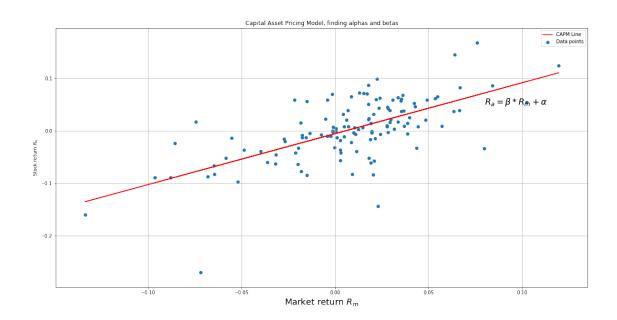
CAPM updated

June 30, 2020

```
[3]: import pandas_datareader as pdr
     from pandas_datareader import data, wb
     from datetime import date
     import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
     risk_free_rate = 0.05
     def capm(start_date, end_date, ticker1, ticker2):
             #qet the data from Yahoo Finance
             stock1 = pdr.get_data_yahoo(ticker1, start_date, end_date)
             stock2 = pdr.get_data_yahoo(ticker2, start_date, end_date)
             #we prefer monthly returns instead of daily returns
             return_stock1 = stock1.resample('M').last()
             return_stock2 = stock2.resample('M').last()
             \#creating a dataFrame from the data - Adjusted Closing Price is used as \sqcup
      \rightarrow usual
             data = pd.DataFrame({'s_adjclose' : return_stock1['Adj Close'],__
      → 'm_adjclose' : return_stock2['Adj Close']}, index=return_stock1.index)
             #natural logarithm of the returns
             data[['s_returns', 'm_returns']] = np.
      →log(data[['s_adjclose', 'm_adjclose']]/data[['s_adjclose', 'm_adjclose']].
      \rightarrowshift(1))
             #no need for NaN/missing values values so let's get rid of them
             data = data.dropna()
             #covariance matrix: the diagonal items are the vairances – off_{\sqcup}
      → diagonals are the covariances
             #the matrix is symmetric: cov[0,1] = cov[1,0] !!!
             covmat = np.cov(data["s_returns"], data["m_returns"])
             print(covmat)
             #calculating beta according to the formula
```

```
beta = covmat[0,1]/covmat[1,1]
       print("Beta from formula:", beta)
        #using linear regression to fit a line to the data [stock returns, ⊔
→market_returns] - slope is the beta
       beta,alpha = np.polyfit(data["m_returns"], data['s_returns'], deg=1)
       print("Beta from regression:", beta)
       #plot
       fig,axis = plt.subplots(1,figsize=(20,10))
       axis.scatter(data["m_returns"], data['s_returns'], label="Data points")
       axis.plot(data["m_returns"], beta*data["m_returns"] + alpha, ___
 plt.title('Capital Asset Pricing Model, finding alphas and betas')
       plt.xlabel('Market return $R_m$', fontsize=18)
       plt.ylabel('Stock return $R_a$')
       plt.text(0.08, 0.05, r'$R_a = \beta * R_m + \alpha$', fontsize=18)
       plt.legend()
       plt.grid(True)
       plt.show()
       #calculate the expected return according to the CAPM formula
       expected_return = risk_free_rate + beta*(data["m_returns"].
→mean()*12-risk_free_rate)
       print("Expected return:", expected_return)
if __name__ == "__main__":
       #using historical data 2010-2017: the market is the S&P500 !!!
       capm('2010-01-01', '2020-06-29','IBM', '^GSPC')
```

[[0.0035266 0.00152343] [0.00152343 0.00156933]] Beta from formula: 0.9707495612765603 Beta from regression: 0.9707495612765604



Expected return: 0.098842002414029

[]: