

# TimeSeries\_Forecast

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## Time Series - Forecast

This document summarizes the learning process on Time Series analysis. The original source of the learning material is from Datacamp.

### Preparation

Load the dataset and the required packages

```
library(tidyverse)
```

```
## -- Attaching packages -----  
  
## v ggplot2 3.3.2      v purrr  0.3.4  
## v tibble  3.0.2      v dplyr  1.0.0  
## v tidyr   1.1.0      v stringr 1.4.0  
## v readr   1.3.1      v forcats 0.5.0  
  
## -- Conflicts -----  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()    masks stats::lag()
```

```
library(lubridate)
```

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

```
library(forecast)
```

```
## Registered S3 method overwritten by 'quantmod':  
##   method      from  
##   as.zoo.data.frame zoo
```

```
library(Hmisc)
```

```
## Loading required package: lattice

## Loading required package: survival

## Loading required package: Formula

##
## Attaching package: 'Hmisc'

## The following objects are masked from 'package:dplyr':
##
##     src, summarize

## The following objects are masked from 'package:base':
##
##     format.pval, units
```

```
library(magrittr)
```

```
##
## Attaching package: 'magrittr'

## The following object is masked from 'package:purrr':
##
##     set_names

## The following object is masked from 'package:tidyr':
##
##     extract
```

```
ptt_price <- read.csv(choose.files())
head(ptt_price)
```

```
##   adm0_id adm0_name adm1_id      adm1_name mkt_id mkt_name cm_id
## 1    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
## 2    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
## 3    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
## 4    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
## 5    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
## 6    205   Rwanda  21973 $West/Iburengerazuba  1045  Birambo  148
##           cm_name cur_id cur_name pt_id pt_name um_id um_name mp_month mp_year
## 1 Potatoes (Irish)    77    RWF    15  Retail    5    KG      11    2010
## 2 Potatoes (Irish)    77    RWF    15  Retail    5    KG      12    2010
## 3 Potatoes (Irish)    77    RWF    15  Retail    5    KG       1    2011
## 4 Potatoes (Irish)    77    RWF    15  Retail    5    KG       2    2011
## 5 Potatoes (Irish)    77    RWF    15  Retail    5    KG       3    2011
## 6 Potatoes (Irish)    77    RWF    15  Retail    5    KG       4    2011
##   mp_price mp_commoditysource
```

```
## 1 157.0000      MINAGRI
## 2 133.3333      MINAGRI
## 3  96.5000      MINAGRI
## 4  97.0000      MINAGRI
## 5 107.8000      MINAGRI
## 6 125.5000      MINAGRI
```

```
# Show unique value in each column
summarise_all(ptt_price, n_distinct)
```

```
##   adm0_id adm0_name adm1_id adm1_name mkt_id mkt_name cm_id cm_name cur_id
## 1      1      1      1      5      5      63      63      1      1      1
##   cur_name pt_id pt_name um_id um_name mp_month mp_year mp_price
## 1      1      1      1      1      1      12      8      1452
##   mp_commoditysource
## 1      1
```

```
# Select column with more than one unique value
ptt_price_selected <- ptt_price %>%
  select_if(summarise_all(., n_distinct) > 1) %>%
  rename(region_id = adm1_id,
         region = adm1_name,
         market_id = mkt_id,
         market = mkt_name,
         month = mp_month,
         year = mp_year,
         price = mp_price) %>%
  mutate(date = ymd(paste(year, month, "01"))) %>%
  select(region_id, market, date, price)

describe(ptt_price_selected)
```

```
## ptt_price_selected
##
## 4 Variables      4320 Observations
## -----
## region_id
##      n missing distinct      Info      Mean      Gmd
##   4320      0      5    0.951    20262    3120
##
## lowest :  2587 21969 21971 21972 21973, highest:  2587 21969 21971 21972 21973
##
## Value      2587 21969 21971 21972 21973
## Frequency   381   883  1030   919  1107
## Proportion 0.088 0.204 0.238 0.213 0.256
## -----
## market
##      n missing distinct
##   4320      0      63
##
## lowest : Base      Birambo  Bugarama  Buhanda  Bumazi
## highest: Rukomo    Rushashi  Rusine   Rwagitima Rwamagana
## -----
```

```
## date
##          n      missing  distinct      Info      Mean      Gmd      .05
##        4320          0         96         1 2012-07-01      933.9 2008-07-01
##          .10         .25         .50         .75         .90         .95
## 2009-02-01 2010-11-01 2012-11-01 2014-05-08 2015-05-01 2015-09-01
##
## lowest : 2008-01-01 2008-02-01 2008-03-01 2008-04-01 2008-05-01
## highest: 2015-08-01 2015-09-01 2015-10-01 2015-11-01 2015-12-01
## -----
## price
##          n      missing  distinct      Info      Mean      Gmd      .05      .10
##        4320          0        1452         1      162.6      47.77      96.25     108.33
##          .25         .50         .75         .90         .95
##      132.50     160.00     192.64     219.14     235.00
##
## lowest : 62.5000 64.5000 66.0000 70.0000 70.6667
## highest: 301.7500 301.8571 390.0000 415.7500 430.0000
## -----
```

```
location <- list.dirs(choose.dir())
allfile <- list.files(path = location, pattern = c(".csv"))

datalist <- list()
for (x in 1:length(allfile)) {
  # Read dataset
  datalist[[x]] <- read_csv(paste(location, allfile[x], sep = "\\"))
}
```

```
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
```

```

## adm1_name = col_character(),
## mkt_id = col_double(),
## mkt_name = col_character(),
## cm_id = col_double(),
## cm_name = col_character(),
## cur_id = col_double(),
## cur_name = col_character(),
## pt_id = col_double(),
## pt_name = col_character(),
## um_id = col_double(),
## um_name = col_character(),
## mp_month = col_double(),
## mp_year = col_double(),
## mp_price = col_double(),
## mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),

```

```

## mp_year = col_double(),
## mp_price = col_double(),
## mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),

```

```

##  cm_id = col_double(),
##  cm_name = col_character(),
##  cur_id = col_double(),
##  cur_name = col_character(),
##  pt_id = col_double(),
##  pt_name = col_character(),
##  um_id = col_double(),
##  um_name = col_character(),
##  mp_month = col_double(),
##  mp_year = col_double(),
##  mp_price = col_double(),
##  mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()
## )
## Parsed with column specification:
## cols(
##   adm0_id = col_double(),
##   adm0_name = col_character(),
##   adm1_id = col_double(),
##   adm1_name = col_character(),
##   mkt_id = col_double(),
##   mkt_name = col_character(),
##   cm_id = col_double(),
##   cm_name = col_character(),
##   cur_id = col_double(),
##   cur_name = col_character(),
##   pt_id = col_double(),
##   pt_name = col_character(),
##   um_id = col_double(),
##   um_name = col_character(),
##   mp_month = col_double(),
##   mp_year = col_double(),
##   mp_price = col_double(),
##   mp_commoditysource = col_character()

```

```
## )
```

```
allfood <- reduce(datalist, bind_rows) %>%
  select_if(summarise_all(., n_distinct) > 1) %>%
  rename(region_id = adm1_id,
         region = adm1_name,
         market_id = mkt_id,
         market = mkt_name,
         food = cm_name,
         month = mp_month,
         year = mp_year,
         price = mp_price) %>%
  mutate(date = ymd(paste(year, month, "01"))) %>%
  select(region_id, market, date, food, price)

describe(allfood)
```

```
## allfood
##
## 5 Variables      27387 Observations
## -----
## region_id
##      n missing distinct      Info      Mean      Gmd
## 27387      0          5      0.951    20332    3003
##
## lowest : 2587 21969 21971 21972 21973, highest: 2587 21969 21971 21972 21973
##
## Value      2587 21969 21971 21972 21973
## Frequency  2316 5843 6631 5804 6793
## Proportion 0.085 0.213 0.242 0.212 0.248
## -----
## market
##      n missing distinct
## 27387      0          63
##
## lowest : Base      Birambo  Bugarama  Buhanda  Bumazi
## highest: Rukomo    Rushashi  Rusine   Rwagitima  Rwamagana
## -----
## date
##      n missing distinct      Info      Mean      Gmd      .05
## 27387      0          96      1 2012-12-03    881.2 2008-09-01
##      .10      .25      .50      .75      .90      .95
## 2009-07-01 2011-07-01 2013-06-01 2014-09-01 2015-06-01 2015-09-01
##
## lowest : 2008-01-01 2008-02-01 2008-03-01 2008-04-01 2008-05-01
## highest: 2015-08-01 2015-09-01 2015-10-01 2015-11-01 2015-12-01
## -----
## food
##      n missing distinct
## 27387      0          9
##
## lowest : Beans (dry)      Cassava      Chili (red)      Maize      Oranges (big si
## highest: Oranges (big size) Peas (fresh)      Potatoes (Irish)  Sorghum      Tomatoes
##
```



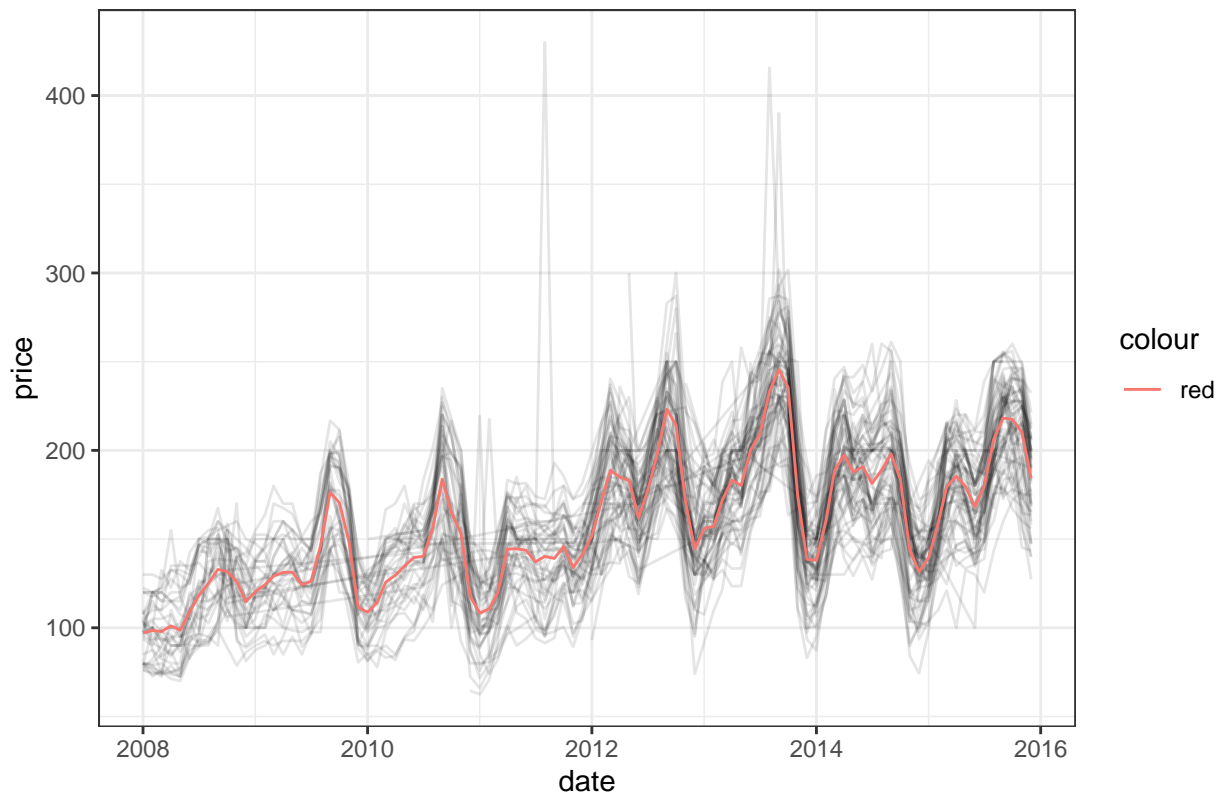
```
## Beans (dry) (4357, 0.159), Cassava (2766, 0.101), Chili (red) (1938, 0.071),
## Maize (4024, 0.147), Oranges (big size) (1843, 0.067), Peas (fresh) (1893,
## 0.069), Potatoes (Irish) (4320, 0.158), Sorghum (4099, 0.150), Tomatoes (2147,
## 0.078)
## -----
## price
##      n missing distinct      Info      Mean      Gmd      .05      .10
## 27387         0      6208         1    385.8    290.3    129.8    150.0
##   .25     .50     .75     .90     .95
## 200.0    280.0    416.0    800.0   1055.6
##
## lowest :   50.0000   62.5000   64.5000   66.0000   67.0833
## highest: 2900.0000 3000.0000 3283.3333 3500.0000 6500.0000
## -----
```

## Exploration

This

```
allfood %>%
  filter(food == "Potatoes (Irish)") %>%
  group_by(year(date), month(date)) %>%
  mutate(average_price = mean(price)) %>%
  ggplot() +
    geom_line(aes(x = date,
                  y = price,
                  group = market),
              alpha = 0.1) +
    geom_line(aes(x = date,
                  y = average_price,
                  color = "red")) +
  theme_bw() +
  labs(title = "Potato Price Trend")
```

## Potato Price Trend



## Forecasting

Forecasting preparation. As mentioned in the source, best way to conduct forecasting is by using a `ts` object. Further research should be done to ease this process.

```
# Make list
food_price_list <- allfood %>%
  group_by(food, date) %>%
  summarise(price = mean(price)) %>%
  group_split(food, .keep = F) %>%
  setNames(unique(allfood$food))

## 'summarise()' regrouping output by 'food' (override with '.groups' argument)

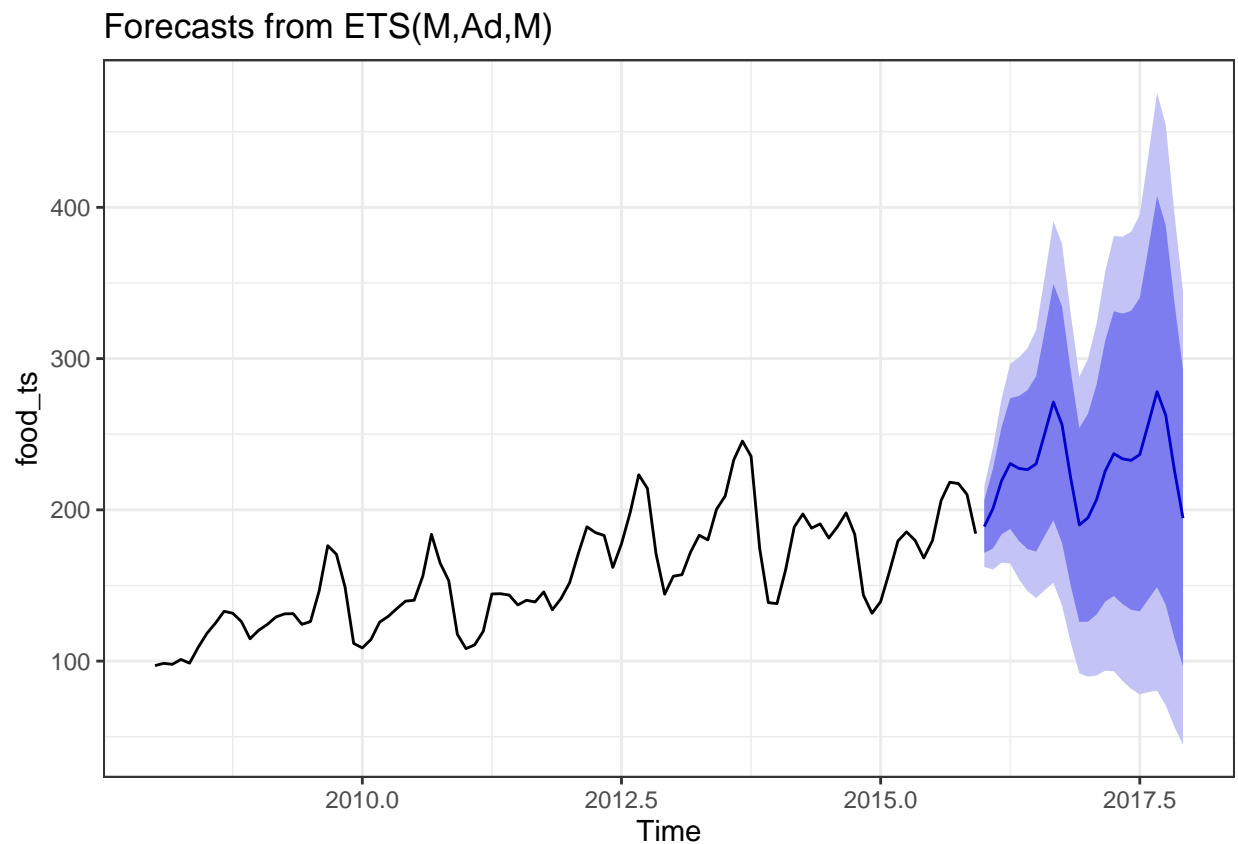
## Warning: ... is ignored in group_split(<grouped_df>), please use
## group_by(..., .add = TRUE) %>% group_split()

# Create ts object for forecasting
food_ts <- food_price_list$`Potatoes (Irish)` %>%
  ts(price,
     start = c(year(min(date)), month(min(date))),
     end = c(year(max(date)), month(max(date))),
     frequency = 12)

food_ts
```

##		Jan	Feb	Mar	Apr	May	Jun	Jul
## 2008		97.00000	98.52273	97.84091	101.09848	98.61364	109.09091	118.24243
## 2009		120.37500	124.25000	129.16667	131.21944	131.35484	124.28472	126.13548
## 2010		108.72656	114.24059	125.73334	129.57393	134.74327	139.62608	140.22014
## 2011		108.22805	110.70811	119.71144	144.38943	144.52988	143.65701	137.11667
## 2012		151.84952	171.07966	188.77012	184.90020	183.04400	161.90685	177.52511
## 2013		156.10126	157.13208	172.10822	183.19044	180.13387	200.35022	209.11998
## 2014		137.98415	160.31585	188.66735	197.28938	187.88636	190.71303	181.32434
## 2015		139.30230	158.76986	179.43795	185.43599	179.71824	168.17693	179.65607
##		Aug	Sep	Oct	Nov	Dec		
## 2008		125.11364	132.91667	131.64815	126.10417	114.74444		
## 2009		146.32930	176.28522	170.61989	148.90484	111.67473		
## 2010		156.15862	183.81879	164.90942	153.19605	117.75421		
## 2011		140.21399	139.08074	145.76252	133.94809	141.34497		
## 2012		198.02347	223.17500	214.16667	170.93782	144.20070		
## 2013		233.07375	245.46842	235.40024	174.39085	138.67057		
## 2014		188.84630	197.97793	183.84359	143.67184	131.59355		
## 2015		206.16530	218.25415	217.41502	210.15317	184.24260		

```
food_forecast <- forecast(object = food_ts)
autoplot(food_forecast) +
  theme_bw()
```



*Simplify by making a function* Using the food\_price\_list

```

plot_predict <- function(dataset, foodLabel = "Food") {
  dataset %>%
    ts(price,
      start = c(year(min(date)), month(min(date))),
      end = c(year(max(date)), month(max(date))),
      frequency = 12) %>%
    forecast() %>%
    autoplot() +
      theme_bw() +
      labs(y = foodLabel)
}

plot_predict(dataset = food_price_list$Cassava, foodLabel = "Cassava")

```

