

# Föreläsning 4

TAMS24 – Statistisk teori

$\chi^2$  - fördelning och t-fördelning

Konfidensintervall

Prediktionsintervall

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$$\underline{X} \quad \theta \quad \hat{\theta}$$

Interval estimation:

$$\theta \in (a, b)$$

e.g.

$$\underline{X} = \{ \text{lifetime of lions} \}$$

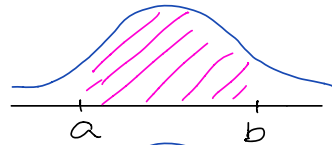
$$\underline{X} \sim \text{Exp}\left(\frac{1}{\mu}\right), \quad E(\underline{X}) = \mu, \quad \text{expected lifetime}$$

$$I_{\mu} = (15 \text{ years}, 25 \text{ years})$$

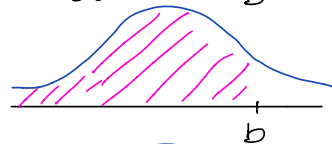
### 1. Preparation

(1)  $\underline{X}$  is continuous

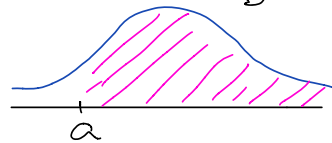
$$P(a \leq \underline{X} \leq b) = \text{Area of}$$



$$P(\underline{X} \leq b) = \text{(Left) area of}$$



$$P(\underline{X} \geq b) = \text{(right) area of}$$



(2)  $\underline{X}_1, \underline{X}_2, \dots, \underline{X}_n$  i.i.d (independent and identically distributed) and each

$$\underline{X}_i \sim N(\mu, \sigma), \quad i = 1, 2, \dots, n$$

Then

$$\bar{X} = \frac{X_1 + \dots + X_n}{n} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

Since

$$E(\bar{X}) = \mu, \quad V(\bar{X}) = \frac{\sigma^2}{n}$$

Note:

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1)$$

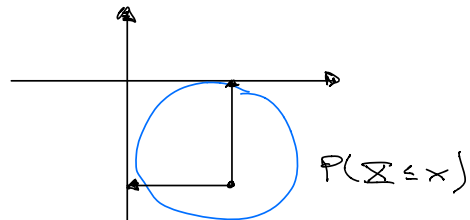
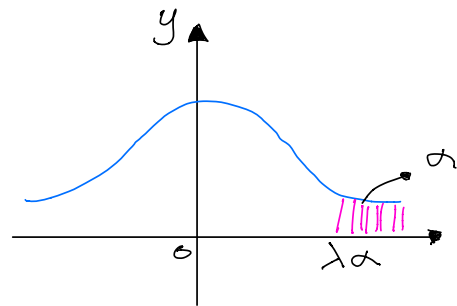
(3) Notations:

$$(3.1) \bar{X} \sim N(0, 1)$$

$$P(\bar{X} \geq \lambda\alpha) = \alpha$$

eg.

$$\lambda_{0.05} = \frac{1.64 + 1.65}{2} = 1.645$$



$$\text{left} = 1 - 0.05 = 0.95$$

$$0.95 = \frac{0.9495 + 0.9505}{2}$$

$$(3.2) \bar{X} \sim t(f):$$

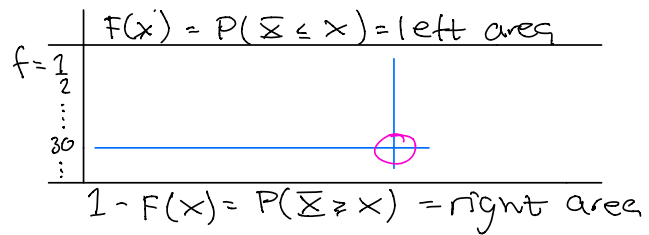
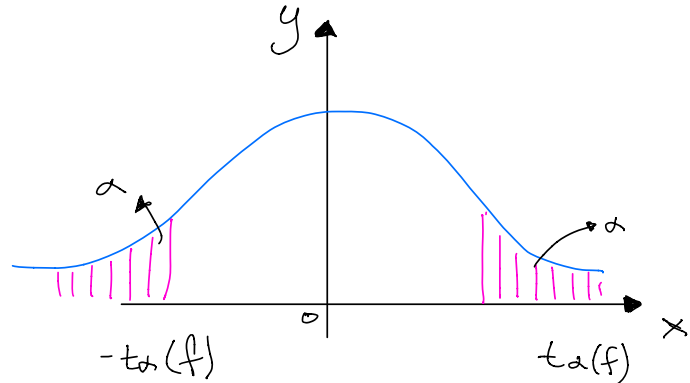
t-distribution.

$f$  = degrees of freedom (frihetsgrader)

$$P(\bar{X} \geq t_{\alpha}(f)) = \alpha$$

$$\alpha = 0.05, f = 30$$

$$t_{0.05}(30) = 1.7$$



(3.3)  $\bar{X} \sim \chi^2(f)$ : Chi-square distribution

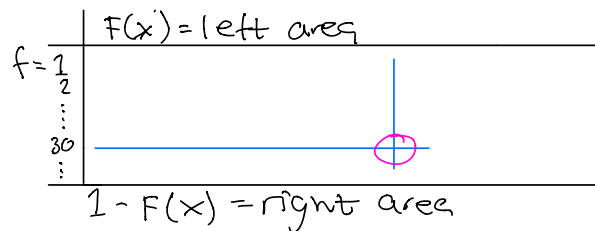
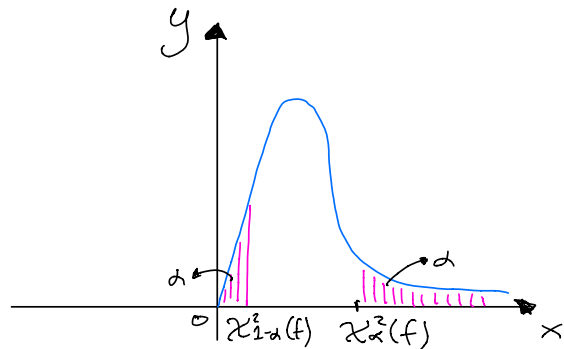
$f$  = degrees of freedom

$$P(\bar{X} \geq \chi_{\alpha}^2(f)) = \alpha$$

$$\alpha = 0.05, f = 30$$

$$\chi_{0.05}^2(30) \stackrel{\text{table}}{\approx} 43.78$$

$$\chi_{1-\alpha}^2(f) = \chi_{0.95}^2(30) = 18.49$$



## 2. Interval Estimation

Type I:

One sample

$$\{\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n\} \sim N(\mu, \sigma)$$

(1)  $(1-\alpha)$

confidence interval for  $\mu$ :  $I_\mu$ .

$$I_\mu = (0, 300 \text{ cm}) \Rightarrow 100\%$$

$$I_\mu = (175, 185) \Rightarrow ?$$

$1-\alpha$  = confidence coefficient (Konfidenzgrad)

e.g.

$$1-\alpha = 95\%$$

95% confidence to say

$\mu \in I_\mu$ :

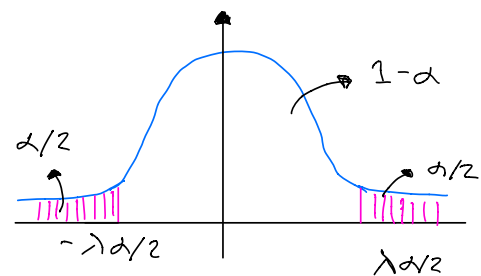
$$100 \text{ samples: } \begin{cases} 95: \mu \in I_\mu \\ 5: \mu \notin I_\mu \end{cases}$$

(1.1) If  $\sigma$  is known, fact:

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$$

$$P\left(-\frac{\lambda}{2} \leq \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq \frac{\lambda}{2}\right) = 1-\alpha$$

$$P\left(-\frac{\lambda \alpha}{2} \cdot \frac{\sigma}{\sqrt{n}} \leq \bar{X} - \mu \leq \frac{\lambda \alpha}{2} \cdot \frac{\sigma}{\sqrt{n}}\right) = 1-\alpha$$

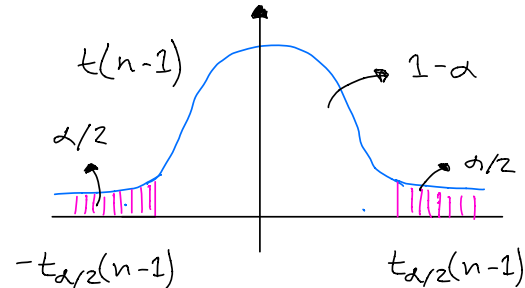


$$P\left(\bar{x} - \frac{\lambda\alpha}{2} \cdot \frac{\sigma}{\sqrt{n}} \leq \mu \leq \bar{x} + \frac{\lambda\alpha}{2} \cdot \frac{\sigma}{\sqrt{n}}\right) = 1-\alpha$$

$$(1-\alpha) \text{ CI: } I_{\mu} = \left(\bar{x} - \frac{\lambda\alpha}{2} \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + \frac{\lambda\alpha}{2} \cdot \frac{\sigma}{\sqrt{n}}\right) = \\ = \bar{x} \pm \frac{\lambda\alpha}{2} \cdot \frac{\sigma}{\sqrt{n}}$$

(1.2)  $\sigma$  is known, fact:

$$\frac{\bar{x} - \mu}{s/\sqrt{n}} \sim t(n-1)$$



$$P\left(-t_{\alpha/2}(n-1) \leq \frac{\bar{x} - \mu}{s/\sqrt{n}} \leq t_{\alpha/2}(n-1)\right) = 1-\alpha$$

$$(1-\alpha) \text{ CI: } I_{\mu} = \bar{x} \pm t_{\alpha/2}(n-1) \cdot \frac{s}{\sqrt{n}}$$

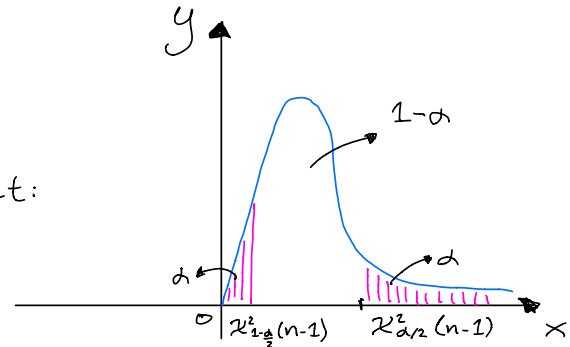
$$I_{\mu} = (a, \infty) = \left(\bar{x} - \lambda\alpha \cdot \frac{\sigma}{\sqrt{n}}, \infty\right)$$

Note:

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

(1.3)  $(1-\alpha)$  CI for  $\sigma^2$  or  $\sigma$ ; fact:

$$\frac{(n-1)s^2}{\sigma^2} \sim \chi^2(n-1)$$



$$P\left(\chi^2_{1-\frac{\alpha}{2}}(n-1) \leq \frac{(n-1)s^2}{\sigma^2} \leq \chi^2_{\frac{\alpha}{2}}(n-1)\right) = 1-\alpha$$

$(1-\alpha)$  CI:

$$I_{\sigma^2} = \left(\frac{(n-1)s^2}{\chi^2_{\frac{\alpha}{2}}(n-1)}, \frac{(n-1)s^2}{\chi^2_{1-\frac{\alpha}{2}}(n-1)}\right)$$

$(1-\alpha)$  CI:

$$I\sigma = \left( \sqrt{\frac{(n-1)S^2}{\chi^2_{\frac{\alpha}{2}}(n-1)}}, \sqrt{\frac{(n-1)S^2}{\chi^2_{1-\frac{\alpha}{2}}(n-1)}} \right)$$

Ex 4.1:

The heights of a certain kind of christmas trees are assumed to be  $N(\mu, \sigma)$

Sample:

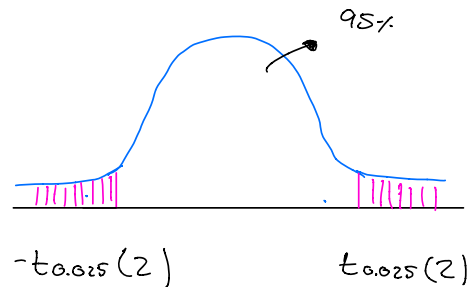
178cm, 182cm, 186cm.

Find confidence interval for expected height of a tree.  $I\mu$ .

Solution:

Fact:

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t(n-1)$$



$$I\mu = \bar{X} \pm t_{0.025}(2) \cdot \frac{S}{\sqrt{n}} = \dots \approx (172, 192)$$

$$\bar{X} = 182 \text{ cm}, t_{0.025}(2) = 4.3, S^2 = 16, S = 4$$

Remark:

- ① two sided (tväsidigt) CI:  $(a, b)$  (uppåt begränsat)
- ② one sided (ensidigt) CI:  $\begin{cases} (-\infty, b), \text{ upper limit CI} \\ (a, \infty), \text{ lower limit CI} \end{cases}$  (nedre begränsat)

### Ex 4.1:

Find 95% upper limit CI for  $\sigma^2$ .

Solution:

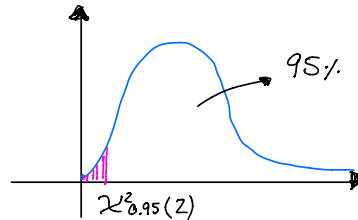
Fact:

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$$

$$I_{\sigma^2} = (-\infty, b) = \left(0, \frac{(n-1)S^2}{\chi^2_{1-\alpha}(n-1)}\right)$$

$$P\left(\frac{(n-1)S^2}{\sigma^2} \geq \chi^2_{0.95}(2)\right) = 95\%$$

$$I_{\sigma^2} = \left(0, \frac{(n-1)S^2}{\chi^2_{0.95}(2)}\right) \approx (0, 320), \quad \chi^2_{0.95}(2) = 0.1$$



Remark:

A short cut to get one-sided from two-sided CI.

I. Get the corresponding form

II Replace  $\frac{\alpha}{2}$  by  $\alpha$ .

Note: If it is upper limit for  $\sigma^2$  for  $\chi^2(f)$ . Use  $(0, b)$ .

### 3. Theorems

(3.1) Fact:

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$$



Theorem 1:

$$X_1, \dots, X_n$$

i.i.d & each

$$X_i \sim N(0, 1),$$

then

$$\sum_{i=1}^n X_i^2 \sim \chi^2(n)$$

Theorem 2:

If

$$X \sim \chi^2(f_1)$$

and

$$Y \sim \chi^2(f_2)$$

are independent, then

$$X + Y \sim \chi^2(f_1 + f_2)$$

Proof:

$$\begin{aligned} \frac{(n-1)S^2}{\sigma^2} &= \frac{(n-1) \cdot \frac{1}{(n-1)} \cdot \sum_{i=1}^n (X_i - \bar{X})^2}{\sigma^2} = \\ &= \frac{\sum_{i=1}^n (X_i - \mu + \mu - \bar{X})^2}{\sigma^2} = \dots = \frac{\sum_{i=1}^n (X_i - \mu)^2}{\sigma^2} - \frac{n(\bar{X} - \mu)^2}{\sigma^2} \end{aligned}$$

$$= \underbrace{\sum_{i=1}^n \left( \frac{x_i - \mu}{\sigma} \right)^2}_{\chi^2(n)} - \underbrace{\left( \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right)^2}_{\chi^2(1)} = \chi^2(n-1)$$