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# Financial Market Analytics Project

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## **Abstract**

The project consist in the creation of portfolios, utilizing FTSE MIB index stocks, using different criteria in order to select the assets of the portfolios with a final analisys of the returns and risks of the portfolios

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## Contents

### 1 Introduction

In this project, the goal consists in creating different portfolios, using FTSE MIB index stocks, using different criteria in order to select the assets of the portfolios. Finally we make some analysis on the created portfolios, such as the returns and the level of risk of the portfolios, in order to see which one is the better performing.

### 2 Dataset

The dataset used consists in past data about the FTSE MIB index, in particular is composed of daily data about the past 5 years of every singular stock present in the FTSE MIB index, obtained using the yahoo finance API.

### 2.1 Dataset Columns

The dataset is composed by the following columns:

- Date: Date relative to the dates of the singular stock.
- Open: Opening price of the stock.
- High: Highest price reached by the stock in the current day.
- Low: Lowest price reached by the stock in the current day.
- Close: Closing price of the stock.
- Volume: Trading volume of the stocks.
- Adjusted Close: Closing price adjusted after accounting for any corporate actions.
- log ret: This column is calculated using the adjusted close prices, is the logarithm of the adjusted close price of the current day of the stock subtracted by the logarithm of the adjusted close price of the previous day.

### 2.2 Data Exploration

The data did not present missing data so all the data is used inside the project.

## 3 Parameters

After obtaining the data, we proceeded to obtain some parameters relative to the single stocks, in order to create the portfolios.

### 3.1 Rolling Regression

In order to obtain some of the parameters needed we proceeded to do a step of rolling regression on every single stock. The rolling regression was made using data about the past 180 days, and repeated for every week. The rolling regression is made using the Security Market Line(SML):

$$r_i = \alpha_i + \beta_i(R_M) + e_i$$

From the various rolling regressions we obtain the following parameters:

- Beta: is the beta coefficient obtained directly from the regression, this parameter indicates the Systematic Risk, in other words the risk that cannot be diversified away.
- Residual Variance( $\sigma_{ei}$ ): Is the variance of the residuals of the regression.
- R-Squared: R-Squared statistic obtained from the rolling regression, indicates how well the regression can fit the data.

We decided to not use also the  $\alpha$  coefficient because in most of the regressions it wasn't statistically significant.

### 3.2 Other parameters

The other parameters used to build the portfolios are:

- Log returns: weekly logarithmic returns of the single stock.
- Risk: weekly risk of the single stock, obtained by calculating the variance of the weekly returns of the stock.

## 4 Portfolio Selection

In this phase we selected five different portfolios using different criteria. In general the portfolios are selected by ordering the weekly stock parameters in decrescent order, and selecting the top and bottom 10% of the ordered stocks. The parameters used for the portfolio creation are:

- Beta Coefficient
- Stock variance
- Stock returns
- Residual variance
- R-Squared

## 5 Portfolios Data

Using the mentioned parameters we obtained 5 different portfolios. For every portfolio we have obtained also the returns.

### 5.1 Portfolio returns

In order to obtain the portfolio returns we used the calculated weekly logarithmic returns, we made the inverse transformation in order to obtain the weekly return and multiplied for the available investment.

$$r_t = r_{t-1} * e^{\log ret_t}$$

### 5.2 Beta Portfolio

This portfolio is obtained by taking the top and bottom 10% of the stocks, ordered by the beta parameter, and is rebalanced every week. The portfolio expected return, on the five years analyzed, is 107.89 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 4.73.

### 5.3 R-Squared Portfolio

This portfolio is obtained by taking the top and bottom 10% of the stocks, ordered by the R-Squared parameter, and is rebalanced every week.

The portfolio expected return, on the five years analyzed, is 102.40 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 4.33.

### 5.4 Stock variance Portfolio

This portfolio is obtained by taking the top and bottom 10% of the stocks, ordered by their stock return variance, and is rebalanced every week.

The portfolio expected return, on the five years analyzed, is 106.62 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 4.50.

### 5.5 Residual variance Portfolio

This portfolio is obtained by taking the top and bottom 10% of the stocks, ordered by their

rolling regression residual variance, and is rebalanced every week.

The portfolio expected return, on the five years analyzed, is 102.40 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 4.67.

## 5.6 Weekly returns Portfolio

This portfolio is obtained by taking the top 20% of the stocks, ordered by their weekly returns, and is rebalanced every week.

This approach is a strategy called Ross Sectional Momentum Strategy, which consists in selecting the stocks of the portfolio by looking at past returns and taking only the best performing stocks, in this case we used the previous week best performing stocks. The portfolio expected return, on the five years analyzed, is 108.99 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 4.76.

## 5.7 FTSE-MIB Portfolio

This portfolio is obtained by following the FTSE-MIB index. The portfolio expected return, on the five years analyzed, is 99.27 euros, with an initial investment of 100 euros. The portfolio risk, calculated on the returns of a 100 euros investment, is equal to 2.52.

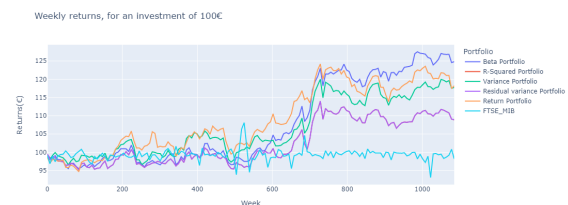
# 6 Results

Finally we made a table and a chart in order to compare all the portfolios.

**Table 1:** Table containing the expected returns and risks of the created portfolios, calculated with an initial investment of 100 euros

/	A. Ret(%)	A. Risk(&)	Eff.(%)
Beta	-2.86	34.07	-9.1
R-Squared	1.2	31.23	3.5
Variance	-3.39	32.47	-10.44
Res. Var.	1.2	33.7	3.56
Return	-1.89	34.27	-5.54
FTSE MIB	0.26	8.63	3.07

**Figure 1:** Line chart of portfolios returns



# 7 Conclusions

As we can see by the results the best performing portfolios are the ones built with Residuals variance and R-Squared as parameters. Many portfolios have negative annual returns, probably this negative annual return is caused because the dataset contains daily data including the years 2020-2022; it is possible that these bad results are caused by the covid 19 pandemic and the war in Ukraine.

# Bibliography

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