

# Stat 250: Stat Principles and Practices

## Central Limit Theorem illustration

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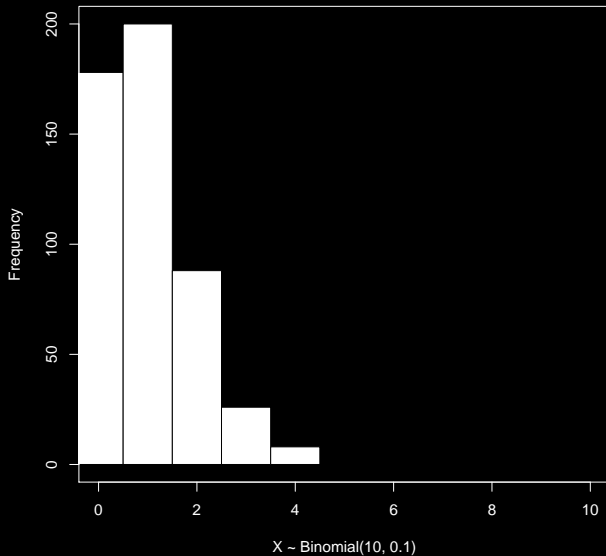
Set-up: *Binomial*(10,  $p$ )

Generate 500 data sets of size  $n$

Illustration with  $n = 5$ ,  $p = 0.1$

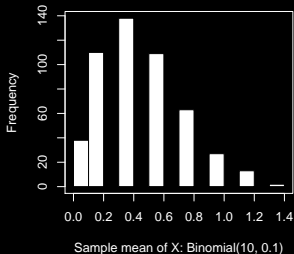
Sample	1	2	3	4	5	$\bar{x}$
Data set 1	0	0	1	2	0	0.6
Data set 2	0	0	1	3	0	0.8
Data set 3	0	1	2	2	2	1.4
Data set 4	0	0	0	3	1	0.8
Data set 5	1	2	2	0	0	1.0
$\vdots$			$\vdots$			$\vdots$

# *Binomial*(10, $p$ ) histogram

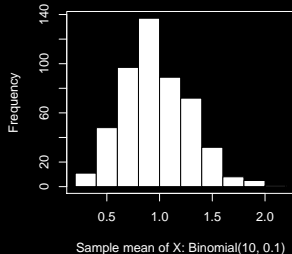


# CLT illustration

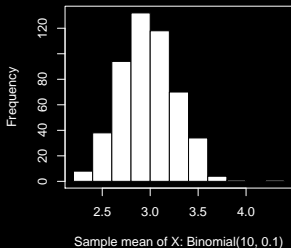
Sample size  $n = 5$



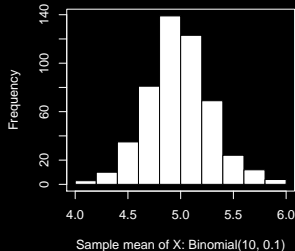
Sample size  $n = 10$



Sample size  $n = 30$

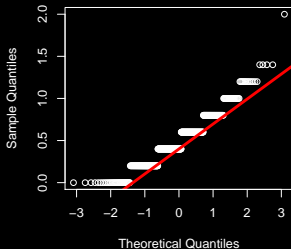


Sample size  $n = 50$

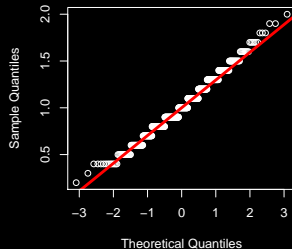


# CLT illustration: normal probability plot

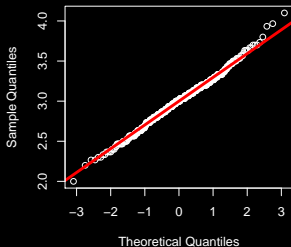
Sample size  $n = 5$



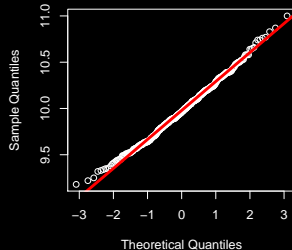
Sample size  $n = 10$



Sample size  $n = 30$



Sample size  $n = 100$



## Foreshadowing: Statistical Inference

- Collect *binomial*( $n, p$ ) random sample:  $X_1, X_2, \dots, X_n$ .
- Normal approximation to the binomial
- Another CLT: For “large  $n$  and reasonable  $p$ ,”

$$\frac{\sum X_i}{n} \sim AN \left( p, \sqrt{\frac{p(1-p)}{n}} \right).$$

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- Use the average to estimate  $p$  and sampling distribution to assess variability.