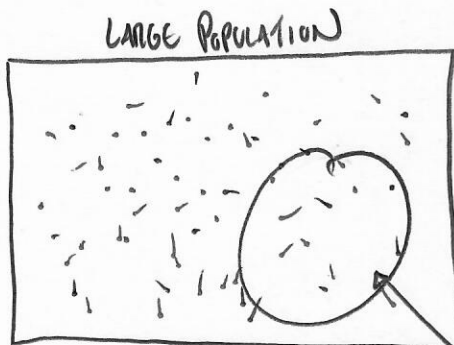


§ 7.6 THE SAMPLING DISTRIBUTION OF THE SAMPLE PROPORTION



- SUCCESS
- FAILURE

POPULATION PROPORTION
(PROPORTION OF SUCCESSSES IN POPULATION)

$$p = \frac{\# \text{ SUCCESSSES}}{N}$$

POPULATION SIZE = N

SAMPLE PROPORTION $\hat{p} = \frac{\# \text{ SUCCESSSES IN SAMPLE}}{\text{SAMPLE SIZE}} = \frac{x}{n}$

NOW TAKE A RANDOM SAMPLE OF SIZE n &
COUNT THE # OF SUCCESSSES IN THE SAMPLE, x .

IF $\frac{n}{N} \leq .05$ THEN x IS APPROXIMATELY BINOMIAL.

(OTHERWISE IT'S HYPERGEOMETRIC)

$\left(\begin{matrix} 00 \\ 000 \end{matrix} \right)$ PICK 2 MARBLES,
 $x = \# \text{ RED.}$

$\left(\begin{matrix} 2 \text{ MILL.} \\ 3 \text{ MILL.} \end{matrix} \right)$ PICK 2 MARBLES
 $x = \# \text{ RED}$

| x | 0 | 1 | 2 |
|--------|-----------------------------|-----------------------------|-----------------------------|
| $P(x)$ | $\frac{C_0^2 C_3^3}{C_2^5}$ | $\frac{C_1^2 C_1^3}{C_2^5}$ | $\frac{C_2^2 C_0^3}{C_2^5}$ |

| x | 0 | 1 | 2 |
|--------|-----------------|-----------------|-----------------|
| $P(x)$ | $C_0^2 p^0 q^2$ | $C_1^2 p^1 q^1$ | $C_2^2 p^2 q^0$ |

BINOMIAL R.V. # TRIALS = n

PROB. SUCCESS p

PROB. FAILURE $q = 1 - p$

$$np \geq 5$$

$$nq \geq 5$$

APPROX NORMAL WITH

MEAN $\mu = np$

STAND. DEV. $\sigma = \sqrt{npq}$

SO, GIVEN A LARGE POPULATION WITH PROPORTION OF SUCCESS p ,

$X = \#$ OF SUCCESSSES IN A SAMPLE OF SIZE n ($\frac{n}{N} \leq .05$)

IS APPROXIMATELY NORMAL WITH MEAN $\mu = np$

STAND. DEV. $\sigma = \sqrt{npq}$

DIV. BY n

DIV BY n

\Rightarrow SAMPLE PROPORTION $\hat{p} = \frac{X}{n}$ \leftarrow # SUCCESSSES IN SAMPLE
 \leftarrow SIZE OF SAMPLE

IS APPROX. NORMALLY DISTRIBUTED WITH MEAN p

AND STANDARD DEV = STANDARD ERROR S.E. $\sqrt{\frac{pq}{n}}$

ex. SUPPOSE A MANUFACTURING PLANT PRODUCES 1,000 DEVICES.
 2% OF THE DEVICES PRODUCED ARE DEFECTIVE.
 IN AN ORDER OF 100 DEVICES, FIND THE PROBABILITY
 THAT LESS THAN 3% OF THE DEVICES ARE DEFECTIVE.

POPULATION: ALL DEVICES

$$p = .02 \quad (\text{DEFECTIVE})$$

$$q = .98 \quad (\text{NOT DEFECTIVE})$$

SAMPLE:

$$\text{SIZE: } n = 100$$

SAMPLE PROPORTION \hat{p}
 "P-HAT"

\hat{p} IS APPROXIMATELY NORMALLY DISTRIBUTED WITH

$$\text{MEAN } p = .02$$

$$\text{STANDARD ERROR S.E.} = \sqrt{\frac{pq}{n}} = \sqrt{\frac{(.02)(.98)}{100}}$$

$$\text{FIND } P(\hat{p} < .03) = P(Z < .71) = .7611$$

$$Z = \frac{\text{RAN. VAR.} - \text{MEAN}}{\text{S.E.}}$$

$$Z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}} = \frac{.03 - .02}{\sqrt{\frac{(.02)(.98)}{100}}}$$

7.45

POPULATION: ALL MEM'S

PROPORTION OF POP.
THAT ARE BROWN

$$\text{POP PROP. } p = .13$$

$$q = .87$$

PACKAGE = RANDOM SAMPLE OF SIZE $n = 55$

$X = \#$ BROWN MEM'S IN PACKAGE

(a)

Z APPROX. NORMAL RANDOM VARIABLE

$$\hat{p} = \frac{X}{n} \quad \text{PROPORTION OF BROWN MEM'S IN THE SAMPLE} \quad \left(\begin{array}{l} \text{SAMPLE} \\ \text{PROPORTION} \end{array} \right)$$

↑ APPROX. NORMAL DISTRIBUTION WITH

MEAN $p = .13$

$$\text{S.E. } \sqrt{\frac{pq}{n}} = \sqrt{\frac{(.13)(.87)}{55}} = .0453$$

$$(b) \quad P(\hat{p} < .2) = P\left(Z < \frac{.2 - .13}{\sqrt{\frac{(.13)(.87)}{55}}}\right)$$

$$Z = \frac{\text{R.V.} - \text{MEAN}}{\text{S.E.}} = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}}$$

$$= P(Z < 1.54) = .9382$$

13.30

$$(c) P(\hat{p} > .35) = P\left(z > \frac{.35 - .13}{\sqrt{\frac{(.13)(.87)}{55}}}\right)$$

$$= P(z > 4.85) = 1 - P(z < 4.85)$$

$$\approx 1 - 1 = \boxed{0} .0000000357$$

$$P(z \leq z_0) = \begin{cases} \approx 0 & \text{IF } z_0 < -3.49 \\ \text{TABLE} & \text{IF } -3.49 \leq z_0 \leq 3.49 \\ \approx 1 & \text{IF } z_0 > 3.49 \end{cases}$$

(d) BY EMPIRICAL RULE

\hat{p} WILL BE WITHIN 2 S.E.'S OF MEAN

$$p - \overset{1.96}{\textcircled{2}} \text{S.E.} \leq \hat{p} \leq p + \overset{1.96}{\textcircled{2}} \text{S.E.}$$

95% OF THE TIME

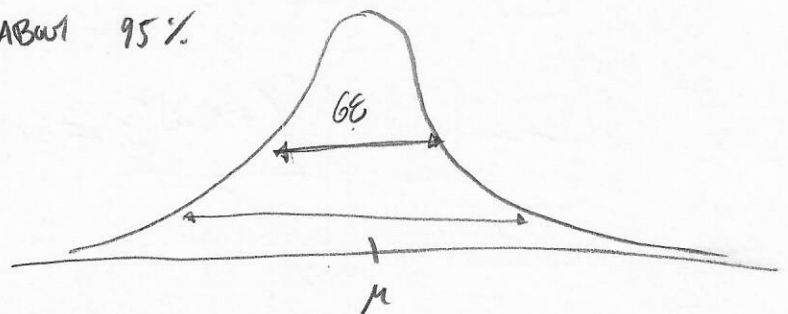
$$.13 - 2(.0453)$$

$$.13 + 2(.0453)$$

$$.0394$$

$$.2206$$

ABOUT 95%



§ 8.3 TYPES OF ESTIMATORS

INFERENCEAL STATISTICS: USING KNOWLEDGE ABOUT A SAMPLE
TO MAKE (INFER) ESTIMATES
ABOUT A POPULATION.

IN REALITY, POPULATION PARAMETERS

- MEAN μ
 - PROPORTION p
 - STANDARD DEV. σ
- } USUALLY UNKNOWN.

SO WE USE SAMPLE STATISTICS

- SAMPLE MEAN \bar{x}
 - SAMPLE PROPORTION \hat{p}
 - SAMPLE STAND. DEV. s
- } TO ESTIMATE THE POPULATION PARAMETERS.

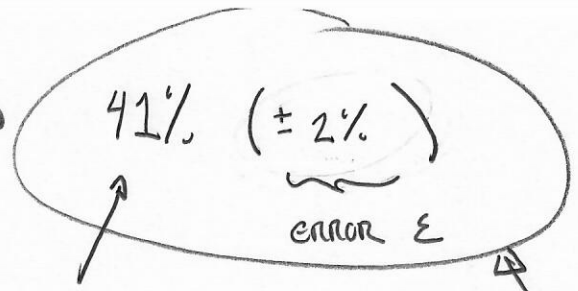
OR WE USE INTERVALS

- $\bar{x} \pm \epsilon$
 - $\hat{p} \pm \epsilon$
 - $s \pm \epsilon$
- ↖ POINT ESTIMATES
- } TO ESTIMATE POPULATION PARAMETERS

"EPSILON"
E FOR ERROR,

↖ INTERVAL ESTIMATES

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SAMPLE PROPORTION \hat{p}

POINT ESTIMATE

INTERVAL ESTIMATE

39% - 43%

§ 8.4 POINT ESTIMATES

Def: AN ESTIMATOR OF A PARAMETER IS UNBIASED IF
THE MEAN OF ITS DISTRIBUTION IS THE SAME AS
THE VALUE OF THE PARAMETER.
OTHERWISE, IT IS BIASED.

WE WANT TO USE UNBIASED ESTIMATORS.

WE KNOW (CENTRAL LIMIT THEOREM)

SAMPLE MEAN \bar{X} HAS A (NORMAL)

DISTRIBUTION WITH THE SAME MEAN AS THE
POPULATION.

§ 8.3 TYPES OF ESTIMATORS

POPULATION PARAMETERS (μ, ρ, σ) USUALLY UNKNOWN, SO WE

USE SAMPLE STATISTICS AS POINT ESTIMATES. (\bar{x}, \hat{p}, s) OR

AS INTERVAL ESTIMATES $(\bar{x} \pm E, \hat{p} \pm E)$.

POP. PARAM. ARE APPROX.
EQUAL TO THESE POINTS

CONFIDENT THAT POP. PARAM.
LIE IN THESE INTERVALS

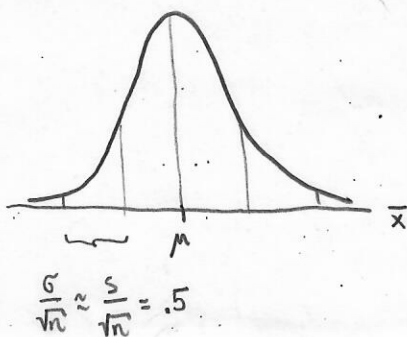
§ 8.4 Point Estimation

Def. AN ESTIMATOR OF A PARAMETER IS UNBIASED IF MEAN OF ITS
(SAMPLE) DISTRIBUTION IS SAME AS TRUE VALUE OF PARAMETER.
OTHERWISE BIASED

WE WANT TO USE UNBIASED ESTIMATORS.
PREFERABLY WITH SMALL VARIANCE.

CLT SAYS SAMPLE MEAN \bar{x}
IS DISTRIBUTION WITH SAME
MEAN AS POPULATION!

e.g. $n = 100$
 $\bar{x} = 18$
 $s = 5$

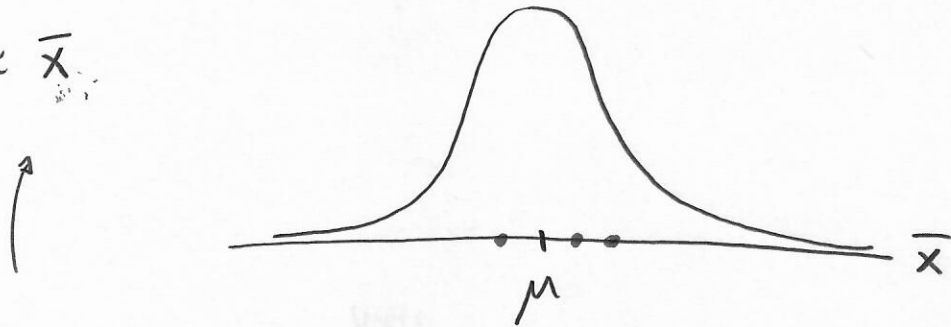


ESTIMATE $\mu \approx 18$

115

eg. You want to estimate the avg. GPA of all CCNY students, by taking a random sample of 50 students.

$$\mu \approx \bar{X}$$

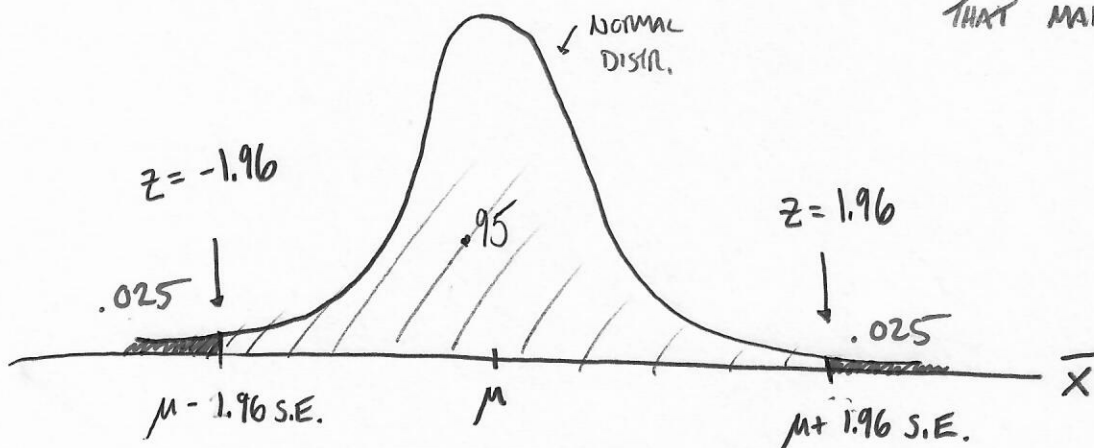


\bar{X} is unbiased because it is equally likely to be overestimate / underestimate and the mean (expected value) of \bar{X} is μ !

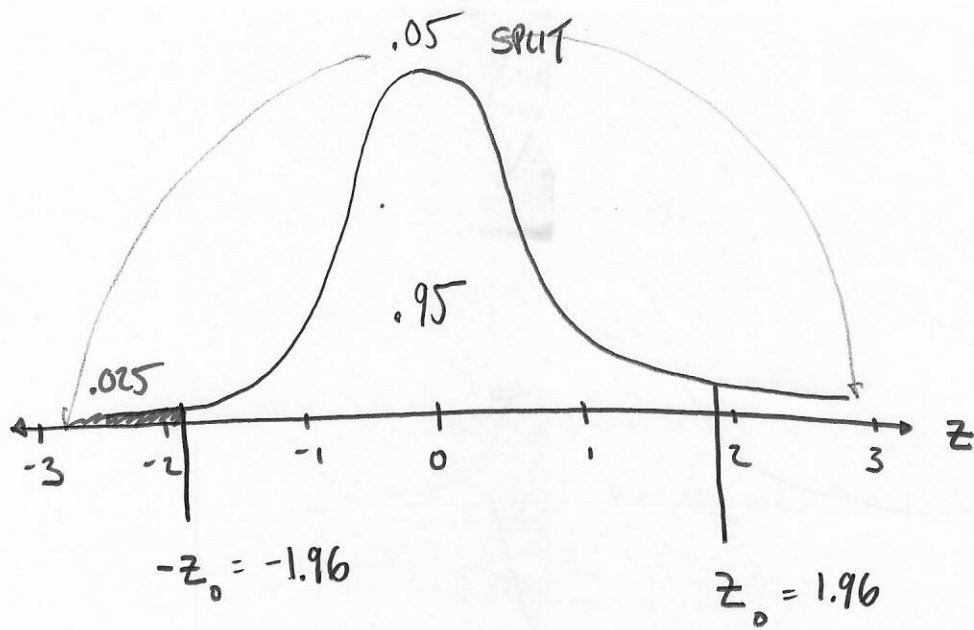
Def. The distance between an estimator & a parameter is called the error of the estimation.

$$\mu \approx \bar{X} \quad , \quad \mu = \bar{X} + \varepsilon$$

error - missing piece that makes equality.



CLT: Sample means \bar{X} are approx. normal with mean = μ , and S.E. = $\frac{\sigma}{\sqrt{n}}$



$$P(-z_0 \leq z \leq z_0) = .95$$

$$P(-1.96 \leq z \leq 1.96) = .95$$

$$z = \frac{\bar{x} - \mu}{\text{S.E.}} \longrightarrow \bar{x} = \mu + z \text{ S.E.}$$

$$P(\mu - 1.96 \text{ S.E.} \leq \bar{x} \leq \mu + 1.96 \text{ S.E.}) = .95$$

POINT ESTIMATE
↓
POPULATION MEAN $\mu \approx \bar{X}$ (SAMPLE MEAN)
↑

HAS S.E. = $\frac{\sigma}{\sqrt{n}} \approx \frac{s}{\sqrt{n}}$

INTERVAL ESTIMATE: $\mu \approx \bar{X} \pm \overbrace{1.96 \text{ S.E.}}^{\text{MARGIN OF ERROR}}$

$$P(\bar{X} \text{ LIES WITHIN } 1.96 \text{ S.E. OF } \mu) = .95$$

$$P(\mu \text{ LIES WITHIN } 1.96 \text{ S.E. OF } \bar{X}) = .95$$

HW THRU.

§7.6