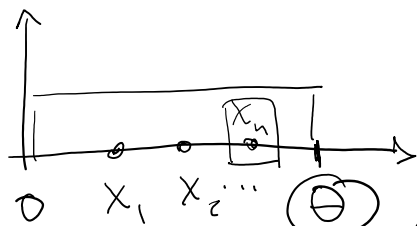


# German tank problem



$$X_1, X_2, \dots, X_n \sim V[0, \theta]$$

$$1) \hat{\theta}^{(1)} = \max((x_1, \dots, x_n))$$

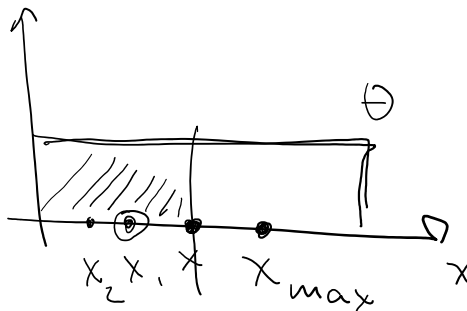
MLE

$$2) \bar{X} = \theta/2 \rightarrow \hat{\theta}^{(2)} = \frac{2}{n} \sum_{i=1}^n X_i$$

MOM estimate

$$1) a) \hat{\theta} = \max(\dots)$$

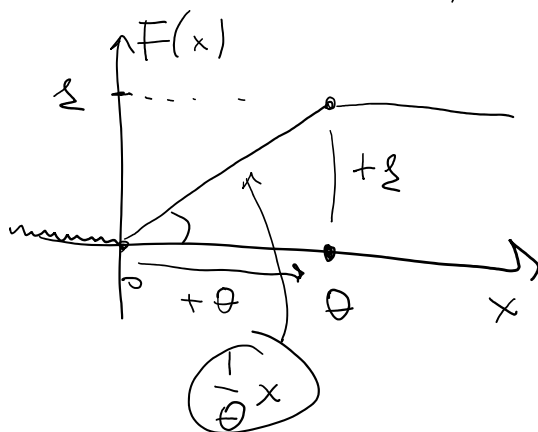
$$\text{bias}(\hat{\theta}) = E(\hat{\theta}) - \theta$$



$$Pr(\hat{\theta} \leq x) =$$

$$Pr(\max(x_1, x_2, \dots, x_n) \leq x) =$$

$$= Pr(x_1 \leq x) Pr(x_2 \leq x) \dots Pr(x_n \leq x)$$



CDF of  $V[0, \theta]$

$$\left(\frac{x}{\theta}\right)^n \leftarrow \text{CDF of } \hat{\theta}$$

if  $0 \leq x \leq \theta$



$$E(\hat{\theta}) = \int x \underbrace{dF}_{\frac{dF}{dx} \cdot dx} = \frac{1}{\theta^n} \int_0^{\theta} \underbrace{x}_{x^n} \cdot \underbrace{\frac{1}{\theta^n} n x^{n-1}}_{\frac{d}{dx} \left( \frac{x^n}{n} \right)} dx = \frac{1}{\theta^n} n \cdot \frac{\theta^{n+1}}{n+1} = \frac{n}{n+1} \theta$$

$\int x^n dx = \frac{x^{n+1}}{n+1} + C$

$\left( \left( \frac{x}{\theta} \right)^n \right)' = \frac{1}{\theta^n} n x^{n-1}$

$$\hat{\theta} = \frac{n}{n+1} \theta$$

$$\text{bias}(\hat{\theta}) = \frac{n}{n+1} \theta - \theta = -\frac{\theta}{n+1} \xrightarrow{n \rightarrow \infty} 0$$

$\frac{n+1-1}{n+1} = 1 - \frac{1}{n+1}$

b) standard error

$$se(\hat{\theta}) = \sqrt{\text{Var}(\hat{\theta})}$$

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

$$\begin{aligned} \int_0^{\theta} x^2 dF &= \int_0^{\theta} x^2 \cdot \frac{1}{\theta^n} n x^{n-1} dx = \frac{n}{\theta^n} \int_0^{\theta} x^{n+1} dx = \\ &= \frac{n}{\theta^n} \frac{\theta^{n+2}}{n+2} = \frac{n}{n+2} \theta^2 \end{aligned}$$

$\frac{x^{n+2}}{n+2} \Big|_0^{\theta}$

$$\text{Var}(\hat{\theta}) = \frac{n}{n+2} \theta^2 - \left( \frac{n}{n+1} \right)^2 \theta^2 \xrightarrow{n \rightarrow \infty} 0$$

$$c) \text{MSE}(\hat{\theta}) = \text{bias}^2(\hat{\theta}) + \text{Var}(\hat{\theta}) \rightarrow 0$$

$$\text{se}^2(\hat{\theta}) \quad n \rightarrow \infty$$


---

$$2) a) \quad \hat{\theta} = 2\bar{X} = \frac{2}{n} \sum_{i=1}^n X_i$$

$$\text{bias}(\hat{\theta}) = \mathbb{E}(\hat{\theta}) - \theta = 0$$

$$\mathbb{E}\left(\frac{2}{n} \sum_{i=1}^n X_i\right) = \frac{2}{n} \sum_{i=1}^n \mathbb{E}(X_i) = \frac{\theta}{n} \cdot n = \theta$$


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$$b) \quad \text{se}(\hat{\theta}) = \sqrt{\text{Var}(\hat{\theta})}$$

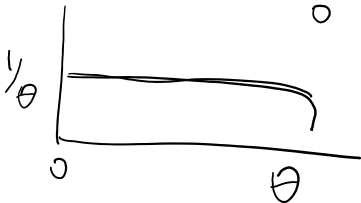
$$\text{Var}\left(\frac{2}{n} \sum_{i=1}^n X_i\right) = \mathbb{E}(\hat{\theta}^2) - \mathbb{E}^2(\hat{\theta})$$

$$\mathbb{E}\left(\frac{4}{n^2} \left(\sum_{i=1}^n X_i\right)^2\right) = \frac{4}{n^2} \left( \sum_{i=1}^n \mathbb{E}(X_i^2) + \sum_{i \neq j} \mathbb{E}(X_i \cdot X_j) \right)$$

$$\underbrace{\sum_{i=1}^n X_i^2}_n + \underbrace{\sum_{i \neq j} X_i X_j}_{\binom{n}{2} = \frac{n(n-1)}{2}}$$

// indep  
 $\mathbb{E}(X_i) \mathbb{E}(X_j)$   
 $\theta/2$

$$\mathbb{E}(X^2) = \int_0^{\theta} x^2 \frac{1}{\theta} dx = \frac{1}{\theta} \frac{\theta^3}{3} = \frac{\theta^2}{3}$$



$$\dots \rightarrow \text{se}(\hat{\theta}) = \theta \sqrt{\frac{1}{3n}} \rightarrow 0$$

$$n \rightarrow \infty$$


---

consistency ( $\hat{\theta}$ ):

$$\hat{\theta} \xrightarrow[n \rightarrow \infty]{P} \theta$$

$$\Pr(|\hat{\theta} - \theta| > \varepsilon) \xrightarrow[n \rightarrow \infty]{} 0$$

Strong consistency:

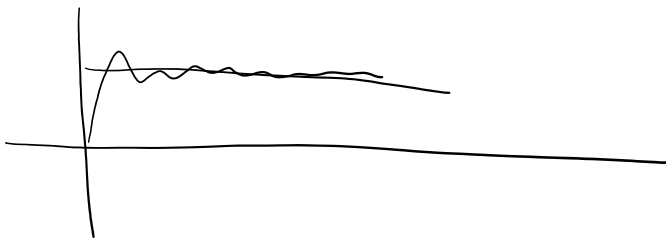
$$\hat{\theta} \xrightarrow[n \rightarrow \infty]{a.s.} \theta$$

$$\Pr(\lim \hat{\theta}_n \neq \theta) = 0$$

$$\Pr(\lim \hat{\theta}_n = \theta) = 1$$

for  $\hat{\theta}_n = 2 \bar{X}$

$$\lim \theta_n = 2 \lim \bar{X}_{(n)}$$



$$\bar{X}_n \xrightarrow{a.s.} E(x)$$

Strong Law  
of Large Numbers

From CLT

$$\hat{\theta}_n - \theta \underset{n \rightarrow \infty}{\sim} \mathcal{N}(0, \frac{\sigma^2}{\sqrt{n}})$$

HW2

MSE  
convergence  
implies  
convergence  
in probability

# HW2

$$1) \quad X_1, X_2, \dots, X_n \sim \text{Be}(p)$$

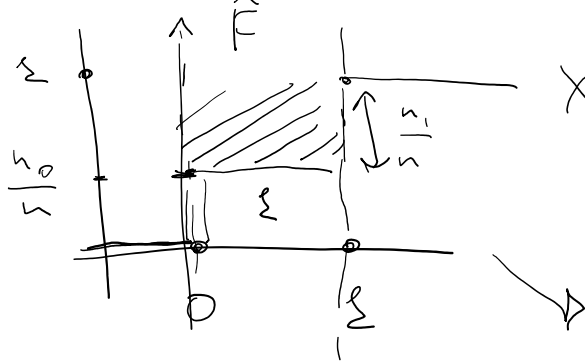
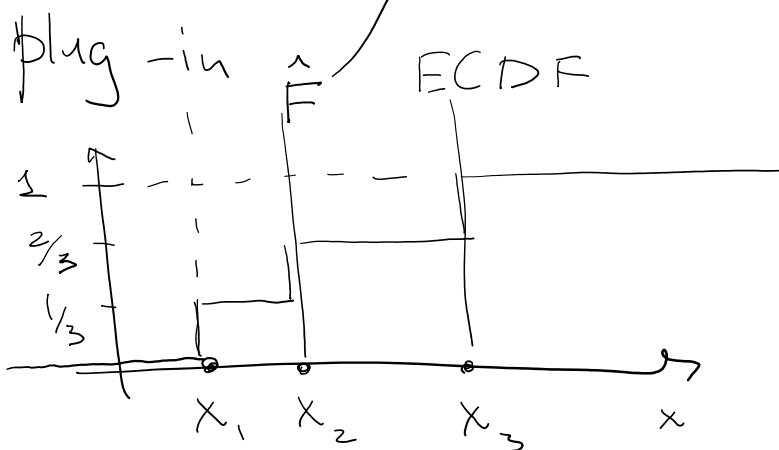
$$Y_1, Y_2, \dots, Y_m \sim \text{Be}(q)$$

estimator for  $p$ ?  $p-q$ ?

plug-in estimator?

↑  
Statistical functionals = functionals of the  $F(x)$   
CDF(x)  
Var, ... Median,

$$E(x) = \int x dF$$



$$X_1, \dots, X_n \rightarrow n_0 \text{ 0-s}$$

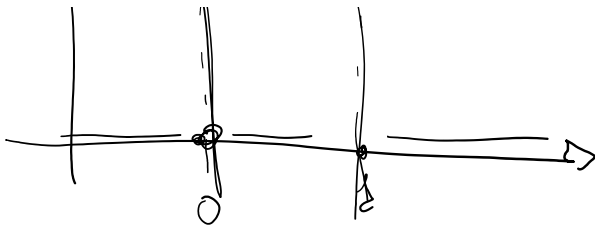
$$X_1, \dots, X_n \rightarrow n_1 \text{ 1-s}$$

$$n_0 + n_1 = n$$

$$p = E(\text{Be}(p)) = \int x dF = \int x d\hat{F} =$$

$$\left| \frac{n_0}{n} \delta(x_0) \right| \left| \frac{n_1}{n} \delta(x_1) \right|$$

$$= 0 \cdot \frac{n_0}{n} + \left( x_1 \cdot \frac{n_1}{n} \right)$$



$$= 0 \cdot \frac{1}{n} + \left( \frac{1}{n} \cdot \frac{1}{n} \right)$$

$$X_1, \dots, X_n = \textcircled{0}, 1, 1, \dots, \textcircled{0}, 1$$

$$\bar{X}_{(n)} = \frac{1}{n} \sum_{i=1}^n x_i = \frac{n_1}{n}$$

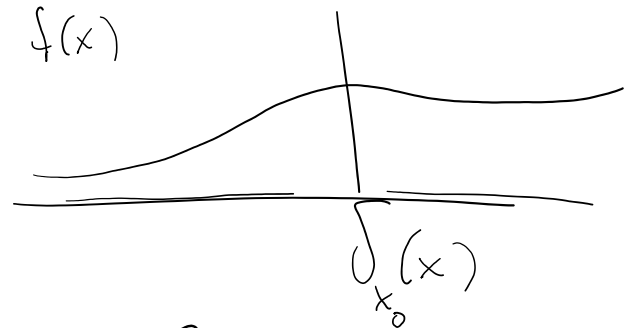
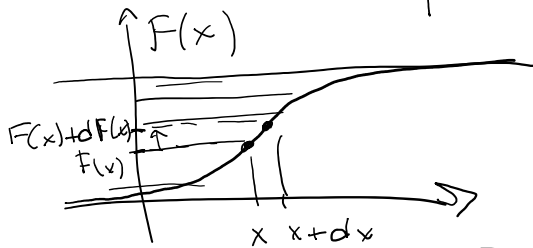
$$\rightarrow \hat{p} = \bar{X}_{(n)}$$

any  
linear  
functional

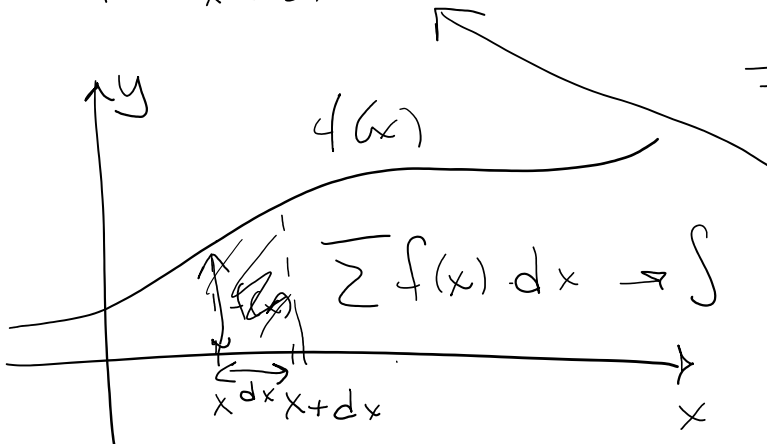
$$\left| \int r(x) dF \right| = \int r(x) d\hat{F} = \left| \frac{1}{n} \sum_{i=1}^n r(x_i) \right|$$

$$\int r(x) dF(x)$$

$\frac{dF}{dx} dx$   
 $p(x)$



$$\int f(x) \cdot \delta_{x_0}(x) dx = f(x_0)$$



$$\int r(x) dF$$

$$\int f(x) dx$$

$$E(e^x) = \int \underbrace{e^x}_{f(x)} dF$$

plug-in  
estimator  
for  $e^x$

$$= \frac{1}{n} \sum_{i=1}^n e^{x_i}$$

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$$\hat{p} = \overline{X}_n$$

$$\text{bias}(\hat{p}) = E(\hat{p}) - p = 0$$

$$E\left(\frac{1}{n} \sum_{i=1}^n x_i\right) = \frac{1}{n} \sum \underbrace{E(x_i)}_p \stackrel{Be(p)}{=} \frac{1}{n} p \cdot n = p$$

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$$2) \quad p - q = \overline{X}_n - \overline{Y}_m$$