# Using tidyverse for trade analysis

School of Economics, Quaid-i-Azam University, Islamabad

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#### Trade Data Analysis

This material is mainly based on UN ESCAPE online training on R for trade analysis. In this post, tidyverse is used instead of base-R (used in the original material) due to convenience and reading codes like english language. Moreover,ggplot2 provides awesome viusalization tools.

#### Loading data

Data are provided on <a href="https://r.tiid.org/">https://r.tiid.org/</a> to be downloaded in csv file and one can read this data directly in R. I have save this data in R as RData and have loaded as follows. For more details on how to read various data formats in R, one can browse on how to import data. An easy way is File->Import Data Set -> then choose relevant file. In case of error, upload relevant packages.

```
library(tidyverse)
load("trade_data.RData")
```

Warning: NAs introduced by coercion

To get idea about the nature of this data, one may get glimpse ,names or usehead etc.

```
# drop missing values
  trade_data <- trade_data[!is.na(trade_data$value),]</pre>
  trade_data
# A tibble: 691,923 x 5
   reporter flow partner
                           year
                                        value
      <dbl> <chr>
                     <dbl> <dbl>
                                        <dbl>
 1
        213 I
                       186
                            2008
                                    156130202
2
        213 E
                       174 2008
                                    117661208
3
        213 E
                       172
                            2008
                                     31986252
 4
                       134 2008
                                   1507966544
        213 E
5
        213 I
                       181
                            2008
                                      2407260
6
        213 I
                       182
                            2008
                                     80414681
7
                       676 2008
        213 I
                                       991884
8
        213 E
                       258
                            2008
                                    107246580
```

# ... with 691,913 more rows

614 I

213 E

9

10

# i Use `print(n = ...)` to see more rows

We notice that there are **5 main variables**, namely: **reporter** indicating destination country, **flow** consists of import and export , **year** for which trade data are included in this file and **value** is USD.

2660760

110 2008 15302709034

648 2008

We can now check the **value** in billion by dividing the previous function by a billion. Notice after the comma we specify the value column. Other than very large exporting/importing countries, one may convert **value** in millions of USD instead of taking it in billions of USD if required.

trade\_data %>% distinct(year) # distinct number of years for which this trade data are av

```
# A tibble: 10 x 1
year
<dbl>
1 2008
2 2009
3 2010
4 2011
5 2012
6 2013
7 2014
```

```
8 2015
9 2016
10 2017
```

```
trade_data<-trade_data %>% mutate(value=value/1000000000)

options(scipen = 999) ## To avoid scientific numbers (like appearing in exponents)
```

#### **Pakistan**

Now we analyze data for Pakistan for year 2016 and learn R for data exploration in this post.

```
trade_data %>% filter(flow=="E"&reporter==564&year==2016&partner==1) %>% select(value)

# A tibble: 1 x 1
  value
  <dbl>
1 20.5
```

So total export by Pakistan to the world (world coder in partner category is 1) is USD 20.8 billion. Now if we calculate Pakistan's export with all countries, it is 86.356242592 USD Billion. This is greater than we calculated earlier with world. This is because there are many categories in partner category which include countries as well as different regions. So need some data wrangling before one can match both the figures. One has to exclude all those partner codes which are not equal to 1 and isCountry==1.

#### Export plot for all the years

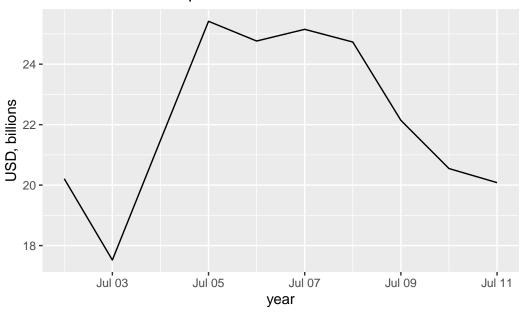
```
library(lubridate)

Attaching package: 'lubridate'

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

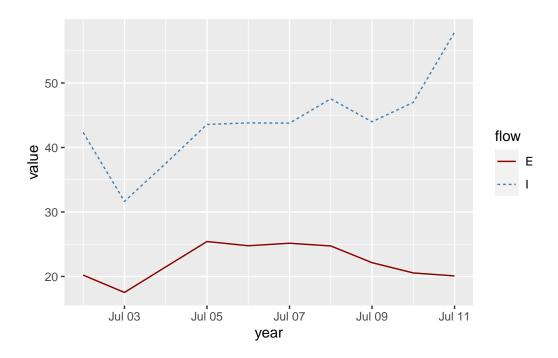
```
PK_X<-trade_data %>% filter(flow=="E"&reporter==564&partner==1)
PK_X$year<-as_date(PK_X$year)
ggplot(PK_X)+aes(x=year,y=value)+geom_line()+labs(y="USD, billions", title = "Pakistan's was a second to be a second
```

# Pakistan's world export



### **Import and Export Together**

```
PK_XM<-trade_data %>% filter(reporter==564&partner==1)
PK_XM$year<-as_date(PK_XM$year)
ggplot(PK_XM, aes(x = year, y = value)) +
    geom_line(aes(color = flow, linetype=flow)) +
    scale_color_manual(values = c("darkred", "steelblue"))</pre>
```



#### Top 10 exporting destination for China

Let's now show in a plot the top 10 countries Pakistan exported to in 2016.

We start by creating a table with Pakistanis exports in 2016 to individual countries:

```
PK_X_countries <- trade_data %>% filter(flow=="E"&reporter==564&year==2016&partner==1)
```

But to get top trade partners we have to extract names of the countries for which following file is downloaded from the IMF. **Partner** codes are available but country names are not given which one has to download from the IMF website.

#### Country code

To select countries, 'isCountry" or code has to be set 1 so other categories like advanced countries and countries not specified are excluded

```
imf_codes <- read_csv("imf_codes.csv")</pre>
```

Rows: 245 Columns: 10
-- Column specification ------

```
Delimiter: ","
chr (5): iso2, country, region, income_level, AP_region
dbl (5): code, isCountry, LLDC, LDC, SIDS
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
  codes<-imf_codes %>% select(code,country,isCountry, region)
  cntry_w <- codes %>% filter(isCountry==1|code==1) %>% select(code)
  cntry_w
# A tibble: 216 x 1
    code
   <dbl>
     512
 1
 2
   799
 3
   914
 4
     612
 5
     1
 6
    859
 7
    614
 8
     312
 9
     311
10
    213
# ... with 206 more rows
# i Use `print(n = ...)` to see more rows
  PK_X_2016<-trade_data %>% filter(flow=="E"&reporter==564&year==2016&partner!=1&partner %in
  PK_X_2016
  PK_X_2016<-trade_data[trade_data$flow=="E"& #select only exports
  trade_data$reporter==564& #from china
  trade_data$year==2016& #in 2016
  trade_data$partner!=1& #where country is not world (code for world is 1)
  trade_data$partner %in% cntry_w, #which is in the list of codes that we filtered earlier
  "value" #variable we're interested in
  sum(PK_X_2016)/1000000000
```

#### Merge country codes and trade\_data

Lets do the same exercise for Pakistan before proceeding further.

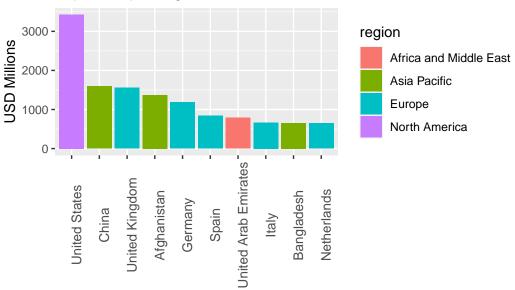
```
PK_X_countries <-trade_data %>% filter(flow=="E"&reporter==564&year==2016&partner!=1)
  PKX_merged<-left_join(PK_X_countries,codes,by=c("partner"="code"))</pre>
  PKX_merged <- PKX_merged %>% select(country, value, isCountry, region)
  PKX_merged
# A tibble: 207 x 4
  country
                          value isCountry region
  <chr>
                           <dbl>
                                     <dbl> <chr>
                      0.0329
1 Somalia
                                         1 Africa and Middle East
2 Albania
                       0.00430
                                         1 Europe
3 Syrian Arab Republic 0.0158
                                        1 Africa and Middle East
4 Morocco
                                       1 Africa and Middle East
                      0.0175
5 Angola
                     0.0171
                                         1 Africa and Middle East
                    0.651
6 Netherlands
                                         1 Europe
                                         1 Africa and Middle East
7 Niger
                      0.0000769
8 Paraguay
                      0.00672
                                        1 Latin America and Carribean
                                       1 Africa and Middle East
9 Gabon
                       0.00155
10 Peru
                       0.0261
                                         1 Latin America and Carribean
# ... with 197 more rows
# i Use `print(n = ...)` to see more rows
  # Only keep countries
  PKX_merged<-PKX_merged %>% filter(isCountry==1)
```

#### Top 10 exporting destinations in case of Pakistan in 2016

Sort export value by descending order using arrange() and then select top 10 exporting destinations.

```
PKX_merged_top10<-PKX_merged %>% arrange(desc(value))
PKX_merged_top10<-head(PKX_merged_top10,10)
PAK<-ggplot(PKX_merged_top10, aes(x = reorder(country,-value), y = value*1000, fill=region
PAK</pre>
```

Top 10 exporting destinations of Pakistan in 2016



#### India

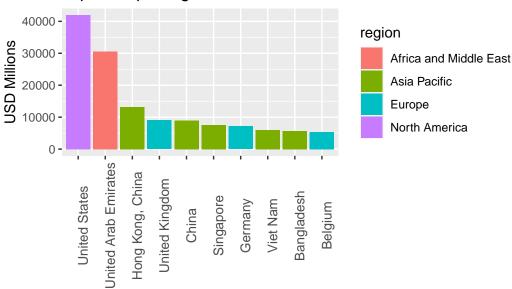
```
PK_X_2016<-trade_data %>% filter(flow=="E"&reporter==564&year==2016&partner!=1&partner %in PK_X_2016

PK_X_2016<-trade_data[trade_data$flow=="E"& #select only exports trade_data$reporter==564& #from china trade_data$year==2016& #in 2016 trade_data$partner!=1& #where country is not world (code for world is 1) trade_data$partner %in% cntry_w, #which is in the list of codes that we filtered earlier "value" #variable we're interested in ]

sum(PK_X_2016)/1000000000
```

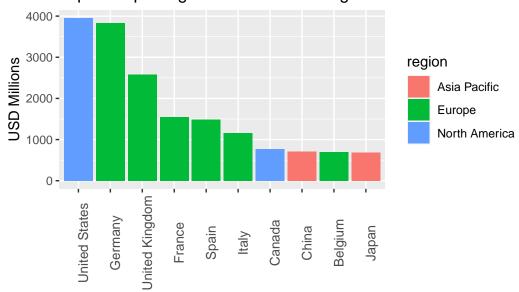
# India

Top 10 exporting destinations of India in 2016



## Bangladesh

Top 10 exporting destinations of Bangladesh in 2016



## Sri Lank

region **USD Millions** 2000 -Africa and Middle East Asia Pacific 1000 -Europe North America 0 United Arab Emirates -United Kingdom United States Netherlands Germany Belgium Japan India Italy

Top 10 exporting destinations of Sri Lank in 2016

#### South Asia

library(gridExtra)

Total export of Bangladesh, India, Pakistan and Sril Lanka in billion of USD is 30.131039072, 261.862, 20.547844736 and 10.045959206 respectively.

```
Attaching package: 'gridExtra'

The following object is masked from 'package:dplyr':

combine

grid.arrange(BD,IND,PAK,SRL, ncol=2)
```

