

# Financial Econometrics - Group Assignment

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## EDA

```
data <- read.csv("FIN 36182-Industry_Portfolios.CSV")
head(data)
sum(is.na(data))
anyDuplicated(data)

data[, 2:6] <- data[, 2:6] / 100 ## Convert returns to decimal form as asked
head(data)
```

## Task 1

1. Report the arithmetic mean of the returns for each of the five industries over the entire sample.

```
monthly_means <- colMeans(data[, c("Cnsmr", "Manuf", "HiTec", "Hlth", "Other")])
annualized_means <- monthly_means * 12
annualized_means <- round(annualized_means, 4)
annualized_means
```

2. Report the standard deviation of the returns for each of the five industries over the entire sample.

```
monthly_std <- apply(data[, c("Cnsmr", "Manuf", "HiTec", "Hlth", "Other")], 2, sd)
annualized_std <- monthly_std * sqrt(12)
annualized_std <- round(annualized_std, 4)
annualized_std
```

3. Report the Sharpe ratio of each industry.

```
sharpe_ratios <- annualized_means / annualized_std
sharpe_ratios <- round(sharpe_ratios, 4)
sharpe_ratios
```

```
results_table1 <- data.frame(
  Mean>Returns = annualized_means,
  Std>Returns = annualized_std,
  Sharpe_Ratio = sharpe_ratios
```

```
)
results_table1
```

5. Provide a table (5×5) with the sample correlation between the returns of the five industries.

```
correlation_matrix <- cor(data[, c("Cnsmr", "Manuf", "HiTec", "Hlth", "Other")])
correlation_matrix <- round(correlation_matrix, 4)
correlation_df <- as.data.frame(correlation_matrix)
correlation_df
```

```
library(corrplot)
corrplot(correlation_matrix, method = "color", type = "upper",
         tl.col = "black", tl.srt = 45,
         addCoef.col = "black",
         number.cex = 0.8,
         number.digits = 2)
```

6. Construct a time series of the simple, non-cumulative returns of a portfolio where capital is allocated equally across the first four industries (excluding Other). Report the arithmetic mean, standard deviation and Sharpe.

```
portfolio_returns <- rowMeans(data[, c("Cnsmr", "Manuf", "HiTec", "Hlth")])
data$Portfolio_Returns <- portfolio_returns
mean_portfolio <- mean(portfolio_returns)
sd_portfolio <- sd(portfolio_returns)

annualized_mean_portfolio <- mean_portfolio * 12
annualized_sd_portfolio <- sd_portfolio * sqrt(12)
sharpe_ratio_portfolio <- annualized_mean_portfolio / annualized_sd_portfolio

portfolio_results <- data.frame(
  Metric = c("Annualized Mean Return", "Annualized Standard Deviation", "Sharpe Ratio"),
  Value = round(c(annualized_mean_portfolio, annualized_sd_portfolio, sharpe_ratio_portfolio), 4)
)
print(portfolio_results)
```

```
library(ggplot2)
data$Date <- as.character(data$Date)
data$Date <- paste0(data$Date, "01")
data$Date <- as.Date(data$Date, format = "%Y%m%d")

plot_data <- data.frame(Date = data$Date, Portfolio_Returns = data$Portfolio_Returns)
ggplot(plot_data, aes(x = Date, y = Portfolio_Returns)) +
  geom_line(color = "blue") +
  labs(title = "Portfolio Returns Over Time",
       x = "Date",
       y = "Portfolio Returns") +
  theme_minimal()
```

## Task 2

In this task you will treat the portfolio you computed in Task 1, point (6), as the market portfolio, denote its returns as  $R_m$ , and will estimate and interpret beta and alpha coefficients in the context of the CAPM.

### 1. Compute the kurtosis and skeweness of Rm

```
library(moments)

skewness_value <- skewness(data$Portfolio_Returns, na.rm = TRUE)
kurtosis_value <- kurtosis(data$Portfolio_Returns, na.rm = TRUE)

kurtosis_value <- round(kurtosis_value, 4)
skewness_value <- round(skewness_value, 4)

portfolio_results2 <- data.frame(
  Metric = c("Kurtosis", "Skewness"),
  Value = c(kurtosis_value, skewness_value)
)
print(portfolio_results2)

library(ggplot2)
ggplot(data = data.frame(returns = data$Portfolio_Returns), aes(x = returns)) +
  geom_histogram(aes(y = ..density..), bins = 30, fill = "blue", color = "black", alpha = 0.7) +
  geom_density(color = "red", size = 1) +
  stat_function(fun = dnorm,
    args = list(mean = mean(portfolio_returns, na.rm = TRUE),
      sd = sd(portfolio_returns, na.rm = TRUE)),
    color = "green", linetype = "dashed", size = 1) +
  labs(title = "Distribution of Portfolio Returns with Normal Distribution Overlay",
    x = "Portfolio Returns",
    y = "Density") +
  theme_minimal()
```

### 3. Repeat point (1), but eliminating the first 70 years of data (i.e. from 199706).

```
data_filtered <- data[data$Date >= as.Date("1997-06-01"), ]
portfolio_returns_filtered <- data_filtered$Portfolio_Returns

skewness_filtered <- skewness(portfolio_returns_filtered, na.rm = TRUE)
kurtosis_filtered <- kurtosis(portfolio_returns_filtered, na.rm = TRUE)

skewness_filtered <- round(skewness_filtered, 4)
kurtosis_filtered <- round(kurtosis_filtered, 4)

portfolio_results3 <- data.frame(
  Metric = c("Kurtosis", "Skewness"),
  Value = c(kurtosis_filtered, skewness_filtered)
)
print(portfolio_results3)

ggplot(data = data.frame(returns = portfolio_returns_filtered), aes(x = returns)) +
  geom_histogram(aes(y = after_stat(density)), bins = 30, fill = "blue", color = "black", alpha = 0.7) +
  geom_density(color = "red", linewidth = 1) +
  stat_function(fun = dnorm,
    args = list(mean = mean(portfolio_returns_filtered, na.rm = TRUE),
      sd = sd(portfolio_returns_filtered, na.rm = TRUE)),
    color = "green", linetype = "dashed", linewidth = 1) +
  labs(title = "Distribution of Filtered Portfolio Returns with Normal Distribution Overlay",
```

```

x = "Portfolio Returns",
y = "Density") +
theme_minimal()

```

#### 4. Compute and report the covariance of the first four industries with Rm

```

cov_Cnsmr_Rm <- cov(data$Cnsmr, data$Portfolio_Returns, use = "complete.obs")
cov_Manuf_Rm <- cov(data$Manuf, data$Portfolio_Returns, use = "complete.obs")
cov_HiTec_Rm <- cov(data$HiTec, data$Portfolio_Returns, use = "complete.obs")
cov_Hlth_Rm <- cov(data$Hlth, data$Portfolio_Returns, use = "complete.obs")

cov_Cnsmr_Rm <- round(cov_Cnsmr_Rm, 4)
cov_Manuf_Rm <- round(cov_Manuf_Rm, 4)
cov_HiTec_Rm <- round(cov_HiTec_Rm, 4)
cov_Hlth_Rm <- round(cov_Hlth_Rm, 4)

covariances <- data.frame(
  Industry = c("Cnsmr", "Manuf", "HiTec", "Hlth"),
  Covariance_with_Rm = c(cov_Cnsmr_Rm, cov_Manuf_Rm, cov_HiTec_Rm, cov_Hlth_Rm)
)

print(covariances)

```

#### 5. Use the results obtained so far to compute the beta values for the first four industries for the full sample

```

# calculate the variance of the portfolio returns (Rm)
var_Rm <- var(data$Portfolio_Returns, use = "complete.obs")
var_Rm <- round(var_Rm, 4)

# calculate the beta values for each industry
beta_Cnsmr <- cov_Cnsmr_Rm / var_Rm
beta_Manuf <- cov_Manuf_Rm / var_Rm
beta_HiTec <- cov_HiTec_Rm / var_Rm
beta_Hlth <- cov_Hlth_Rm / var_Rm

beta_Cnsmr <- round(beta_Cnsmr, 4)
beta_Manuf <- round(beta_Manuf, 4)
beta_HiTec <- round(beta_HiTec, 4)
beta_Hlth <- round(beta_Hlth, 4)

beta_values <- data.frame(
  Industry = c("Cnsmr", "Manuf", "HiTec", "Hlth"),
  Beta = c(beta_Cnsmr, beta_Manuf, beta_HiTec, beta_Hlth)
)

print(beta_values)

```

#### 6. Compute the beta values for the first four industries for the sample starting from 199706. Briefly comment on how results compare with those in point (5)

```

var_Rm_filtered <- var(portfolio_returns_filtered, use = "complete.obs")

```

```

beta_Cnsmr_filtered <- cov(data_filtered$Cnsmr, data_filtered$Portfolio_Returns, use = "complete.obs")
beta_Manuf_filtered <- cov(data_filtered$Manuf, data_filtered$Portfolio_Returns, use = "complete.obs")
beta_HiTec_filtered <- cov(data_filtered$HiTec, data_filtered$Portfolio_Returns, use = "complete.obs")
beta_Hlth_filtered <- cov(data_filtered$Hlth, data_filtered$Portfolio_Returns, use = "complete.obs")

beta_Cnsmr_filtered <- round(beta_Cnsmr_filtered, 4)
beta_Manuf_filtered <- round(beta_Manuf_filtered, 4)
beta_HiTec_filtered <- round(beta_HiTec_filtered, 4)
beta_Hlth_filtered <- round(beta_Hlth_filtered, 4)

beta_values_filtered <- data.frame(
  Industry = c("Cnsmr", "Manuf", "HiTec", "Hlth"),
  Beta_Filtered = c(beta_Cnsmr_filtered, beta_Manuf_filtered, beta_HiTec_filtered, beta_Hlth_filtered)
)
print(beta_values_filtered)

comparison <- data.frame(
  Industry = c("Cnsmr", "Manuf", "HiTec", "Hlth"),
  Beta_Full_Sample = c(beta_Cnsmr, beta_Manuf, beta_HiTec, beta_Hlth), # point 5
  Beta_Filtered = c(beta_Cnsmr_filtered, beta_Manuf_filtered, beta_HiTec_filtered, beta_Hlth_filtered)
)
print(comparison)

```

7. Assuming a risk-free rate of 5%, compute Jensen's alpha for each of the first four industries (on the full sample). Report the alpha in percentage terms.

```

annualized_returns_industries <- annualized_means
risk_free_rate <- 0.05

alpha_Cnsmr <- (annualized_returns_industries["Cnsmr"] - (risk_free_rate + beta_Cnsmr * (annualized_mean_portfolio_returns - risk_free_rate)))
alpha_Manuf <- (annualized_returns_industries["Manuf"] - (risk_free_rate + beta_Manuf * (annualized_mean_portfolio_returns - risk_free_rate)))
alpha_HiTec <- (annualized_returns_industries["HiTec"] - (risk_free_rate + beta_HiTec * (annualized_mean_portfolio_returns - risk_free_rate)))
alpha_Hlth <- (annualized_returns_industries["Hlth"] - (risk_free_rate + beta_Hlth * (annualized_mean_portfolio_returns - risk_free_rate)))

jensens_alpha <- data.frame(
  Industry = c("Cnsmr", "Manuf", "HiTec", "Hlth"),
  Alpha = round(c(alpha_Cnsmr, alpha_Manuf, alpha_HiTec, alpha_Hlth), 4)
)
print(jensens_alpha)

```

### Task 3

Use the `lm()` function to run a few regressions:

1. Regress  $R_m(t)$  on an intercept and on  $R_m(t-1)$ . Report estimates and t-statistics. Briefly interpret the results

```

library(dplyr)
data <- data %>%
  mutate(Rm_lag = lag(Portfolio_Returns, 1))
head(data)

```

```
model <- lm(Portfolio_Returns ~ Rm_lag, data = data)
summary(model)
```

```
library(ggplot2)
ggplot(data, aes(x = Rm_lag, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) + # Scatter plot of returns
  geom_smooth(method = "lm", color = "red", se = FALSE) + # Regression line without confidence intervals
  labs(title = "Regression of Portfolio Returns on Lagged Returns",
       x = "Lagged Portfolio Returns (R_m(t-1))",
       y = "Portfolio Returns (R_m(t))") +
  theme_minimal()
```

2. Regress  $R_m(t)$  on an intercept and on  $(P_m(t-1)/P_m(t-13)-1)$ . Report estimates and t-statistics. Briefly interpret the results

```
data$Pm <- NA # initializing the price series
data$Pm[1] <- 1 # arbitrary starting price

# calculate Pm(t) iteratively based on Rm(t) = (Pm(t)/Pm(t-1)) - 1
for (i in 2:nrow(data)) {
  data$Pm[i] <- data$Pm[i - 1] * (1 + data$Portfolio_Returns[i])
}
```

```
data$Pm_lag1 <- dplyr::lag(data$Pm, 1)
data$Pm_lag13 <- dplyr::lag(data$Pm, 13)
data$Pm_ratio <- (data$Pm_lag1 / data$Pm_lag13) - 1
head(data, 20)
```

```
model2 <- lm(Portfolio_Returns ~ Pm_ratio, data = data)
summary(model2)
```

```
ggplot(data, aes(x = Pm_ratio, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) + # Scatter plot of Portfolio Returns vs. Pm_ratio
  geom_smooth(method = "lm", color = "red", se = FALSE) + # Regression line without confidence intervals
  labs(title = "Regression of Portfolio Returns on Pm_ratio",
       x = "Pm_ratio (Pm(t-1) / Pm(t-13) - 1)",
       y = "Portfolio Returns (R_m(t))") +
  theme_minimal()
```

4. Regress  $R_m(t)$  on an intercept and on  $\text{abs}(R_m(t))$ . Report estimates and t-statistics. Briefly interpret the results

```
data <- data %>%
  mutate(abs_Rm = abs(Portfolio_Returns))
head(data)
```

```
model_3 <- lm(Portfolio_Returns ~ abs_Rm, data = data)
summary(model_3)
```

```
ggplot(data, aes(x = abs_Rm, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) + # Scatter plot of returns
  geom_smooth(method = "lm", color = "red", se = FALSE) + # Regression line without confidence intervals
  labs(title = "Regression of Portfolio Returns on Absolute Returns",
       x = "Absolute Returns (abs(R_m(t)))",
```

```
y = "Portfolio Returns (R_m(t))" +
theme_minimal()
```

5. Repeat (1) on data from 199706. (Delete all data prior to 199706, then compute lagged returns). Comment briefly

```
data_filtered <- data %>%
  filter(Date >= as.Date("1997-06-01"))

data_filtered <- data_filtered %>%
  mutate(Rm_lag = lag(Portfolio_Returns, 1))

head(data_filtered)

model_filtered <- lm(Portfolio_Returns ~ Rm_lag, data = data_filtered)
summary(model_filtered)

ggplot(data_filtered, aes(x = Rm_lag, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) +
  geom_smooth(method = "lm", color = "red", se = FALSE) +
  labs(title = "Regression of Portfolio Returns on Lagged Returns (Post-1997)",
       x = "Lagged Portfolio Returns (R_m(t-1))",
       y = "Portfolio Returns (R_m(t))") +
  theme_minimal()
```

6.Repeat (2) on data from 199706. (Delete all data prior to 199706, then compute lagged returns). Comment briefly

```
head(data_filtered, 20)

model_2 <- lm(Portfolio_Returns ~ Pm_ratio, data = data_filtered)
summary(model_2)

ggplot(data_filtered, aes(x = Pm_ratio, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) +
  geom_smooth(method = "lm", color = "red", se = FALSE) +
  labs(title = "Regression of Portfolio Returns on Pm_ratio",
       x = "Pm_ratio (Pm(t-1) / Pm(t-13) - 1)",
       y = "Portfolio Returns (R_m(t))") +
  theme_minimal()
```

7. Repeat (4) on data from 199706. Comment briefly

```
data_filtered <- data_filtered %>%
  mutate(abs_Rm = abs(Portfolio_Returns))
head(data_filtered)

model_3 <- lm(Portfolio_Returns ~ abs_Rm, data = data_filtered)
summary(model_3)

ggplot(data_filtered, aes(x = abs_Rm, y = Portfolio_Returns)) +
  geom_point(color = "blue", alpha = 0.6) +
  geom_smooth(method = "lm", color = "red", se = FALSE) +
  labs(title = "Regression of Portfolio Returns on Absolute Returns (Post-1997)",
```

```
x = "Absolute Returns (abs(R_m(t)))",  
y = "Portfolio Returns (R_m(t))" +  
theme_minimal()
```