# Project 1

2025-03-09

### Project 1 - Team 12

## Needed packages

## Exercise 4: Market Portfolio Returns

Download monthly Fama-French data

## 5 1927-05-01 0.0544 0.003

```
start_date <- "1927-01-01"
end_date <- "2024-12-31"
#monthly
factors_ff_monthly_raw <- download_french_data("Fama/French 3 Factors")</pre>
## New names:
## New names:
## * `` -> `...1`
# first data set is the monthly data
mkt_rf_ff_monthly <- factors_ff_monthly_raw$subsets$data[[1]] %>%
  mutate(
    # transform the data to be first day of the month and of the form "yyyy-mm-dd"
   date = floor_date(ymd(paste0(date, "01")), "month"),
    # data in percentage format, therefore convert them to decimal form
   across(c(RF, `Mkt-RF`, SMB, HML), ~as.numeric(.) / 100),
    # drop all other columns
    .keep = "none"
  # rename all columns for better readability
 rename_with(str_to_lower) %>%
  rename(mkt_excess = `mkt-rf`) %>%
  # filter for the given time frame
  filter(date >= start_date & date <= end_date) %>%
  # Rearranging columns
  select(date, mkt_excess, rf)
head(mkt_rf_ff_monthly)
## # A tibble: 6 x 3
    date
          mkt_excess
##
     <date> <dbl> <dbl>
## 1 1927-01-01 -0.0006 0.0025
## 2 1927-02-01 0.0418 0.0026
## 3 1927-03-01 0.0013 0.003
## 4 1927-04-01 0.0046 0.0025
```

#### Calculate Market Portfolio Returns

The mkt-excess data set in the Fama-French data is the market excess return. Therefore, we need to add the riskfree rate "rf" to receive the market portfolio returns.

```
mkt_ret_monthly <- mkt_rf_ff_monthly %>%
  mutate(
    mkt_ret = mkt_excess + rf
    ) %>%
  select(-mkt_excess,-rf)
# Function to calculate rolling cumulative returns
# width = the number of data we want to cumulate over --> roll monthly
# Fun = calculates the cumulative return
# align = "right" --> assure we take last month of the window
rolling_cum_return <- function(return_series, window) {</pre>
  rollapply(return_series, width = window, FUN = function(x) prod(1 + x) - 1, align = "right", fill = N
# Compute rolling returns
mkt_ret_rolling <- mkt_ret_monthly %>%
  arrange(date) %>%
  mutate(
   ret_1yr = rolling_cum_return(mkt_ret, 12), # 1 year
   ret_10yr = rolling_cum_return(mkt_ret, 120), # 10 years
    ret_20yr = rolling_cum_return(mkt_ret, 240) # 20 years
  ) %>%
  select(-mkt ret)
# 1. Average if rolling every month
# Compute average return for each investment duration
avg_ret_1yr_m <- round(mean(mkt_ret_rolling$ret_1yr, na.rm = TRUE),4) *100
avg_ret_10yr_m <- round(mean(mkt_ret_rolling$ret_10yr, na.rm = TRUE),4) *100
avg_ret_20yr_m <- round(mean(mkt_ret_rolling$ret_20yr, na.rm = TRUE),4) *100
# 2. Average if rolling every year
mkt_ret_rolling_y <- mkt_ret_rolling %>%
 filter(month(date) == 1)
avg_ret_1yr_y <- round(mean(mkt_ret_rolling_y$ret_1yr, na.rm = TRUE),4) *100
avg_ret_10yr_y <- round(mean(mkt_ret_rolling_y$ret_10yr, na.rm = TRUE),4) *100
avg_ret_20yr_y <- round(mean(mkt_ret_rolling_y$ret_20yr, na.rm = TRUE),4) *100
results <- data.frame(
  Rolling_Type = c("Monthly", "Yearly"),
  Average_1_year_Return = c(avg_ret_1yr_m, avg_ret_1yr_y),
  Average_10_year_Return = c(avg_ret_10yr_m, avg_ret_10yr_y),
  Average_20_year_Return = c(avg_ret_20yr_m, avg_ret_20yr_y)
)
```

## Average Returns of the Market Portfolio [1927-2024] (in %)

Rolling Type	Avg. 1 Year Ret.	Avg. 10 Year Ret.	Avg. 20 Year Ret.
Monthly	12.14	198.14	797.02
Yearly	11.97	197.51	793.07

### Annualize Market Portfolio Returns

```
#function for annualizing returns with given input of the months = T
annualize_returns <- function(return,T){</pre>
  (1+return)^(12/T) - 1
# Annualize returns in the df mkt_ret_rolling using the yearly rolling timeframe
mkt_ret_rolling_y_ann <- mkt_ret_rolling_y %>%
  mutate(
    ret_10yr_ann = annualize_returns(ret_10yr,120),
    ret_20yr_ann = annualize_returns(ret_10yr,240)
  ) %>%
  select(-ret_10yr,-ret_20yr)
avg_ret_10yr_y_ann <- round(mean(mkt_ret_rolling_y_ann$ret_10yr_ann, na.rm = TRUE),4) *100
avg_ret_20yr_y_ann <- round(mean(mkt_ret_rolling_y_ann$ret_20yr_ann, na.rm = TRUE),4) *100
results <- data.frame(</pre>
  Rolling_Type = c("Yearly"),
 Average_1_year_Return = c(avg_ret_1yr_y),
 Average_10_year_Return = c(avg_ret_10yr_y_ann),
  Average_20_year_Return = c(avg_ret_20yr_y_ann)
colnames(results) <- c("Rolling Type", "Avg. 1 Year Ret.", "Avg. 10 Year Ret.", "Avg. 20 Year Ret.")</pre>
```

Rolling Type	Avg. 1 Year Ret.	Avg. 10 Year Ret.	Avg. 20 Year Ret.
Yearly	11.97	10.5	5.09

### Resample from the actual data and calculate the bootstrap standard error

```
# We want to sample from the actual return data series
# Repeat the sample 10,000 times and calculate the mean for every sample
# standard error is the sd of the sample means
# We use the annualized return series
bootstrap_se <- function(returns, num_samples = 10000) {</pre>
  boot_means <- replicate(num_samples, mean(sample(returns, size = length(returns), replace = TRUE)))
  return(sd(boot_means))
}
se_1yr <- bootstrap_se(na.omit(mkt_ret_rolling_y_ann$ret_1yr))</pre>
se_10yr <- bootstrap_se(na.omit(mkt_ret_rolling_y_ann$ret_10yr_ann))</pre>
se_20yr <- bootstrap_se(na.omit(mkt_ret_rolling_y_ann$ret_20yr_ann))</pre>
result_se <- data.frame(</pre>
 Metric = c("Bootstrap SE"),
  Average_1_year_Return = round(se_1yr,4) *100 ,
  Average_10_year_Return = round(se_10yr,4)*100,
  Average_20_year_Return = round(se_20yr,4)*100
colnames(result_se) <- c("Rolling Type", "Avg. 1 Year Ret.", "Avg. 10 Year Ret.", "Avg. 20 Year Ret.")</pre>
result_se %>%
  gt() %>%
  tab_style(
    style = cell_borders(sides = "right",
                          color = "grey80",
                          weight = px(2)),
```

## Bootstrap Standard Error of Average Annualized Returns (Ret. in %)

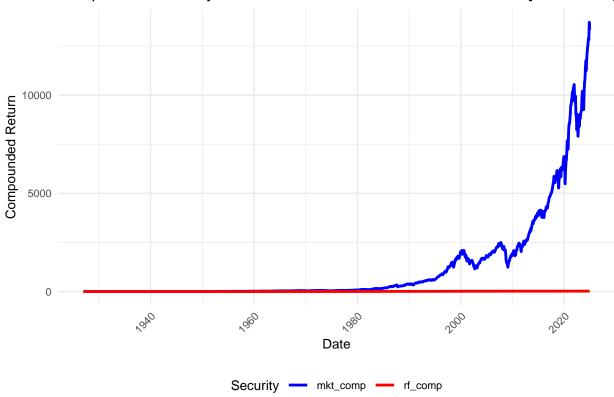
Rolling Type	Avg. 1 Year Ret.	Avg. 10 Year Ret.	Avg. 20 Year Ret.
Bootstrap SE	2.01	0.56	0.27

### Comparison of the returns of investment of the MP with riskfree rate

```
mkt_ret_monthly <- mkt_rf_ff_monthly %>%
 mutate(
   mkt_ret = mkt_excess + rf
   ) %>%
  select(-mkt_excess)
mkt_ret_monthly <- mkt_ret_monthly %>%
  mutate(
   mkt comp = cumprod(1 + mkt ret) - 1,
   rf_{comp} = cumprod(1 + rf) - 1
# Reshape data for plotting
mkt_ret_monthly_long <- na.omit(mkt_ret_monthly) %>%
  select(date, mkt_comp , rf_comp) %>%
  pivot_longer(cols = c(mkt_comp,rf_comp),
              names_to = "Security",
              values_to = "Compound_Return")
# Plot the compounded returns for Market Portfolio and riskfree rate
ggplot(mkt_ret_monthly_long, aes(x = date, y = Compound_Return, color = Security)) +
  geom_line(linewidth = 1) + # Line plot for compounded returns
   title = "Compounded Monthly Returns Market Portfolio and Riskfree Rate [1927-2024]",
   x = "Date",
   y = "Compounded Return",
   color = "Security"
 ) +
  theme_minimal() + # Clean theme
  theme(
   legend.position = "bottom",
   text = element_text(size = 10),
   axis.text.x = element_text(angle = 45, hjust = 1)
```



## Compounded Monthly Returns Market Portfolio and Riskfree Rate [1927-2024]



### Housekeeping for Git

```
save.image(file = "Pj1_Task4.RData")
```

### # Exercise 5: Find a proxy for the Market Portfolio

As Alnylam Pharmaceuticals, Inc. is a publicly traded large cap corporation in the US and it is part of the Russel 1000 and Russel 3000, we think that it is suitable to use the wider Russel 3000 index as the market portfolio.

### First download the data from Excel

The price data for the Russel 3000 and for Alnylam Pharmaceuticals is in an Excel file, which we downloaded from Bloomberg.

```
ALNY_PX <- read.xlsx("Daily_Data_Case_1.xlsx", sheet = 2)
ALNY_PX$Date <- as.Date(ALNY_PX$Date,origin = "1899-12-30")

RUS_3_PX <- read.xlsx("Daily_Data_Case_1.xlsx", sheet = 1)
RUS_3_PX$Date <- as.Date(RUS_3_PX$Date,origin = "1899-12-30")

# Observe NA values

ALNY_NA <- sum(is.na(ALNY_PX$Last.Price))
```

```
RUS_3_NA <- sum(is.na(RUS_3_PX$Last.Price))
# One NA in ALNY_PX --> Lin. Interpolate
na <- which(is.na(ALNY_PX$Last.Price))
ALNY_PX$Last.Price[na] <- (ALNY_PX$Last.Price[na-1] + ALNY_PX$Last.Price[na+1]) * 0.5</pre>
```

#### Calculate Arithmic Average Annual Return and Compound Annual Return

```
ALNY_PX_xts <- xts(ALNY_PX$Last.Price, order.by = as.Date(ALNY_PX$Date))
RUS_3_PX_xts <- xts(RUS_3_PX$Last.Price, order.by = as.Date(RUS_3_PX$Date))
PX_data <- list(ALNY_PX_xts,RUS_3_PX_xts)</pre>
time <- c("daily", "weekly", "monthly")</pre>
time_2 \leftarrow c(252, 52, 12)
type <- c("Alnylam", "Russel 3000 Index")</pre>
ret <- list() # List to store returns</pre>
avg_an_ret <- vector("numeric", 3)</pre>
comp_an_ret <- vector("numeric", 3)</pre>
summary_ret <- list()</pre>
for (j in 1:2){
  for (i in 1:3) {
    # Calculate returns
    ret[[i]] <- coredata(periodReturn(PX_data[[j]]["/2019-01-01"], period = time[i], type = "arithmetic</pre>
    # Average Arithmetic Annual Return
    avg_an_ret[i] <- (1 + mean(ret[[i]])) ^ time_2[i] - 1</pre>
    # Compound Annual Return
    comp_returns <- cumprod(1 + ret[[i]])</pre>
    # Annualize the return by adjusting for the number of periods
    comp_an_ret[i] <- (comp_returns[length(comp_returns)]) ^ (time_2[i] / length(comp_returns)) - 1</pre>
summary_ret[[j]] <- data.frame(</pre>
  "Period" = time,
  "Art_Avg_An_Ret (in %)" = round(avg_an_ret,4)*100,
  "CAGR (in %)" = round(comp_an_ret,4)*100
colnames(summary_ret[[j]]) <- c("Period", "Arithmetic Average Annual Return (in %)", "Compound Annual R
}
summary_ret[[1]] %>%
    gt() %>%
    tab_style(
      style = cell_borders(sides = "right",
                             color = "grey80",
                             weight = px(2)),
```

## Arithmetic Average Annualized Returns vs Compound Annual Returns of Alnylam

Period	Arithmetic Average Annual Return (in $\%$ )	Compound Annual Return (in %)
daily	33.00	11.25
weekly	32.66	11.16
monthly	30.51	11.23

Arithmetic Average Annualized Returns vs Compound Annual Returns of Russel 3000 Index

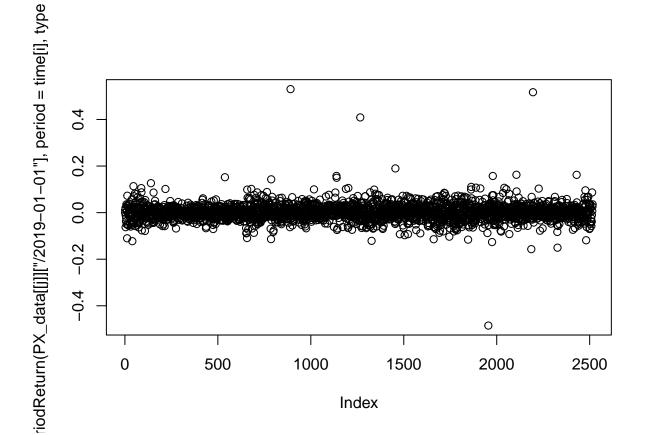
Period	Arithmetic Average Annual Return (in $\%)$	Compound Annual Return (in $\%$ )
daily	12.26	10.63
weekly	12.00	10.56
monthly	11.74	10.63

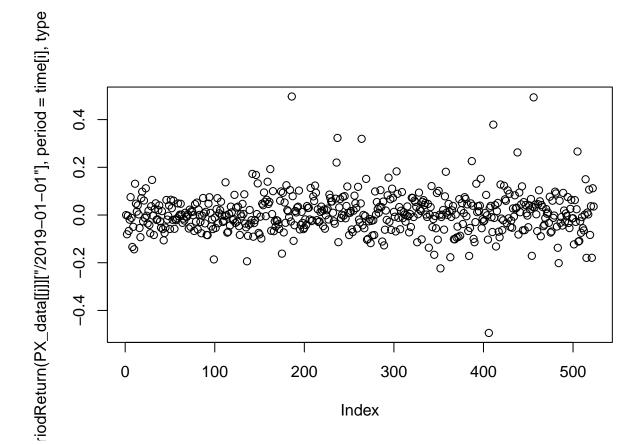
```
locations = cells_body(columns = everything())
    ) %>%
   tab_style(
     style = cell_borders(sides = "right",
                           color = "grey80",
                           weight= px(2)),
     locations = cells_column_labels(columns = everything())
   tab_header(title = paste0("Arithmetic Average Annualized Returns vs Compound Annual Returns of ", t
summary_ret[[2]] %>%
   gt() %>%
   tab style(
      style = cell_borders(sides = "right",
                           color = "grey80",
                           weight = px(2)),
     locations = cells_body(columns = everything())
   ) %>%
   tab_style(
     style = cell_borders(sides = "right",
                           color = "grey80",
                           weight= px(2)),
     locations = cells_column_labels(columns = everything())
   tab_header(title = paste0("Arithmetic Average Annualized Returns vs Compound Annual Returns of ", t
```

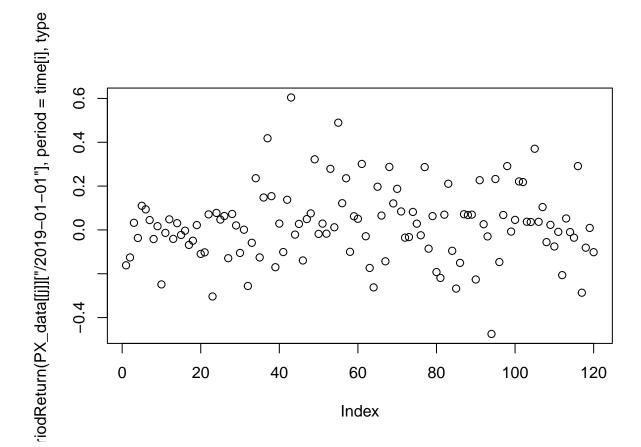
### **Estimate Return Volatility**

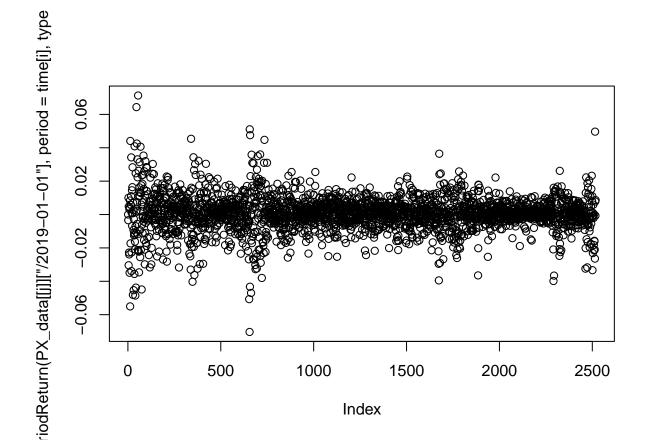
```
st_dev <- vector("numeric",3)
summary_dev <- data.frame(
    "Period" = time
)
for(j in 1:2){</pre>
```

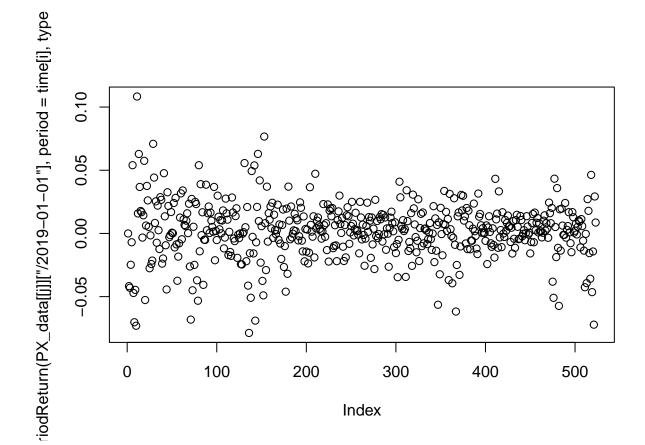
```
for (i in 1:3){
    st_dev[i] <- round(sd(coredata(periodReturn(PX_data[[j]]["/2019-01-01"], period = time[i], type = ".
    print(plot(coredata(periodReturn(PX_data[[j]]["/2019-01-01"], period = time[i], type = "arithmetic"
}
summary_dev <- cbind(summary_dev, st_dev)
}</pre>
```

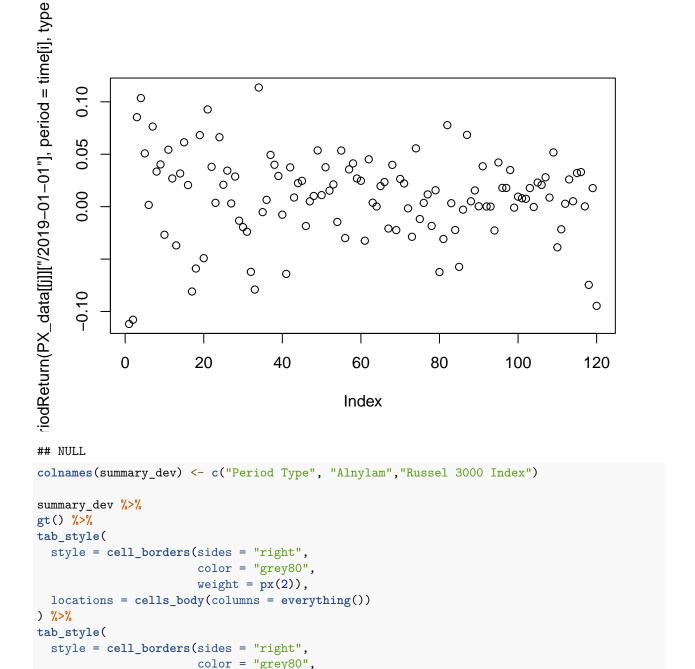












### Plot the rolling 2 years (104 weeks) Return Volatility for your firm and your

Market Portfolio with weekly returns from 1/1/2009 to 12/31/2024

weight= px(2)),

```
returns_weekly_ALNY <- periodReturn(PX_data[[1]], period = "weekly", type = "arithmetic")
returns_weekly_RUS <- periodReturn(PX_data[[2]], period = "weekly", type = "arithmetic")</pre>
```

locations = cells\_column\_labels(columns = everything()) # Apply to column headers

tab\_header(title = "Estimated Return Volatility for Alnylam and Russel 3000 Index")

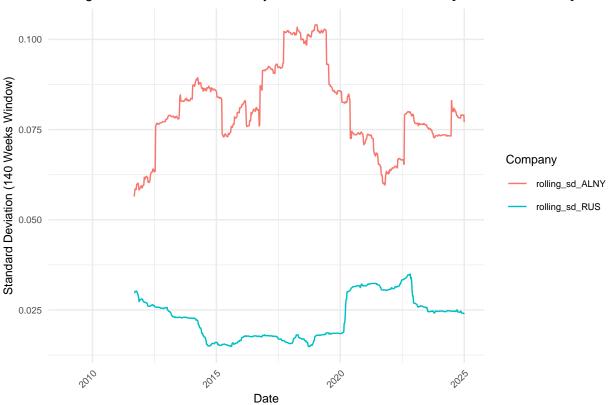
### Estimated Return Volatility for Alnylam and Russel 3000 Index

Period Type	Alnylam	Russel 3000 Index
daily	0.0380	0.0108
weekly	0.0832	0.0223
monthly	0.1667	0.0410

```
returns_weekly_ALNY_df <- data.frame(Date = index(returns_weekly_ALNY)), coredata(returns_weekly_ALNY))
returns weekly RUS df <- data.frame(Date = index(returns weekly RUS), coredata(returns weekly RUS))
returns_weekly <- inner_join(returns_weekly_ALNY_df, returns_weekly_RUS_df, by = "Date")
colnames(returns_weekly) <- c("Date", "ALNY", "Russel3000")</pre>
returns_weekly <- returns_weekly[-1,]</pre>
returns_weekly$rolling_sd_ALNY <- rollapply(returns_weekly$ALNY, width = 140, FUN = sd, align = "right"
returns_weekly$rolling_sd_RUS <- rollapply(returns_weekly$Russel3000, width = 140, FUN = sd, align = "r
returns_weekly$rolling_cor <- runCor(</pre>
  x = returns_weekly$ALNY,
 y = returns_weekly$Russel3000,
  n = 140
# For Plotting, put into long format
sd_long <- returns_weekly %>%
  pivot_longer(cols = c(rolling_sd_ALNY, rolling_sd_RUS),
               names_to = "Company",
               values_to = "Rolling_SD")
ggplot(sd_long, aes(x = Date, y = Rolling_SD, color = Company)) +
  geom line() +
  labs(title = "Rolling Standard Deviation of Alnylam and Russel 3000 Index [09/2011-21/2024]",
       x = "Date",
       y = "Standard Deviation (140 Weeks Window)") +
  theme_minimal() +
  theme(
    axis.text.x = element_text(angle = 45, hjust = 1),
        text = element_text(size = 9))
```

## Warning: Removed 278 rows containing missing values or values outside the scale range
## (`geom\_line()`).

## Rolling Standard Deviation of Alnylam and Russel 3000 Index [09/2011–21/2024]



## Warning: Removed 139 rows containing missing values or values outside the scale range
## (`geom\_line()`).

