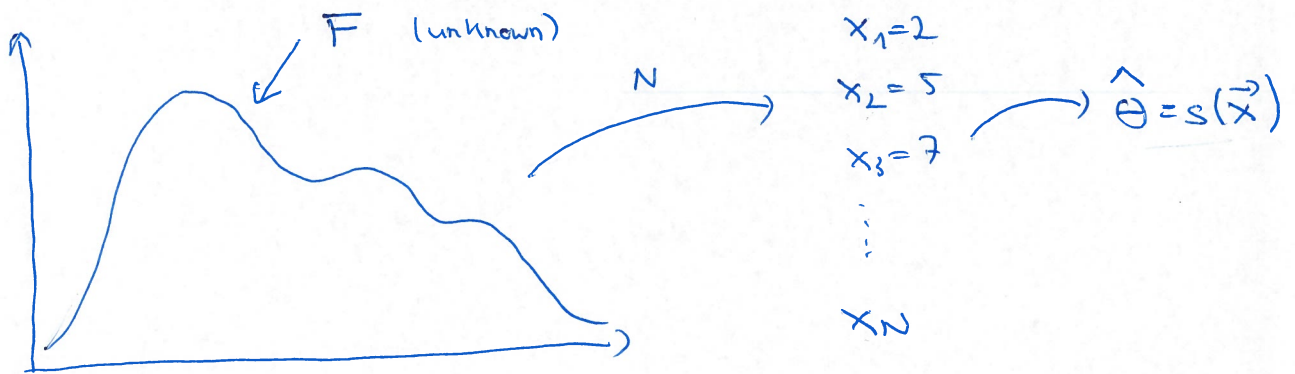


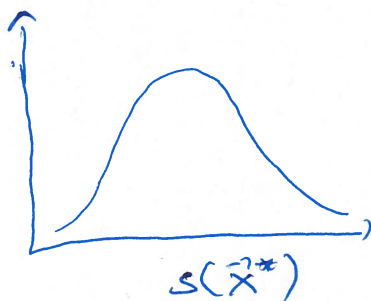
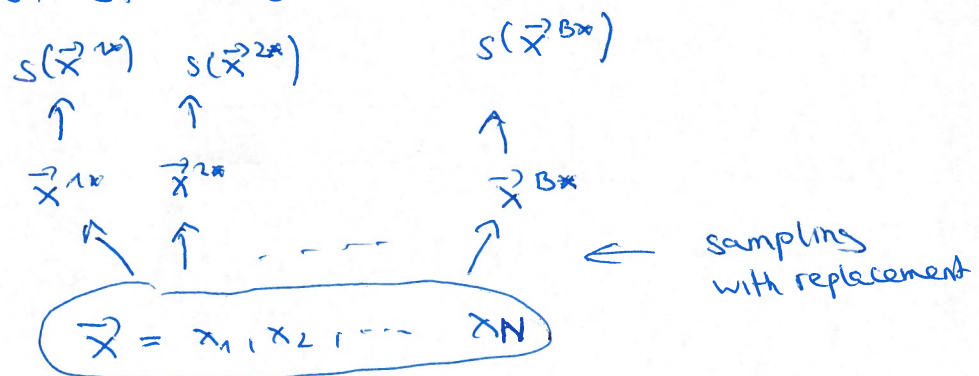
## Bootstrapping - rough idea



$$\bar{x}^{1*} = \underbrace{(x_{11}, x_{21}, x_{31}, \dots, x_{N1})}_N$$

$$s(\bar{x}^{1*}) = \hat{\theta}^{1*}$$

repeat a lot of times:



$$G_B^2 = \frac{\sum_{i=1}^B (s(x^{i*}) - \overline{s(x^*)})^2}{B-1}$$

$$\lim_{N \rightarrow \infty} \left(1 - \frac{1}{N}\right)^N = e^{-1} \approx 0.367 \Rightarrow 0.633$$

to be included  
in the bootstrap  
sample

$$\Theta = E(X)$$

$$\hat{\Theta} = s(\bar{X}) = \frac{1}{n} \sum_{i=1}^n x_i = \bar{X}$$

$$\hat{\Theta}^* = s(\bar{X}^*) = \frac{1}{n} \sum_{i=1}^n x_i^* = \bar{X}^*$$

1 ideal bootstrap estimate (using the empirical distribution function):

$$SD_{\hat{F}}(\hat{\Theta}^*) = SD_{\hat{F}}\left(\frac{1}{n} \sum_{i=1}^n x_i^*\right)$$

$$= \sqrt{\text{Var}_{\hat{F}}\left(\frac{1}{n} \sum_{i=1}^n x_i^*\right)}$$

$$= \sqrt{\frac{1}{n^2} \sum_{i=1}^n \text{Var}_{\hat{F}}(x_i^*)} = \sqrt{\frac{1}{n} \text{Var}_{\hat{F}}(x_i^*)}$$

$$\text{Var}_{\hat{F}}(x_i^*) = \sum_{i=1}^n (x_i - \bar{x})^2 \cdot \frac{1}{n}$$

$\uparrow$   $\text{E}_{\hat{F}}(x_i^*)$   $\underbrace{\frac{1}{n}}_{p(x_i)}$

Each bootstrap draw can take  $n$  possible values  $(x_1, \dots, x_n)$  and each with prob  $\frac{1}{n}$

$$\Rightarrow SD_{\hat{F}}(\hat{\Theta}^*) = \sqrt{\frac{1}{n^2} \sum_{i=1}^n (x_i - \bar{x})^2} = \sqrt{\frac{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}{n}}$$

Note: Here, we could compute the ideal bootstrap estimate analytically. However, in most cases we cannot do this

Note: Except for the factor  $\frac{1}{n}$  instead of  $\frac{1}{n-1}$  this is the usual estimate for the standard error.