

Configuration Manual

Impact and Implications Analysis of Fintech and International Trade & Supply Chain Finance

> MSc Research Project Programme Name: MSC FINTECH

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MSc Project Submission Sheet

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Configuration Manual Fintech and International Trade & Supply Chain Finance Narasimhan Kannan Student ID: X18100040

1 General Information

Aim of this manual is to assist the other user to replicate the model developed in this study. The user's manual includes the modus operandi and algorithm processed in "R" studio. For this study two datasets are used, one from Kaggle trade statistics and other created virtual data sets from reliable sources. This manual consists of data background, application of CRISP-DM approach, model building, predictions, graph plotting.

2 Getting Started

2.1 Application of Time series for data set 1 -for Traditional banking trade flow

2.2 Data sets and background.

This data set is from Kaggle database and includes the imports and exports volumes of commodities of various countries in last 2 decades. The attributes consist of year, country, commodity, export, and import trade flow in USD, etc (10 columns and > 0.82 million rows). The rational for choosing this data set is to arrive a benchmark for the trade flows and presume that these flows were transacted through banking channels where 80% to 90% utilizing trade finance facility (Xin Fin Organization, 2018) by buyers and sellers.

2.3 Install Library:

Before the data sets imported into "R" following package to be loaded.

Loading the libraries required for time series install. packages('dplyr') library(timeSeries) library(timeDate) library(tseries) library(forecast) library(dplyr) library(ggplot2) install.packages("psych") library(psych)

2.4 Ratio Split and Scientific notation (E)

In order to Train and Test the data sets, ratios of 80:20 is assigned and used "options (scipen=999) "for scientific notation.

2.5 Read the data into R

As the data is large, data is split into two files and merged using "rbind "function in "R"

```
tradedata1 <- read.csv(file="D:/Research/Data/Data_part1.csv",header=TRUE, sep=",")
tradedata2 <- read.csv(file="D:/Research/Data/Data_part2.csv",header=TRUE,sep=",")
tradedata <- rbind(tradedata2,tradedata1)
```

2.6 Exploratory Data Analytics

2.6.1 To check Missing data and Header:

```
which(is.na(tradedata$country_or_area))
which(is.na(tradedata$year))
which(is.na(tradedata$trade_usd))
describe(na.omit(tradedata))
head(tradedata)
```

2.6.2 Identify the number of rows and columns

nrow(tradedata1) + nrow(tradedata2)

It is noted that this data set consists of 10 columns and 821707 observations and null missing values for key variables such as country, Year USD and Trade flow.

2.6.3 Descriptive statistics

summary(tradedata\$trade_usd)
str(tradedata)
names(tradedata)

2.7 Identify of Top 10 countries including import and export in Trade - USD Aggregated.

Since the data includes the import and export in USD, values are aggregated based on the countries to find top 10 countries. The result is in USD billion (descending order) and displaced using Bar Chart.

• # top 10 countries including import and export in Trade USD Aggregated.

Code

tradedata_top<-tradedata%>% group_by(country_or_area) %>% summarise(USD=round(sum(trade_usd)/100000000)) %>% arrange(USD)%>% tail(10)

• # to Display Top 10 countries

tradedata_top

• # To display in GG plot for 10 countries

```
p <- ggplot(df, aes(x= reorder(country_or_area,-USD), y=USD, fill=country_or_area)) +
geom_bar(stat="identity")+theme(aspect.ratio = 1)+
ggtitle("Top 10 countries in terms of trade flow in USD")
plot(p)</pre>
```

2.8 Year wise aggregation of trade flow for Top 5 countries

• #For Canada:

tradedata_sum<-tradedata%>% group_by(country_or_area, year)%>% summarise (USD=round(sum(trade_usd)/1))%>% arrange(USD)%>% filter(country_or_area == 'Canada') tradedata_canada<-data.frame(year=tradedata_sum\$year,usd =round(tradedata_sum\$USD/100000000))

similarly, replace the Canada to other 4 countries namely EU-28, China, Germany, and France

• #To check data frame: class(tradedata_eu28) "data.frame"

2.9 Converting to Timeseries for each of top 5 countries

As data identified on above steps is a data frame format and in order to pass the data into the timeseries model, the data is converted using <u>ts</u> function from start of year 1999. However, for EU, the start date is from 2000.

tradedata_canada.ts <- ts(tradedata_canada\$usd,start=1989,end=2016) tradedata_france.ts <- ts(tradedata_france\$usd,start=1994,end=2016) tradedata_germany.ts <- ts(tradedata_germany\$usd,start=1991,end=2016) tradedata_china.ts <- ts(tradedata_china\$usd,start=1992,end=2016) tradedata_eu28.ts <- ts(tradedata_eu28\$usd,start=2000,end=2015)

to check whether Timeseries converted

class(tradedata_canada.ts)

"ts"

2.10 Plot the Time Series

In order to identify the original trends for each country, the data are plotted using **plot. ts.**

plot.ts(tradedata_canada.ts, main="Canada - Time Series Original") plot.ts(tradedata_france.ts, main="France - Time Series Original") plot.ts(tradedata_germany.ts, main="Germany - Time Series Original") plot.ts(tradedata_china.ts, main="China - Time Series Original") plot.ts(tradedata_eu28.ts, main="EU-28 Time Series Original")

3 Model Building

To build the model, this data is trained and tested as defined by ratio (80:20) and performed for top 5 countries such as Canada, France, Germany, China, and EU-28 and then applied **Auto Arima model**

```
• # Time series Modelling using Train and Test
#Canada
samplesize = ratio * nrow(tradedata_canada)
trainrows <- floor(samplesize)
testrows <- nrow(tradedata_canada) - trainrows
```

train_can <- trainrows
test_can <- testrows</pre>

• # train and test split based on ratio. train.ts.d <- ts(tradedata_canada.ts[1:trainrows],1988+1,end=1988+trainrows) test.ts.d<ts(tradedata_canada.ts[trainrows+1:(trainrows+testrows)],start=1988+trainrows+1,end=1988 +trainrows+testrows)

```
• #Auto Fitting with seasonal
auto.fit1 <- auto.arima(train.ts.d,seasonal=T)
summary(auto.fit1)
auto.fcast1 <- forecast(auto.fit1,h=testrows)
plot(auto.fcast1,main="Canada")
auto.fcast1
```

```
• # root mean square error
pred_canada = auto.fcast1$mean
RMSE_canada = sqrt(sum(test.ts.d-pred_canada)^2/testrows)
test.ts.d
pred_canada
RMSE_canada
```

similarly, replace the Canada to other 4 countries namely EU-28, China, Germany, and France and list the Performance Matrix

4 Prediction for next 10 years trade flow using Auto ARIMA Model

Based on the past trends and using the auto ARIMA model project the future sales for next 10 years of trad flow.

Predection for next 10 years #Auto Fitting with seasonal # Canada auto.fit1 <- auto.arima(tradedata_canada.ts,seasonal=T) summary(auto.fit1) auto.fcast1 <- forecast(auto.fit1,h=10) plot(auto.fcast1,main="Canada - Next 10 years forecast") auto.fcast1

Similarly, replace the Canada to other 4 countries namely EU-28, China, Germany, and France

Data Set 2

5 Fitech data set 1 -To identify the interest rate to decide profitability (World Bank Data)

5.1 Read the values

filedata <-read.csv(file="D:/Research/Data/ARIMA .csv",sep=",",header=TRUE)

5.2 To decide plot per frame

par(mfrow=c(1,1))

5.3 To assign value to create line

x <- Fintechdata\$Year y0 <- c(11.0, 10.0, 9.0, 8.5, 8.0, 7.0, 7.0, 6.0, 6.0, 6.0) y1 <- Fintechdata\$FinTechRate y2 <- Fintechdata\$GlobalAverage y3 <- Fintechdata\$MoneyTranOper

5.4 To Create a first and Subsequent line

Add a first line

lines(x, y1, pch = 20, col = "red", type = "b", lty = 2)

Add a second line

lines(x, y2, pch = 20, col = "blue", type = "b", lty = 2)

Add a third line

lines(x, y3, pch = 20, col = "green", type = "b", lty = 2)

5.5 # Add a legend to the plot

6 Data Set2 – For Checking Liquidity

6.1 Read the values

filedata <-read.csv (file="D:/Research/Data/Data set 2 Fintech.csv",sep=",",header=TRUE)

6.2 For Global Online Disbursement

Create a first line

plot(x, y0, type = "l", frame = FALSE,

col = "white", xlab = "Year", ylab = "Disbursement",main="Global online Disbursement USD (Million)")

Add a first line

lines(x, y1, pch = 20, col = "red", type = "b", lty = 2)

Add a second line

lines(x, y2, pch = 20, col = "blue", type = "b", lty = 2)

Add a second line

lines(x, y3, pch = 20, col = "green", type = "b", lty = 2)

Add a second line

lines(x, y4, pch = 20, col = "brown", type = "b", lty = 2)

Add a legend to the plot

legend("topleft", legend=c("EU-28", "France", "Germany", "Canada"), col=c ("red", "blue", "green", "brown"), lty = 1:2, cex=0.8)

6.3 To plot china separately - due to wide range in values where China is in Billion and other countries in Million

to decide plot per frame
par(mfrow=c(1,1))

x <- Fintechdata1\$Year y0 <- c(0, 10000, 20000, 24000, 30000, 31000, 28000, 500, 550, 520000) y5 <- Fintechdata1\$China_Disb

Create a first line

lines(x, y5, pch = 20, col = "black", type = "b", lty = 2)

Add a legend to the plot

6.4 For Invoice Trading Disbursement

```
# to decide plot per frame
par(mfrow=c (1,1))
```

To assign value to create line

x <- Fintechdata1\$Year y0 <- c(0, 100, 200, 240, 300, 400, 450, 500, 550, 780) y1 <- Fintechdata1\$EU28_InvTrade y2 <- Fintechdata1\$France_InvTrade y3 <- Fintechdata1\$Germany_InvTrade y4 <- Fintechdata1\$Canada_InvTrade # Create a first line plot(x, y0, type = "l", frame = FALSE, col = "white", xlab = "Year", ylab = "Disbursement",main="Invoice Trading Disbursement USD (Million)")

Add a first line

lines(x, y1, pch = 20, col = "red", type = "b", lty = 2)

Add a second line lines(x, y2, pch = 20, col = "blue", type = "b", lty = 2)

Add a second line lines(x, y3, pch = 20, col = "green", type = "b", lty = 2)

Add a second line
lines(x, y4, pch = 20, col = "brown", type = "b", lty = 2)

Add a legend to the plot

6.5 To plot china- due to wide range in values where China is in Billion and other countries in Million

to decide plot per frame
par(mfrow=c(1,1))

x <- Fintechdata1\$Year y0 <- c(0, 1000, 2000, 2400, 3000, 3100, 2800, 500, 550, 5000) y5 <- Fintechdata1\$China_InvTrade

Create a first line

lines(x, y5, pch = 20, col = "black", type = "b", lty = 2)

Add a legend to the plot

References

Xin Fin Orginization, 2018. Blockchain technology for global trade and finance. Xin Fin Orginization