

**WESTMINSTER**  
International University in Tashkent

# Week 5

# Sampling methods and sampling distribution

*By*

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Office hours: Tuesday, 09:00 – 11:00 (ATB 216)

# AGENDA

1. Probability vs Non-Probability sampling
2. Probabilistic Sampling Methods
3. Sampling Distribution
4. Central Limit Theorem (CLT)

# Reasons to sample

1. To contact the whole population would be time-consuming.

*E.g. Contacting each voter for election survey.*

2. The cost of studying all the items in a population may be out of the budget constraint. *E.g. Product test using 37 million population in Uzbekistan.*

3. The physical impossibility of checking all items in the population.

*E.g. Studying the population of birds, fish, etc.*

4. The destructive nature of some tests. *E.g. Car crash test.*

5. The sample results are adequate.

*E.g. To study the price of bread, we do not need to include all stores in the nation.*

# **Sampling methods**

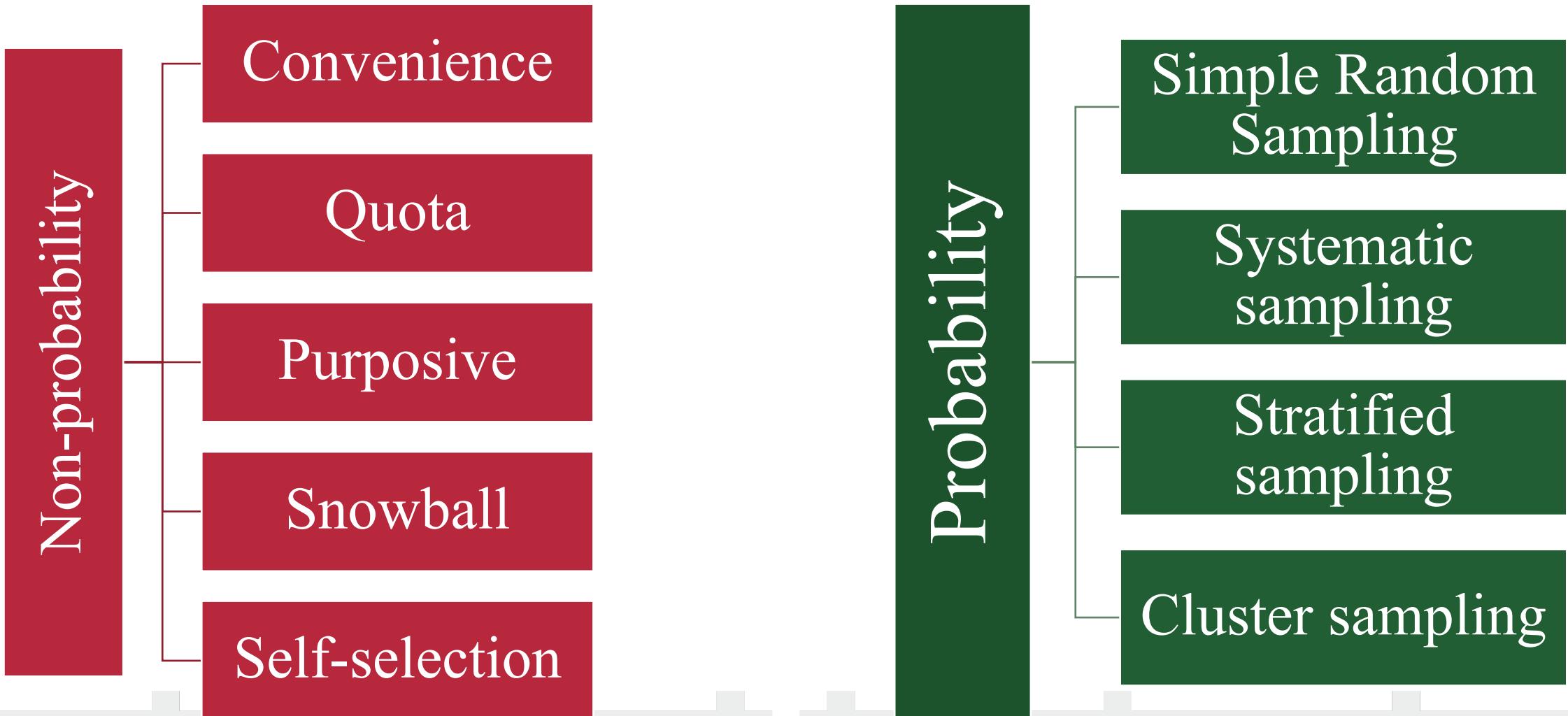
## **Probability Sampling**

- selected at random
- everyone has known chance
- useful for diverse populations
- represents the population accurately
- finding respondent is difficult and expensive

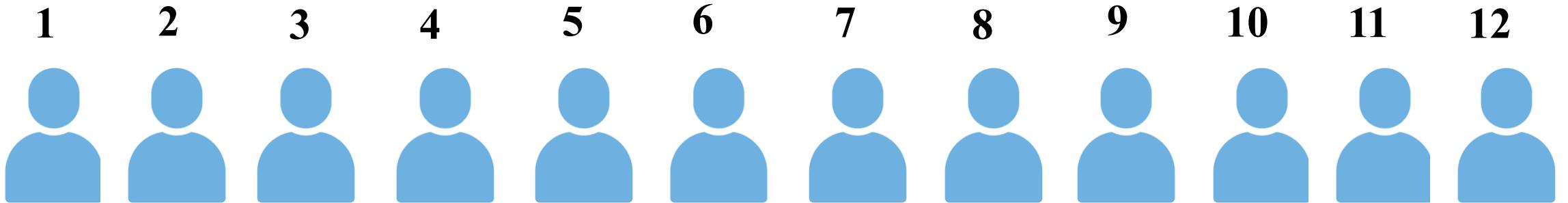
## **Non-probability Sampling**

- selected based on judgement of the researcher
- everyone does not have known chance
- useful for populations with similar traits
- unrepresentative sample
- finding respondent is easy and inexpensive

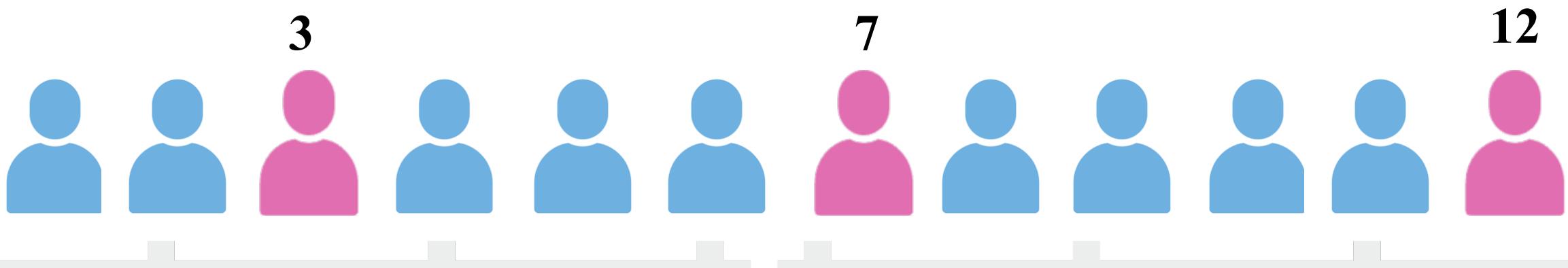
# Sampling methods



**SIMPLE RANDOM SAMPLE** A sample selected so that each item or person in the population has the same chance of being included.

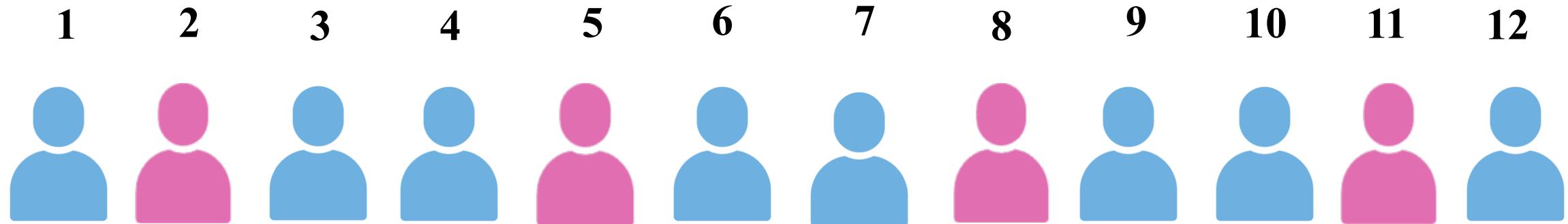


**Random selection: 3, 7, 12**

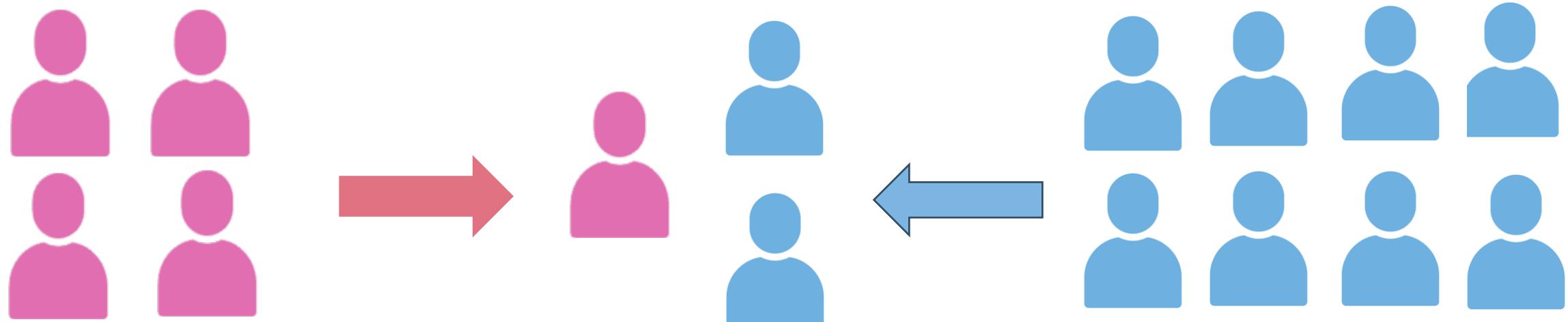


**SYSTEMATIC RANDOM SAMPLE** A random starting point is selected, and then every  $k$ th member of the population is selected.

Starting point is randomly selected (1-3): 2



**STRATIFIED RANDOM SAMPLE** A population is divided into subgroups, called strata, and a sample is randomly selected from each stratum.

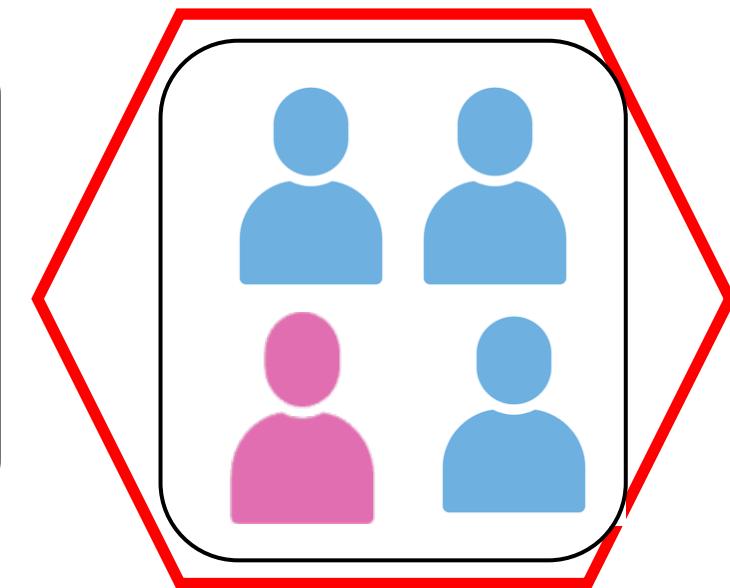
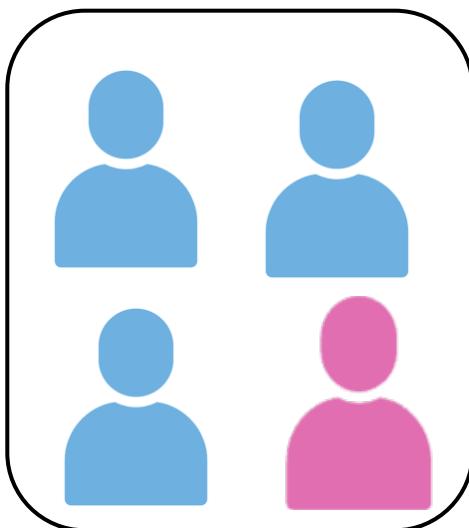
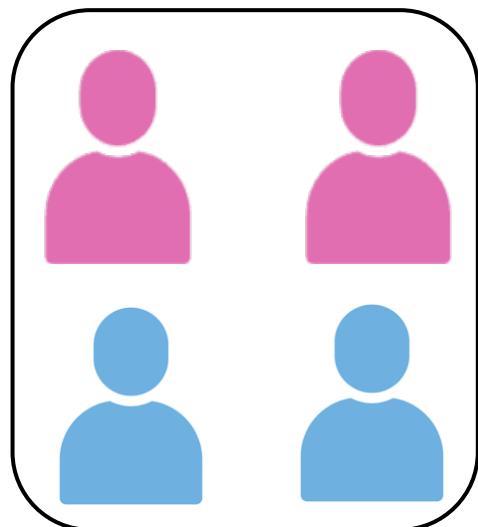


# Example: Stratified sampling

A university wants to survey student satisfaction with online learning. The sample size needs to be **200** students. There are **5,000** students in total, divided by faculty as follows:

Faculty	Total students	Percentage in population	Sample size
Business	2,000	40%	80
Engineering	1,500	30%	60
Medicine	1,000	20%	40
Arts	500	10%	20
<b>TOTAL</b>	<b>5,000</b>	<b>100%</b>	<b>200</b>

**CLUSTER SAMPLING** A population is divided into clusters using naturally occurring geographic or other boundaries. Then, clusters are randomly selected and a sample is collected by randomly selecting from each cluster.



# Example: Cluster sampling

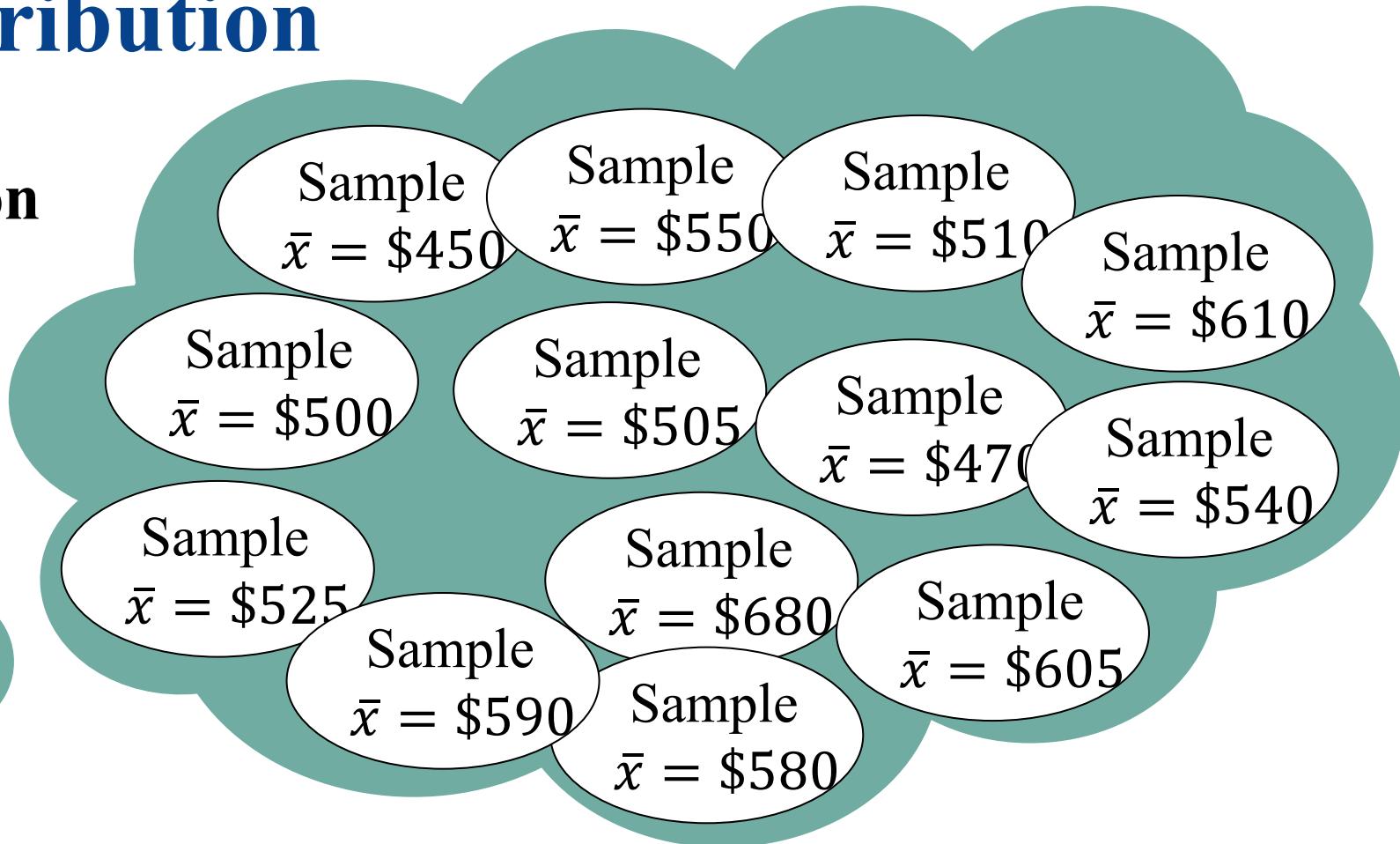
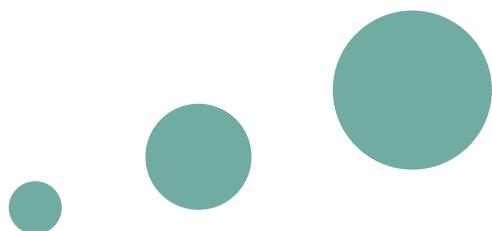
Your company has offices in all 14 regions across Uzbekistan and you want to study employee productivity in your company. You randomly select 3 offices using SRS. Then study all employees in these 3 offices.



# Sampling distribution

**Population**

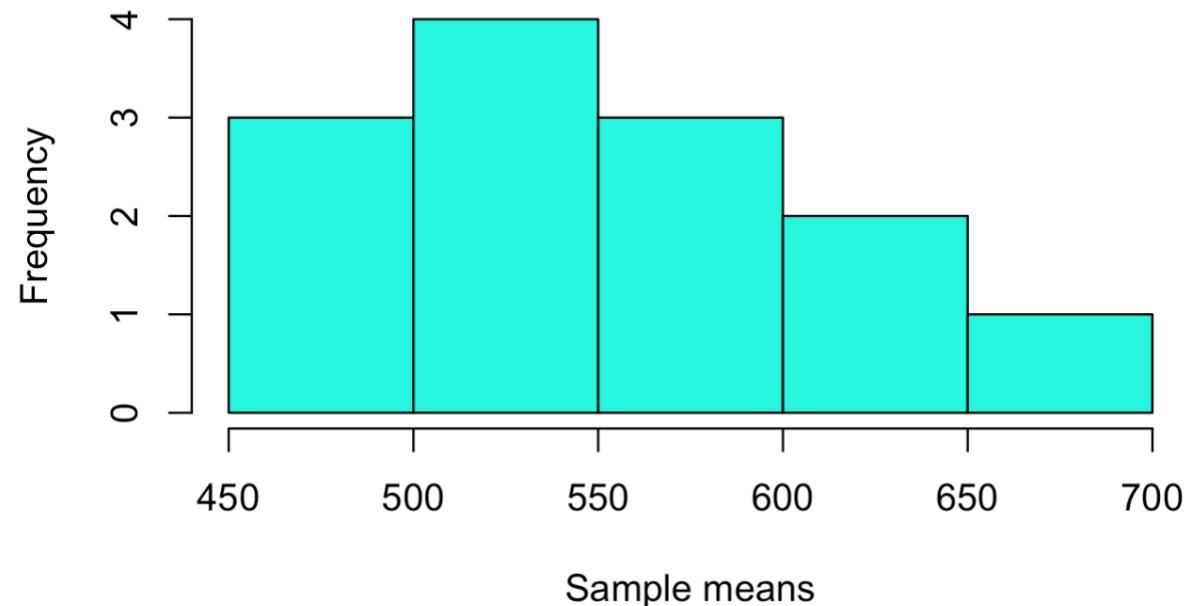
Average income  
in Uzbekistan,  
 $\mu = ?$



A **sampling distribution** is the probability distribution of a sample statistic that is formed when samples of size  $n$  are repeatedly taken from a population.

# Sampling distribution

If the sample statistic is the sample mean, then the distribution is the **sampling distribution of sample means.**



**SAMPLING ERROR** The difference between a sample statistic and its corresponding population parameter.

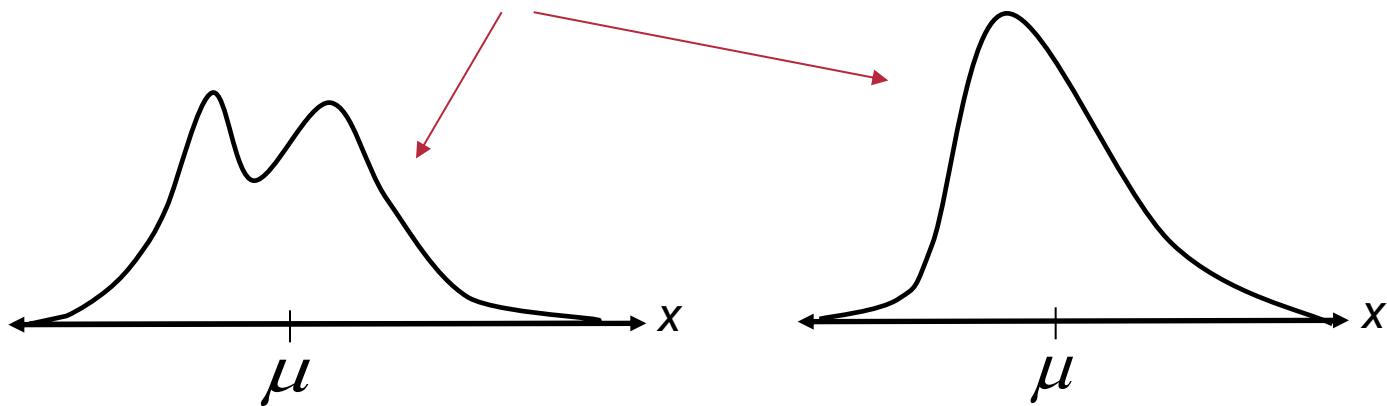
# Example: Sampling distribution

Example: Suppose we have a small population of the following values: **3, 5, 10, 12** where,  $\mu = \frac{3+5+10+12}{4} = 7.5$  (population parameter). Then you have the following sampling distribution for  $n = 2$  (without replacement) and their means:

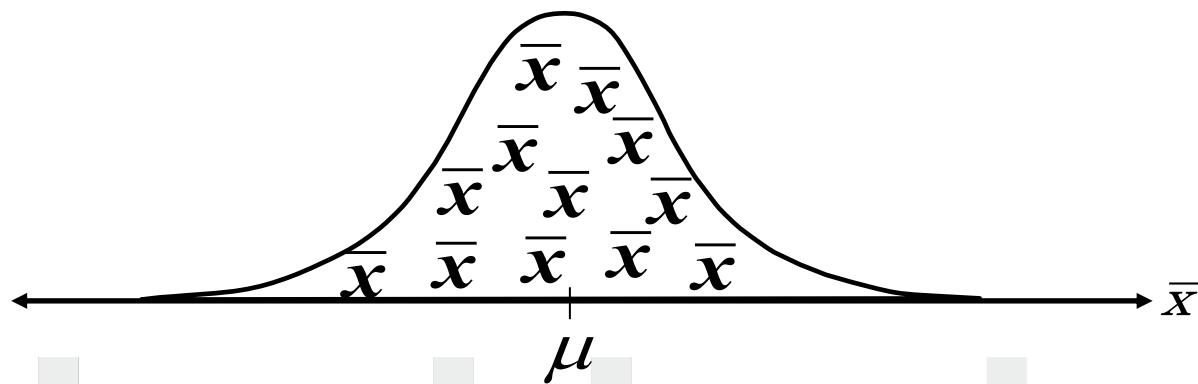
Sampling Distribution	Sampling Distribution of sample means	Sample statistics
3, 5	4	$\bar{x}_1$
3, 10	6.5	$\bar{x}_2$
3, 12	7.5	$\bar{x}_3$
5, 10	7.5	$\bar{x}_4$
5, 12	8.5	$\bar{x}_5$
10, 12	11	$\bar{x}_6$
$\mu_x$	7.5	

# Central Limit Theorem (CLT)

If a sample of size  $n \geq 30$  is taken from a population with *any type of distribution* that has a mean =  $\mu$  and standard deviation =  $\sigma$ ,

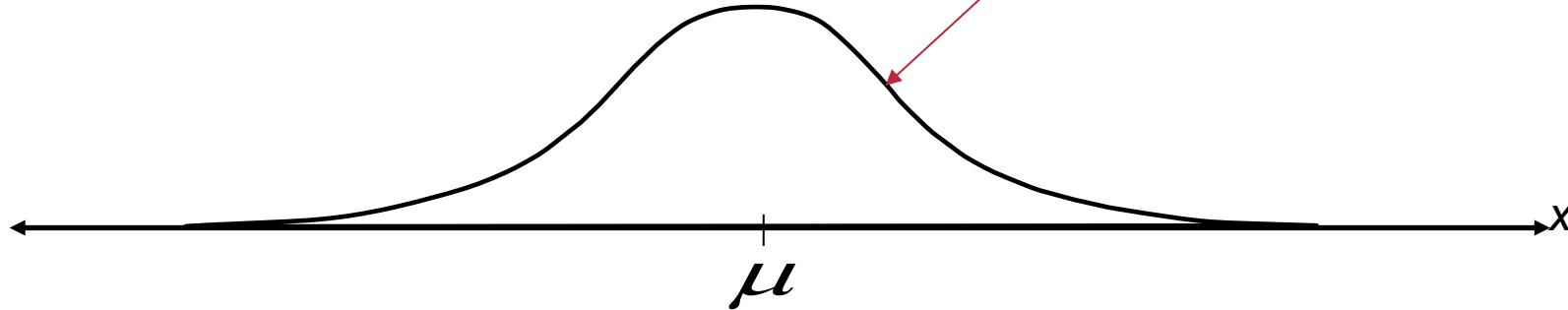


the *sample means* will have a **normal distribution**.

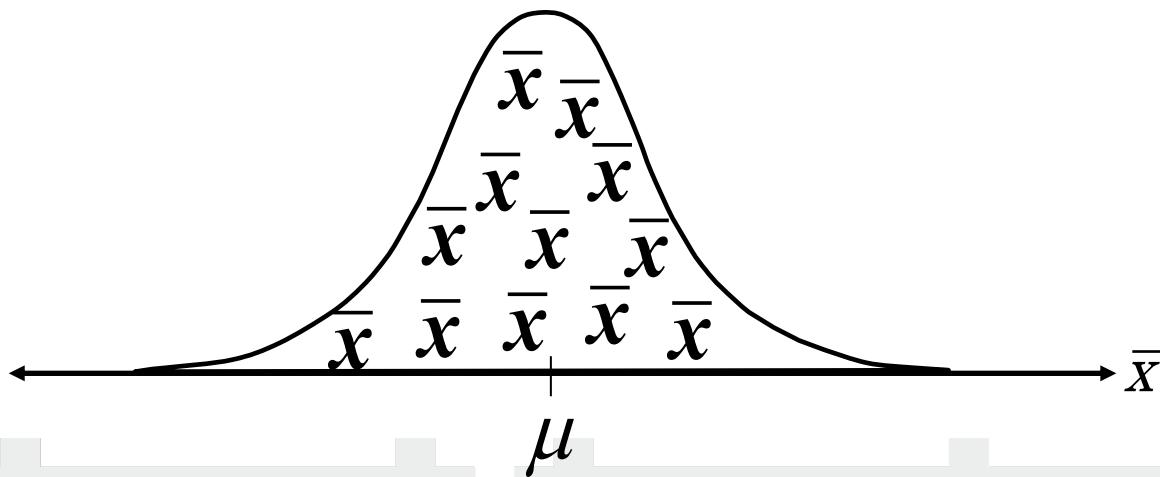


# Central Limit Theorem (CLT)

If the population itself is *normally distributed*, with mean =  $\mu$  and standard deviation =  $\sigma$ ,



the *sample means* will have a **normal distribution** for *any* sample size  $n$ .



# Central Limit Theorem (CLT)

In either case, the sampling distribution of sample means has a mean equal to the population mean.

$$\mu_{\bar{x}} = \mu \quad \text{Mean of the sample means}$$

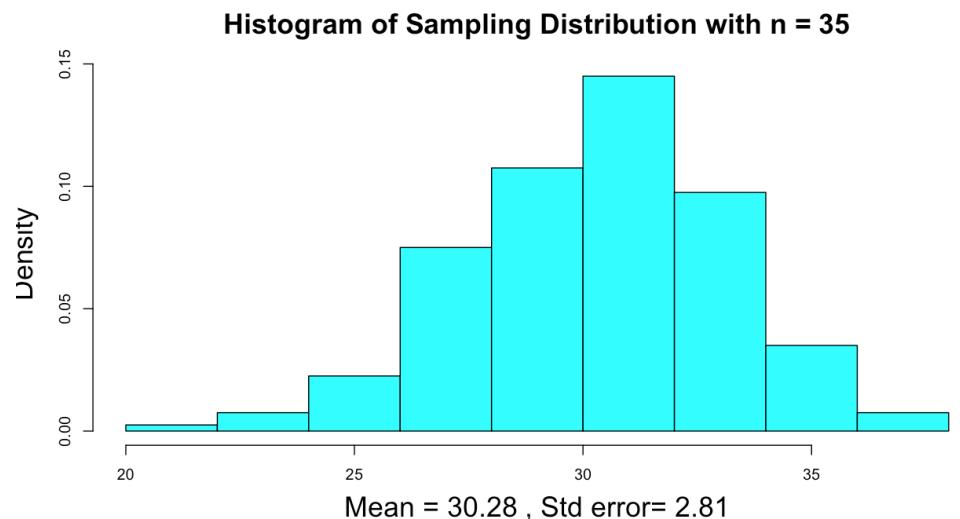
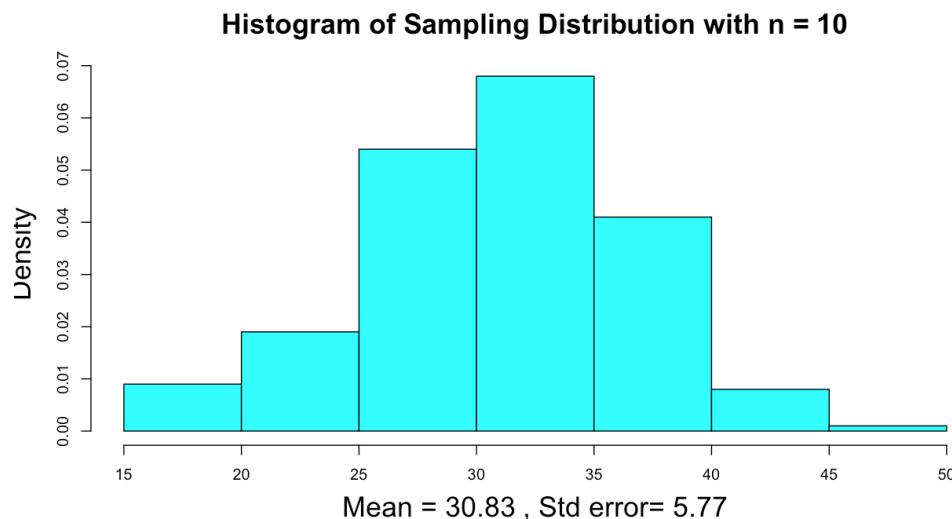
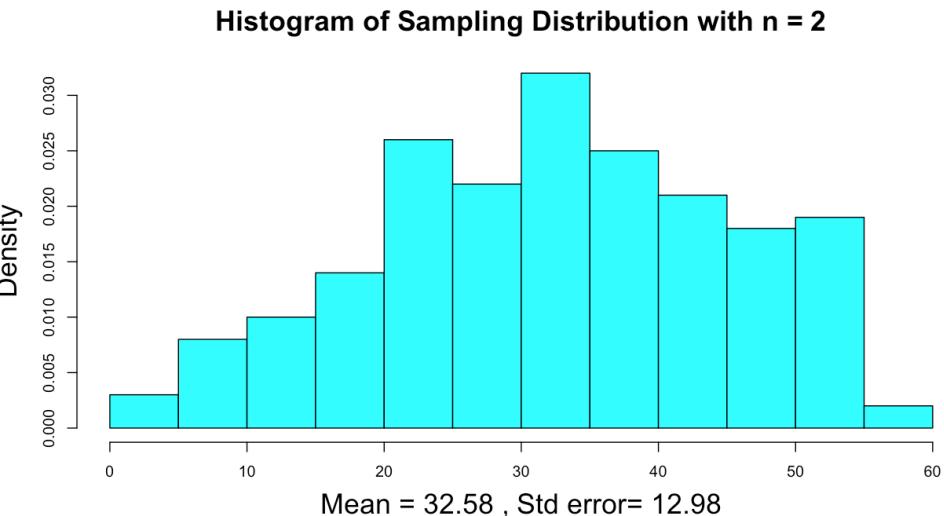
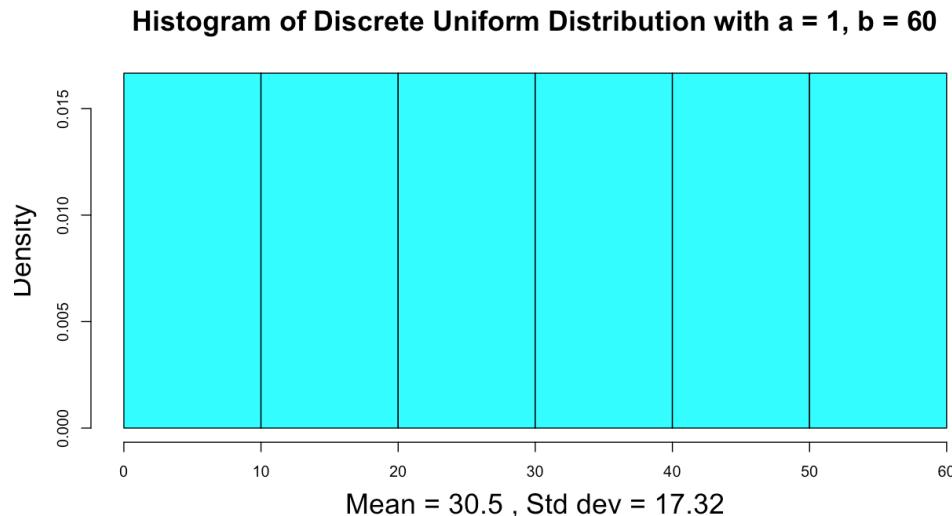
The sampling distribution of sample means has a standard deviation equal to the population standard deviation divided by the square root of  $n$ .

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

Standard deviation of the sample means (standard error)

This is also called the **standard error of the mean**.

# Simulation in R



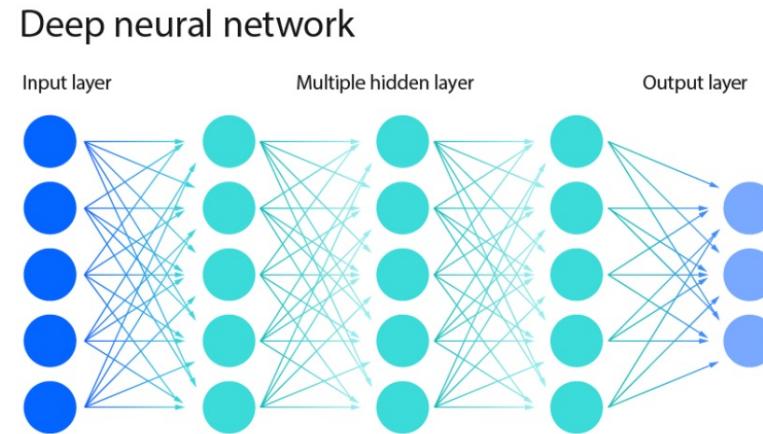
# FINDING THE $z$ VALUE OF $\bar{x}$ WHEN THE POPULATION STANDARD DEVIATION IS KNOWN

$$z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$$

## Example.

Training time for a neural network model varies due to hardware usage and random initialization and it follows a normal distribution. Assume population mean is 200 seconds and  $\sigma$  is 40 seconds.

- a. Find the probability that the *average training time* from 36 runs exceeds 210 seconds.
- b. Compare this with the probability that *a single training time* exceeds 210 seconds.
- c. Find the 90th percentile of the sampling distribution of the *average training time*.



# Solution: CLT

Training time for a neural network model varies due to hardware usage and random initialization and it follows a normal distribution. Assume population mean is 200 seconds and  $\sigma$  is 40 seconds.

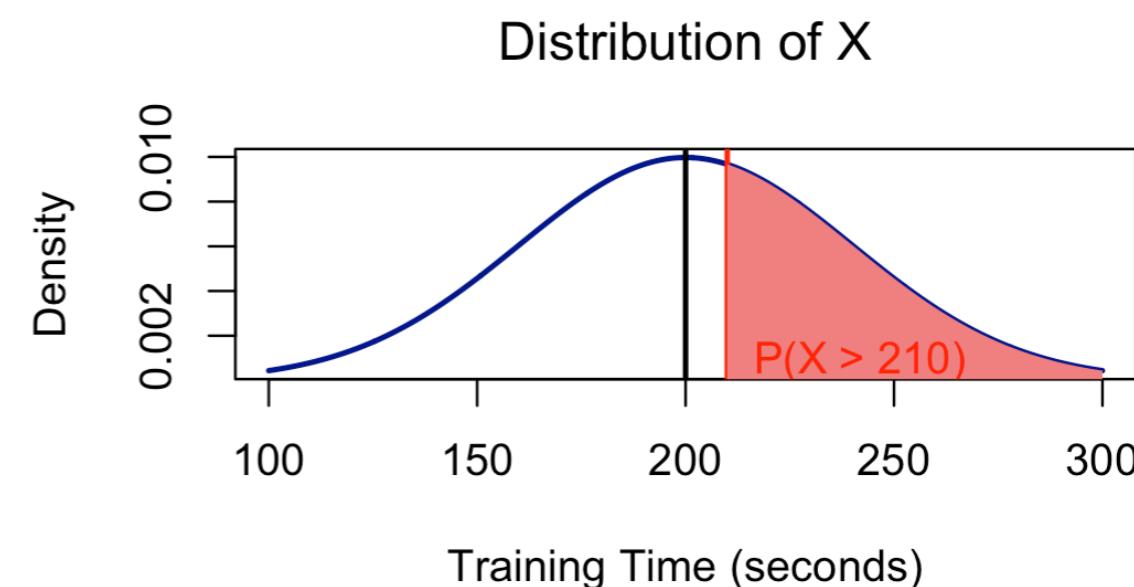
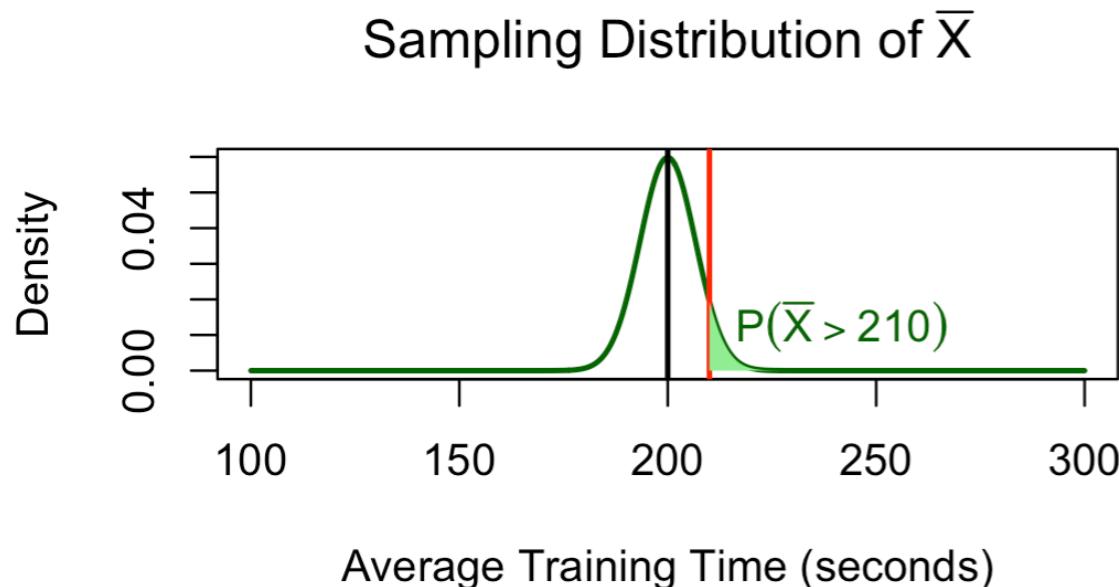
$$\sigma_{\bar{x}} = \frac{40}{\sqrt{36}} \approx 6.67$$

- a. Find the probability that the *average training time* from 36 runs exceeds 210 seconds.

$$P(\bar{x} > 210) = P(z > 1.50) = 1 - P(z < 1.50) = 0.0668$$

- b. Compare this with the probability that *a single training time* exceeds 210 seconds.

$$P(X > 210) = P(z > 0.25) = 0.4013$$

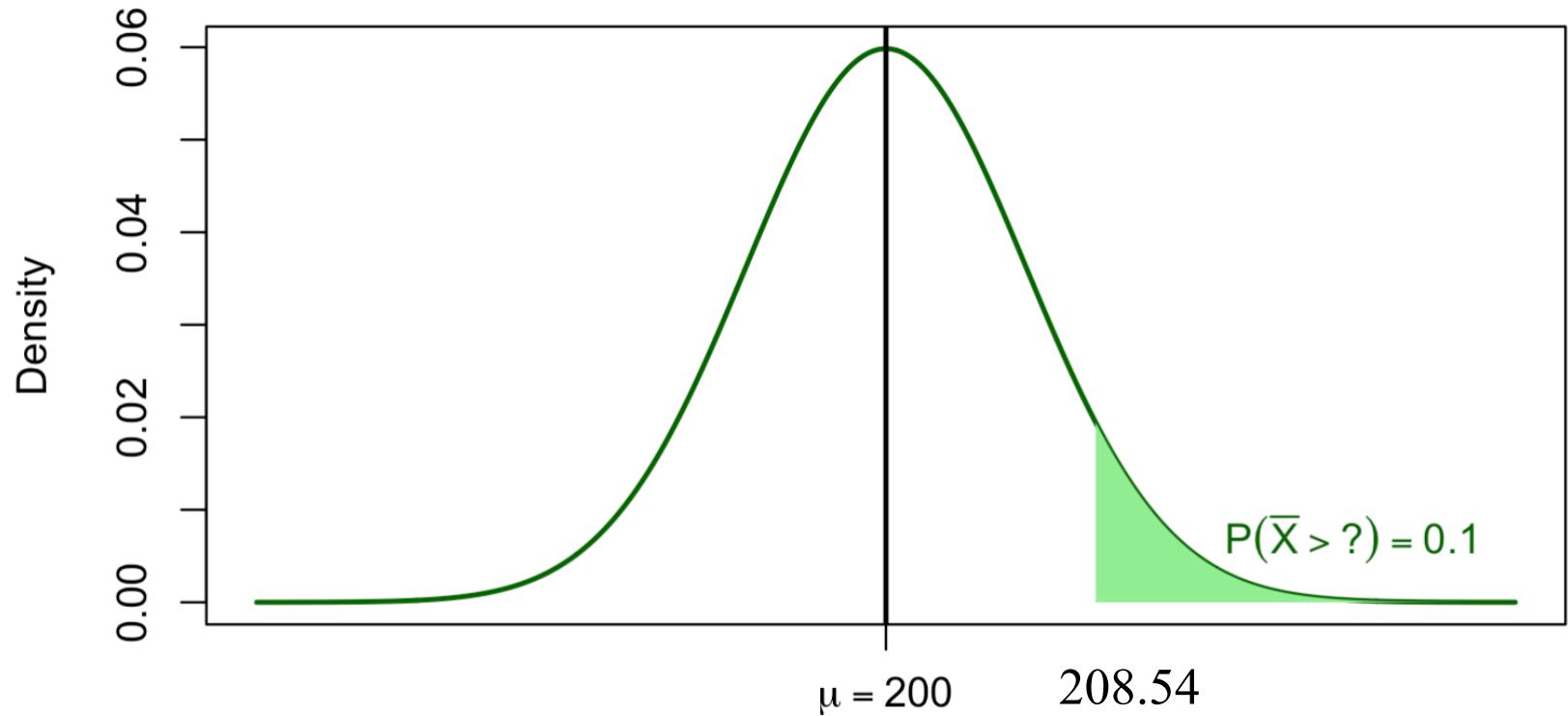


# Solution: CLT

c. Find the 90th percentile of the sampling distribution of the *average training time*.

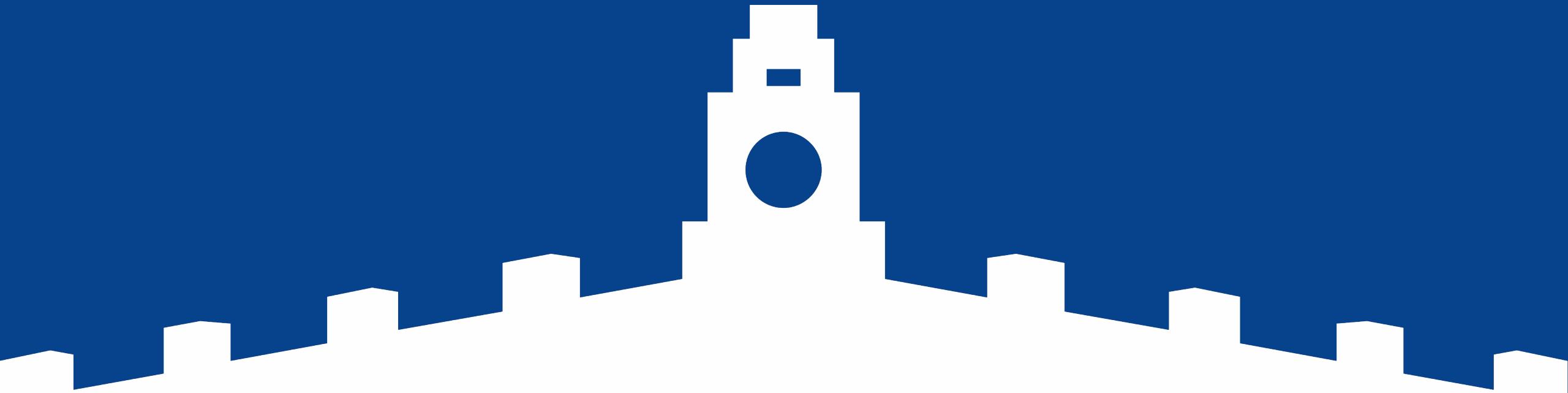
$$z = \frac{\bar{x} - 200}{6.67} = 1.28$$

$$\begin{aligned}\bar{x} &= 200 + 1.28 * 6.67 = \\ &= 208.54\end{aligned}$$



# REFERENCES

1. Lind et al. (ISBN 978-1-260-18750-2), Chapter 8.
2. McClave & Sincich (ISBN 978-0-321-75593-3), Chapter 6.
3. Ott & Longnecker (ISBN 978-0-495-01758-5), Chapter 2.



# Thank You!