PS1

Applied Stats 1

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Question 1

Section 1

Since number of observations = 25 which is smaller than 30, we will use a t-distribution with df = 24 to calculate confidence interval. We calculate the sample mean, sampling error and t-value to construct the CI.

```
1 #Load data
2 y <- c(105, 69, 86, 100, 82, 111, 104, 110, 87, 108, 87, 90, 94, 113, 112, 98, 80, 97, 95, 111, 114, 89, 95, 126, 98)
3 #Calculate lower and upper bound of confidence interval
4 sampling_mean <- mean(y)
5 sampling_sd <- sd(y)/sqrt(25)
6 t_value <- qt((1-0.9)/2 ,lower.tail = FALSE, df = 25 -1 )
7 CI <- c(sampling_mean - t_value*sampling_sd, sampling_mean + t_value*sampling_sd )</pre>
```

We have the result for 90% CI

93.95993 102.92007

Section 2

We follow the 5 step of hypothesis testing:

Step 1: Our assumption is that that the data is randomly selected, the sample is relatively small (25 < 30) and the data is quantitative. And according the question, this will be an one-sided t-test

Step 2: Setting null and alternative hypothesis H_0 : The school mean IQ = 100

 H_a : The school mean IQ > 100

Step 3: Calculating test statistics

```
\begin{array}{l} \text{sample\_mean} < - \text{ mean}(y) \\ \text{3 \#calculate } t\text{-statistic} \\ \text{4 } t = (\text{sample\_mean} - 100)/(\text{sd}(y)/\text{sqrt}(25)) \end{array}
```

We have t = -0.596

Step 4: Calculating p-value (in the direction of the alternative hypothesis)

```
#calculae p-val
pt(t, df = 24, lower.tail = FALSE)
```

We have p-value = 0.72 which is much larger than 0.05.

Step 5: Conclusion

We do not have enough evidence to reject the null hypothesis that the school student's mean IQ = 100

Question 2

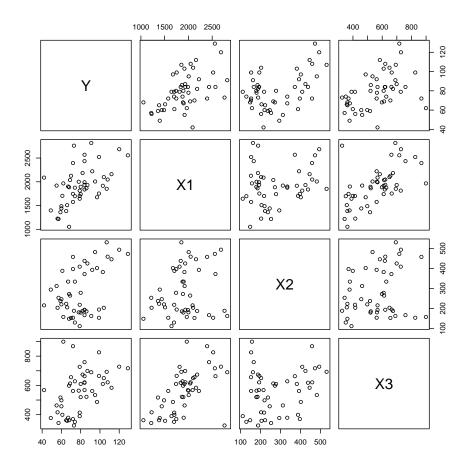
Section 1

We create scatterplot pairwise and examine variables relationship

As seen from Figures 1:

- X1 and Y seem to be positively correlated, as X1 increases, Y generally increases
- X2 and Y do not seem to be correlated, as X2 increase, some observations have Y increase, while others have Y decrease
- X3 and Y seem to be positively correlated
- X1 and X2 do not seem to be correlated
- X1 and X3 seem to be positively correlated
- X2 and X3 do not seem to be correlated

Figure 1: Scatterplot between variables.



Section 2
We use boxplot to examine relationship between Y and Region

```
pdf("/Users/tpminh/Desktop/trinity asds/stat analysis 1/ps1/boxplot.pdf")
boxplot(expenditure$Y ~ expenditure$Region)
dev.off()
```

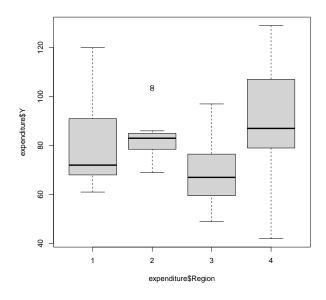
We can see that Region 4 has the highest average expenditures from Figures 2. But just to be sure, lets calculate the actual average of each region

```
\frac{aggregate(list(avg_exp = expenditure\$Y), list(Region = expenditure\$Region),}{FUN=mean}
```

Which give the result:

```
Region avg_exp
1 1 79.44444
2 2 83.91667
3 3 69.18750
```

Figure 2: Boxplot of expeniture by region.



4 4 88.30769

In conclusion Region 4 have he highest average expenditure of 88.3

Section 3

Lets examine relationship between Y and X1

plot (expenditure \$X1, expenditure \$Y)

As we have examined in the previous section, Y and X1 seem to be positively correlated (Figure 3). As X1 increase, Y will generally increase. Figure 4 changes the colour and symbol of scatterplot

```
plot ( expenditure$X1, expenditure$Y, col = expenditure$Region, pch =
    expenditure$Region)
```

Figure 3: Scatterplot between X1 and Y.

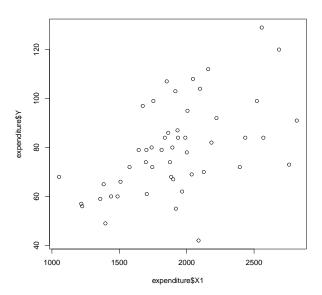


Figure 4: Scatterplot between X1 and Y with new colour and symbol.

