Review Session 5

API-201, 10.15.21 Sophie Hill

Agenda

• Review midterm Q2 sensitivity analysis

Sampling distributions

Build a hypothesis test worksheet in Excel!

Calculate the range of vaccination rates that would indicate that getting the vaccine is effective at preventing the spread of COVID.

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$$\frac{0.74}{X} = \frac{0.26}{(1-X)}$$

$$0.74 - 0.74X = 0.26X$$

$$0.74 = X$$

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If the vaccine is effective, then the vaccinated should be a *smaller* proportion of the **infected** than the **population**... i.e., the vaccinated should be a *larger* proportion of the **population** than the **infected**.

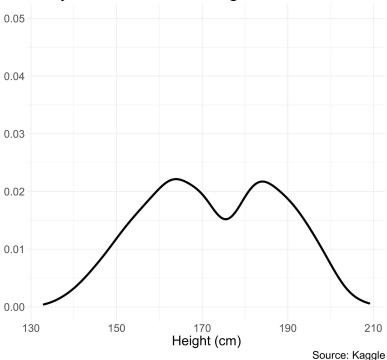
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If the vaccine is effective, then the vaccinated should be a *smaller* proportion of the **infected** than the **population**... i.e., the vaccinated should be a *larger* proportion of the **population** than the **infected**.

So the vaccine is effective if the vaccination rate in the population is above 74% (and has no effect when it is equal to 74%).

Sampling distribution

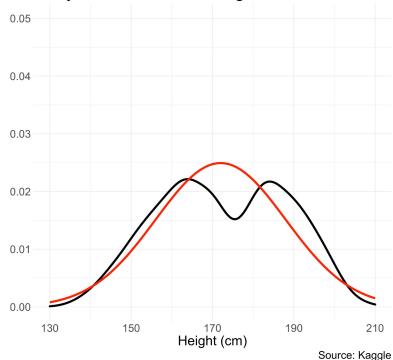
Height distribution of Olympic athletes in Gymnastics and Rowing



Not normally distributed!

Why?

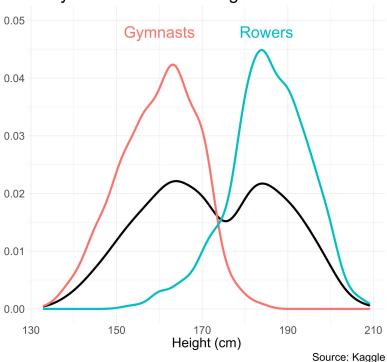
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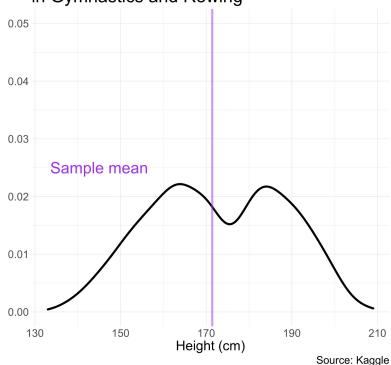
Gymnasts tend to be much shorter than Rowers...

Height distribution of Olympic athletes in Gymnastics and Rowing



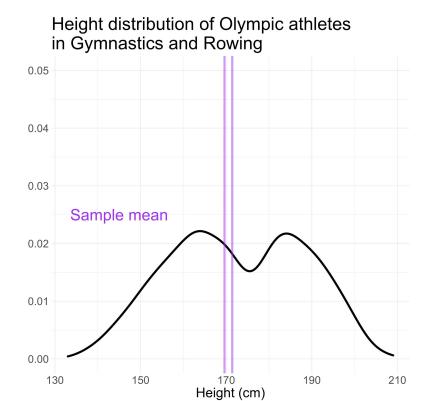
Now let's take a random sample of 10 athletes and calculate the sample mean.





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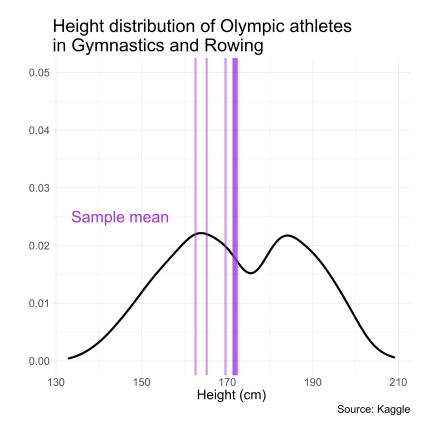
... and again.



Source: Kaggle

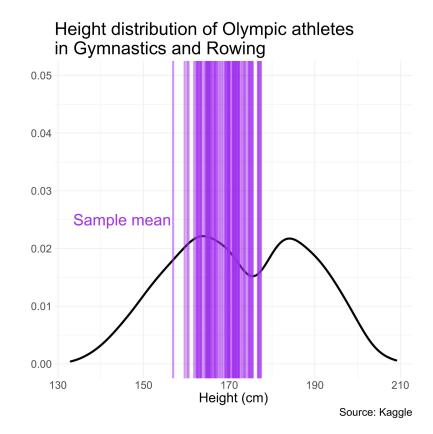
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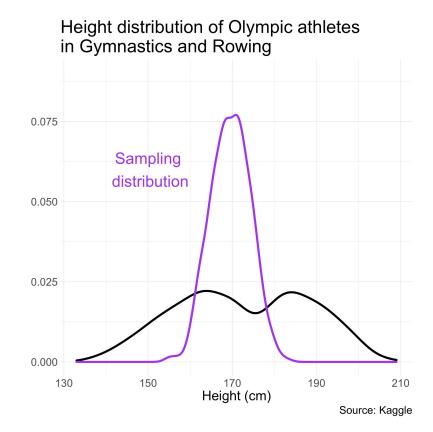
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... and again.



If we did this lots of times, we could plot the PDF of the sample means, i.e., the sampling distribution.

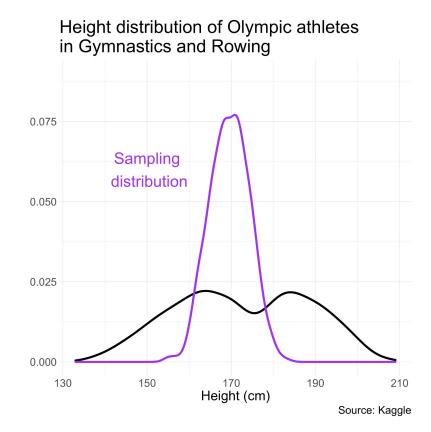
What do you notice about its shape, compared to the PDF of the raw data?



Key takeaway:

The sampling distribution is approximately normal, even though the raw distribution is *not*.

This is the magic of the CLT!!



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Let's make a hypothesis test worksheet in Excel (link to template on Canvas).

HYPOTHESIS TESTING FOR PROPORTIONS

Inputs		
Null hypothesis	q =	0.50
Significance level	alpha =	0.05
Sample estimate	q-hat =	0.47
Sample size	n =	800

Outputs		
Does the CLT app	oly?	Yes
Standard error	SE =	0.0177
Z-score	Z =	-1.6971
p-value	p =	0.0897
Is p-value < alpha	a?	FALSE
Do we reject the	null?	Fail to reject

FAQs

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```
= 2 * (1 - NORM.S.DIST(ABS(Z), TRUE))
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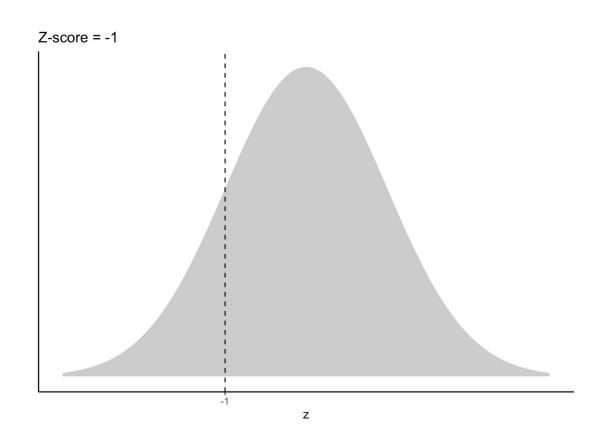
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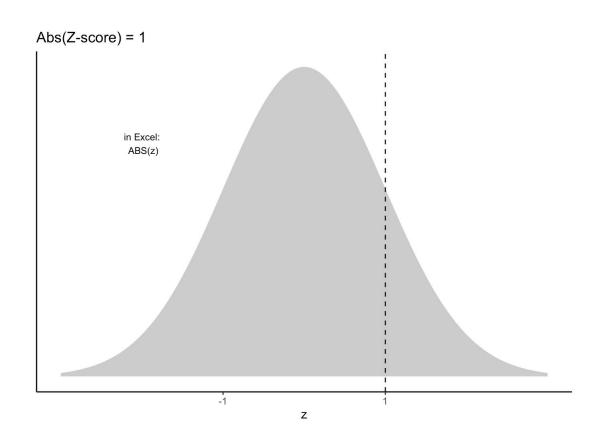
But what does it actually mean??

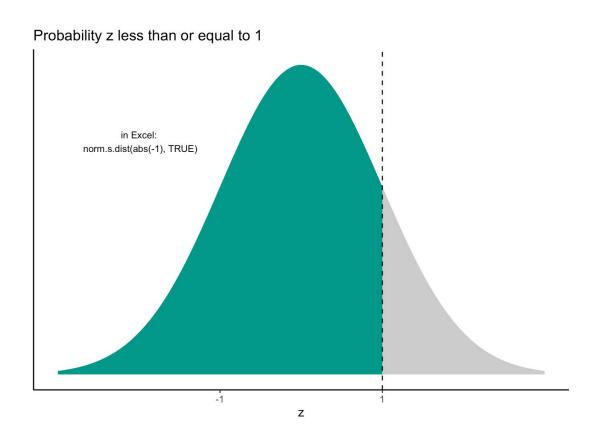
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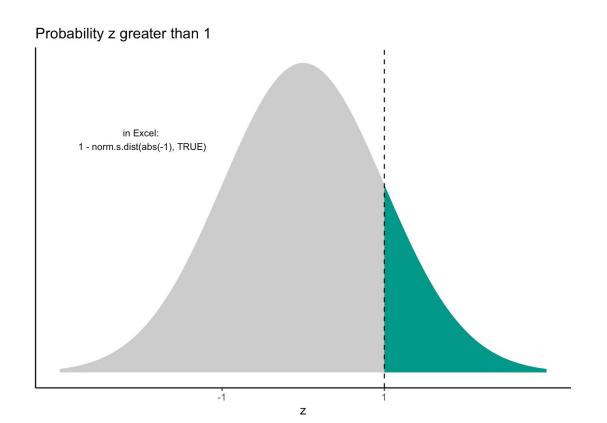
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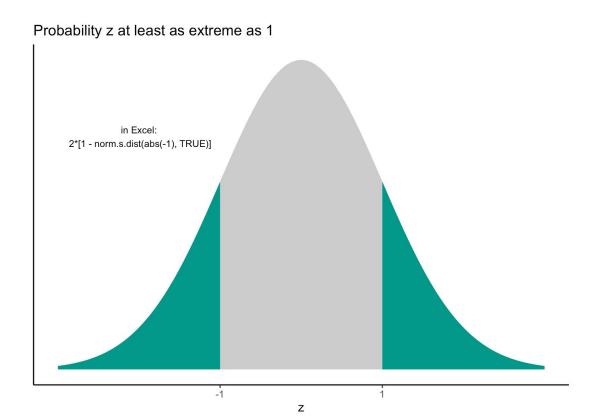
But what does it actually mean?? Let's break it down step-by-step.

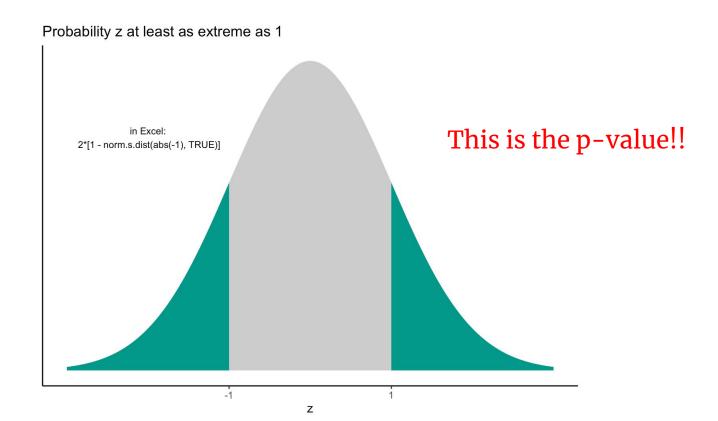












How does the proportion and sample size affect the SE?

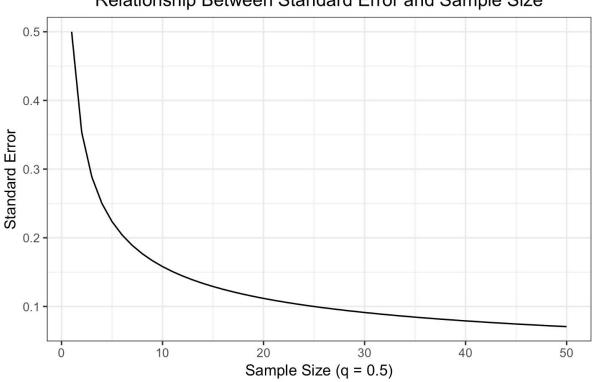
$$\frac{SE}{\sqrt{\frac{q * (1-q)}{n}}}$$

What happens to SE as n increases (holding q fixed)?

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