

# Food Inflation Study

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```
knitr::opts_chunk$set(echo = TRUE)

library(glmnet)
```

```
## Loading required package: Matrix
```

```
## Loaded glmnet 4.1-8
```

```
library(MASS)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:MASS':
##
##      select
```

```
## The following objects are masked from 'package:stats':
##
##      filter, lag
```

```
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union
```

```
# library(emergency_back_up_brain)
# library(coffee)
```

## R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <http://rmarkdown.rstudio.com>.

In this R document I will be specifically just cleaning data and correlating data points that are relevant to modeling, looking at the distribution of the data and saving those back to a new CSV file to be visually assessed in Python and modeled in R.

Below is the loaded CSV files needed to start cleaning for modeling and analysis work. The link to get the food inflation information is <https://microdata.worldbank.org/index.php/catalog/4483> from December 2019

to December 2023 to complete 4 years of data on international food inflation. “Monthly food price estimates by product and market 25 countries, 1353 markets, 2007/01/01-2023/12/01, version 2023/12/11” -The World Bank of Microdata website. There are two other factors in the analysis work, that is conflict (gun violence and war) and the US stock Market prices from December 2019 to December 2023.

The current HO: “The occurrence of conflict in key food-producing regions, coupled with fluctuations in the US stock market, significantly influences international food inflation rates. Higher instances of conflict and volatility in the US stock market are expected to correlate with increased food inflation on a global scale.” The current Alternative HO : “The impact of conflict and US stock market fluctuations on international food inflation rates may not exhibit a significant correlation.”

```
# Food inflation data
food_data <- read.csv("Food_inflation_2019_2023.csv", header=TRUE, sep=",")
# time to clean this up
clean_food_data <- subset(food_data, select = -X)
clean_food_data <- na.omit(clean_food_data)
clean_food_data <- subset(clean_food_data, select = -Market)
clean_food_data <- subset(clean_food_data, select = -Currency)
```

Narrowing down the top most conflicted countries in the past 4 years according to Wikipedia [https://en.wikipedia.org/wiki/List\\_of\\_ongoing\\_armed\\_conflicts](https://en.wikipedia.org/wiki/List_of_ongoing_armed_conflicts) sited. Mexico, Ukraine, Afghanistan, Syria (Syrian Arab Republic), Ethiopia, Yemen

```
conflicted_countries_food_data <- clean_food_data %>%
  filter(Country %in% c("Afghanistan", "Syrian Arab Republic", "Mexico", "Yemen", "Ukraine", "Ethiopia"))
str(conflicted_countries_food_data)
```

```
## 'data.frame': 145236 obs. of 8 variables:
## $ Country: chr "Afghanistan" "Afghanistan" "Afghanistan" "Afghanistan" ...
## $ Region : chr "Badakhshan" "Badakhshan" "Badakhshan" "Badakhshan" ...
## $ Product: chr "bread" "bread" "bread" "bread" ...
## $ Date : chr "2019-12-01" "2019-12-01" "2020-01-01" "2020-01-01" ...
## $ Open : num 49.8 49.8 49.8 49.8 49.9 ...
## $ High : num 49.9 49.9 49.9 49.9 50 ...
## $ Low : num 49.7 49.7 49.7 49.7 49.8 ...
## $ Close : num 49.8 49.8 49.9 49.9 49.9 ...
```

```
conflicted_countries_food_data <- conflicted_countries_food_data %>%
  filter(Open != 0)
conflicted_countries_food_data <- conflicted_countries_food_data %>% filter(Close != 0)
conflicted_countries_food_data <- conflicted_countries_food_data %>% filter(High != 0)
```

*#Isolate Food Price Index*

```
conflicted_countries_food_price_index <- conflicted_countries_food_data[conflicted_countries_food_data$Product == "food_price_index", ]
conflicted_countries_food_data <- conflicted_countries_food_data %>%
  filter(Product != "food_price_index")
str(conflicted_countries_food_price_index)
```

```
## 'data.frame': 9898 obs. of 8 variables:
## $ Country: chr "Afghanistan" "Afghanistan" "Afghanistan" "Afghanistan" ...
## $ Region : chr "Badakhshan" "Badakhshan" "Badakhshan" "Badakhshan" ...
## $ Product: chr "food_price_index" "food_price_index" "food_price_index" "food_price_index" ...
## $ Date : chr "2019-12-01" "2019-12-01" "2020-01-01" "2020-01-01" ...
```

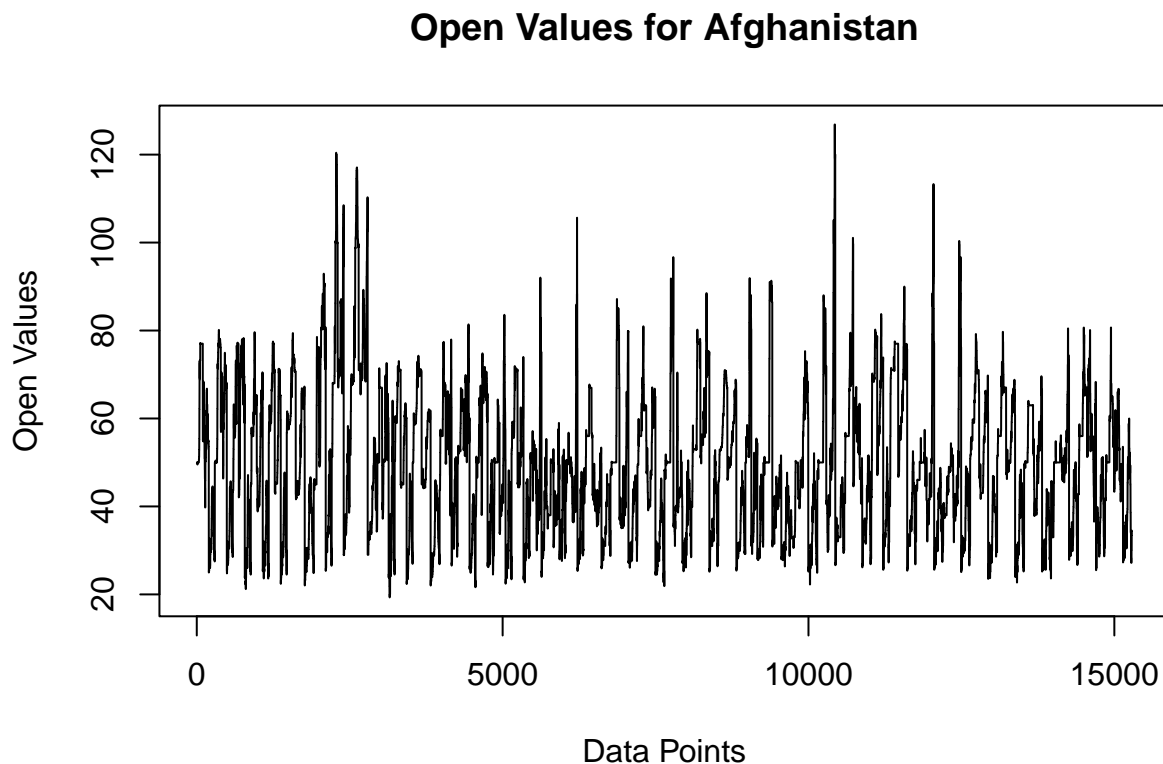
```
## $ Open   : num  1.12 1.12 1.13 1.13 1.13 1.13 1.13 1.13 1.23 1.23 ...
## $ High   : num  1.13 1.13 1.14 1.14 1.14 1.14 1.2 1.2 1.25 1.25 ...
## $ Low    : num  1.11 1.11 1.12 1.12 1.12 1.12 1.12 1.12 1.21 1.21 ...
## $ Close  : num  1.13 1.13 1.13 1.13 1.13 1.13 1.2 1.2 1.25 1.25 ...
```

Afghanistan data on Food Inflation open and close, the difference between the two. Its important to look at the distribution of the data to do further analysis and cleaning of the data. If the differences between open and close prices themselves follow a normal distribution, it could suggest a certain level of regularity and randomness in price movements.

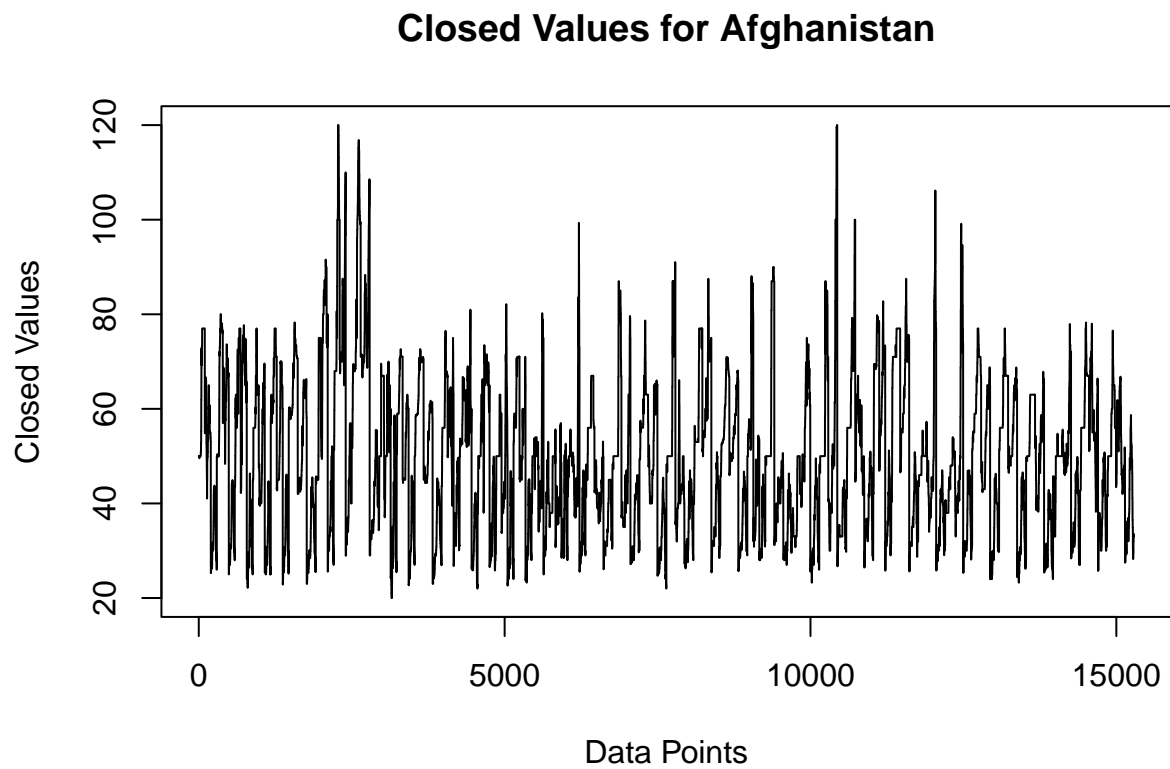
```
data_afghanistan <- conflicted_countries_food_data[conflicted_countries_food_data$Country == "Afghanistan", ]

open_values_afghanistan <- data_afghanistan$Open
closed_values_afghanistan <- data_afghanistan$Close
difference_afghanistan <- open_values_afghanistan - closed_values_afghanistan

# Create a sequence for x-axis (assuming you want a sequence of numbers as x-axis)
x <- seq(length(open_values_afghanistan))
x2 <- seq(length(closed_values_afghanistan))
x3 <- seq(length(difference_afghanistan))
# Plotting Open values for Afghanistan
plot(x, open_values_afghanistan, type = "l", xlab = "Data Points", ylab = "Open Values", main = "Open Values for Afghanistan")
```

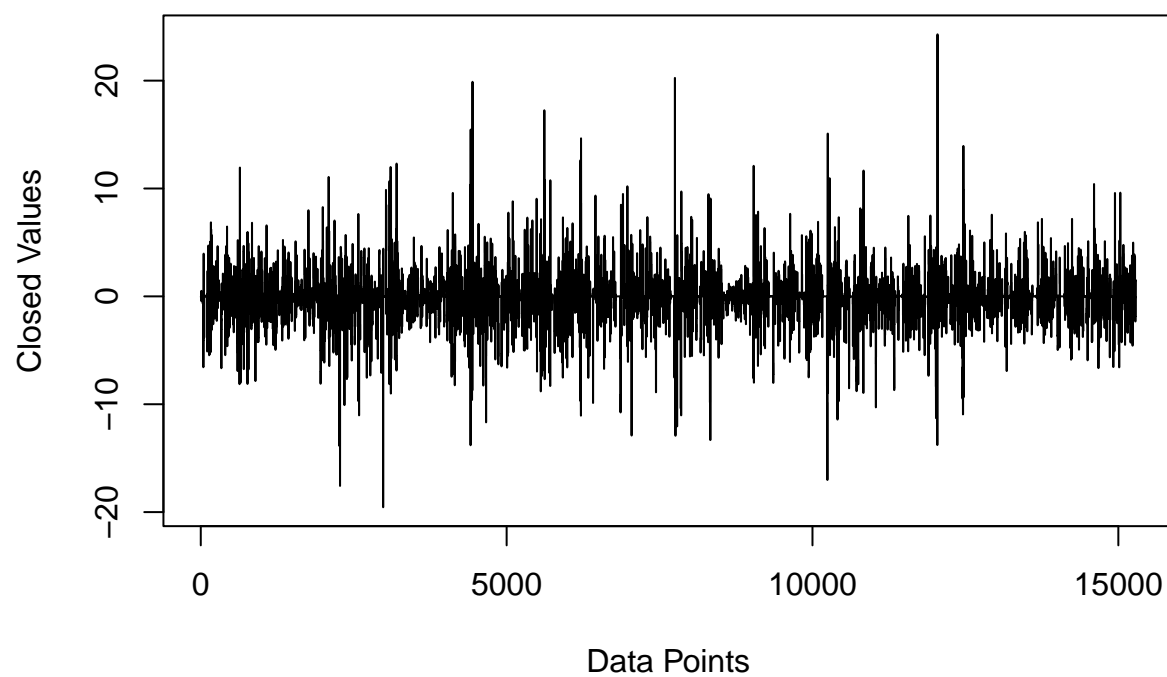


```
plot(x2, closed_values_afghanistan, type = "l", xlab = "Data Points", ylab = "Closed Values", main = "C
```

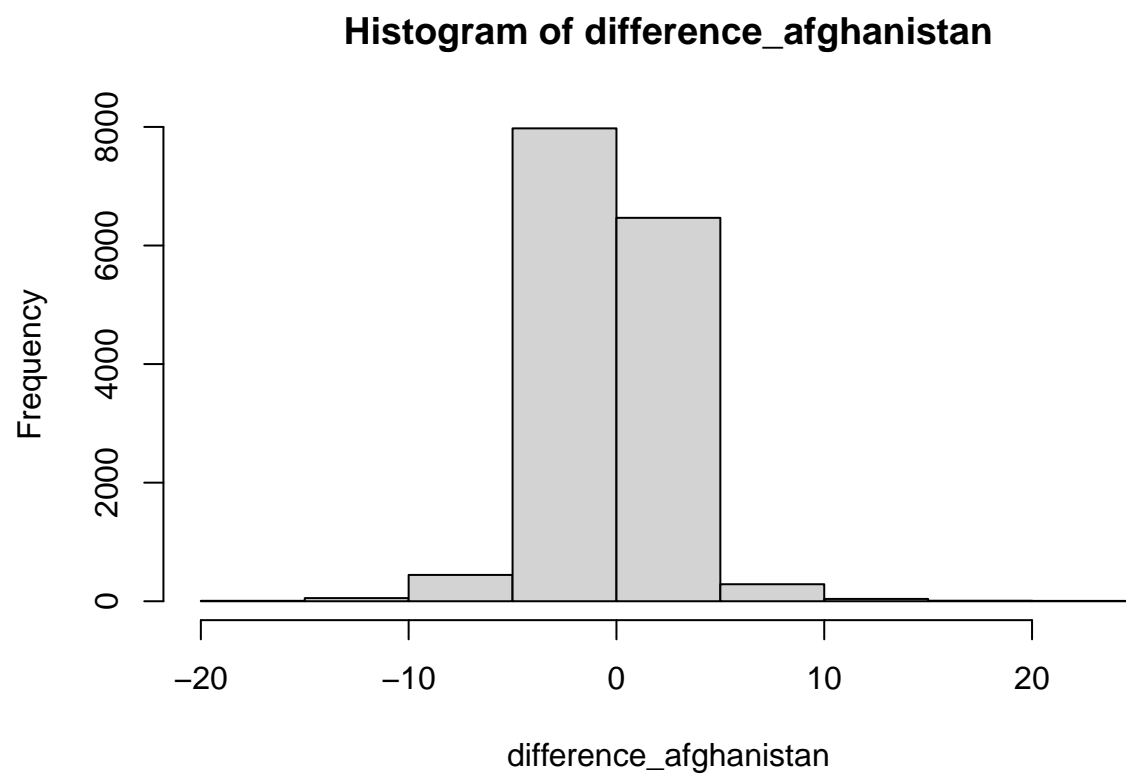


```
plot(x3, difference_afghanistan, type = "l", xlab = "Data Points", ylab = "Closed Values", main = "Clos
```

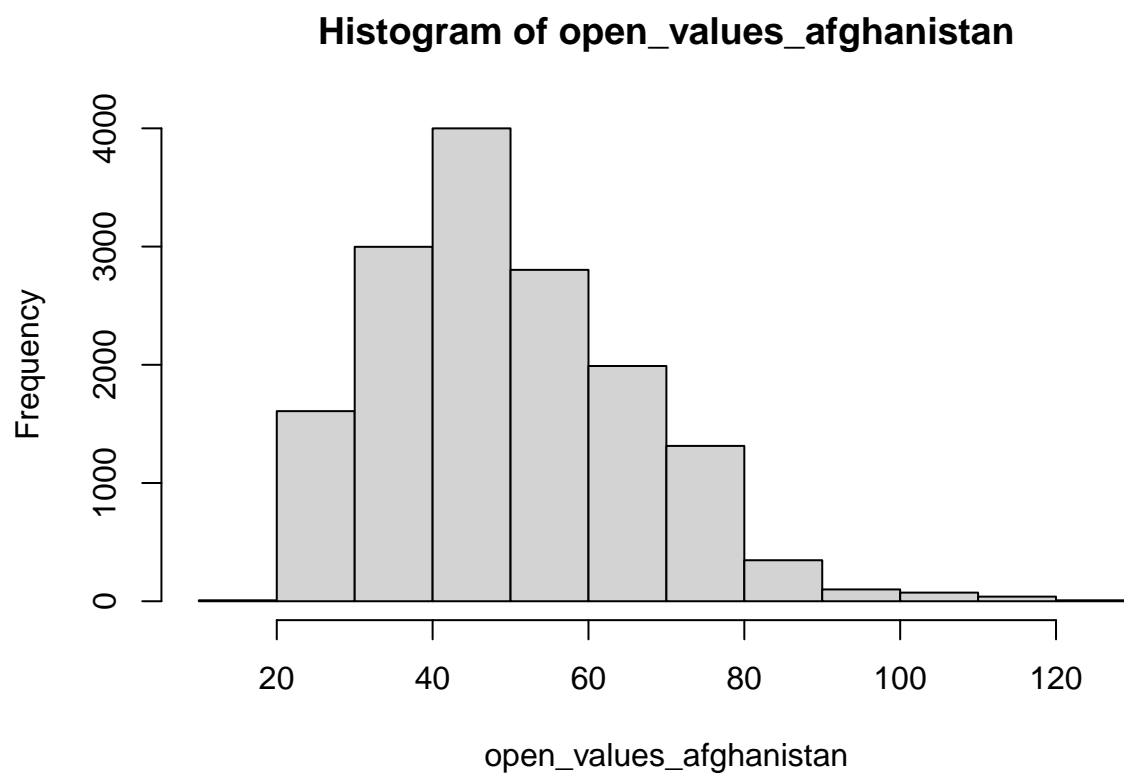
## Closed Values for Afghanistan



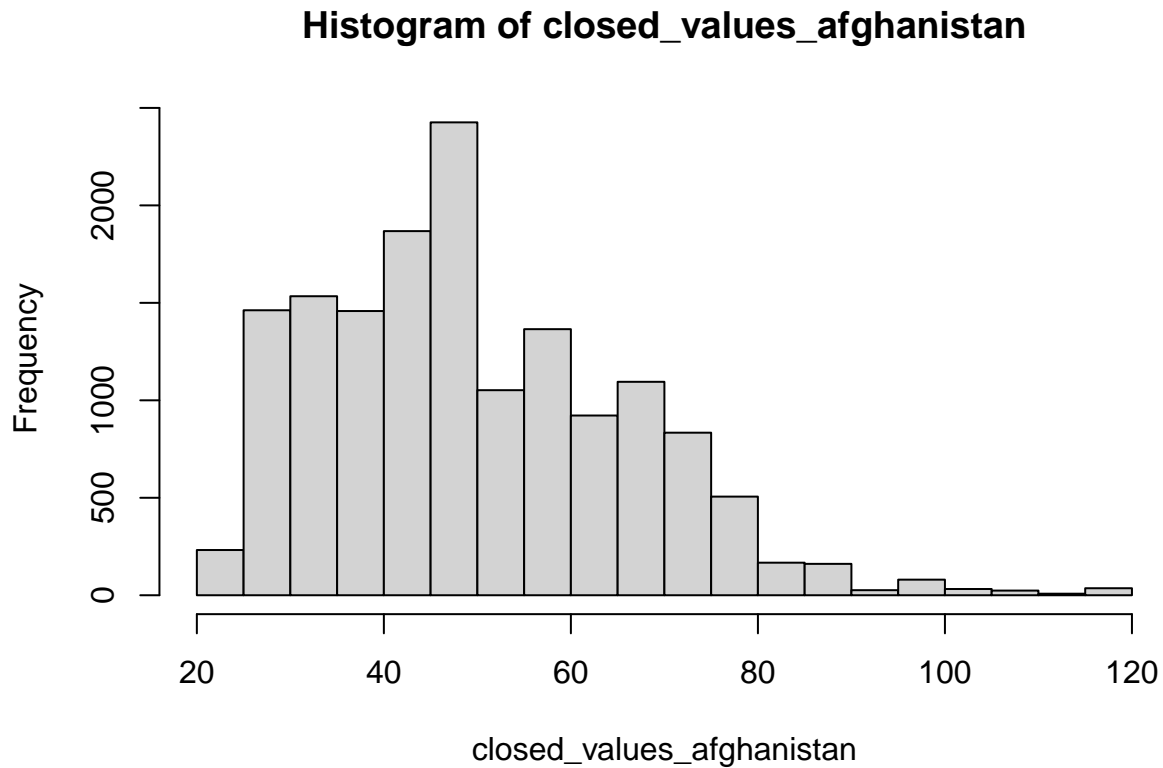
*#seems to be normally distributed, this helps narrow down appropriate models to examine Afghanistan spe*  
`hist(difference_afghanistan)`



```
# open and close are distributed to the right  
hist(open_values_afghanistan)
```



```
hist(closed_values_afghanistan)
```



In financial markets, a normal distribution of price index differences might support the idea of market efficiency, suggesting that market prices reflect all available information, and arbitrage opportunities might be limited. Furthermore, the questioning factor still remains that the open and close histograms are right skewed, which indicate otherwise. Its considered a bullish market indication.

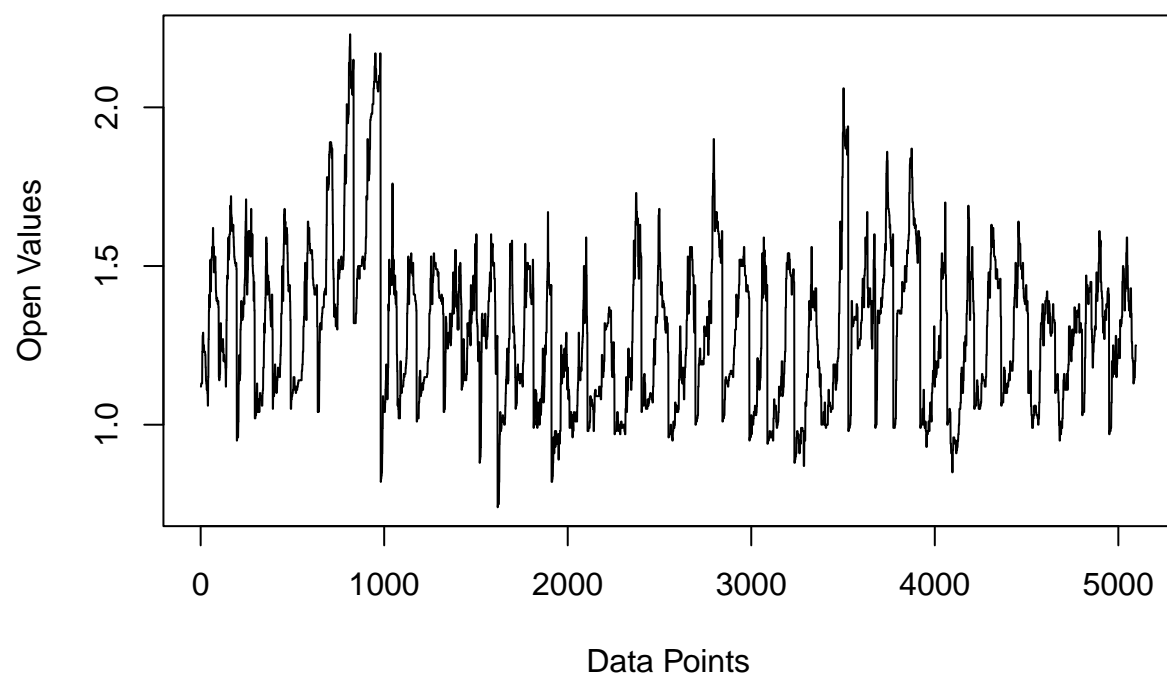
```
# check to values of the Food Index Prices and see if they reflect a normal distribution or not
# Food Price Index
index_afghanistan_data <- conflicted_countries_food_price_index[conflicted_countries_food_price_index$Country == "Afghanistan", ]
index_afghanistan_open <- index_afghanistan_data$Open
index_afghanistan_close <- index_afghanistan_data$Close
index_afghanistan_difference <- index_afghanistan_open - index_afghanistan_close

x1_index <- seq(length(index_afghanistan_open))
x2_index <- seq(length(index_afghanistan_close))
x3_index <- seq(length(index_afghanistan_difference))

plot(x1_index, index_afghanistan_open, type = "l", xlab = "Data Points", ylab = "Open Values", main = "Open Values")
```

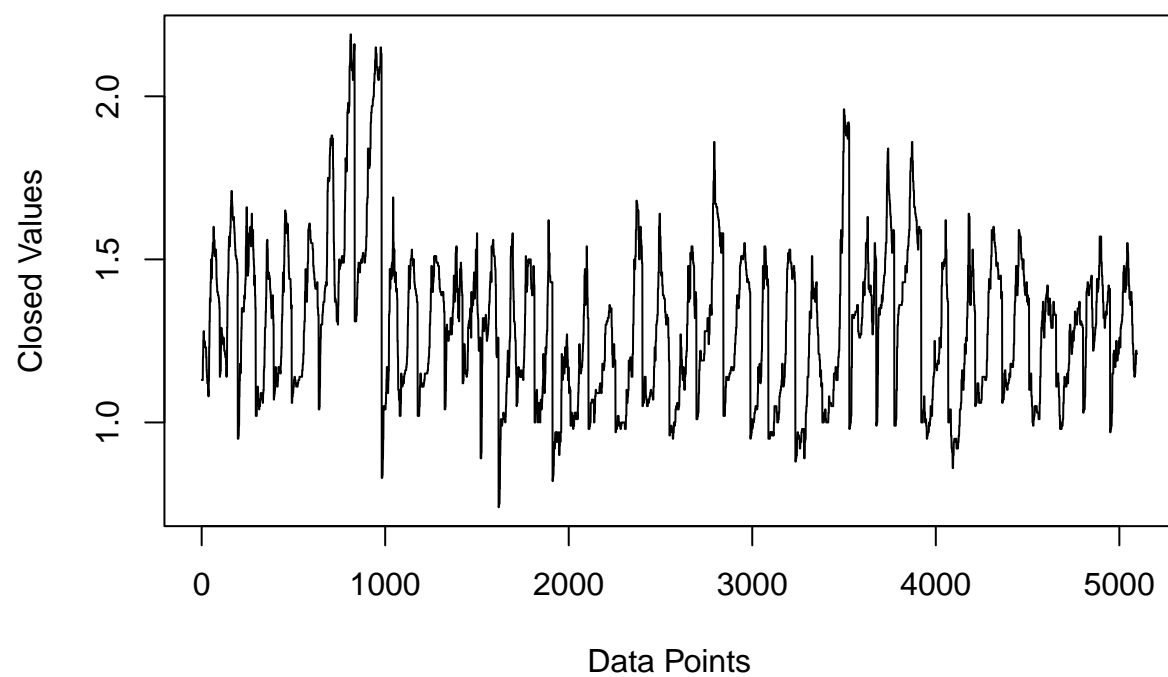


## Open Values for Food Index Prices Afghanistan



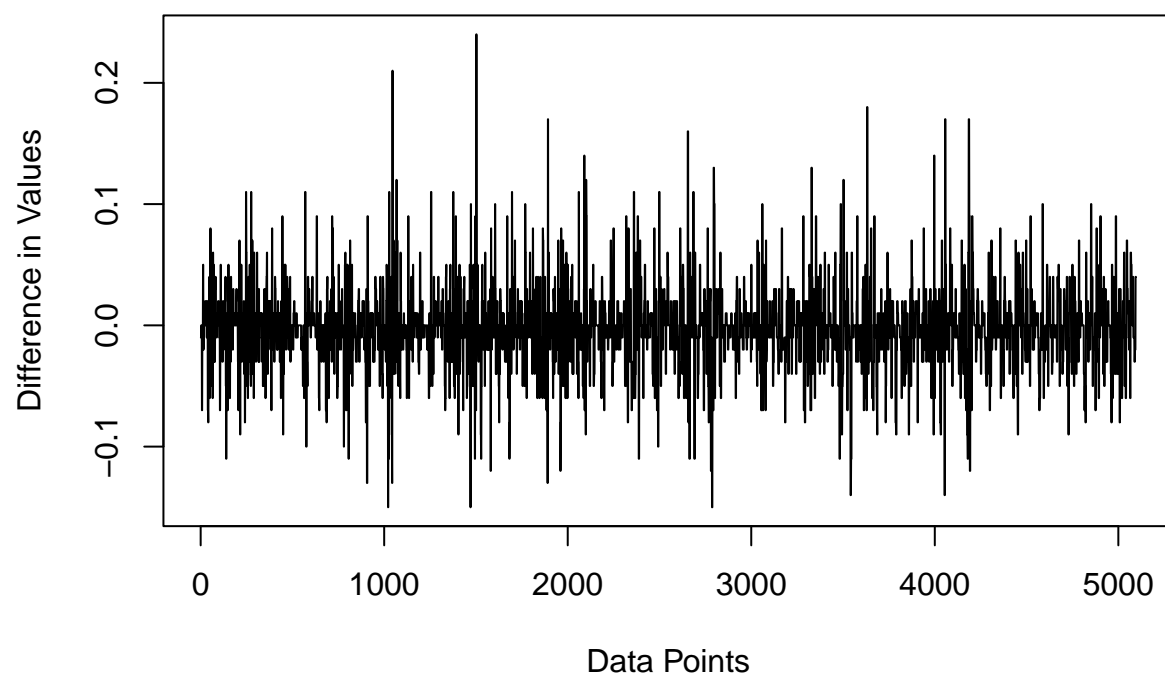
```
plot(x2_index, index_afghanistan_close, type = "l", xlab = "Data Points", ylab = "Closed Values", main = "Closed Values for Food Index Prices Afghanistan")
```

## Open Values for Food Index Prices Afghanistan

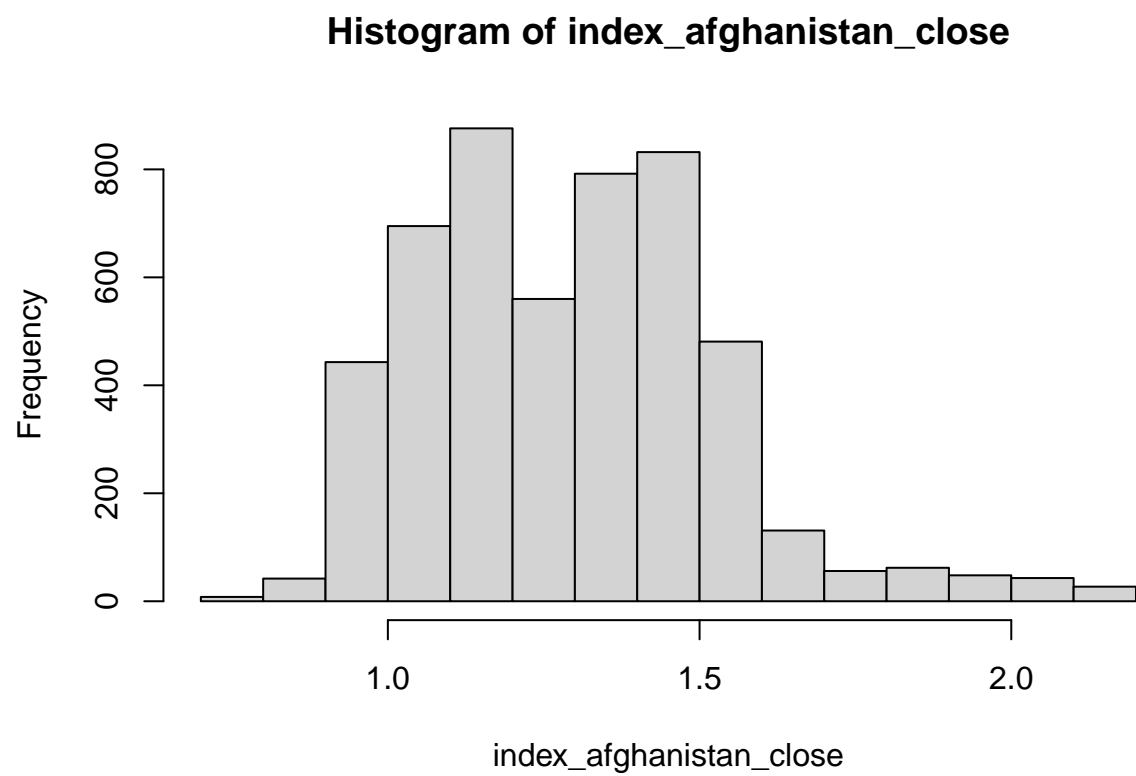


```
plot(x3_index, index_afghanistan_difference, type = "l", xlab = "Data Points", ylab = "Difference in Va
```

## Difference in Values for Food Index Prices Afghanistan

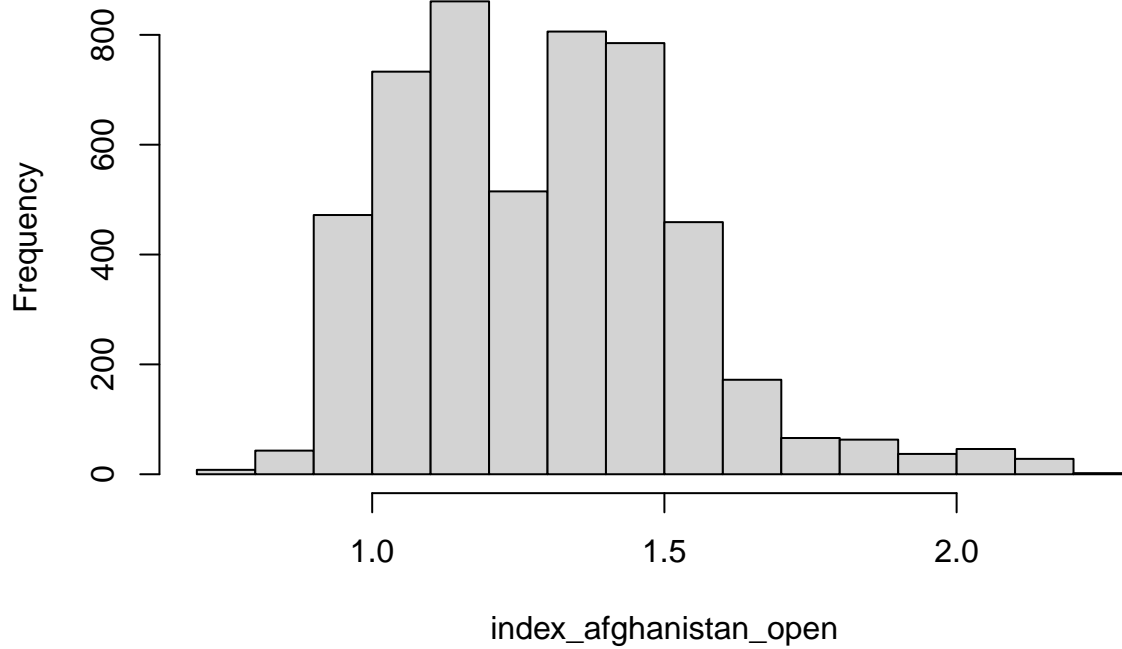


```
# these are skewed to the right to the higher values  
hist(index_afghanistan_close)
```



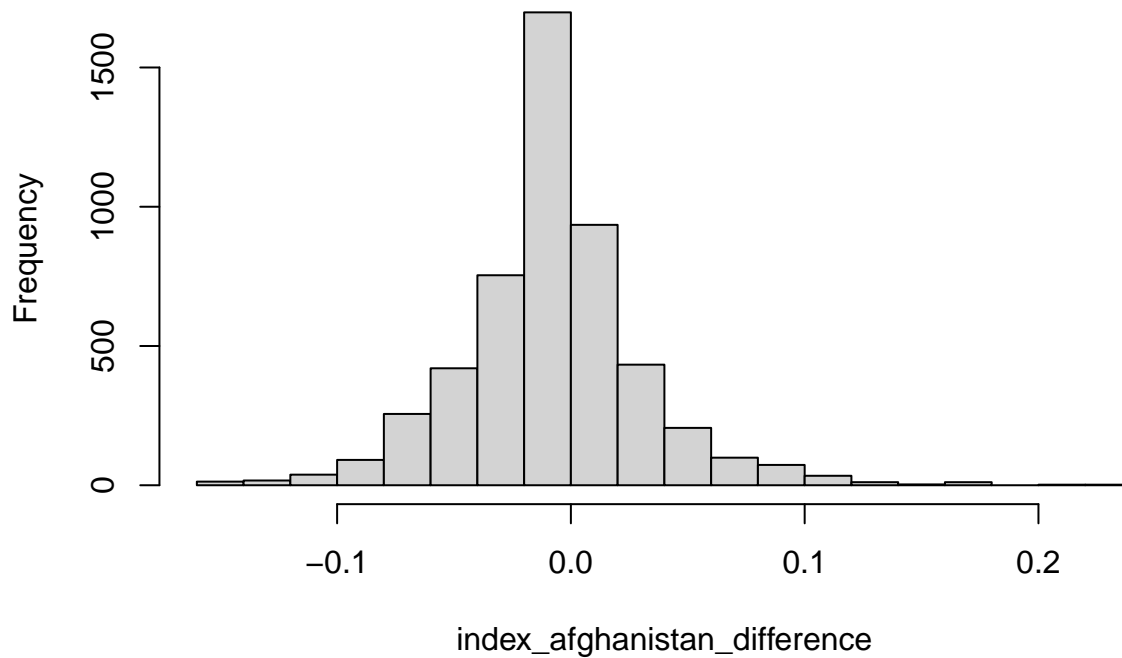
```
hist(index_afghanistan_open)
```

**Histogram of index\_afghanistan\_open**



```
# this seems to hold a normal distribution which is probably best to work with-  
hist(index_afghanistan_difference)
```

## Histogram of index\_afghanistan\_difference



Below is the conflict data which is from <https://acleddata.com/data-export-tool/>. This requires a access key to get this data, it seems to be the most up to date data on conflicts. That coincides with the Dates from the Food inflation.

```
# Conflict data
```

Below is the the US Stock Market Data which is from <https://www.ers.usda.gov/data-products/wheat-data/> and Yahoo Stocks for rough rice <https://finance.yahoo.com/quote/ZR%3DF/history?period1=1575158400&period2=1703030400&interval=1d&filter=history&frequency=1d&includeAdjustedClose=true>.

```
# US Stock Market Data
```