!pip install pandas-datareader statsmodels

```
Requirement already satisfied: pandas-datareac
Requirement already satisfied: statsmodels in
Requirement already satisfied: lxml in /usr/lc
Requirement already satisfied: pandas>=0.23 ir
Requirement already satisfied: requests>=2.19.
Requirement already satisfied: numpy<3,>=1.22.
Requirement already satisfied: scipy!=1.9.2,>=
Requirement already satisfied: patsy>=0.5.6 ir
Requirement already satisfied: packaging>=21.3
Requirement already satisfied: python-dateutil
Requirement already satisfied: pytz>=2020.1 ir
Requirement already satisfied: tzdata>=2022.7
Requirement already satisfied: charset-normali
Requirement already satisfied: idna<4,>=2.5 ir
Requirement already satisfied: urllib3<3,>=1.2
Requirement already satisfied: certifi>=2017.4
Requirement already satisfied: six>=1.5 in /us
```

import pandas as pd
import pandas_datareader.data as web
import statsmodels.formula.api as sm

```
# Download Fama-French 3-factor data (monthly)
ff_data = web.DataReader('F-F_Research_Data_Factor
ff_data = ff_data[0] # Get the monthly data (the
ff_data.index = ff_data.index.to_timestamp() # Cor
ff_data = ff_data / 100 # Convert percentages to c
```

FloatingPointError

F •••

```
import yfinance as yf
end_date = datetime.date.today().strftime('%Y-%m-%d
apple = yf.Ticker("AAPL")
AAPL = apple.history(start = "2020-01-01", end= end
AAPL.head()
\rightarrow
                    Open
                               High
                                           Low
                                                   Close
         Date
      2020-01-
         02
                71.721011 72.776591 71.466805 72.716064
      00:00:00-
       05:00
      2020-01-
         03
                71.941321 72.771737 71.783954 72.009109
      00:00:00-
       05:00
      2020-01-
         06
                71.127873 72.621654 70.876083 72.582916
      00:00:00-
       05:00
 Next
         Generate code with AAPL
                                  View recommended plo
 steps:
```

Get the Balance Sheet and Income Statements

```
balance_sheet = apple.balance_sheet
print("Balance Sheet:")
print(balance_sheet.head())

income_statement = apple.financials
print("\nIncome Statement:")
print(income_statement.head())

# Information about Apple:
info = apple.info
print(f"\nCompanv: {info['longName']}")
```

print(f"Sector: {info['sector']}")
print(f"Industry: {info['industry']}")
print(f"Market Cap: {info['marketCap']}")
print(f"P/E Ratio: {info['trailingPE']}")

dividend data
dividends = apple.dividends
print("Dividends:")
print(dividends.tail())

→ Balance Sheet:

2024-09-30	24
NaN	
15116786000.0	15550
15116786000.0	15550
76686000000.0	81123
106629000000.0	111088
	NaN 15116786000.0 15116786000.0 76686000000.0

2021-09-30 2020-09

2024 00 20

20

Treasury Shares Number
Ordinary Shares Number
Share Issued
Net Debt
Total Debt
NaN
16426786000.0
89779000000.0
136522000000.0

Income Statement:

Tax Effect Of Unusual Items
Tax Rate For Calcs
Normalized EBITDA
Net Income From Continuing Operation Net Minor
Reconciled Depreciation

Tax Effect Of Unusual Items
Tax Rate For Calcs
Normalized EBITDA
Net Income From Continuing Operation Net Minor
Reconciled Depreciation

Tax Effect Of Unusual Items
Tax Rate For Calcs
Normalized EBITDA
Net Income From Continuing Operation Net Minor
Reconciled Depreciation

Tax Effect Of Unusual Items

Tax Rate For Calcs Normalized EBITDA Net Income From Continuing Operation Net Minor Reconciled Depreciation Company: Apple Inc. Sector: Technology Industry: Consumer Electronics Market Cap: 2590110646272 P/E Ratio: 27.368254 Dividends: Date 2024-02-09 00:00:00-05:00 0.24 2024-05-10 00:00:00-04:00 0.25 2024-08-12 00:00:00-04:00 0.25 2024-11-08 00:00:00-05:00 0.25 2025-02-10 00:00:00-05:00 0.25 Name: Dividends, dtype: float64 apple = yf.Ticker("AAPL") tickers = ["SPY", "AAL", "ZM", "NFLX", "META", 'A/ end_date = datetime.date.today().strftime('%Y-%m-9 apple = yf.Ticker("AAPL") AAPL = apple.history(start = "2020-01-01", end= "2 for ticker in tickers: globals()[ticker] = yf.Ticker(ticker) globals()[ticker] = globals()[ticker].history

print(META.Close.mean()) META.describe()



→ 299.9754145219389

	Open	High	Low	
count	1257.000000	1257.000000	1257.000000	1257.(
mean	299.811910	304.029698	295.797743	299.9
std	124.745251	125.702634	123.419958	124.6
min	89.657445	90.035660	87.676766	88.4
25%	207.860343	210.607414	205.541276	208.7
50%	277.850522	283.892024	274.984004	279.5
75%	345.003997	350.448357	341.570169	344.6
max	630.430194	637.318496	626.147483	631.6

Now, let us keep only the closing prices for our analysis.

keep only column close for all tickers
for ticker in tickers:
 globals()[ticker] = globals()[ticker].Close

SPY



Close

Date	
2020-01-02 00:00:00-05:00	300.291504
2020-01-03 00:00:00-05:00	298.017792
2020-01-06 00:00:00-05:00	299.154602
2020-01-07 00:00:00-05:00	298.313507
2020-01-08 00:00:00-05:00	299.903351
2024-12-23 00:00:00-05:00	592.906433
2024-12-24 00:00:00-05:00	599.496582
2024-12-26 00:00:00-05:00	599.536499
2024-12-27 00:00:00-05:00	593.225464
2024-12-30 00:00:00-05:00	586.455811
1257 rows × 1 columns	

dtype: float64



Gemini

Gemini is a powerful AI tool built by Google that helps you use Colab. Not sure what to ask? Try a suggested prompt below

How do I filter a Pandas DataFrame?

How can I create a plot in Colab?

Show me a list of publicly available datasets

keep only column close for all tickers
for ticker in tickers:

Assign the 'Close' column to the ticker var:
globals()[ticker] = globals()[ticker]['Close']

SPY

 $\overline{\Sigma}$

print(AAPL)

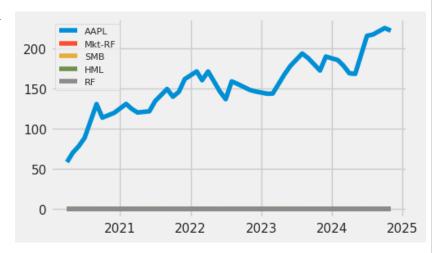
print(AAPL.shift(1))

Close

Date	
2020-01-02 00:00:00-05:00	300.291504
2020-01-03 00:00:00-05:00	298.017700
2020-01-06 00:00:00-05:00	299.154633
2020-01-07 00:00:00-05:00	298.313507
2020-01-08 00:00:00-05:00	299.903381
2024-12-23 00:00:00-05:00	592.906433
2024-12-24 00:00:00-05:00	599.496582
2024-12-26 00:00:00-05:00	599.536499
2024-12-27 00:00:00-05:00	593.225464
2024-12-30 00:00:00-05:00	586.455811
1257 rows × 1 columns	
dtype: float64	

```
plt.style.use('fivethirtyeight')
plt.figure(figsize=(5, 3))
plt.plot(df, label=df.columns)
plt.legend(loc='upper left',fontsize=8)
plt.show()
```





For financial analysis, we require the log returns (daily), rather than the raw stock prices. The formula for log returns is:

log(Today's Price/yesterday's price - 1)

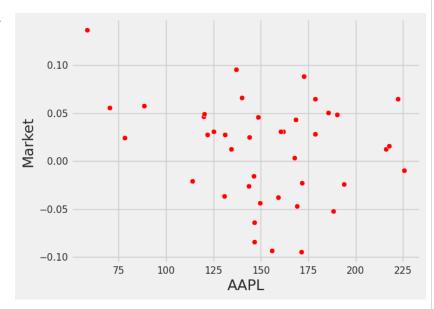
Find the betas of the stocks. The formula is shown below:

beta_aapl = (df[['Mkt-RF','AAPL']].cov()/df['Mkt-F

Calculate beta using regression line.

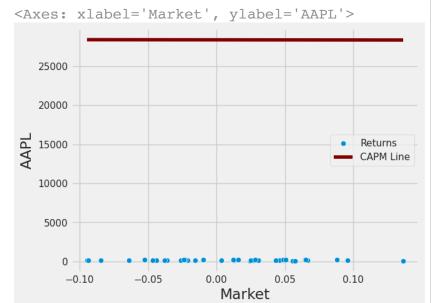
```
# plt.axvline(0, color='grey', alpha = 0.5)
# plt.axhline(0, color='grey', alpha = 0.5)
sns.scatterplot(y = 'Market', x = 'AAPL', data = (
plt.show()
```





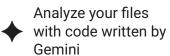
sns.scatterplot(y = 'AAPL', x = 'Market', data = (sns.lineplot(x = data['Market'], y = alpha + (data))





Convert Daily Stock Market Returns to
Annualized Returns (assuming 252 trading days in a year).

```
rm = data['Market'].mean()*252
rm
cov = data[['Market','AAPL']].cov() *252
cov_aapl_market = cov.iloc[0,1]
cov_aapl_market
market_var = data['Market'].var()*252
market var
AAPL_beta_annual = cov_aapl_market / market_var
print('The annualized beta will equal the one cald
rf = 0.025
riskpremium = rm - rf
## CAPM
AAPL capm return = rf + AAPL beta annual*riskprem:
print(f"The annualized CAPM return of AAPL is {AAF
→▼ The annualized beta will equal the one calcula
    The annualized CAPM return of AAPL is -57061.1
sharperatio = (rm-rf)/(data['AAPL'].std()*np.sqrt()
sharperatio
print(f"Sharpe Ratio: {round(sharperatio,4)}")
→ Sharpe Ratio: 0.0051
Start coding or generate with AI.
Start coding or generate with AI.
Start coding or generate with AI.
```



Upload

Enter a prompt here



0 / 2000

Gemini can make mistakes so double-check responses and use code with caution. <u>Learn more</u>