

Lecture 15 Sampling Distribution And The Central Limit Theorem

BIO210 Biostatistics

Xi Chen

Fall, 2023

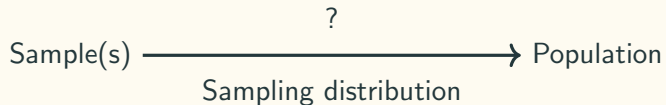
School of Life Sciences

Southern University of Science and Technology

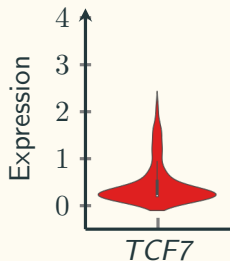
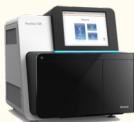
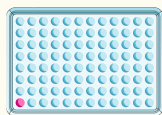


南方科技大学生命科学学院
SUSTech · SCHOOL OF
LIFE SCIENCES

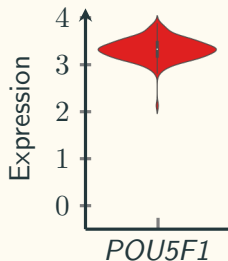
Use Sample Statistics To Estimate Population Parameters



Intuition of Sampling Distribution



→ Not
T
cells



→ Probably
ES cells

Experiment_name

```
graph LR
    plate_1[plate_1] --- cell_1[cell_1]
    plate_1 --- cell_2[cell_2]
    plate_1 --- cell_3[cell_3]
    plate_1 --- dots1[...]
    plate_2[plate_2] --- cell_4[cell_1]
    plate_2 --- cell_5[cell_2]
    plate_2 --- cell_6[cell_3]
    plate_2 --- dots2[...]
```

plate_1

cell_1

cell_2

cell_3

⋮

plate_2

cell_1

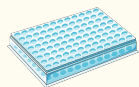
cell_2

cell_3

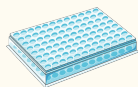
⋮

Intuition of Sampling Distribution

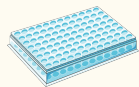
100 plates (samples)



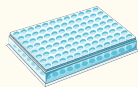
$n = 96$
 $\bar{x} = 3.07$



$n = 96$
 $\bar{x} = 2.97$

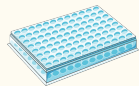


$n = 96$
 $\bar{x} = 2.80$



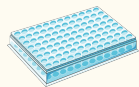
$n = 96$
 $\bar{x} = 4.95$

\vdots

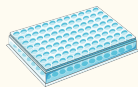


$n = 96$
 $\bar{x} = 3.60$

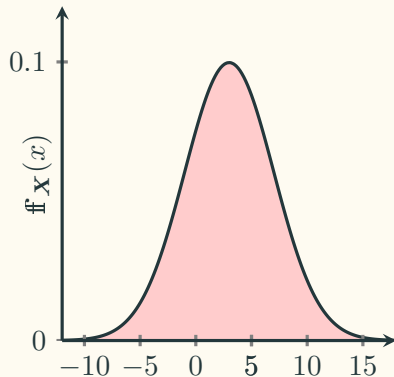
\vdots



$n = 96$
 $\bar{x} = 1.05$

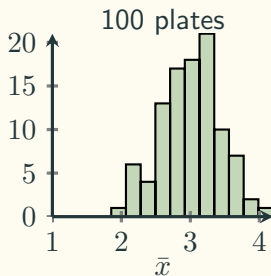
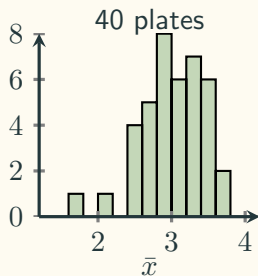
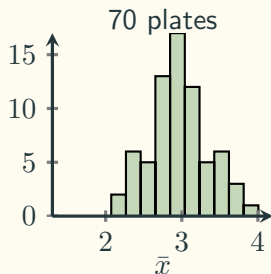
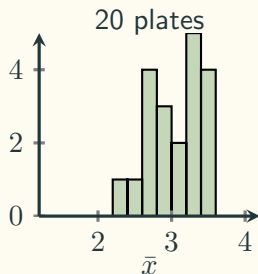
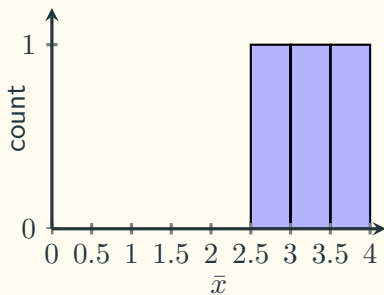
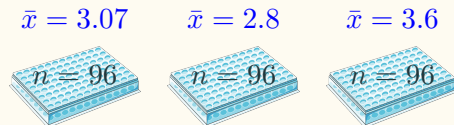


$n = 96$
 $\bar{x} = 3.27$

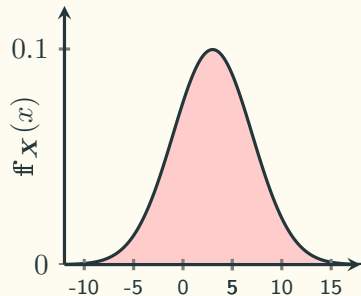


$$POU5F1 \sim \mathcal{N}(\mu = 3, \sigma^2 = 4^2)$$

Intuition of Sampling Distribution

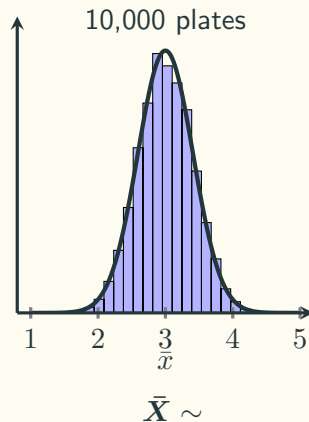


Sampling Distribution of The Sample Mean



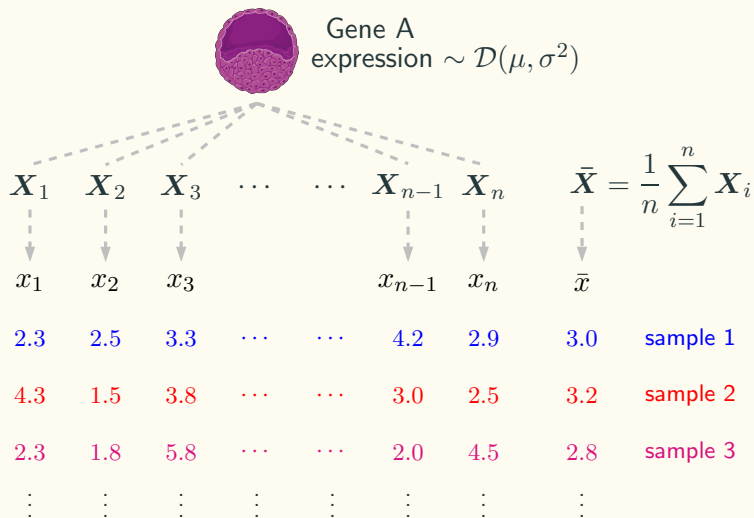
$$POU5F1 \sim \mathcal{N}(\mu = 3, \sigma^2 = 4^2)$$

Sample	Sample mean
$n = 96$	$\mapsto \bar{x} = 3.05$
$n = 96$	$\mapsto \bar{x} = 3.13$
$n = 96$	$\mapsto \bar{x} = 3.15$
$n = 96$	$\mapsto \bar{x} = 2.95$
$n = 96$	$\mapsto \bar{x} = 2.87$
$n = 96$	$\mapsto \bar{x} = 3.20$
\vdots	\vdots
$n = 96$	$\mapsto \bar{x} = 2.50$
\vdots	\vdots



**Sampling distribution
of the sample mean**

i.i.d. Random Variables



X_1, X_2, \dots, X_n are independent and identically distributed (**i.i.d.**) random variables.

$$X_1 \sim \mathcal{D}(\mu, \sigma^2)$$

$$X_2 \sim \mathcal{D}(\mu, \sigma^2)$$

$$X_3 \sim \mathcal{D}(\mu, \sigma^2)$$

\vdots

$$X_{n-1} \sim \mathcal{D}(\mu, \sigma^2)$$

$$X_n \sim \mathcal{D}(\mu, \sigma^2)$$

$$\bar{X} \sim ?(?, ?)$$

The Central Limit Theorem

By Pierre Simon de Laplace in 1810.

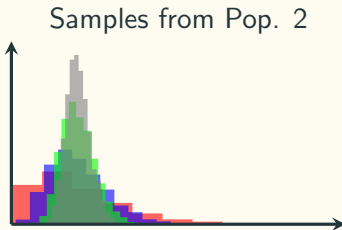
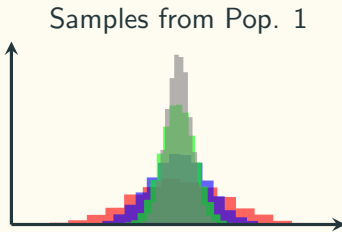
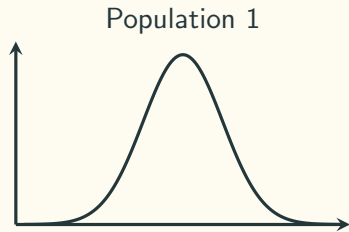
Theorem

The sampling distribution of the sample mean of n independent and identically distributed (i.i.d.) random variables is approximately normal, even if original variables themselves are not normally distributed, provided that n is large enough.

$$\bar{X} \sim \mathcal{N}(\mu_{\bar{X}}, \sigma_{\bar{X}}^2), \text{ where } \mu_{\bar{X}} = \mu, \sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$$

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}: \text{ standard error.}$$

The Central Limit Theorem



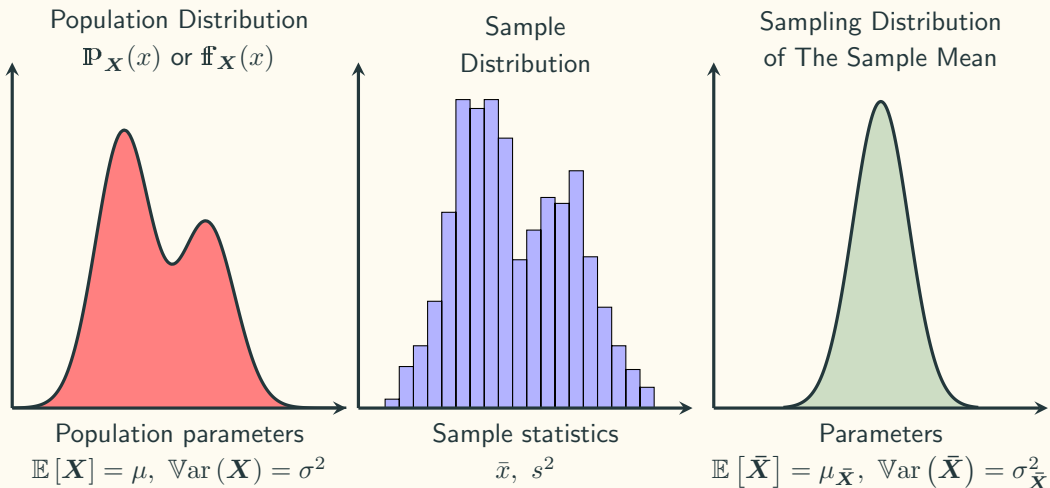
$n = 2$

$n = 5$

$n = 15$

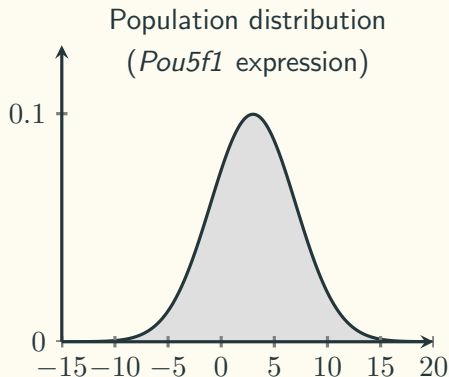
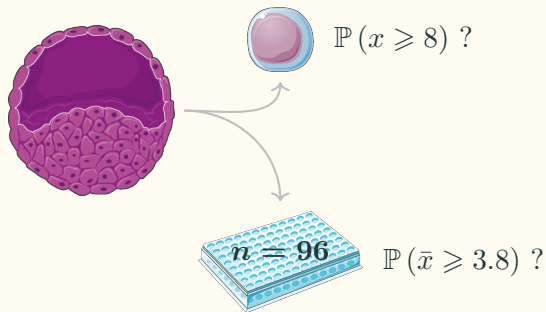
$n = 30$

Three Distributions

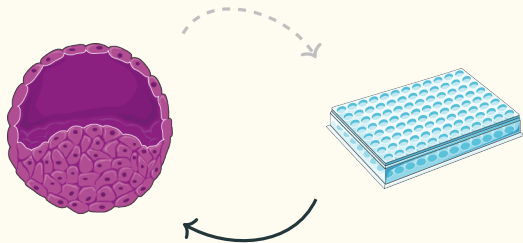


Practice: *Pou5f1* Expression

Based on the previous research, the expression of *Pou5f1* in all ES cells follow a normal distribution with $\mu = 3$ and $\sigma^2 = 4^2$.



Estimation



Use info. from the sample
to do a **point estimation**

Population parameter
 μ, σ^2

Sample statistics
 \bar{x}, s^2

- **Estimator**

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

- **Estimate**

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

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Unbiased Estimator

We say the following estimators are unbiased estimators:

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

Because:

$$\mathbb{E} [\bar{X}] = \mu$$

$$\mathbb{E} [S^2] = \sigma^2$$