

# Sampling People, Records, & Networks

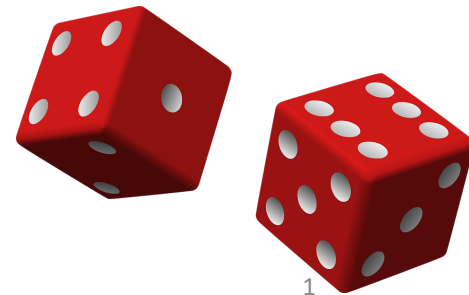
Jim Lepkowski, PhD

Professor & Research Professor *Emeritus*

Institute for Social Research, University of Michigan

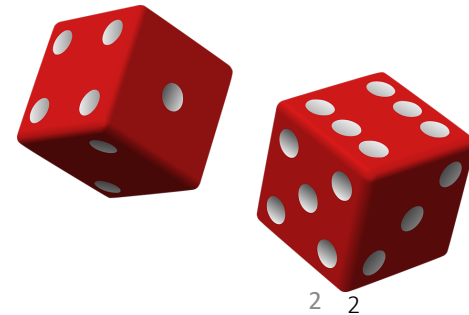
Research Professor,

Joint Program in Survey Methodology, University of Maryland

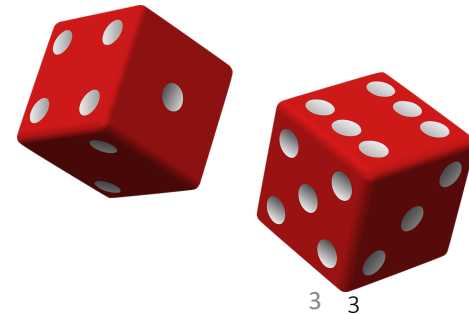


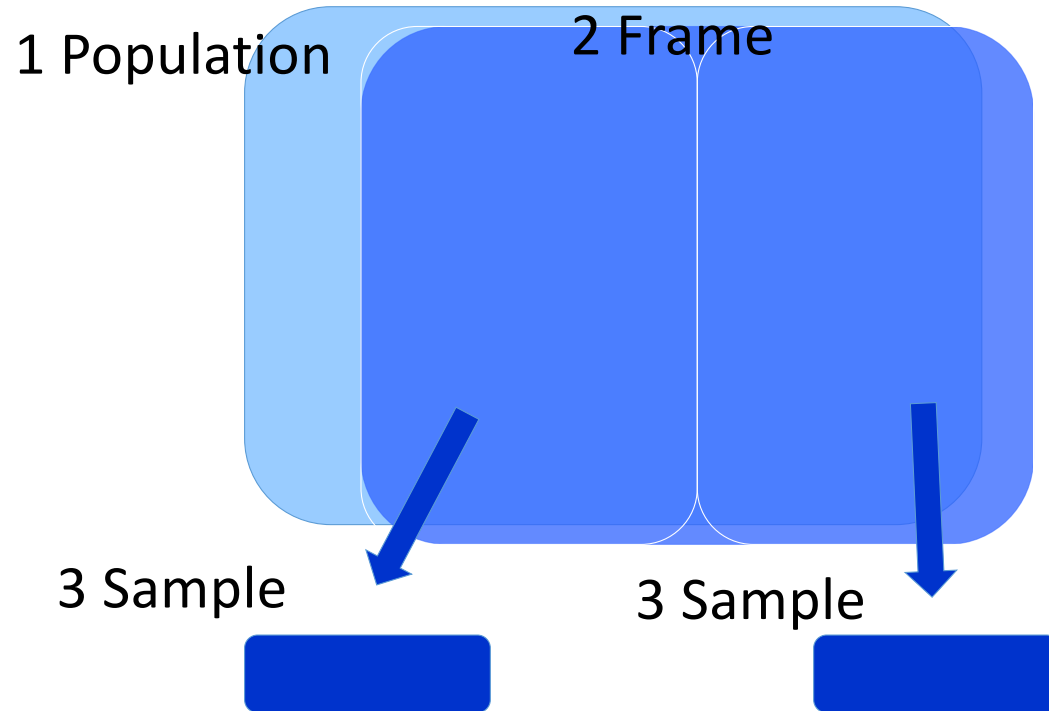
## Unit 4

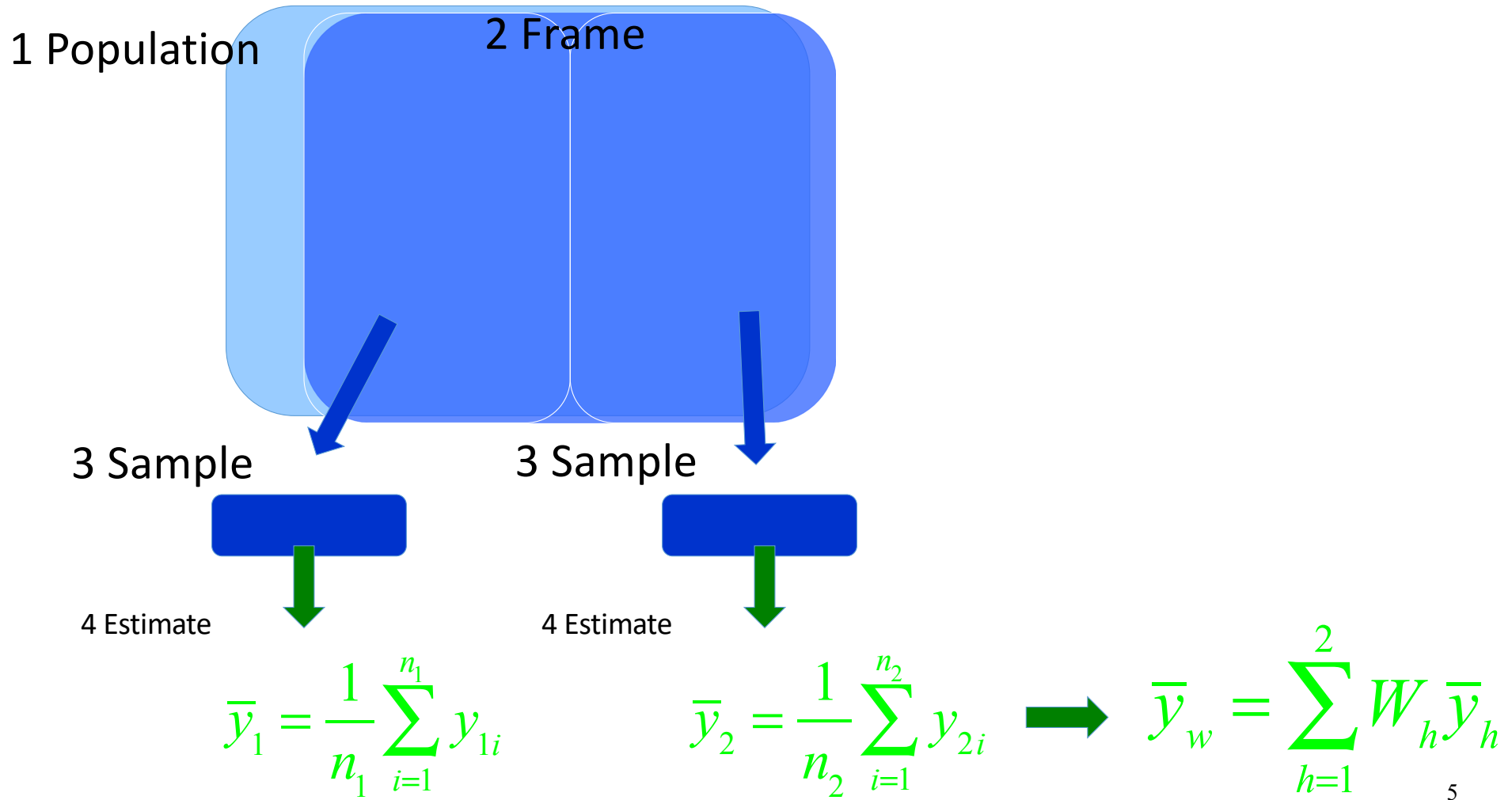
- 1 Forming groups
  - 2 Sampling variance
  - 3 More on grouping
  - 4 Allocate sample
  - 5 Other allocations
  - 6 Weights
- Unit 1: Sampling as a research tool
  - Unit 2: Mere randomization
  - Unit 3: Saving money
  - **Unit 4: Being more efficient**
    - Forming groups
    - Sampling variance
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    - Other allocations
    - Weights to combine across strata
  - Unit 5: Simplifying sampling
  - Unit 6: Some extensions & applications



- Principles
  - Example
  - Confidence interval
  - Design effect
  - Effective sample size
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- In theory,

$$Var(\bar{y}) = \sum_{h=1}^H W_h^2 Var(\bar{y}_h)$$



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- **Estimate** this variance by

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- And what is  $var(\bar{y}_h)$  ?

- For SRS within strata,  $var(\bar{y}_h) = \frac{(1-f_h)}{n_h} S_h^2$





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- And what is  $var(\bar{y}_h)$  ?
- For SRS within strata,  $var(\bar{y}_h) = \frac{(1-f_h)}{n_h} S_h^2$
- We thus need the within stratum variances:



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h	Stratum	$N_h$	$W_h$	$n_h$	$f_h$	$\bar{y}_h$	$s_h^2$
1	Assistant	115	0.2875	23	0.2	50	125
2	Associate	75	0.1875	15	0.2	70	250
3	Full	210	0.5250	42	0.2	90	500
Total		400	1.0000	80	0.2	\$74.75	

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  - The calculation is really beyond the scope of this course, but since we've come this far ...



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$$\bullet \text{ } var(\bar{y}) = \frac{(0.2875)^2 (0.8) (125)}{23} + \frac{(0.1875)^2 (0.8) (250)}{15} + \frac{(0.5250)^2 (0.8) (500)}{42} = 3.453$$



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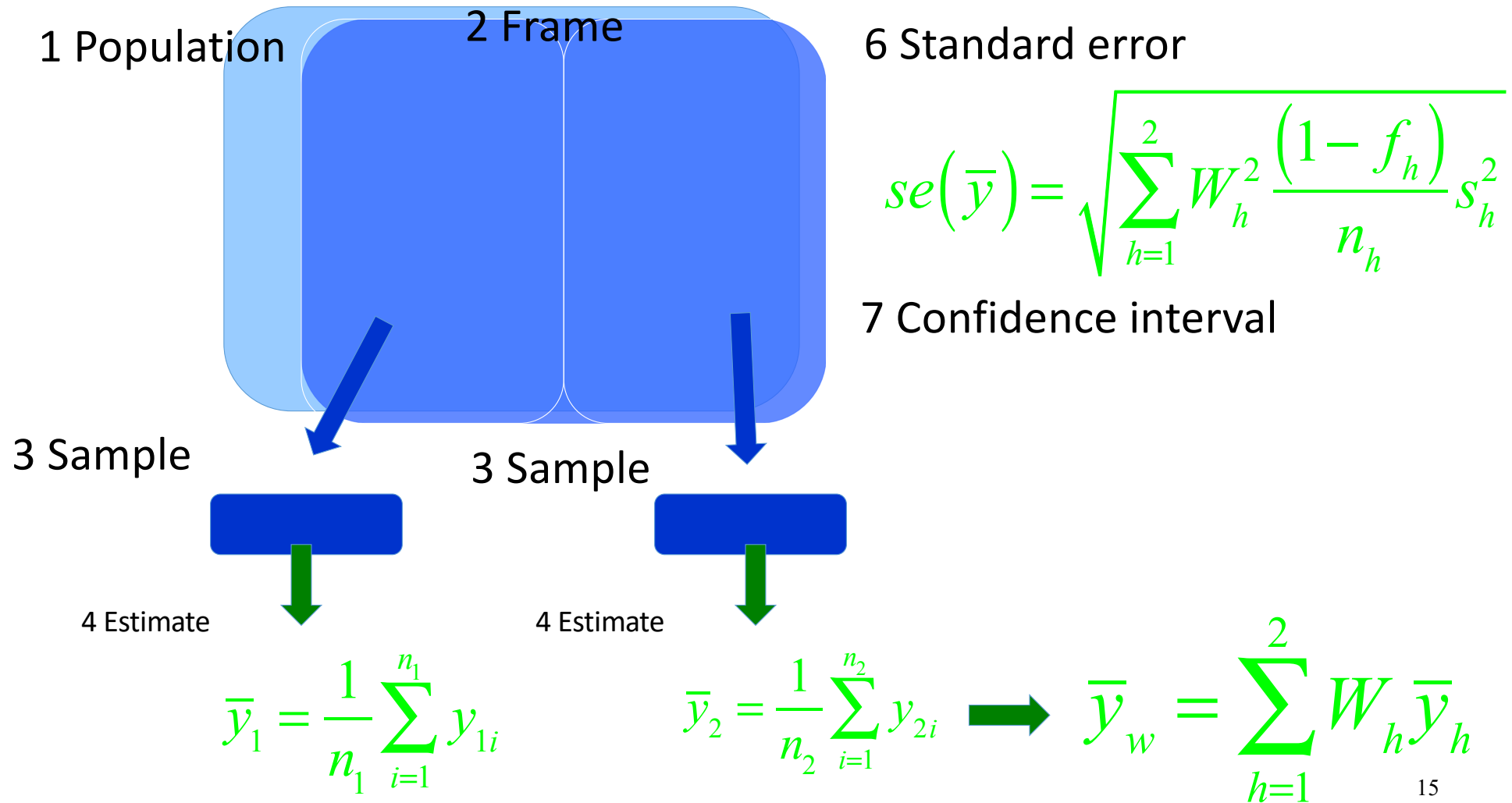
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- And  $se(\bar{y}) = 1.858$



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- Thus, we have completed **step 6a & 6b** – within stratum sampling variances and combining across strata





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- The last step, **step 7**, is the confidence interval
  - Here, let's use the  $t$ -distribution
  - We have  $n_h - 1$  degrees of freedom for each stratum and  $n - H$  for combining across strata
  - For a 95% confidence interval, then,  $t_{(0.975, 80-3)} = 1.991$





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- For a 95% confidence interval, then,  $t_{(0.975, 80-3)} = 1.991$
- The 95% confidence interval is then
- $(\bar{y} - t_{(0.975, 77)}se(\bar{y}), \bar{y} + t_{(0.975, 77)}se(\bar{y}))$
- $(74.75 - 1.991 \times 1.858, 74.75 + 1.991 \times 1.858)$
- $(71.05, 78.45)$



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- Here, as in cluster sampling, there is another issue to be addressed – how does the sampling variance from stratified sampling compare to simple random sampling?
  - That is, what is the design effect,  $deff$ ?

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- Here, as in cluster sampling, there is another issue to be addressed – how does the sampling variance from stratified sampling compare to simple random sampling?
  - That is, what is the design effect,  $deff$ ?
  - For SRS, we need to calculate  $var(\bar{y}) = (1 - f) \frac{s^2}{n}$
  - From a separate calculation,  $s^2 = 647.8$

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- For our sample of size  $n = 80$ , from a population of size  $N = 400$ , or a sampling fraction  $f = 0.2$ ,  
$$var(\bar{y}) = (1 - 0.2) \frac{647.8}{80} = 6.478$$

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$$var(\bar{y}) = (1 - 0.2) \frac{647.8}{80} = 6.478$$

- Compared to the stratified sample,  $deff(\bar{y}) = \frac{var(\bar{y})}{var_{SRS}(\bar{y})} = \frac{3.453}{6.478} = 0.5331$

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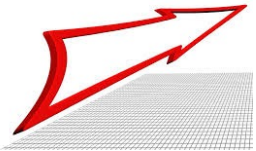
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- A **47% reduction** in sampling variance

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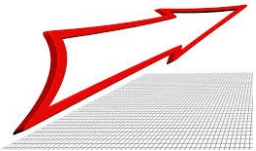
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- For example, the actual sample size of 400 in the stratified proportionately allocated sample is the equivalent of having a simple random sample of

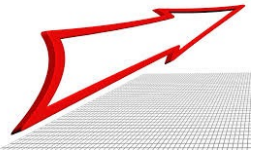
$$n_{eff} = \frac{400}{0.5331} = 750$$





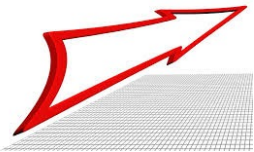
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$$n_{eff} = \frac{400}{0.5331} = 750$$
- That is, the gains in precision are the equivalent of adding 350 cases to the sample

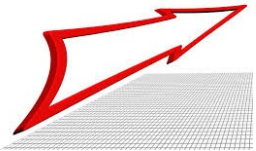
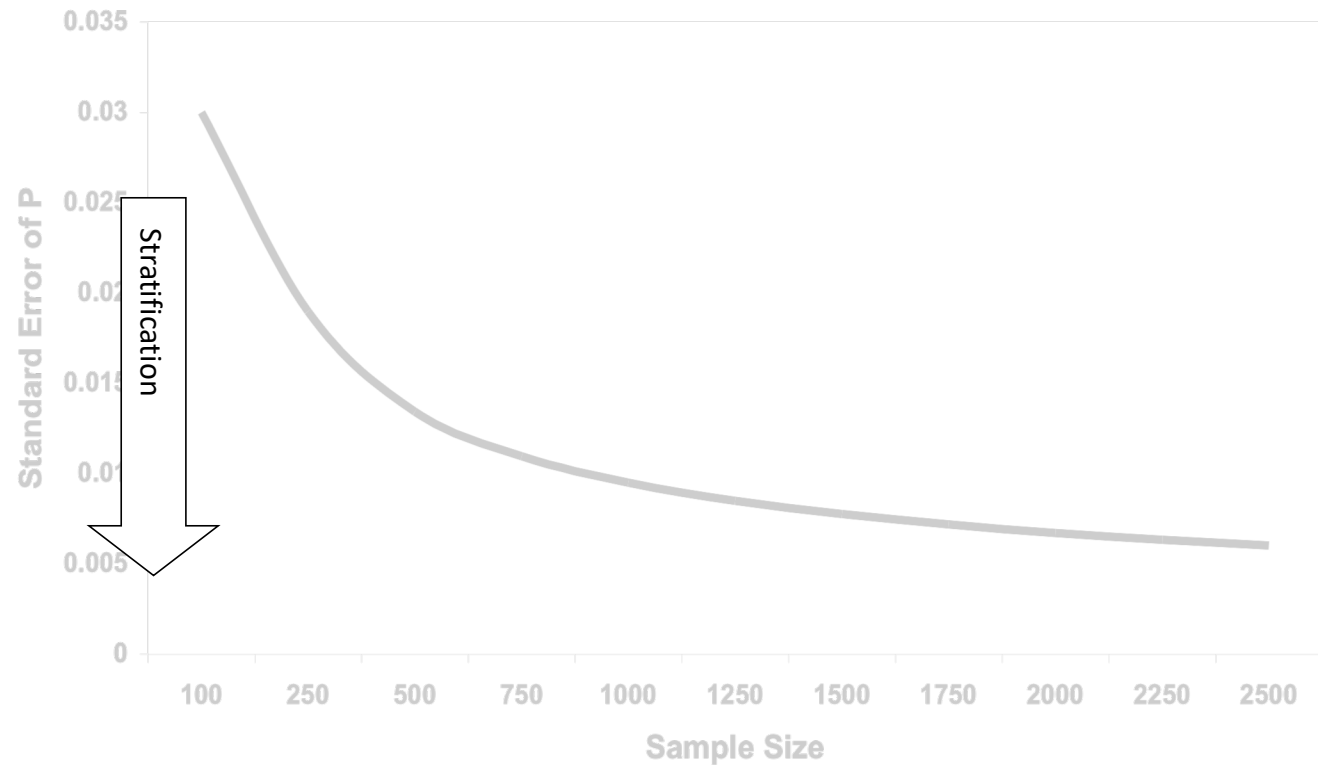


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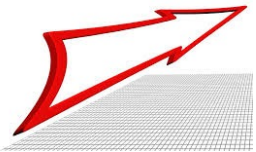
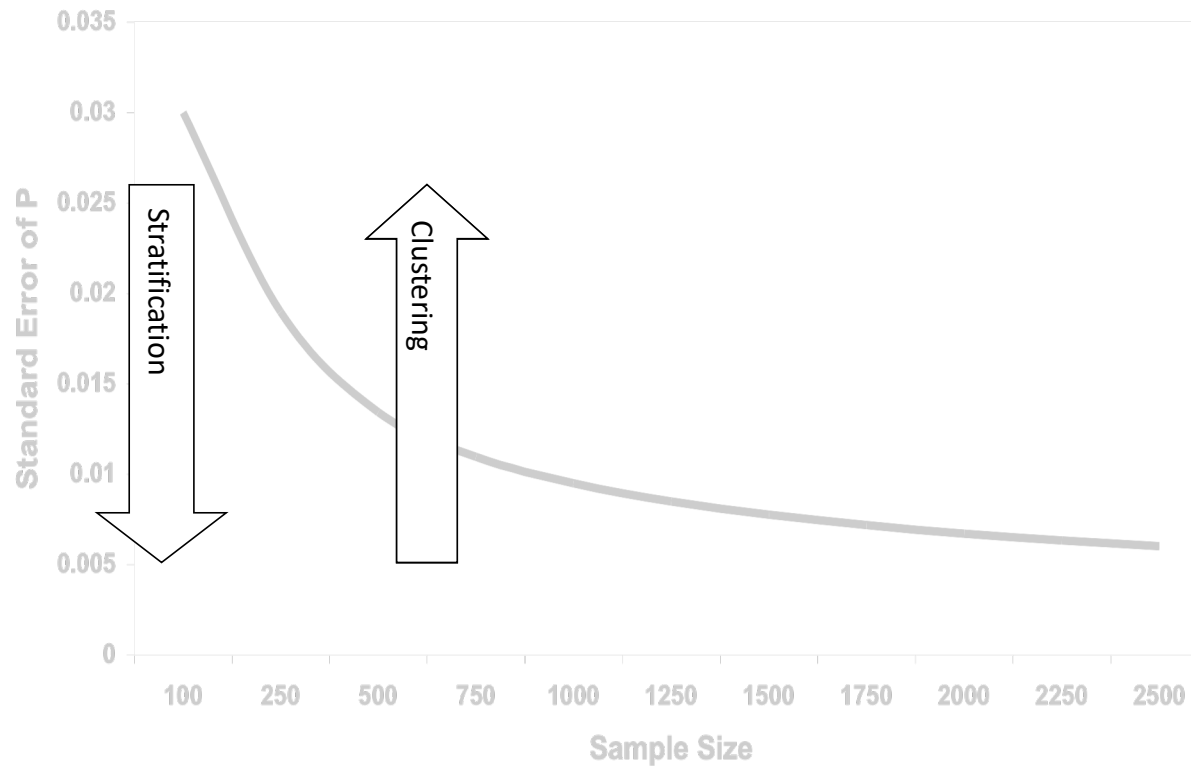
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- Alternatively, the standard errors are smaller and the confidence intervals are narrower by a factor of
$$1 - \sqrt{0.5331} = 1 - 0.7301 = 0.2699, 27\%$$



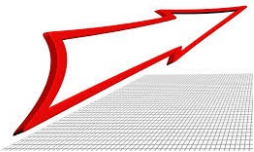
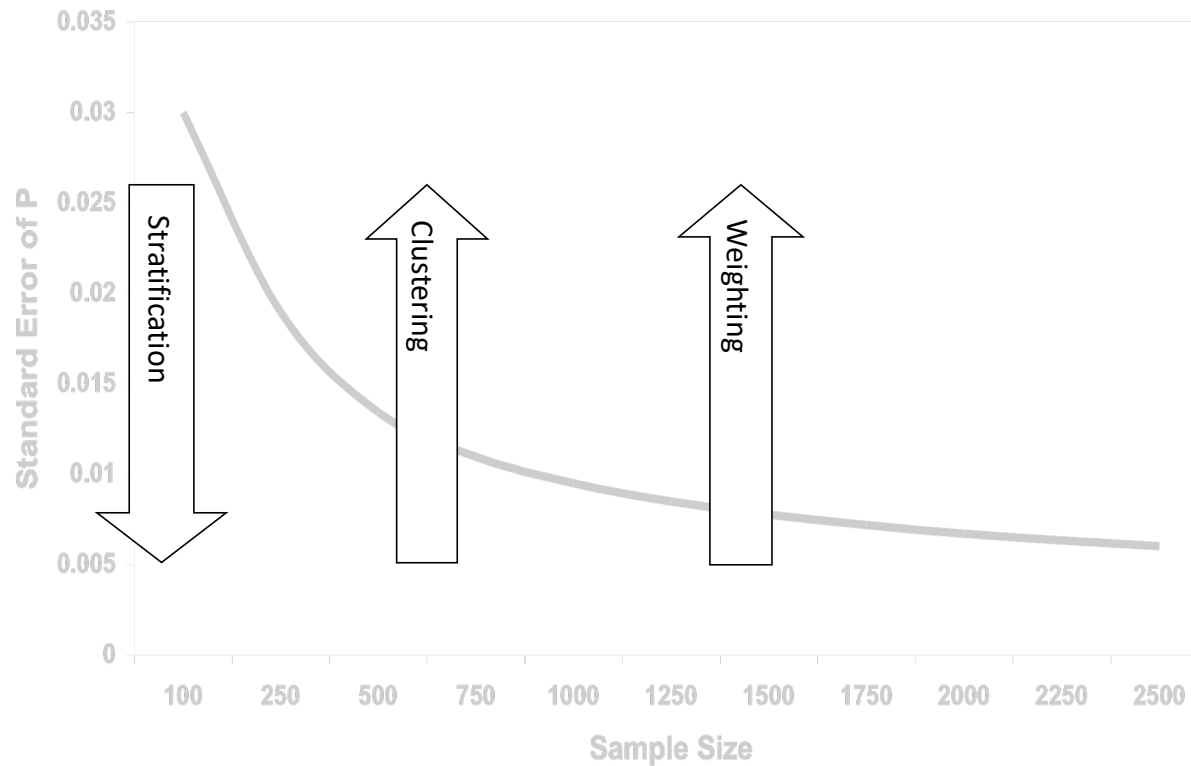
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