## FIN511 Week 1

First code chunk used to load the necessary libraries in R if you don't have these libraries, please uncomment the code and install them or install them in R-Studio.

# load libraries

```
# install readxl and tidyverse (or just ggplot2 and dyplr)
# install.packages("readxl", "qqplot2", "dplyr")
library(ggplot2)
library(readxl)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(reshape2)
The following code chunk shows a very "dumb" way to load excel spreadsheets in R directly, skipping lines if
necessary, like in this case and use the function "head" to see if the spreadsheet is loaded properly
# Load all the datasets and separate sheets, clean up the data and leave a single row for column names.
# You will still need the original excel sheet to read the instructions at the top or additional inform
# modify the files paths to match your needs
# I didn't set working directories or anything else to keep the customization to a minimum
df1_mom_returns <- read_xlsx("/Users/ataru074/Desktop/Education/MBA/03 2020 Fall/FIN511 Investments/213
df1_portfolio_strategies <- read_xlsx("/Users/ataru074/Desktop/Education/MBA/03 2020 Fall/FIN511 Invest:
## New names:
## * `` -> ...3
## * `` -> ...7
rt vol risky and rf <- read xlsx("/Users/ataru074/Desktop/Education/MBA/03 2020 Fall/FIN511 Investments
## New names:
## * `` -> ...2
## * `` -> ...9
## * `` -> ...12
rt_vol_risky_assets <- read_xlsx("/Users/ataru074/Desktop/Education/MBA/03 2020 Fall/FIN511 Investments
```

```
## * `` -> ...10
## * `` -> ...13
## * `` -> ...15
head(df1_mom_returns)
## # A tibble: 6 x 12
                          `3`
                                        `5`
                    `2`
                                 `4`
                                               `6`
                                                      `7`
                                                              .8.
    Year
            Low
                                                                          High
##
     <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <
                                             <dbl> <dbl> <dbl> <dbl> <dbl> <
                  9.07 17.2 26.3
## 1 1927
            18.5
                                     32.7
                                             25.1
                                                    35.4
                                                            27.2
                                                                   39.4
                                                                          67.4
## 2 1928
           14.7 23.2
                        21.4 26.1
                                      25.5
                                             34.3
                                                    41.9
                                                           42.8
                                                                   48.7
## 3 1929 -57.6 -47.8 -34.2 -21.3
                                     -2.32 -13.3
                                                     5.95
                                                           1.32
                                                                    0.25 - 29.9
## 4 1930 -54.6 -48.4 -50.4 -49.6 -31.7 -31.7 -32.4 -19.7 -22.9 -20.9
## 5 1931 -64.5 -58.3 -60.6 -52.9 -53.9 -49.7 -49.4 -42.5 -34.8 -24.2
          25.7 -2.46 -12.2 -7.47 9.31 -1.33 -17.1 -14.4 -12.4 -25.8
## 6 1932
## # ... with 1 more variable: `High-Low` <dbl>
Lets replicate the "two risky assets" spreadsheet
Step 1, let's get all the information together and organize it in a dataframe (a structure similar to a
spreadsheet)
# Risk Free rate
RF <- 3
# The large Weight is a sequence that goes from -1 to 2 in 0.01 steps
Large_Weight \leftarrow seq(-1,2,0.01)
# The Small Weight is simply 1 - Large Weight
Small_Weight <- 1 - Large_Weight</pre>
# Large return
LR <- 8
# Small return
SR. <- 15
# Correlation (using CR because cor is an R funcion therefore reserved word)
CR < -0.4
# Large Standard Deviation
LSD <- 25
# Small Standard Deviation
SSD <- 50
# lets create the dataframe. Note, R automatically recycle variables to fit the length of the dataframe
# therefore is important that the first variable is the correct lenght.
df <- data.frame("RF" = RF ,"Large_Weight" = Large_Weight, "Small_Weight" = Small_Weight, "LR" = LR, "S
                 "CR" = CR, "LSD" = LSD, "SSD" = SSD)
# We could have also avoided all the variable declarations and just do:
\# df \leftarrow data.frame("Large_Weight" = seq(-1,2,0.01), "Small_Weight" = 1 - seq(-1,2,0.01), "LR" = 8)
# Now let's add the calculated columns (note that column can be added just using df "$" name_of_the_col
# and we use the dollar sign to select the column by name as well
```

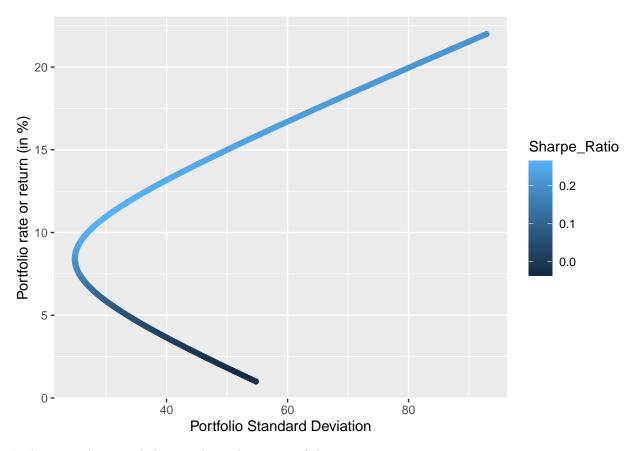
## New names: ## \* ` -> ...2

# Important Don't use spaces in the name of columns or variables or anything in general, it's just bad

```
# practice at best and in most cases the software will give you an error
df$Portfolio_SD <- sqrt(df$Large_Weight^2 * df$LSD^2 + df$Small_Weight^2 * df$SSD^2 +
                          2*df$CR*df$Large Weight*df$LSD*df$Small Weight*df$SSD)
df$Portfolio_Return <- df$Large_Weight*df$LR + df$Small_Weight*df$SR
df$Sharpe_Ratio <- (df$Portfolio_Return - df$RF)/df$Portfolio_SD
# let's verify that our dataframe is created correctly
head(df)
##
     RF Large_Weight Small_Weight LR SR CR LSD SSD Portfolio_SD Portfolio_Return
## 1 3
                             2.00 8 15 0.4 25
              -1.00
                                                        92.87088
                                                 50
                                                                            22.00
## 2 3
               -0.99
                             1.99 8 15 0.4 25
                                                 50
                                                        92.42680
                                                                            21.93
## 3 3
               -0.98
                             1.98 8 15 0.4 25
                                                 50
                                                        91.98288
                                                                            21.86
## 4 3
               -0.97
                             1.97 8 15 0.4 25
                                                 50
                                                        91.53913
                                                                            21.79
## 5 3
              -0.96
                             1.96 8 15 0.4 25
                                                 50
                                                        91.09555
                                                                            21.72
## 6 3
               -0.95
                             1.95 8 15 0.4 25 50
                                                        90.65215
                                                                            21.65
##
    Sharpe_Ratio
## 1
       0.2045851
## 2
       0.2048107
## 3
       0.2050382
## 4
       0.2052674
## 5
       0.2054985
## 6
       0.2057315
tail(df)
       RF Large_Weight Small_Weight LR SR CR LSD SSD Portfolio_SD
## 296 3
                                                          52.72867
                  1.95
                              -0.95 8 15 0.4 25 50
## 297
       3
                  1.96
                              -0.96 8 15 0.4
                                               25 50
                                                          53.13568
## 298 3
                  1.97
                              -0.97 8 15 0.4 25 50
                                                          53.54356
## 299 3
                  1.98
                              -0.98 8 15 0.4 25 50
                                                          53.95229
## 300 3
                  1.99
                              -0.99 8 15 0.4 25 50
                                                          54.36187
## 301 3
                  2.00
                              -1.00 8 15 0.4 25 50
                                                          54.77226
##
      Portfolio_Return Sharpe_Ratio
## 296
                  1.35 -0.03129227
## 297
                   1.28 -0.03236997
## 298
                   1.21 -0.03343073
## 299
                   1.14 -0.03447490
## 300
                   1.07 -0.03550283
## 301
                   1.00 -0.03651484
Now that the data has been created, let's create the charts in ggplot2 protip: download the ggplot2 cheat
sheet https://rstudio.com/wp-content/uploads/2015/03/ggplot2-cheatsheet.pdf
# Let's use GGPLOT to create our charts, it gives us more customization abilities
# for example we can encode the sharpe ratio as color along the curve
```

ggplot(df, aes(x=Portfolio\_SD, y=Portfolio\_Return)) +

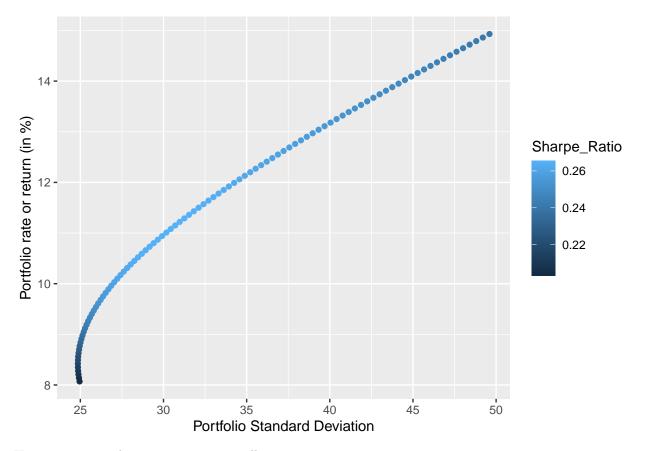
geom\_point(aes(color=Sharpe\_Ratio)) +
xlab("Portfolio Standard Deviation") +
ylab("Portfolio rate or return (in %)")



Let's create the second chart without the option of shorting

```
# Careful, I'm using the filter function from the package dplyr, which is different from the standard
# one, therefore, load dyplr as library.

ggplot(filter(df, Large_Weight < 1, Small_Weight < 1), aes(x=Portfolio_SD, y=Portfolio_Return)) +
    geom_point(aes(color=Sharpe_Ratio)) +
    xlab("Portfolio Standard Deviation") +
    ylab("Portfolio rate or return (in %)")</pre>
```



How to generate the answers automatically

First let's create a second dataframe with the modified parameters and call it df2

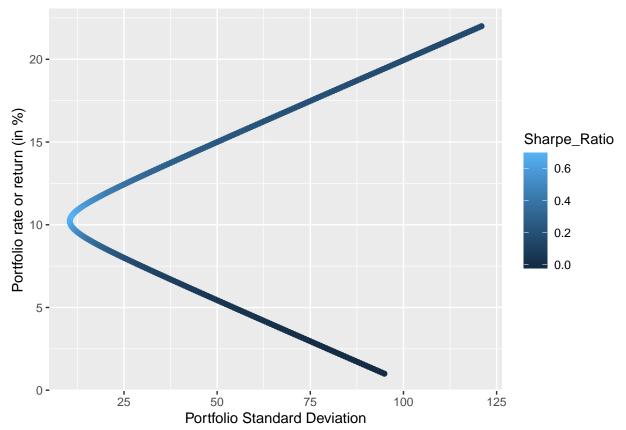
```
# Risk Free rate
RF <- 3
# The large Weight is a sequence that goes from -1 to 2 in 0.01 steps
Large_Weight \leftarrow seq(-1,2,0.01)
# The Small Weight is simply 1 - Large Weight
Small_Weight <- 1 - Large_Weight</pre>
# Large return
LR <- 8
# Small return
SR <- 15
# Correlation (using CR because cor is an R funcion therefore reserved word)
CR < -0.8
# Large Standard Deviation
LSD <- 25
# Small Standard Deviation
SSD <- 50
# lets create the dataframe. Note, R automatically recycle variables to fit the lenght of the dataframe
# therefore is important that the first variable is the correct lenght.
df2 <- data.frame("RF" = RF ,"Large_Weight" = Large_Weight, "Small_Weight" = Small_Weight, "LR" = LR, "
```

```
"CR" = CR, "LSD" = LSD, "SSD" = SSD)
# We could have also avoided all the variable declarations and just do:
# df \leftarrow data.frame("Large_Weight" = seq(-1,2,0.01), "Small_Weight" = 1 - seq(-1,2,0.01), "LR" = 8)
# Now let's add the calculated columns (note that column can be added just using df "$" name_of_the_col
# and we use the dollar sign to select the column by name as well
# Important Don't use spaces in the name of columns or variables or anything in general, it's just bad
# practice at best and in most cases the software will give you an error
df2$Portfolio_SD <- sqrt(df2$Large_Weight^2 * df2$LSD^2 + df2$Small_Weight^2 * df2$SSD^2 +
                         2*df2$CR*df2$Large_Weight*df2$LSD*df2$Small_Weight*df2$SSD)
df2$Portfolio_Return <- df2$Large_Weight*df2$LR + df2$Small_Weight*df2$SR
df2$Sharpe_Ratio <- (df2$Portfolio_Return - df2$RF)/df2$Portfolio_SD
# let's verify that our dataframe is created correctly
head(df2)
    ## 1 3
              -1.00
                           2.00 8 15 -0.8 25 50
                                                      120.9339
                                                                         22.00
## 2 3
                           1.99 8 15 -0.8 25 50
                                                                         21.93
              -0.99
                                                      120.2207
                           1.98 8 15 -0.8 25 50
## 3 3
              -0.98
                                                      119.5075
                                                                         21.86
## 4 3
              -0.97
                           1.97 8 15 -0.8 25 50
                                                      118.7944
                                                                         21.79
## 5 3
              -0.96
                           1.96 8 15 -0.8 25 50
                                                      118.0813
                                                                         21.72
                           1.95 8 15 -0.8 25 50
## 6 3
              -0.95
                                                      117.3683
                                                                         21.65
##
   Sharpe_Ratio
## 1
       0.1571107
## 2
       0.1574604
## 3
       0.1578143
## 4
       0.1581724
## 5
       0.1585348
## 6
       0.1589015
tail(df2)
      RF Large_Weight Small_Weight LR SR
##
                                        CR LSD SSD Portfolio_SD
## 296 3
                 1.95
                            -0.95 8 15 -0.8 25 50
                                                        91.31162
## 297 3
                 1.96
                            -0.96 8 15 -0.8 25 50
                                                        92.02282
## 298 3
                 1.97
                            -0.97 8 15 -0.8 25 50
                                                        92.73410
## 299 3
                 1.98
                            -0.98 8 15 -0.8 25
                                                 50
                                                        93.44544
## 300 3
                 1.99
                            -0.99 8 15 -0.8 25
                                                 50
                                                        94.15685
                            -1.00 8 15 -0.8 25 50
                                                       94.86833
## 301 3
                 2.00
##
      Portfolio_Return Sharpe_Ratio
## 296
                 1.35 -0.01806999
                 1.28 -0.01869102
## 297
## 298
                  1.21 -0.01930250
## 299
                  1.14 -0.01990466
## 300
                  1.07 -0.02049771
## 301
                  1.00 -0.02108185
Let see the modified chart
```

# Let's use GGPLOT to create our charts, it gives us more customization abilities

# for example we can encode the sharpe ratio as color along the curve

```
ggplot(df2, aes(x=Portfolio_SD, y=Portfolio_Return)) +
  geom_point(aes(color=Sharpe_Ratio)) +
  xlab("Portfolio Standard Deviation") +
  ylab("Portfolio rate or return (in %)")
```



And now overlay them. In R this is not very automated, please follow the code This example is simplified by the fact that the portfolio rate of return does not depend on the correlation coefficient but only by the weights, so we can use it "as is" for our Y axis

```
# first we need to create a dataframe with the information we need.
# the portfolio rate of return is our Y and cor_04 and cor_min08 are the standard deviations
# create a simple dataframe with only the variables of interest (this step is not necessary but it is e
overlay_df <- data.frame(Y=df$Portfolio_Return, cor_04=df$Portfolio_SD, cor_min08=df2$Portfolio_SD)
# melt the dataframe to stack the variables in a single column
overlay_df.m <- melt(overlay_df, id.vars = "Y", measure.vars = c("cor_04", "cor_min08"))
# double check the stacking is correct
head(overlay_df.m)
##
         Y variable
                       value
## 1 22.00
            cor 04 92.87088
## 2 21.93
            cor 04 92.42680
## 3 21.86
            cor_04 91.98288
            cor_04 91.53913
## 4 21.79
```

## 5 21.72

## 6 21.65

cor\_04 91.09555

cor\_04 90.65215

## tail(overlay\_df.m) Y variable ## value ## 597 1.35 cor\_min08 91.31162 ## 598 1.28 cor\_min08 92.02282 ## 599 1.21 cor\_min08 92.73410 ## 600 1.14 cor\_min08 93.44544 ## 601 1.07 cor\_min08 94.15685 ## 602 1.00 cor\_min08 94.86833 # plot reversing the axis (we put Y where X is expected and so on) ggplot(overlay\_df.m, aes(value, Y, colour = variable)) + geom\_point() + xlab("Portfolio Standard Deviation") + ylab("Portfolio rate or return (in %)") 20 -Portfolio rate or return (in %) variable cor\_04 cor\_min08 5 -0 -25 50 75 100 125

Question 1: Find the composition of the portfolio at minimum variance

```
# the function filter from dyplr is very useful in this case.
filter(df2, Portfolio_SD == min(Portfolio_SD))
```

```
## RF Large_Weight Small_Weight LR SR CR LSD SSD Portfolio_SD Portfolio_Return
## 1 3 0.68 0.32 8 15 -0.8 25 50 10.47855 10.24
## Sharpe_Ratio
## 1 0.6909353
```

Portfolio Standard Deviation

Question 2:

Question 3 shows something that is very common... what if there isn't a portfolio return exactly equal to 25.

Note: -check the numbers, I'm giving the template, not the solution can be approached in 2 ways. First, a brute force approach using the filter function from dyplr Second, we can calculate exactly the

## Question 3:

```
# Test if there is a return exactly equal to 25.
filter(df2, Portfolio_Return == 25)
##
   [1] RF
                       Large_Weight
                                       Small_Weight
                                                       LR
##
   [5] SR
                                       LSD
                                                       SSD
                       CR
                       Portfolio_Return Sharpe_Ratio
  [9] Portfolio_SD
## <0 rows> (or 0-length row.names)
# then we just filter the portfolio around the value and we select the correct row
filter(df2, Portfolio_Return > 24.5 & Portfolio_Return < 25.5)</pre>
##
   [1] RF
                       Large_Weight
                                                       LR
                                       Small_Weight
   [5] SR
##
                                       LSD
                                                       SSD
## [9] Portfolio_SD
                       Portfolio_Return Sharpe_Ratio
## <0 rows> (or 0-length row.names)
# or we just use a neat function "which min" which returns the row where the absolute value
# of the difference between the Portfolio_Return and 20 is at its minimum
row.number <- which.min(abs(df2$Portfolio_Return - 25))</pre>
# and we verify that is correct.
df2[row.number,]
    ## 1 3
                             2 8 15 -0.8 25 50
                                                     120.9339
##
    Sharpe_Ratio
## 1
       0.1571107
```