## Stats with R Assignment 1

#### Amanda

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```
#input data
Kiva_Example <- read.csv("Kiva_Sample.csv", header = TRUE)</pre>
## Install tidyverse
#install.packages("rmarkdown")
#install.packages("tidyverse")
#install.packages("knitr")
library("tidyverse")
## -- Attaching packages -----
                                                                    ----- tidyverse 1.3.0
## v ggplot2 3.3.2
                    v purrr
                             0.3.4
## v tibble 3.0.3
                             1.0.2
                    v dplyr
## v tidyr 1.1.2
                    v stringr 1.4.0
## v readr
         1.3.1
                    v forcats 0.5.0
                                              ----- tidyverse_conflicts()
## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
library("rmarkdown")
library("knitr")
```

#### Filter the data set to just the three countries that you selected.

The countries I selected were the United States, Puerto Rico and the Vigin Islands.

My initial country filters were not successful, so I ran the individual countries one at a time to see how many observarions were in the dataset for each country. Then upon discovering that there were 0 observations for the Virgin Islands I substituted Israel for the Virgin Islands.

```
UnitedStates <-filter(Kiva_Example, country =="United States") #1069 observations
PuertoRico <- filter(Kiva_Example, country == "Puerto Rico") #10 observations
VirginIslands <- filter(Kiva_Example, country == "Virgin Island") #0 observations

#Used a substitution country for the Virgin Islands
Israel <- filter(Kiva_Example, country == "Israel") #37 observations
```

#### Give the data set a name.

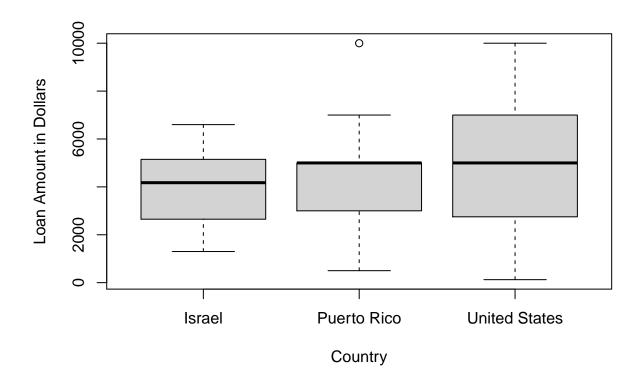
The professor called the data set "small", but I choose to call the dataset KivaComparison to make it easier to identify.

```
KivaComparison <- filter(Kiva_Example, country == "United States" | country == "Puerto Rico" | country = KivaComparison$loan_amount <- as.numeric(KivaComparison$loan_amount) #converts from character to numeri
```

### Create a graphic to explore the business question

Looking at the boxplot below we can infer that: 1. Loans in Israel typically fall within in \$2500 to \$5000. 2. Loans in Puerto Rico typically fall within a similar range to Israel, however there is a single outlier for \$10,000. 3. Loans in the U.S. typically see a wider range and higher amounts up to \$10,000 are part of the upper range.

boxplot(KivaComparison\$loan\_amount ~ KivaComparison\$country, xlab = "Country", ylab = "Loan Amount in D



## Is there a difference in the loan amounts for a crowdsourced microlending organization (KIVA) between the three countries?

To answer this question with real numbers, I ran the summary statistics on the combine country data using the KivaComparison dataset. Then I also ran the summary statistics on the individual countries.

```
""r
#The summary statistics for all groups
summary(KivaComparison$loan_amount)
##
      Min. 1st Qu.
                     Median
                                Mean 3rd Qu.
                                                 Max.
##
       125
               2750
                       5000
                                5045
                                        6581
                                                10000
# Min. 1st Qu.
                 Median
                           Mean 3rd Qu.
                                            Max.
# 125
         2750
                  5000
                          5045
                                   6581
                                          10000
#The summary statistics for the United States
UnitedStates$loan_amount <- as.numeric(UnitedStates$loan_amount)</pre>
summary(UnitedStates$loan_amount)
```

```
##
      Min. 1st Qu.
                     Median
                                Mean 3rd Qu.
##
       125
               2750
                       5000
                                5089
                                         7000
                                                10000
# Min. 1st Qu.
                Median
                           Mean 3rd Qu.
# 125
         2750
                  5000
                           5089
                                   7000
                                           10000
#The summary statistics for the Puerto Rico
PuertoRico$loan_amount <- as.numeric(PuertoRico$loan_amount)</pre>
summary(PuertoRico$loan_amount)
##
      Min. 1st Qu.
                     Median
                                Mean 3rd Qu.
                                                 Max.
##
       500
               3500
                       5000
                                4600
                                         5000
                                                10000
# Min. 1st Qu. Median
                            Mean 3rd Qu.
                                             Max.
# 500
         3500
                  5000
                           4600
                                   5000
                                           10000
#The summary statistics for the Israel
Israel$loan_amount <- as.numeric(Israel$loan_amount)</pre>
summary(Israel$loan_amount)
      Min. 1st Qu.
##
                     Median
                                Mean 3rd Qu.
                                                 Max.
##
      1300
               2650
                       4175
                                3907
                                         5150
                                                 6600
#Min. 1st Qu.
                Median
                           Mean 3rd Qu.
                                            Max.
#1300
         2650
                  4175
                           3907
                                   5150
                                            6600
```

# Is there a difference in the loan amounts for a crowdsourced microlending organization (KIVA) between the three countries?

When looking at the summary statistics for the United States loan amounts we can see there is a strong resemblance to the summary statistics for all three countries. This is because out of the 1116 observations, 1069 of them are for loans in the United States.

- 1. So the first observation is that there are more loans in the United States than either Puerto Rico or Israel
- 2. The United States has the largest range of loan amounts from \$175 to \$10,000.
- 3. The United States has the smallest loan amount of \$175. The minimum loan given for Puerto Rico was \$500 and for Israel was \$1300.
- 4. The United States and Israel tied for the highest loan amount of \$10,000 and matched median loan amounts of \$5,000.

#### Conduct a one way ANOVA test.

Make sure to include the parameters, hypothesis, assumptions, test statistic, p-value and conclusion.

#Hypothesis Null Hypothesis: That loan amounts in all three countries, the United States, Puerto Rico and Israel are equal. Alternative Hypothesis: The loan amounts in all three countries, The United States, Puerto Rico and Israel are not equal.

#Assumptions 1. Normal Distribution: We are not able to assume a normal distribution of data since there are only 10 Puerto Rico loans, which is under the 30 observation threshold. 2. Outliers: By looking at the boxplot we see there is a single outlier in the Puerto Rico dataset. 3. Randomization: Since the dataset is all loans, the indivdual amounts are randomized.

4. Independence: The loans in one country are independent from the loans in any other country.

```
'''r
#Anova
```

```
LoanAmount <-aov(KivaComparison$loan_amount ~ KivaComparison$country)
anova(LoanAmount)

## Analysis of Variance Table

##

## Response: KivaComparison$loan_amount

## Df Sum Sq Mean Sq F value Pr(>F)

## KivaComparison$country 2 51891424 25945712 3.0153 0.04943 *

## Residuals 1113 9577066848 8604732

## ---

## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

#### **Interputating Anova Results**

Test statistic 3.0153 P-value .04943

conclusion: With a p-value under .05, we have evidence that at least one of the population mean loan amounts is different from the others.

#### Limitations

Due to the assumptions of a normal distribution not being properly met for Puerto Rico