

# CONFIDENCE INTERVAL ESTIMATION USING BOOTSTRAPPING METHODS AND MAXIMUM LIKELIHOOD ESTIMATE

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# Introduction

**According to Petty (2012), confidence interval (CI) is an interval estimate of a parameter of a population (e.g., a mean) calculated from a sample drawn from the population.**

- Confidence level

**Important:**

- in statistical inferences about a population parameter and**
- evaluate the accuracy of its estimator**

- more information than point estimates (Das, 2019)

95% confidence interval using the sample's mean and standard deviation assuming a normal distribution





# Objectives

- to present the steps to calculate confidence interval using SPSS
- to calculate and maximum likelihood estimate, percentile bootstrap, and bias-corrected and accelerated methods.





# Methodology

## Bootstrap methods

- Efron (1979), which can be used to increase the sample size by applying nonparametric resampling.
- The samples are used to test the statistical characteristics of the unknown distribution, such as mean, variance, standard deviation, and confidence interval (Zhang et al., 2019).
- Bootstrap is not commonly used because this method is complex to calculate (Doğan, 2017).
- The advantages: valid for small samples, and it is a convenient tool.



# Methodology

## Bootstrap methods

According to Thai et al., (2013), let  $B$  be the number of bootstrap samples to be drawn from the original dataset, a general bootstrap algorithm is:

1. Generate a bootstrap sample by resampling from the data and/or from the estimated model (Sample  $n$  elements with replacement from original sample data)
2. Obtain the estimates for all parameters of the model for the bootstrap sample eg. **mean, median etc.**
3. Repeat steps 1-2  $B$  times to obtain the bootstrap distribution of parameter estimates and then compute mean, standard deviation, and 95% confidence interval of this distribution

Replication is very important

Diciccio and Efron (1996) - 2000

Efron & Tibshirani, 1994 - than  $n^n$

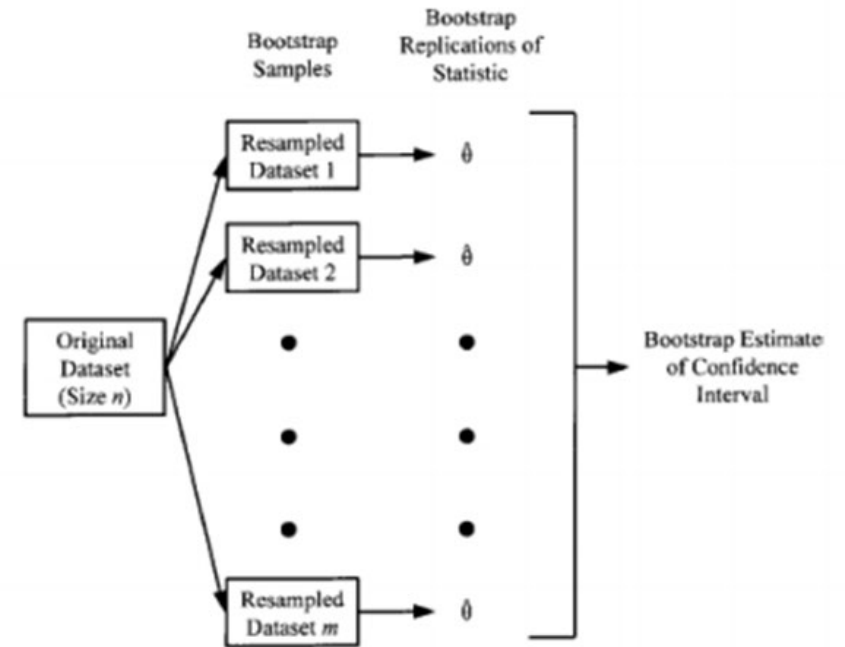


Figure 1: Description of the steps in bootstrapping.





# Methodology

**Two types of confidence interval estimation which are**

- **Likelihood ratio method is based on maximum likelihood estimation (MLE)**
  - Used in SD and built in SD software
  - The likelihood ratio method – computationally efficient but required strong assumptions about data
  - Assumptions are frequently violated in SD
  - Advantage of likelihood ratio – convenience & easy using software
  - Assumptions that are often violated

The confidence interval construction is based on asymptotic normality of the MLE (Kreutz et al., 2013).

- **Traditional CI**
- **Bootstrapping (percentile bootstrap method and BCa)**



# Methodology

Confidence interval estimation which are

- Traditional CI (Dogan, 2017)

$$\bar{x} \pm t_{\alpha/2} * \left( \frac{s}{\sqrt{n}} \right)$$

$$\bar{x} \pm 1.96 * \left( \frac{s}{\sqrt{n}} \right)$$

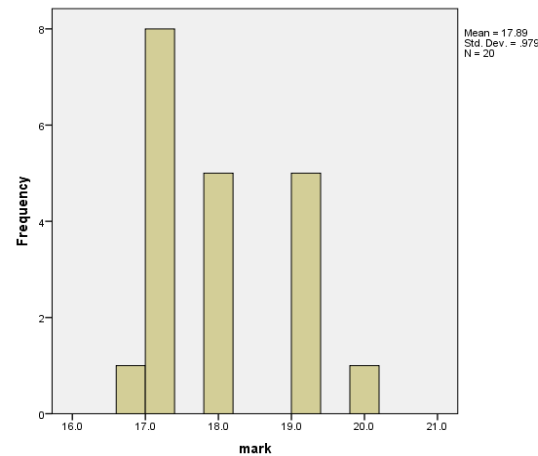
- Bootstrapping (percentile bootstrap method and BCa)
- The different methods available for estimating bootstrap confidence intervals for estimated parameters (Mesabbah et al., 2015).





# Real Data Example

20, 19, 17, 18, 17, 17, 17, 17, 17, 18, 18, 19, 17, 19, 18, 17, 19, 18, 19 and 16.8



# Findings

Table 1. percentile bootstrap (SPSS Output)

Descriptive Statistics						
		Statistic	Bootstrap <sup>a</sup>			
			Bias	Std. Error	95% Confidence Interval	
					Lower	Upper
mark	N	20	0	0	20	20
	Mean	17.890	.001	.214	17.490	18.300
	Std. Deviation	.9787	-.0345	.1195	.7016	1.1760
Valid N (listwise)	N	20	0	0	20	20

a. Unless otherwise noted, bootstrap results are based on 1000 bootstrap samples

Table 2. BCa bootstrap (SPSS Output)

Descriptive Statistics						
		Statistic	Bootstrap <sup>a</sup>			
			Bias	Std. Error	BCa 95% Confidence Interval	
					Lower	Upper
mark	N	20	0	0	.	.
	Mean	17.890	-.012	.212	17.540	18.240
	Std. Deviation	.9787	-.0308	.1133	.8013	1.1014
Valid N (listwise)	N	20	0	0	.	.

a. Unless otherwise noted, bootstrap results are based on 1000 bootstrap samples



# Findings

Methods	Point estimation		Confidence interval	Interval length
	Estimate	Dev.		
Traditional CI	17.89	0.9787	[17.5116, 18.2684]	0.7568
percentile bootstrap	17.89	0.9787	[17.490, 18.300]	0.8100
BCa	17.89	0.9787	[17.540, 18.240]	0.7000

Based on the bootstrap method, the original samples were randomly sampled  $n=1000$  times with replacement, and 1000 bootstrap samples were obtained.

The point estimation of the original sample is estimated to be  $\hat{\mu} = 17.89$ , and the 95% confidence interval for percentile bootstrap is between 17.490 and 18.300.

BCa bootstrap, the point estimation of the original sample is estimated to be  $\hat{\mu} = 17.89$ , and the 95% confidence interval between 17.540 and 18.240.





# Conclusion

-3 methods

-advantages of bootstrapping are assumptions on bootstrap are less restrictive, and more easily checked, than the assumptions on MLE.

-The bootstrap also can be applied to situations where MLE may be difficult or impossible to find.





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