

Probability Review 2

Lecture 2

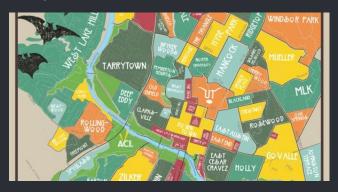
STA 371G

Find out the average house price in Austin.

Find out the average house price in Austin. How would you do that?

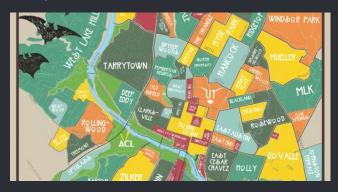


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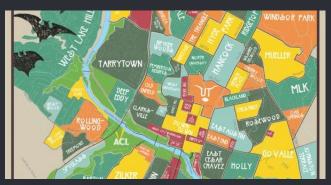
Look at each house price?

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360,000 houses in Austin!
Can we do something smarter?

A smarter approach:

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Just like making polls to predict election results!

	Population	Sample	
Members	all house prices	prices you picked	
Average	population mean	sample mean	
Variance	population variance	sample variance	

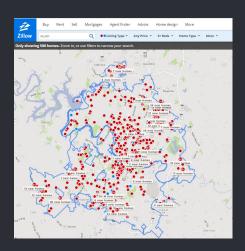
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Estimating a population parameter (population mean) based on a sample statistic (sample mean).

Collecting a sample

Zillow.com, "Austin, TX."

- Click "More Map"
- Select 15 houses, note their prices in an R script.
- Do not discard any price, use the first 15
- Try to represent different regions



Collecting a sample

Your R script should look like this

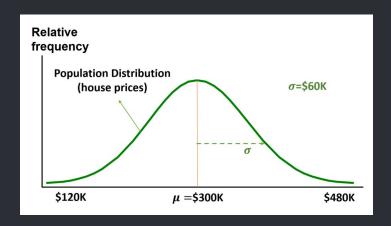
```
# Create a vector of house prices (You should have 15 price data)
sample house prices <- c(327000,276000,513000)</pre>
# Calculate sample statistics
sample mean <- mean(sample house prices)</pre>
sample variance <- var(sample house prices)</pre>
sample standard deviation <- sd(sample house prices)</pre>
# Sample mean of first 5 houses
sample mean 5 <- mean(sample house prices[1:5])</pre>
# Print them to console
cat("Sample Mean", sample mean)
cat("Sample Variance", sample variance)
cat("Sample Standard Deviation", sample standard deviation)
cat("Sample Mean of first 5 houses", sample mean 5)
```

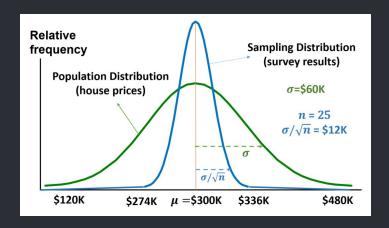
On Learning Catalytics, enter your results.

And here is what they look like...

Distribution of your answers → Sampling distribution

Statistic	Population	Sample Mean
Mean	μ	μ
Standard Deviation	σ	σ/√n





Assume $\mu = $300K$, $\sigma = $60K$.

	n	σ/√n	3 std. dev. range (99.7%)
Survey 1	25	\$12K	\$264K \$336K
Survey 2	100	\$6K	\$282K \$318K
Survey 3	3600	\$1K	\$297K \$303K

Let's compare sample mean of 5 houses vs 15 houses.

What do you expect to see?

t Distribution

We often do not know population variance and use sample variance instead.

In that case, the sample mean will have a *t* distribution.

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Questions, questions...

• Would you reject the hypothesis? Why?

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- Would you be more comfortable with your conclusion if you had 1000 houses in your survey?

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- Would you reject the hypothesis? Why?
- Is it possible that, out of bad luck, you picked the cheapest houses?
- Would you be more comfortable with your conclusion if you had 1000 houses in your survey?
- When should you reject the hypothesis? When not?

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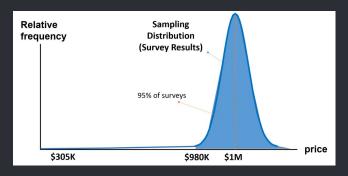
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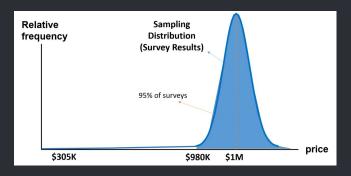
 α is usually chosen as 0.05 prior to sampling.

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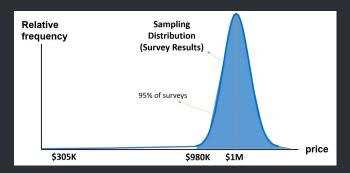


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P-value is smaller than 10^{-100} , while $\alpha = 0.05$. Rather than thinking you are cursed, you simply reject the hypothesis!

Learning Catalytics...

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Learning Catalytics...

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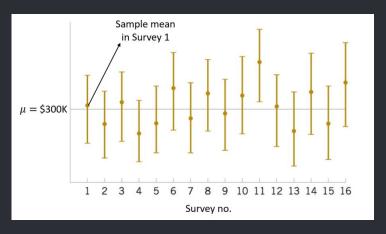
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Add the following to your R script

```
# Calculate 95% confidence interval (default)
avg_price_ci_95 <- t.test(sample_house_prices)
# Calculate 99% confidence interval
avg_price_ci_99 <- t.test(sample_house_prices, conf.level = 0.99)
# Display results
cat("95% confidence interval is:", avg_price_ci_95$conf.int)
cat("99% confidence interval is:", avg_price_ci_99$conf.int)</pre>
```

Enter your results on Learning Catalytics.