## Sok-1005 assignment 4

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rm(list=ls())

```
library(tidyverse)
Warning: package 'tidyverse' was built under R version 4.2.2
Warning: package 'ggplot2' was built under R version 4.2.2
Warning: package 'tidyr' was built under R version 4.2.2
Warning: package 'readr' was built under R version 4.2.2
Warning: package 'purrr' was built under R version 4.2.2
Warning: package 'dplyr' was built under R version 4.2.2
Warning: package 'stringr' was built under R version 4.2.2
Warning: package 'forcats' was built under R version 4.2.2
Warning: package 'lubridate' was built under R version 4.2.2
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr
        1.1.0
                   v readr
                                 2.1.4
v forcats 1.0.0
                    v stringr
                                 1.5.0
v ggplot2 3.4.1
                   v tibble
                                3.1.8
v lubridate 1.9.2
                     v tidyr
                                 1.3.0
v purrr
           1.0.1
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()
                masks stats::lag()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
```

# library(lubridate) library(quantmod)

```
Warning: package 'quantmod' was built under R version 4.2.2
Loading required package: xts
Warning: package 'xts' was built under R version 4.2.2
Loading required package: zoo
Warning: package 'zoo' was built under R version 4.2.2
Attaching package: 'zoo'
The following objects are masked from 'package:base':
   as.Date, as.Date.numeric
# We noticed you have dplyr installed. The dplyr lag() function breaks how
# base R's lag() function is supposed to work, which breaks lag(my_xts).
# Calls to lag(my_xts) that you enter or source() into this session won't
                                                                       #
# work correctly.
# All package code is unaffected because it is protected by the R namespace
                                                                       #
# mechanism.
# Set `options(xts.warn_dplyr_breaks_lag = FALSE)` to suppress this warning.
```

Attaching package: 'xts'

# dplyr from breaking base R's lag() function.

# You can use stats::lag() to make sure you're not using dplyr::lag(), or you #
# can add conflictRules('dplyr', exclude = 'lag') to your .Rprofile to stop #

```
The following objects are masked from 'package:dplyr':
    first, last
Loading required package: TTR
Warning: package 'TTR' was built under R version 4.2.2
Registered S3 method overwritten by 'quantmod':
  as.zoo.data.frame zoo
  library(janitor)
Attaching package: 'janitor'
The following objects are masked from 'package:stats':
    chisq.test, fisher.test
  library(plotly)
Warning: package 'plotly' was built under R version 4.2.3
Attaching package: 'plotly'
The following object is masked from 'package:ggplot2':
    last_plot
The following object is masked from 'package:stats':
    filter
The following object is masked from 'package:graphics':
    layout
```

```
library(knitr)
library(dplyr)
```

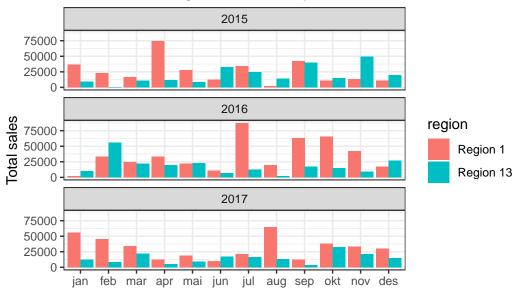
You can add options to executable code like this

```
df <- read.csv("https://raw.githubusercontent.com/uit-sok-1005-v23/uit-sok-1005-v23.github
    clean names()
  table1 <- df %>%
    mutate(order_date = as.Date(order_date)) %>%
    mutate(year = year(order_date),
           month = month(order_date),
           day = day(order_date)) %>%
    filter(year=="2017",
           month >= 10,
           customer_segment %in% c("Corporate", "Consumer"),
           region %in% c("Region 1", "Region 9")) %>%
    group_by(region, month, customer_segment) %>%
    summarize(sales = sum(sales)) %>%
    rename("Region" = "region", "Month" = "month", "Customer segment" = "customer_segment")
`summarise()` has grouped output by 'region', 'month'. You can override using
the `.groups` argument.
  figure1 <- df %>%
    mutate(order_date = as.Date(order_date)) %>%
    mutate(year = year(order_date), month = month(order_date, label = TRUE), day = day(order_date)
    filter(year %in% c("2015", "2016", "2017"), region %in% c("Region 1", "Region 13")) %>%
    group_by(region, month, year) %>%
    summarize(sales = sum(sales)) %>%
      mutate(date = make_date(year,month))
`summarise()` has grouped output by 'region', 'month'. You can override using
the `.groups` argument.
  fig1 <- figure1 %>%
    ggplot(aes(x=month, y=sales))+
    geom_col(aes(fill=region), position="dodge")+
    labs(x="", y="Total sales", title="Total sales in region 1 and 13 by month", color="Regi
```

```
facet_wrap(~year, nrow=3) +
  theme_bw()

fig1
```

#### Total sales in region 1 and 13 by month



```
table2 <- figure1[c(41,50,52,54,61,67,71), ]
table2
```

```
# A tibble: 7 x 5
# Groups:
            region, month [6]
            month year sales date
  region
  <chr>
            <ord> <dbl>
                         <dbl> <date>
                   2016 55632. 2016-02-01
1 Region 13 feb
2 Region 13 mai
                   2016 22822. 2016-05-01
3 Region 13 jun
                   2015 32307. 2015-06-01
4 Region 13 jun
                   2017 17430. 2017-06-01
                   2015 39825. 2015-09-01
5 Region 13 sep
6 Region 13 nov
                   2015 49686. 2015-11-01
                   2016 26890. 2016-12-01
7 Region 13 des
```

```
table3 <- df %>%
      mutate(year = year(order_date), month = month(order_date, label = TRUE), day = day(or
    filter(region %in% c("Region 1", "Region 2", "Region 4", "Region 6", "Region 7", "Region
  filter(year > 2016)
  table3 <- table3[c(5,7:10)]
  table3 <- table3 %>%
    group_by(customer_segment, product_category, region, year) %>%
      summarize(avg_profit = mean(profit))
`summarise()` has grouped output by 'customer_segment', 'product_category',
'region'. You can override using the `.groups` argument.
  #Small Business within Technology in Region 11
  #was the customer segment with the highest average
  #profit for 2017 with 3585.120000.
  xom <- data.frame(getSymbols("XOM", src = "yahoo", from = "2010-1-04", to = "2022-12-31",</pre>
  xom <- tibble::rownames_to_column(xom, var = "Date")</pre>
  crudeoil <- data.frame(getSymbols("DCOILBRENTEU", src = "FRED", from = "2010-1-04", to = "</pre>
  crudeoil <- tibble::rownames_to_column(crudeoil, var = "Date")</pre>
  xom_ <- xom %>%
    mutate(Date = as.Date(Date)) %>%
    mutate(year_month = format(Date, "%Y-%m")) %>%
    group_by(year_month) %>%
    mutate(exxon = weighted.mean(XOM.Adjusted))
  crudeoil_ <- crudeoil %>%
    mutate(Date = as.Date(Date)) %>%
    mutate(year_month = format(Date, "%Y-%m")) %>%
    group_by(year_month) %>%
    mutate(oil = mean(DCOILBRENTEU))
  crudeoil_ <- na.omit(crudeoil_)</pre>
```

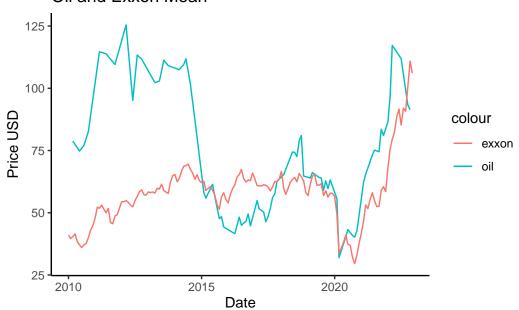
```
crudeoil_ <- crudeoil_ %>%
    filter(Date >= "2010-01-04" & Date < "2022-12-01")

xom_ <- xom_ %>%
    select(year_month,exxon) %>%
    mutate(year_month = ym(year_month)) %>%
    distinct()

crudeoil_ <- crudeoil_ %>%
    select(year_month,oil) %>%
    mutate(year_month = ym(year_month)) %>%
    mutate(year_month = ym(year_month)) %>%
    distinct()

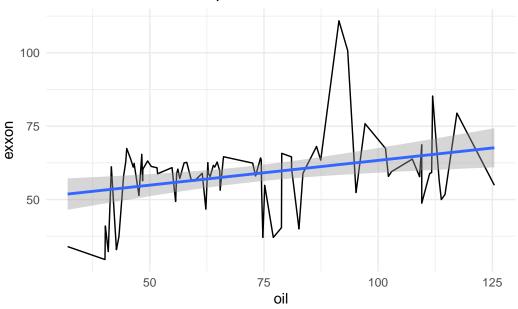
ggplot() +
    geom_line(data = crudeoil_ , aes(x=year_month,y=oil, col = "oil")) +
    geom_line(data = xom_,aes(x=year_month,y=exxon,col="exxon"))+ xlab("Date") + ylab("Pritheme_classic())
```

#### Oil and Exxon Mean



```
#the red line shows us the mean by weight of exxon stocks.
  #The blue line shows us the mean of oil prices.
  #We obsere a growth of exxon stock prices that seem
  #to follow the rising price of oil.
  oil_exxon <- merge(xom_,crudeoil_)</pre>
  lm(exxon ~ oil, data = oil_exxon)
Call:
lm(formula = exxon ~ oil, data = oil_exxon)
Coefficients:
(Intercept)
                     oil
   46.4749
                0.1688
  oil_exxon %>%
    ggplot(aes(x=oil,y=exxon)) +
    geom_line() +
    theme_minimal() +
    xlab("oil") +
    ylab("exxon") +
    ggtitle("Exxon Stocks and Oil price") +
    geom_smooth(method = lm)
`geom_smooth()` using formula = 'y ~ x'
```

### Exxon Stocks and Oil price



# (Intercept) oil # 46.4749 0.1688

# the intercept tells us the average price of the stocks, and the other coefficient tells
#us how much more value is added if the oil prices were to be raised by one.

#We can tell there is a clear corrolation between oil prices #and the prices of exxon stocks.