^{-1}(y) \right| f_X(g^{-1}(y)) \quad (4)$$

### Expected value of function $g$

$$E[g(X)] = E[Y] = \int_{-\infty}^{\infty} f_Y(y) dy = \int_{-\infty}^{\infty} y \left| \frac{d}{dy} g^{-1}(y) \right| f_X(g^{-1}(y)) dy \quad (5)$$

$$= \int_{-\infty}^{\infty} g(x) f_X(x) dx \quad (6)$$

### Variance of a random variable

$$\text{Var}[X] = E[(X - E[X])^2] \quad (7)$$

$\sigma = \sqrt{\text{Var}[X]}$  is called standard deviation.

### Huygens formula

$$\text{Var}[X] = E[X^2] - E[X]^2 \quad (8)$$

### Chebyshev's inequality

Let  $X$  be any random variable with mean  $\mu$  and variance  $\sigma^2$ . For any constant  $c > 0$ , we have

$$P(|X - \mu| \leq c\sigma) \leq \frac{1}{c^2} \quad (9)$$

Chebyshev's inequality helps interpret variance and gives some upper bound for the deviation from the mean.

**Non-linearity of variance**

Let  $X$  be any random variable, and let  $a$  and  $b$  be real numbers. Then

$$\text{Var}[aX + b] = a^2 \text{Var}[X] \quad (10)$$