

Topic	Data Story-2	
Class Description	Students learn to remove the outliers from the data using the IQR method and reach a conclusion at the end.	
Class	C113	
Class time	45 mins	
Goal	 Remove the outliers using the IQR method. Calculate the z score. Complete the data story. 	
Resources Required	 Teacher Resources Google Colaboratory (Colab) Laptop with internet connectivity Earphones with mic Notebook and pen Student Resources Google Colaboratory (Colab) Laptop with internet connectivity Earphones with mic Notebook and pen 	
Class structure	Warm Up Teacher-led Activity Student-led Activity Wrap up	5 mins 15 min 15 min 5 min

CONTEXT

• Review the concepts learned in the earlier classes

Class Steps	Teacher Action	Student Action
Step 1: Warm Up (5 mins)	Hi <student name=""> What all did we learn in our previous classes?</student>	ESR: -We took the data of people who were reminded to save and who weren't.

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CHALLENGE • Learn about the Inter Quartile Range (IQR) • Calculate the IQR and remove the outliers. Step 2: Can you tell me what you can reacher-led understand from the word varied		
Teacher Initiates Screen Share		
	Let's get started then.	-
	z score of the data. Are you up for it?	
	In today's class we'll learn how to calculate the spread of the data, remove the outliers and then find the	
	Yeswe got to know that the data was scattered and had many outliers so our output was not proper.	ESR: varied
		median and mode of the people who were reminded to save and who weren't reminded to save and we saw that the difference was massive. - We also calculated the correlation between the age and the savings and we saw that the data was not correlated.

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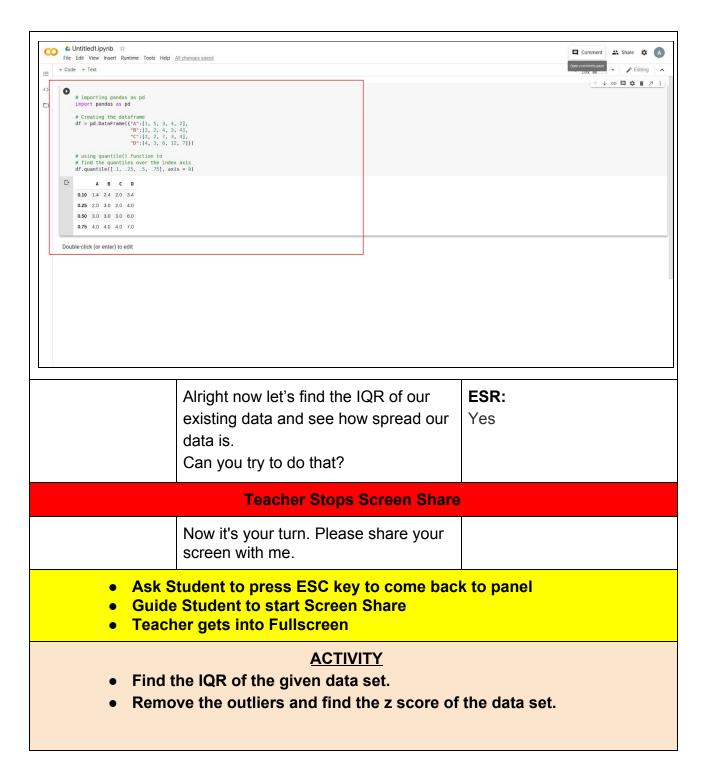


So Interquartile Range or IQR is used to measure how spread out the data points in a set are from the mean of the data set. The higher the IQR, the more spread out the data points; in contrast, the smaller the IQR, the more bunched up the data points are around the mean.	
To calculate the IQR there are 4 simple steps. 1. Order the data from least to greatest. 2. Find the mean. 3. Calculate the median of both the lower and upper half of the data. 4. The IQR is the difference between the upper and lower medians.	
Pandas has a quantile function which does all this for us. Pandas dataframe.quantile() this function returns the data frame where the index is q, columns of self (It will take the names of the columns as they are) and values will be the quantiles.	
Let's see an example to understand better. <teacher 1="" activity="" and="" colab="" from="" notebook="" opens="" td="" teacher="" the="" the<="" writes=""><td></td></teacher>	



```
following code in the code cell and
runs it>
Code:-
# importing pandas as pd
import pandas as pd
# Creating the dataframe
df = pd.DataFrame(
{"A":[1, 5, 3, 4, 2],
"B":[3, 2, 4, 3, 4],
"C":[2, 2, 7, 3, 4],
 "D":[4, 3, 6, 12, 7]})
# using quantile() function to
# find the quantiles over the index
axis
df.quantile([.1, .25, .5, .75], axis = 0)
As we see here we created a sample
dataframe.
Using the quantile function we get
the new data consisting of the
quantiles of .1, .25, .5, .75 on the
given values at axis 0.
```







Step 3:

Step 3. Student-Led Activity (15 min)	new colab notebook from previous class.	notebook from previous class.(Student Activity 1)
	Till now we have seen that in our data the mean differences were significant and there was no correlation between the age and the savings.	Student imports the seaborn library and plots the quant_saved data on it.
	To deal with this, we can remove the outliers. To remove the outliers we'll use the IQR method.	
	Let's first check the outliers. To check the outliers we'll use the box plot from the seaborn library. We'll first import the library and then plot the quant_saved data on it.	
<pre>[] import seaborn as sns sns.boxplot(data=df, x=df["quant_saved"])</pre>		
pandas.util.testing	on3.6/dist-packages/statsmodels/tools/_testing.py:19: is deprecated. Use the functions in the public API at #bplots.AxcsSubplot at 0x7fe9b6304 #bplots.AxcsSubplot at 0x7fe9b6304 #bplots.AxcsSubplot at 0x7fe9b6304	

Teacher helps the student to open a

Student opens the colab

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5000 6000

7000

3000 4000

1000 2000



Here, we can see that the majority of the data is represented by a thick, long black line. We then have a thin long black line and finally we have a horizontal line, which is the rest of the data.

Here, the thin black vertical line is the line that separates the data we should use to do our analysis. Let's remove the outliers first.

Code:-

iqr = q3-q1

q1 = df["quant_saved"].quantile(0.25) q3 = df["quant_saved"].quantile(0.75)

print(f"Q1 - {q1}")
print(f"Q3 - {q3}")
print(f"IQR - {iqr}")

lower_whisker = q1 - 1.5*iqr upper_whisker = q3 + 1.5*iqr

print(f"Lower Whisker {lower_whisker}")
print(f"Upper Whisker {upper_whisker}")

#Creating a new DataFrame
new_df = df[df["quant_saved"] <
upper_whisker]</pre>

Student codes to find the outliers.

Student codes to find the quantile of the quant_saved data.

Student codes to find the quantile of the 3rd quantile and then finds iqr by subtracting q3-q1 print these values.



```
[ ] q1 = df["quant saved"].quantile(0.25)
    q3 = df["quant saved"].quantile(0.75)
    iqr = q3-q1
    print(f"Q1 - {q1}")
    print(f"Q3 - {q3}")
    print(f"IQR - {iqr}")
    lower whisker = q1 - 1.5*iqr
    upper whisker = q3 + 1.5*iqr
    print(f"Lower Whisker - {lower whisker}")
    print(f"Upper Whisker - {upper whisker}")
    #Creating a new DataFrame
    new_df = df[df["quant saved"] < upper whisker]</pre>
Q1 - 2.28400000000000003
    Q3 - 86.514
    IOR - 84.22999999999999
    Lower Whisker - -124.06099999999998
    Upper Whisker - 212.8589999999998
```

From here, we can say that anyone who saved roughly more than 213 is an outlier in this data.

Now that we've dealt with the outliers, let's see what our mean, median and modes are and also, if we have a normal distribution now. Let's also check the standard deviation now.

Teacher helps the student find the standard deviation.

Code:-

#Mean, median and mode of savings

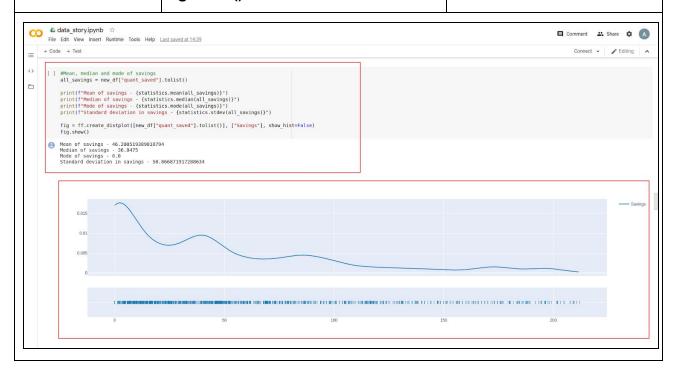
Student codes to find the mean, median, mode and standard deviation of the quant_data. and plot it on the graph.



all_savings =
new_df["quant_saved"].tolist()

print(f"Mean of savings {statistics.mean(all_savings)}")
print(f"Median of savings {statistics.median(all_savings)}")
print(f"Mode of savings {statistics.mode(all_savings)}")
print(f"Standard deviation in
savings {statistics.stdev(all_savings)}")

fig =
ff.create_distplot([new_df["quant_s
aved"].tolist()], ["Savings"],
show_hist=False)
fig.show()





Here, now our mean is much closer to the median than before, and the mode still remains to be 0 since there are people in our dataset who haven't saved at all! Student codes to find the mean of random 100 data points and plot it on the graph.

The standard deviation has also significantly reduced from before, but we still do not have a normal distribution curve.

To make this a normal distribution curve now, we can simply draw random samples from the population to create a sampling (normal distribution).

You already know how to get the 100 data points, find its mean and plot it on the graph.

Code:-

#Collecting 1000 samples of 100 data points each, saving their averages in a list import random

sampling_mean_list = []
for i in range(1000):
 temp_list = []
 for j in range(100):

temp_list.append(random.choice(al l_savings))



sampling_mean_list.append(statist
ics.mean(temp_list))

mean_sampling =
statistics.mean(sampling_mean_lis
t)

fig =

ff.create_distplot([sampling_mean_
list], ["Savings (Sampling)"],
show_hist=False)

fig.add_trace(go.Scatter(x=[mean_
sampling, mean_sampling], y=[0,
0.1], mode="lines",
name="MEAN"))

fig.show()





What do we see?

Perfect now let's find the standard deviation of the sampling distribution.

ESR:We see the normal distribution bell curve.

Code:-

print(f"Standard deviation of the sampling data - {statistics.stdev(sampling_mean_list)}")

Student codes to find the standard deviation of the sampling distribution.

print(f"Standard deviation of the sampling data - {statistics.stdev(sampling mean list)}")

8

Standard deviation of the sampling data - 5.091787150333248

What do we find here?

Now we'll compare the mean of the sampling and the population.

Code:-

print(f"Mean of Population {statistics.mean(all_savings)}")
print(f"Mean of Sampling
Distribution - {mean_sampling}")

ESR:

The sampling standard deviation comes out to be roughly 1/10th of the standard deviation of the population.

Student codes to find the mean of the population and the sampling.

print(f"Mean of Population - {statistics.mean(all_savings)}")
print(f"Mean of Sampling Distribution - {mean_sampling}")

0

Mean of Population - 46.200519389818794 Mean of Sampling Distribution - 45.99224013709402

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	What do we see here?	ESR: We see that they come out exactly the same.
	Let's do one final check, by trying to find the correlation between the age and the savings with the new data from which we have removed the outliers!	
	We'll find the correlation between the age and the savings data from which the outliers were removed.	Student codes to find the new correlation between the age and the savings data from which the outliers were
	Code:-	removed.
	#temp_df will have the rows where age is not 0	
	temp_df = new_df[new_df.age != 0]	
	age = temp_df["age"].tolist() savings = temp_df["quant_saved"].tolist()	
	correlation = np.corrcoef(age, savings)	
	print(f"Correlation between the age	
	of the person and their savings is - {correlation[0,1]}")	
[#temp_df will have temp_df = new_df[ne	the rows where age is not 0 w_df.age != 0]	
	<pre>].tolist() quant_saved"].tolist() rrcoef(age, savings)</pre>	

 $print(f"Correlation\ between\ the\ age\ of\ the\ person\ and\ their\ savings\ is\ -\ \{correlation[0,1]\}")$

Correlation between the age of the person and their savings is - 0.08561544120342093



What do we get?

Now, let's get back to the real question. Does reminder have any impact on the savings of the people? How can we achieve this statistically?

With the z-test! We will take a sample from the people who were given reminders and see if that sample fits with the population of people who were not given reminders. If it fits in well, we can say that the reminders had no significant effect.

Let's start with first filtering out the data of people who were given notifications and those who weren't given notifications.

To filter the data we just have to check if the value under that person is 0 or 1.

Code:-

reminded_df =
new_df.loc[new_df["rem_any"] ==
1]
not_reminded_df =
new_df.loc[new_df["rem_any"] ==
0]

print(reminded_df.head())
print(not_reminded_df.head())

ESR:

We get a correlation of 0.085, which is not significant.

Student codes to filter the data of people who were given notifications and those who weren't given notifications.

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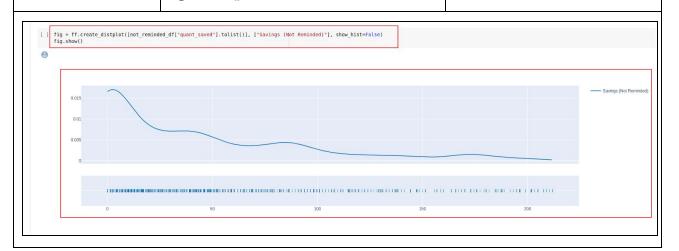
```
reminded df = new df.loc[new df["rem any"] == 1]
not_reminded_df = new_df.loc[new_df["rem_any"] == 0]
print(reminded df.head())
print(not reminded df.head())
               female highschool completed
                                                      wealthy
   quant saved
                                             rem any
       13.0908
                                                               28.0
                                                   1
       39.2724
                                                                0.0
3
       58.9086
                    1
                                                   1
                                                                0.0
4
                                                                0.0
       78.5448
                                          1
                                                   1
                    1
       39.2724
                                                   1
                                                               43.0
    quant_saved female highschool_completed rem_any
                                                      wealthy
                                                                 age
                                                                26.0
11
       39.2724
                                                    0
        58.9086
12
                                                                 0.0
14
       78.5448
                     1
                                                    0
                                                             0 32.0
        2.2840
                                                    0
                                                             1 29.0
31
                     1
                                            1
         2.2840
                                                              1 28.0
```

Now, let's plot the distribution graph for the people who were not reminded to save and see if it follows a normal distribution.

Code:-fig =

ff.create_distplot([not_reminded_df
["quant_saved"].tolist()], ["Savings
(Not Reminded)"],
show_hist=False)
fig.show()

Student codes to plot the not reminded to save the money.





What can we observe?

Correct, so we will take sampling data and plot it and also use it for z test.

We need to find the mean and standard deviation of the sampling data.

Code:-

not_reminded_savings =
not_reminded_df["quant_saved"].t
olist()

sampling_mean_list_not_reminded
= []
for i in range(1000):
 temp_list = []
 for j in range(100):

temp_list.append(random.choice(n ot_reminded_savings))

sampling_mean_list_not_reminded
.append(statistics.mean(temp_list))

mean_sampling_not_reminded =
statistics.mean(sampling_mean_lis
t_not_reminded)
stdev_sampling_not_reminded =
statistics.stdev(sampling_mean_lis
t_not_reminded)

print(f"Mean of Sampling (Not
Reminded) ->
{mean_sampling_not_reminded}")
print(f"Standard Deviation of
Sampling (Not Reminded) ->
{stdev_sampling_not_reminded}")

ESR:

This doesn't follow normal distribution.

Student codes to find the mean and standard deviation of the sampling.

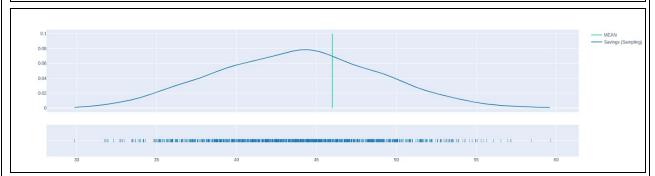
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fig =

ff.create_distplot([sampling_mean_
list_not_reminded], ["Savings
(Sampling)"], show_hist=False)
fig.add_trace(go.Scatter(x=[mean_
sampling, mean_sampling], y=[0,
0.1], mode="lines",
name="MEAN"))
fig.show()

```
not reminded savings = not reminded df["quant saved"].tolist()
sampling mean list not reminded = []
for i in range(1000):
  temp list = []
  for j in range(100):
    temp list.append(random.choice(not reminded savings))
  sampling mean list not reminded.append(statistics.mean(temp list))
mean_sampling_not_reminded = statistics.mean(sampling_mean_list_not_reminded)
stdev sampling not reminded = statistics.stdev(sampling mean list not reminded)
print(f"Mean of Sampling (Not Reminded) -> {mean_sampling_not_reminded}")
print(f"Standard Deviation of Sampling (Not Reminded) -> {stdev sampling not reminded}")
fig = ff.create_distplot([sampling_mean_list_not_reminded], ["Savings (Sampling)|"], show_hist=False)
fig.add trace(go.Scatter(x=[mean sampling, mean sampling], y=[0, 0.1], mode="lines", name="MEAN"))
fig.show()
Mean of Sampling (Not Reminded) -> 43.79363006979631
Standard Deviation of Sampling (Not Reminded) -> 4.998539302693592
```



Now we have a resemblance of a normal distribution, and we also have the mean and the standard deviation of the sampling data for the people who were not notified to save.

Student codes to find the first, second and the third deviation.

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Now, assuming that we started notifying people to see if there was an effect in people's saving, let's see how it turned out!

First, let's calculate the first, second and third standard deviations for our sampling distribution.

Teacher helps the student with the code.

Code:-

first_std_deviation_start =
mean_sampling_not_reminded-std
ev_sampling_not_reminded
first_std_deviation_end =
mean_sampling_not_reminded+std
ev_sampling_not_reminded
print(f"First (start) {first_std_deviation_start} and First
(end) - {first_std_deviation_end}")

second_std_deviation_start =
mean_sampling_not_reminded-(2*s
tdev_sampling_not_reminded)
second_std_deviation_end =
mean_sampling_not_reminded+(2*
stdev_sampling_not_reminded)
print(f"Second (start) {second_std_deviation_start} and
Second (end) {second_std_deviation_end}")



third_std_deviation_start =
mean_sampling_not_reminded-(3*s
tdev_sampling_not_reminded)
third_std_deviation_end =
mean_sampling_not_reminded+(3*
stdev_sampling_not_reminded)
print(f"Third (start) {third_std_deviation_start} and
Third (end) {third_std_deviation_end}")

```
first_std_deviation_start = mean_sampling_not_reminded-stdev_sampling_not_reminded
first_std_deviation_end = mean_sampling_not_reminded+stdev_sampling_not_reminded
print(f"First (start) - {first_std_deviation_start} and First (end) - {first_std_deviation_end}")

second_std_deviation_start = mean_sampling_not_reminded-(2*stdev_sampling_not_reminded)
second_std_deviation_end = mean_sampling_not_reminded+(2*stdev_sampling_not_reminded)
print(f"Second (start) - {second_std_deviation_start} and Second (end) - {second_std_deviation_end}")

third_std_deviation_start = mean_sampling_not_reminded-(3*stdev_sampling_not_reminded)
third_std_deviation_end = mean_sampling_not_reminded+(3*stdev_sampling_not_reminded)
print(f"Third (start) - {third_std_deviation_start} and Third (end) - {third_std_deviation_end}")

First (start) - 38.795090767102714 and First (end) - 48.7921693724899
Second (start) - 33.79655146440912 and Second (end) - 53.790708675183495
Third (start) - 28.798012161715533 and Third (end) - 58.78924797787708
```

Now that we have everything we need for our z-test, let's quickly take a sampling distribution of the people who were reminded for saving.

Teacher helps the student with the code.

Code:-

reminded_savings =
reminded_df["quant_saved"].tolist(
)

sampling_mean_list_reminded = []
for i in range(1000):
 temp_list = []

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for j in range(100):

temp_list.append(random.choice(reminded_savings))

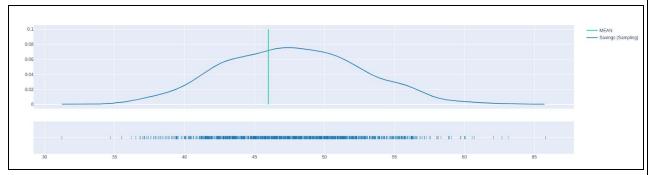
sampling_mean_list_reminded.app
end(statistics.mean(temp_list))

mean_sampling_reminded =
statistics.mean(sampling_mean_lis
t_reminded)
stdev_sampling_reminded =
statistics.stdev(sampling_mean_lis
t_reminded)

print(f"Mean of Sampling
(Reminded) ->
{mean_sampling_reminded}")
print(f"Standard Deviation of
Sampling (Reminded) ->
{stdev_sampling_reminded}")
fig =
ff.create_distplot([sampling_mean_
list_reminded], ["Savings
(Sampling)"], show_hist=False)
fig.add_trace(go.Scatter(x=[mean_
sampling, mean_sampling], y=[0,
0.1], mode="lines",
name="MEAN"))
fig.show()



```
reminded savings = reminded df["quant saved"].tolist()
sampling_mean_list_reminded = []
for i in range(1000):
  temp list = []
  for j in range(100):
   temp_list.append(random.choice(reminded_savings))
  sampling_mean_list_reminded.append(statistics.mean(temp_list))
mean_sampling_reminded = statistics.mean(sampling_mean_list_reminded)
stdev_sampling_reminded = statistics.stdev(sampling_mean_list_reminded)
print(f"Mean of Sampling (Reminded) -> {mean_sampling_reminded}")
print(f"Standard Deviation of Sampling (Reminded) -> {stdev sampling reminded}")
fig = ff.create distplot([sampling mean list reminded], ["Savings (Sampling)"], show hist=False)
fig.add_trace(go.Scatter(x=[mean_sampling, mean_sampling], y=[0, 0.1], mode="lines", name="MEAN"))
fig.show()
Mean of Sampling (Reminded) -> 47.71121101865382
Standard Deviation of Sampling (Reminded) -> 4.91807596219437
```



Now we have the mean of sampling distribution of people who were given reminders and people who were not given reminders. We also have the standard deviation of the people who were not given reminders.

We find the z score to see how significantly reminders impacted savings.

Code:-

z_score =

(mean_sampling_reminded mean_sampling_not_reminded) /
stdev_sampling_not_reminded
print(f"Z-Score is - {z_score}")

Student codes to find the z score of the data.

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[| z_score = (mean_sampling_reminded - mean_sampling_not_reminded) / stdev_sampling_not_reminded
 print(f"Z-Score is - {z_score}")

Z-Score is - 0.7837451526581821

From these tests what can we conclude?

ESR:

Given all the analysis that we have done, we saw that there was not much difference between the people who got reminded about saving v/s people who were not reminded of saving.

With our Z-Score test, we confirmed that there is not much difference. The Z-Score is around 0.8; it is only when the Z-Score is greater than 2 that we consider the change as significant. Although people who were reminded had a little increase in savings, there's still no significant impact.

We also saw that there was no direct correlation between the age of the people and the money that they save.

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Teacher Guides Student to Stop Screen Share FEEDBACK Appreciate the student for their efforts Identify 2 strengths and 1 area of progress for the student So, in this data story class we ESR: Step 4: Wrap-Up reviewed the concepts we have varied (5 min) learned so far. How was your experience? Amazing. While working on this data story, we also made sure that we are at the top of all the concepts we have acquired so far. Next class, we will be learning new concepts and building new projects. **x** End Class **Teacher Clicks** Additional Encourage the student to write The student uses the Activities reflection notes in their reflection markdown editor to write her/his reflection in a journal using markdown. reflection journal. Use these as guiding questions:

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Activity	Activity Name	Links
Teacher Activity 1	Colab notebook link	https://colab.research.google.com/notebooks/intro.ipynb#recent=true
Teacher Activity 2	Colab Reference (final code)	https://colab.research.google.com/dr ive/1jkTo912MZUAmkwdx8OQ9thk1 ME4LyHrf#scrollTo=XWwnrgavkVp8
Student Activity 1	Colab notebook link	https://colab.research.google.com/n otebooks/intro.ipynb#recent=true
Student Activity 2	savings data file	https://raw.githubusercontent.com/w hitehatjr/datasets/master/savings_d ata_final.csv