

# PARAMETRIC BOOTSTRAP

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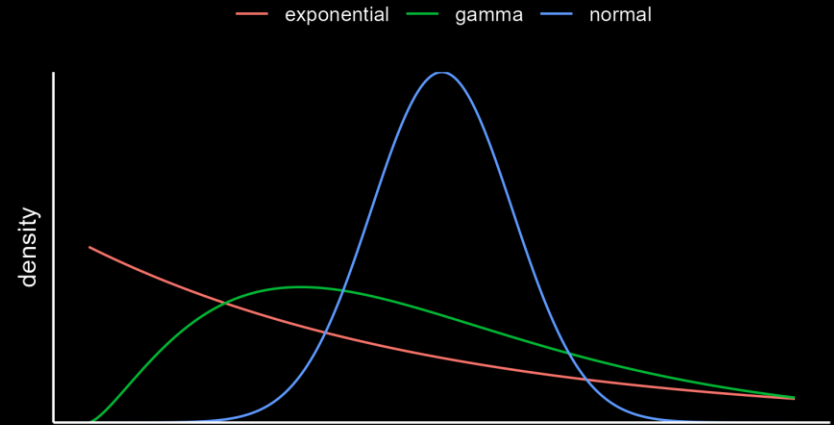
- Resampling method
- Variability of an estimator or test statistic
- Simulating data from an assumed **parametric** model

## Assumptions:

- Independent and identically distributed (iid)
- Correct model specification

# PARAMETRIC BOOTSTRAP

1. Estimate a parametric distribution function
  - Normal distribution, gamma, exponential, binomial, Poisson etc.
2. Simulate from the estimated distribution
3. Calculate a quantity of interest from each simulated data
4. Aggregate across bootstrap samples



# ESTIMATE BIAS OF THE VARIANCE ESTIMATOR

$$X_1, X_2, \dots, X_n \sim iid N(\mu, \sigma^2)$$

$$S_n^2 = \frac{1}{n} \sum (X_i - \bar{X})^2$$

$$Bias(S_n^2) = E(S_n^2) - \sigma^2 = \frac{n-1}{n} \sigma^2 - \sigma^2 = -\frac{\sigma^2}{n}$$

iid: Independent and Identically Distributed

# ESTIMATE BIAS OF THE VARIANCE ESTIMATOR

- $n = 50$
- $\mu = 10$
- $\sigma^2 = 4$

$$\text{Bias}(S_{50}^2) = -\frac{\sigma^2}{n} = -\frac{4}{50} = -0.08$$

# ESTIMATE BIAS OF THE VARIANCE ESTIMATOR

1. Estimate parameters

$$\bar{x}_{50} = 10.21 \text{ and } s_{50}^2 = 3.91$$

2. Simulate from the estimated distribution

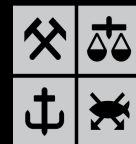
$$x_{jb} \sim N(10.21, 3.91), \\ j = 1, \dots, 50, \quad b = 1, \dots, 10\,000.$$

3. Calculate a quantity of interest from each simulated data

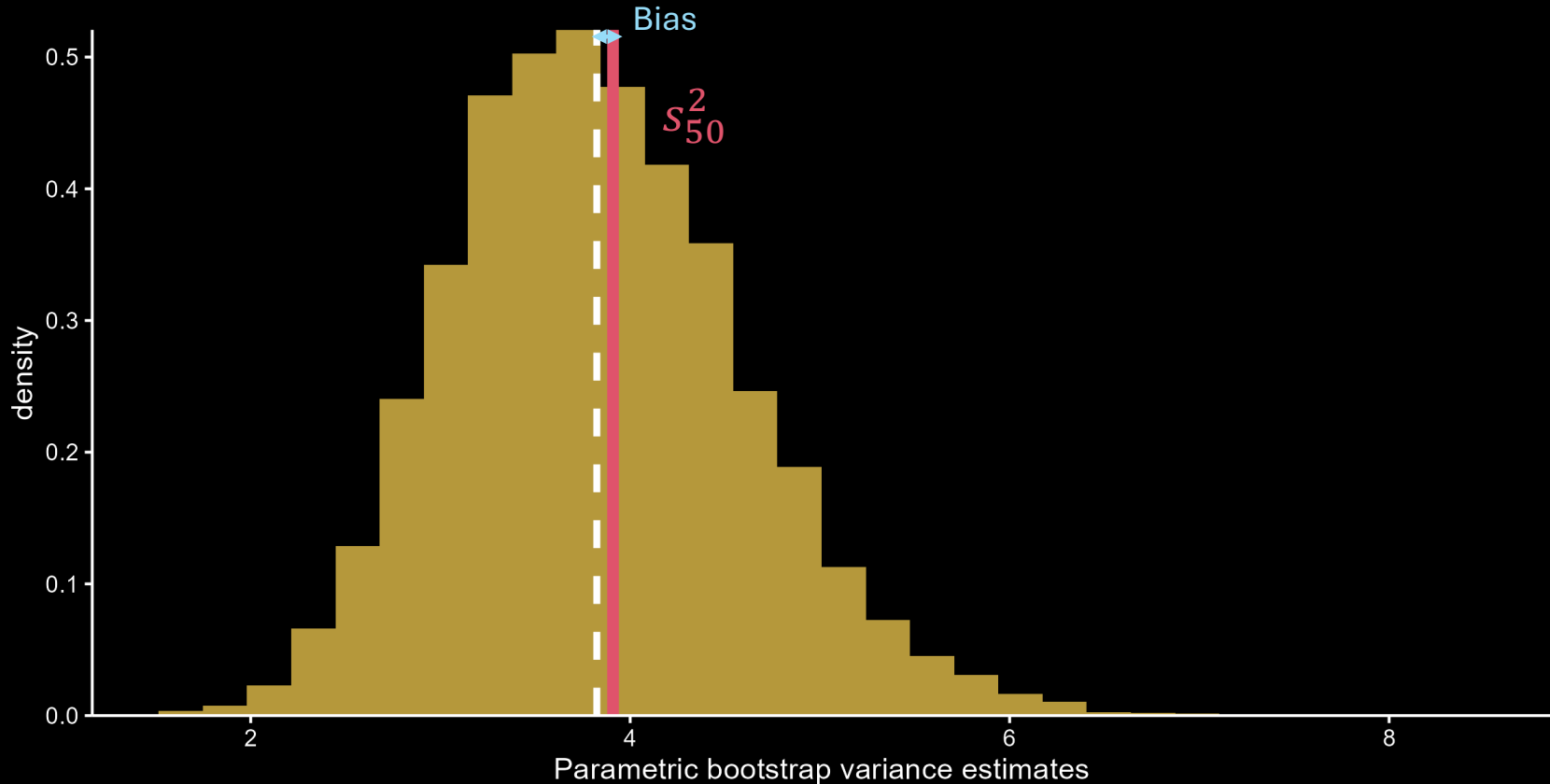
$$s_{50,b}^2 = \frac{1}{50} \sum_{j=1}^{50} (x_{jb} - \bar{x}_{50,b})^2,$$

4. Aggregate across bootstrap samples

$$\begin{aligned} & \widehat{Bias}(S_{50}^2) \\ &= \frac{1}{10\,000} \sum s_{50,b}^2 - s_{50}^2 \\ &= 3.82 - 3.91 = -0.085 \end{aligned}$$



# ESTIMATE BIAS OF THE VARIANCE ESTIMATOR



# PARAMETRIC BOOTSTRAP OR NOT?

Parametric bootstrap:

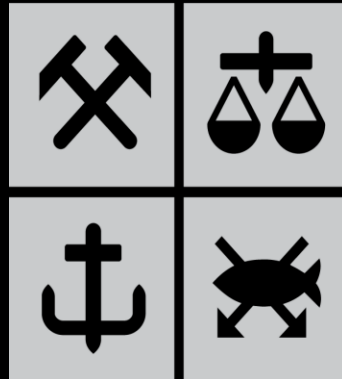
- Stronger assumptions on the data generating process
- Depend on quality of parameter estimates

Non-parametric bootstrap:

- Minimal assumptions on the data generating process



# NHH TECH3



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