

# Metody Statystyczne

wykład 3

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- rozkład normalny
- estymacja punktowa
- estymacja przedziałowa
- procesy stochastyczne
  - łańcuchy Markowa

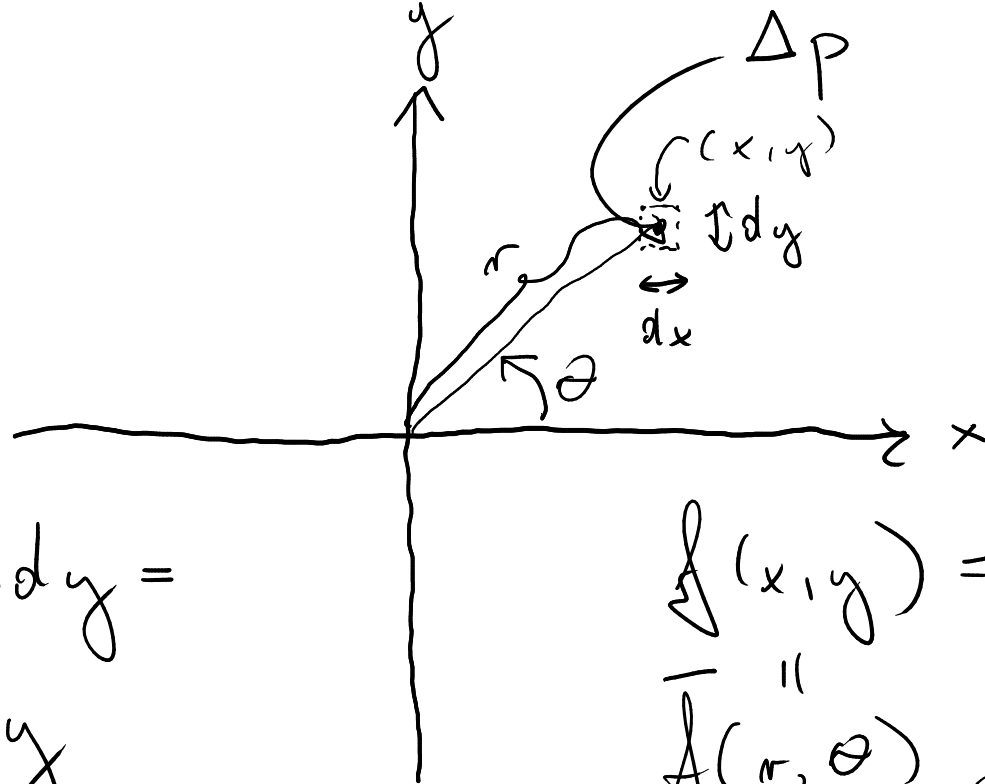
# rozkład normalny

$X, Y$  - niezależne zmienne losowe

FGP jest takie same dla  $X$  oraz  $Y$

$$f(x) \rightarrow 0 \text{ gdy } x \rightarrow \pm \infty$$

$$f(-x) = f(x)$$



$$\Delta p = f(x) dx f(y) dy =$$

$$= \underbrace{f(x, y)} dx dy$$

F6P 2D  
dla  $x, y$

$$f(x, y) = f(x) f(y)$$

$$\underbrace{f(r, \theta)}$$

$$y = r \sin \theta$$

$$x = r \cos \theta$$

$$0 = \underbrace{\frac{\partial f}{\partial \theta}}_{\text{zależność}}(r, \theta) = f(x) \frac{\partial f}{\partial \theta}(y) + f(y) \frac{\partial f}{\partial \theta}(x)$$

$$0 = f(x) \frac{\partial f}{\partial \theta}(r \sin \theta) + f(y) \frac{\partial f}{\partial \theta}(r \cos \theta)$$

$$0 = \underbrace{f(x)}_{f(r \cos \theta)} \underbrace{f'(r \sin \theta)}_{f'(y)} \underbrace{x}_{r \cos \theta} +$$

$$\underbrace{f(r \sin \theta)}_y \underbrace{f'(r \cos \theta)}_x (-\underbrace{r \sin \theta}_y)$$

$$0 = f(x) f'(y) x - f(y) f'(x) y$$

$$f(y) f'(x) y = f(x) f'(y) x$$

$$\underbrace{\frac{f'(x)}{x f(x)} = \frac{f'(y)}{y f(y)}}_{\text{separazione delle variabili}}$$

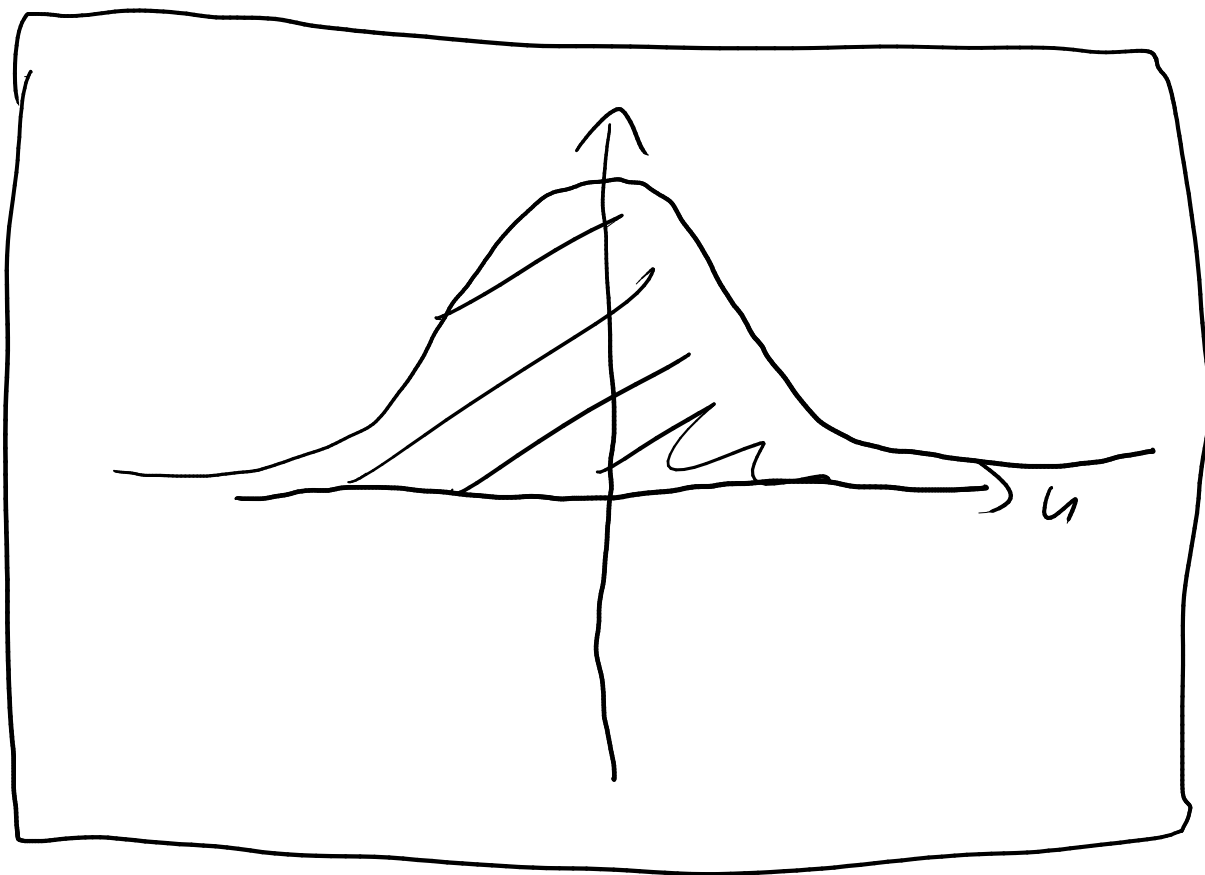
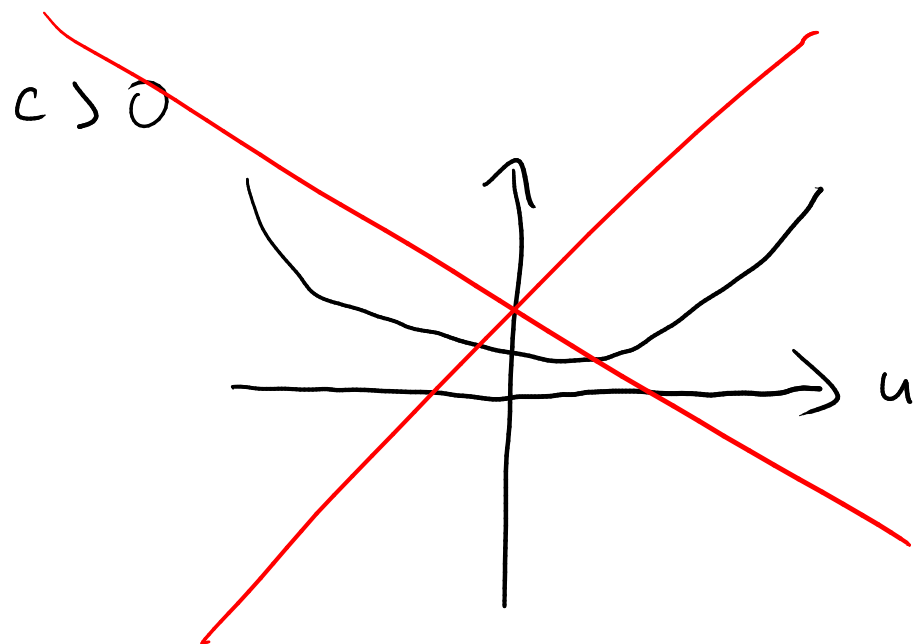
$$\frac{f'(x)}{x f(x)} = c, \quad \frac{f'(y)}{y f(y)} = c$$

$$f'(x) = c x \cdot f(x)$$

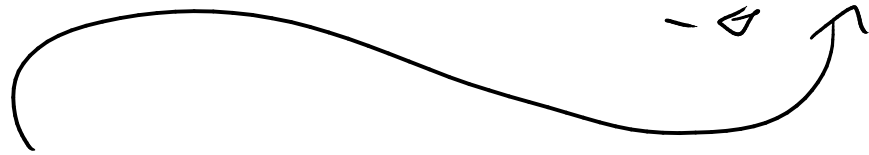
$$f(x) = A \cdot e^{\frac{c}{2} x^2}$$

$$\begin{aligned} f'(x) &= A \cdot e^{\frac{c}{2} x^2} \cdot \frac{c}{2} \cdot 2x \\ &= c \cdot x \cdot \underbrace{A e^{\frac{c}{2} x^2}}_{f(x)} \end{aligned}$$

$$\int_{-\infty}^{+\infty} du f(u) = 1$$



$$\int_{-\infty}^{+\infty} f(u) du = 1$$



$$f(u) = \frac{1}{\sigma \sqrt{2\pi}} \exp\left(-\frac{1}{2} \frac{(u - \mu)^2}{\sigma^2}\right)$$

$A, C \rightarrow \sigma, \mu$

wartość oczekiwana

$\sigma > 0$

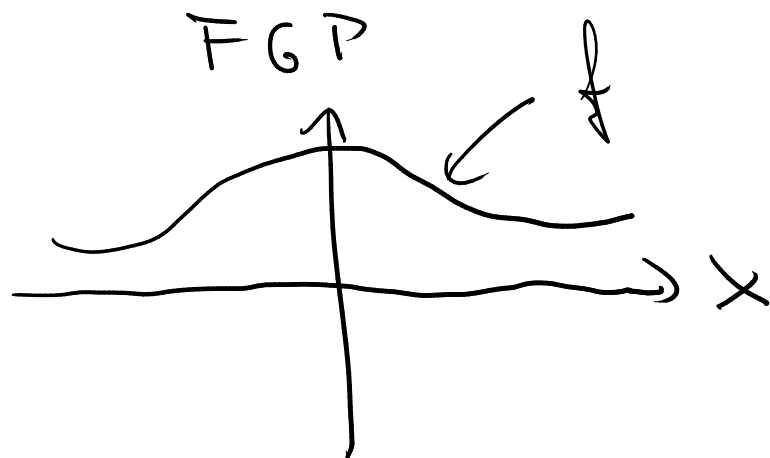
odchylenie standardowe



$$Z = X + X$$

The diagram illustrates the components of the equation  $Z = X + X$ . The first  $X$  is associated with the label "normalized Cauchy's" via a bracket. The second  $X$  is associated with the label "normalized normality" via a bracket. Additionally, the first  $X$  is associated with the label "normalized normality" via an arrow, and the second  $X$  is associated with the label "normalized Cauchy's" via an arrow.

# estymacja punktowa

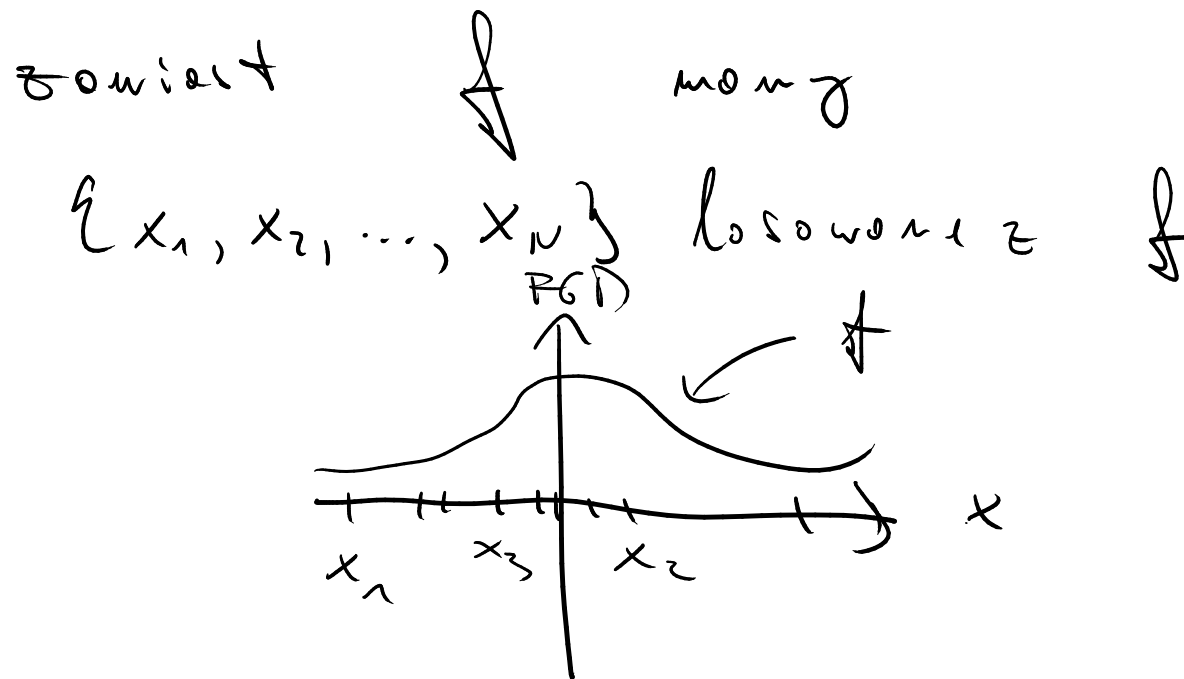


$$E(X) = \int_{-\infty}^{+\infty} dx \, x \, f(x)$$

$$\text{Var}(X) = \int_{-\infty}^{+\infty} dx \, f(x) (x - E(X))^2$$

$$\int_{-\infty}^{+\infty} dx \, x$$

w praktyce:



jest oszacować  $E(X)$ ,  $Var(X)$ ,  $\rho_{x,x}$

konstanty i estymatorów  
jest zmienną losową

$$\frac{1}{n}(\theta) = \hat{\theta}$$

$\uparrow$   $\uparrow$   
 $\underline{E}, \text{Var}$

losowa domyślnie

$$B(\hat{\theta}) = E(\hat{\theta}) - \theta$$

← "meile"

↑  
foss

$$B(\hat{\theta}) = 0$$

estimator  
unbiased

↓

$$\hat{\theta} = \bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n}$$

θ

↓  
describing

$$\frac{E(X_i)}{1}$$

$$B(\hat{\theta}) = E(\hat{\theta}) - \theta = E(\bar{X}) - \theta = E(X_i) - \theta = E(X_i) - E(X_i) = 0$$

$$\hat{\theta}_1 = X_1$$

↑  
unbiasedness

$$\hat{\theta} = \frac{X_1 + X_2 + \dots + X_n}{N}$$

↑

$$\begin{aligned} B(\hat{\theta}_1) &= E(\hat{\theta}_1) - \theta = \\ &= E(X_1) - \theta = \underline{0} \end{aligned}$$

mean      error

↓      ↓

$$MS E(\hat{\theta}) = E((\hat{\theta} - \theta)^2)$$

↑  
squared

$$\hat{\theta}_1 = X_1$$

$$\hat{\theta}_2 = \frac{X_1 + X_2 + \dots + X_n}{N} = \bar{X}$$

$$\begin{aligned} \text{MSE}(\hat{\theta}_1) &= E\left((\hat{\theta}_1 - \theta)^2\right) = E\left((X_1 - E(X_1))^2\right) = \\ &= \text{Var}(X_1) = \sigma^2 \end{aligned}$$

$$\begin{aligned} \text{MSE}(\hat{\theta}_2) &= E\left((\hat{\theta}_2 - \theta)^2\right) = E\left((\bar{X} - \theta)^2\right) = \\ &= \underbrace{\text{Var}(\bar{X} - \theta)}_{\text{Var}(\bar{X})} + E(\bar{X} - \theta)^2 = \frac{\sigma^2}{n} \end{aligned}$$

$$\lim_{N \rightarrow \infty} P(\hat{\theta}_N - \theta > \epsilon) = 0 \quad \text{dla dowolnego } \epsilon$$

$\uparrow$   
i. estymator

estymator konsekwentny



$$\sigma^2 = E((X - \underbrace{\mu}_{\text{среднее}})^2)$$

$$\hat{\sigma}^2 = \frac{1}{N} \sum_{k=1}^N (x_k - \underbrace{\mu}_{\bar{x}})^2$$

$$\begin{aligned} s^2 &= \frac{1}{N} \sum_{k=1}^N (x_k - \bar{x})^2 = \\ &= \frac{1}{N} \left( \sum_{k=1}^N x_k^2 - N \bar{x}^2 \right) \end{aligned}$$

$$\underline{E(X_i) = \mu}$$

$$\underline{Var(X_i) = \sigma^2}$$

$$B(\bar{S}^2) = E(\bar{S}^2) - \sigma^2$$

$$E(\bar{X}^2) = E(\bar{X})^2 + Var(\bar{X}) = \mu^2 + \frac{\sigma^2}{n}$$

$$E(\bar{S}^2) = \frac{1}{N} \left( \sum_{k=1}^N E(X_k^2) - n E(\bar{X}^2) \right) =$$

$$= \frac{1}{N} \left( N(\mu^2 + \sigma^2) - n \left( \mu^2 + \frac{\sigma^2}{n} \right) \right) =$$

$$= \frac{N-1}{N} \sigma^2$$

$$B(\bar{S}^2) = E(\bar{S}^2) - \sigma^2 = \left[ -\frac{\sigma^2}{N} \right]$$

$$B\left(\underbrace{\frac{1}{N-1} \sum_{k=1}^n (X_k - \bar{X})^2}_{\text{estimator variance}}\right) = ?$$

# estymacja predyktora

$$T_w(\theta) = \hat{\theta} \in$$

↓

$$\left[ T^L(\theta), T^R(\theta) \right]$$

predyktorzy dla  $\theta$

$$P(T_N^L(\theta) \leq \theta \leq T_N^R(\theta)) = \gamma$$

$\uparrow$   
 position of median

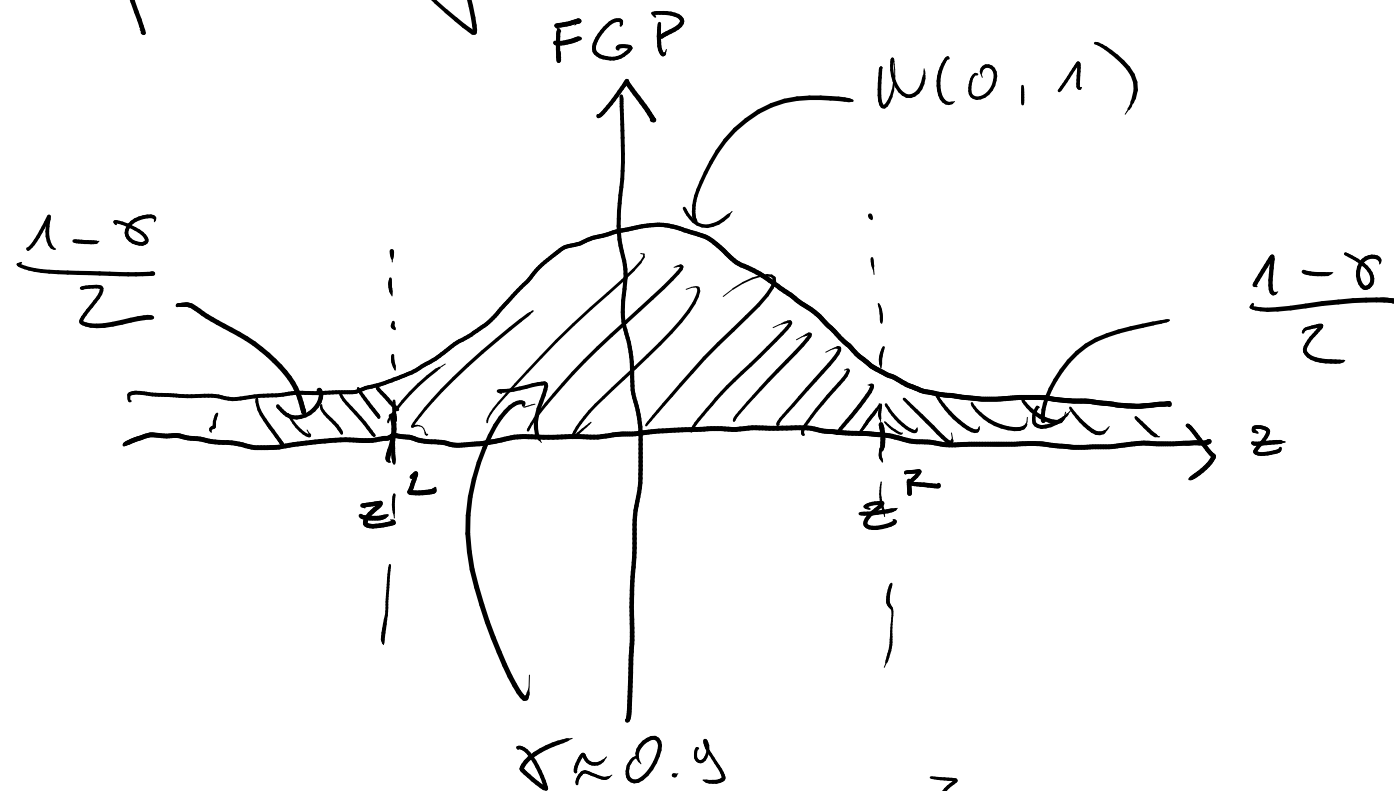
## worst case of algorithm

$$\sigma^2 \leftarrow \text{variance} \rightarrow \sigma(X)$$

$$z = \frac{\sqrt{\frac{1}{N} \sum_{i=1}^N x_i^2} \cdot \frac{\bar{X} - E(X)}{\sigma(\bar{X})}}{\sigma(X)} = \frac{(\bar{X} - E(X)) \sqrt{N}}{\sigma(X)}$$

$$FBP(z) \leftarrow \text{normalized normally } N(0, 1)$$

$\gamma \approx 0.9 \leftarrow$  position of mass



$$z_L = z_{\frac{1-\gamma}{2}} = -z_{\frac{\gamma+1}{2}}$$

$$z_R = z_{\frac{1-\gamma}{2} + \gamma} = z_{\frac{\gamma+1}{2}}$$

$$P(z^L \leq z \leq z^R) = \gamma =$$

$$T_N^R(\bar{R}(x))$$

$$= P\left(z^L \leq \frac{(\bar{x} - \bar{E}(x))\sqrt{N}}{\sigma(x)} \leq z^R\right) =$$

$$= P\left(\frac{z^L \sigma(x)}{\sqrt{N}} \leq \bar{x} - \bar{E}(x) \leq \frac{z^R \sigma(x)}{\sqrt{N}}\right) =$$

$$= P\left(\frac{z^L \sigma(x)}{\sqrt{N}} - \bar{x} \leq -\bar{E}(x) \leq \frac{z^R \sigma(x)}{\sqrt{N}} - \bar{x}\right) =$$

$$= P\left(\bar{x} - \frac{z^L \sigma(x)}{\sqrt{N}} \geq \bar{E}(x) \geq \bar{x} - \frac{z^R \sigma(x)}{\sqrt{N}}\right) =$$

$$\underbrace{T_N^L(\bar{E}(x))}_{=} = P\left(\underbrace{\bar{x} - \frac{z_{\frac{1-L}{2}} \sigma(x)}{\sqrt{N}}}_{\leq \bar{E}(x)} \leq \bar{E}(x) \leq \underbrace{\bar{x} + \frac{z_{\frac{1+L}{2}} \sigma(x)}{\sqrt{N}}}_{\geq \bar{E}(x)}\right)$$

nie znamy  $\sigma(x)$  ( $\sigma^2$ )  
N-pomiarów

$$t = \frac{(\bar{x} - \mathbb{E}(x)) \sqrt{N}}{s(x)}$$

$\uparrow \quad \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$

rozkład studenta o  $N-1$  stopniach swobody



$$P\left(\overline{X} - \frac{1}{\sqrt{N}} S(x) t_{\frac{1+\delta}{2}} \leq E(x) \leq \overline{X} + \frac{1}{\sqrt{N}} S(x) t_{\frac{1+\delta}{2}}\right) = \gamma$$

$$\left[ \frac{1}{\sqrt{N}}^L(E(x)) \quad , \quad \frac{1}{\sqrt{N}}^R(E(x)) \right]$$

$$\chi^2 = \frac{(N-1) S^2}{\sigma^2(x)} \quad S^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2$$

$\uparrow$   
 $\hookrightarrow$  me FGD

$\chi^2_{N-1}$   
 stopniach swobodny

$$\begin{aligned}
 & P( \dots ( \chi^2 < \dots ) ) = \gamma \\
 & P \left( \sqrt{\frac{(N-1) S^2}{(\chi^2_{N-1})_{\frac{1+\gamma}{2}}}} \leq \sigma(x) \leq \sqrt{\frac{(N-1) S^2}{(\chi^2_{N-1})_{\frac{1-\gamma}{2}}}} \right) = \gamma
 \end{aligned}$$

#

$$\begin{array}{ccc}
 X & & \\
 \uparrow & & \\
 X: \Omega \rightarrow \mathbb{R} & & \\
 \uparrow & \text{lib. rezymu} & \\
 X(\square) = z & \text{zbiór zolom' elementarny} &
 \end{array}$$

$$\begin{array}{c}
 \downarrow \omega \in \Omega \\
 X(t, \omega) \\
 \uparrow \text{czas}
 \end{array}$$