```
!pip install yfinance
Requirement already satisfied: yfinance in /usr/local/lib/python3.11/dist-packages (0.2.56)
    Requirement already satisfied: pandas>=1.3.0 in /usr/local/lib/python3.11/dist-packages (from yfinance) (2.2.2)
    Requirement already satisfied: numpy>=1.16.5 in /usr/local/lib/python3.11/dist-packages (from yfinance) (2.0.2)
    Requirement already satisfied: requests>=2.31 in /usr/local/lib/python3.11/dist-packages (from yfinance) (2.32.3)
    Requirement already satisfied: multitasking>=0.0.7 in /usr/local/lib/python3.11/dist-packages (from yfinance) (0.0.11)
    Requirement already satisfied: platformdirs>=2.0.0 in /usr/local/lib/python3.11/dist-packages (from yfinance) (4.3.7)
    Requirement already satisfied: pytz>=2022.5 in /usr/local/lib/python3.11/dist-packages (from yfinance) (2025.2)
    Requirement already satisfied: frozendict>=2.3.4 in /usr/local/lib/python3.11/dist-packages (from yfinance) (2.4.6)
    Requirement already satisfied: peewee>=3.16.2 in /usr/local/lib/python3.11/dist-packages (from yfinance) (3.17.9)
    Requirement already satisfied: beautifulsoup4>=4.11.1 in /usr/local/lib/python3.11/dist-packages (from yfinance) (4.13.4)
    Requirement already satisfied: soupsieve>1.2 in /usr/local/lib/python3.11/dist-packages (from beautifulsoup4>=4.11.1->yfinance) (2.7)
    Requirement already satisfied: typing-extensions>=4.0.0 in /usr/local/lib/python3.11/dist-packages (from beautifulsoup4>=4.11.1->yfinance) (4.13.2)
    Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.3.0->yfinance) (2.9.0.post0)
    Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.3.0->yfinance) (2025.2)
    Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests>=2.31->yfinance) (3.4.1)
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests>=2.31->yfinance) (3.10)
    Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests>=2.31->yfinance) (2.4.0)
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests>=2.31->yfinance) (2025.1.31)
    Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas>=1.3.0->yfinance) (1.17.0)
import pandas as pd
import yfinance as yf
import plotly.io as pio
import plotly.graph objects as go
pio.templates.default ="plotly white"
tickers={
     "apple ticker": "AAPL",
     "google ticker": "GOOGL",
     "facebook ticker": "META",
     "intel ticker":"INTC",
     "microsoft ticker":"MSFT",
     "reliance_ticker":"RELIANCE.NS",
     "Tesla ticker": "TSLA",
     "Amazon ticker": "AMZN"
start_date="2024-10-01"
end date="2024-12-31"
apple data=yf.download(tickers["apple ticker"],start=start date, end=end date)
google data=yf.download(tickers["google ticker"],start=start date, end=end date)
facebook data=yf.download(tickers["facebook ticker"],start=start date, end=end date)
```

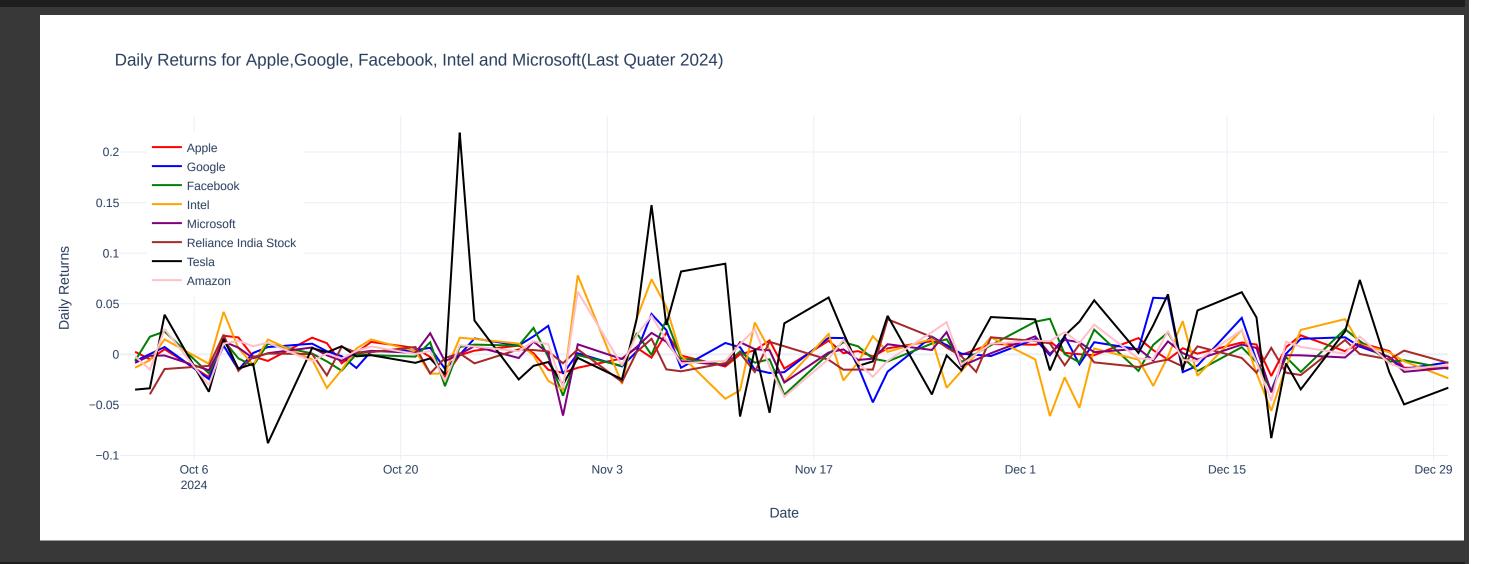
 $1 ext{ of } 8 ext{ } 30/04/25, 17:49$

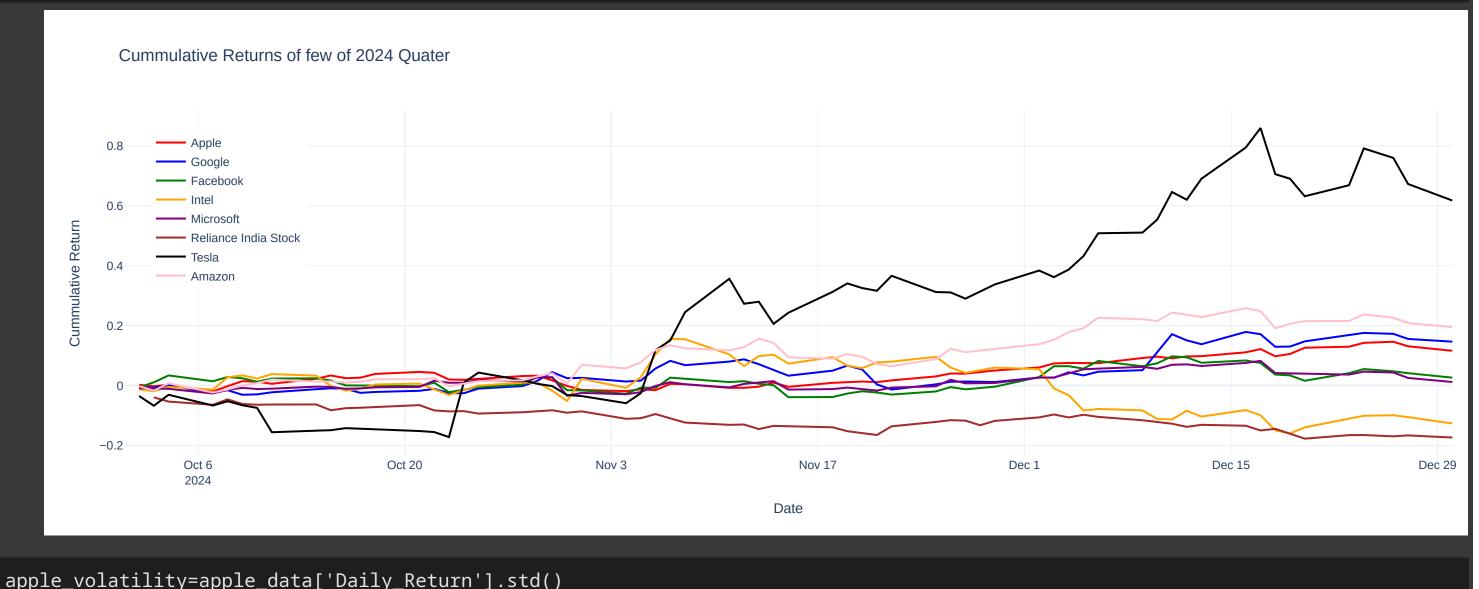
```
Intel_uata-yr.uownloau(tickers[ intel_ticker ],Start-Start_uate, enu-enu_uate)
microsoft data=yf.download(tickers["microsoft ticker"],start=start date, end=end date)
reliance data=yf.download(tickers["reliance ticker"], start=start date, end=end date)
tesla_data=yf.download(tickers["Tesla_ticker"],start=start_date, end=end_date)
amazon data=yf.download(tickers["Amazon ticker"],start=start date, end=end date)
YF.download() has changed argument auto_adjust default to True
   [********** 100%********************** 1 of 1 completed
   [********* 1 of 1 completed
   apple_data['Daily_Return'] = apple_data['Close'].pct_change()
google_data['Daily_Return'] = google_data['Close'].pct_change()
facebook_data['Daily Return']=facebook_data['Close'].pct_change()
Intel_data['Daily Return']=Intel_data['Close'].pct_change()
microsoft_data['Daily Return']=microsoft_data['Close'].pct_change()
reliance_data['Daily Return']=reliance_data['Close'].pct_change()
tesla_data['Daily Return']=tesla_data['Close'].pct_change()
amazon_data['Daily Return']=amazon_data['Close'].pct_change()
fig=go.Figure()
fig.add_trace(go.Scatter(x=apple_data.index,y=apple_data['Daily_Return'],mode='lines',name='Apple',
                      line=dict(color='red',width=2)))
fig.add_trace(go.Scatter(x=google_data.index,y=google_data['Daily_Return'],mode='lines',name='Google',
                      line=dict(color='blue',width=2)))
fig.add_trace(go.Scatter(x=facebook_data.index,y=facebook_data['Daily Return'],mode='lines',name='Facebook',
                      line=dict(color='green',width=2)))
fig.add_trace(go.Scatter(x=Intel_data.index,y=Intel_data['Daily Return'],mode='lines',name='Intel',
                      line=dict(color='orange', width=2)))
fig.add_trace(go.Scatter(x=microsoft_data.index,y=microsoft_data['Daily Return'],mode='lines',name='Microsoft',
                      line=dict(color='purple',width=2)))
fig.add_trace(go.Scatter(x=reliance_data.index,y=reliance_data['Daily Return'], mode='lines',name='Reliance India Stock',
                      line=dict(color='brown',width=2)))
fig.add_trace(go.Scatter(x=tesla_data.index,y=tesla_data['Daily Return'],mode='lines',name='Tesla',
                      line=dict(color='black',width=2)))
fig.add_trace(go.Scatter(x=amazon_data.index,y=amazon_data['Daily Return'],mode='lines',name='Amazon',
                      line=dict(color='pink',width=2)))
```

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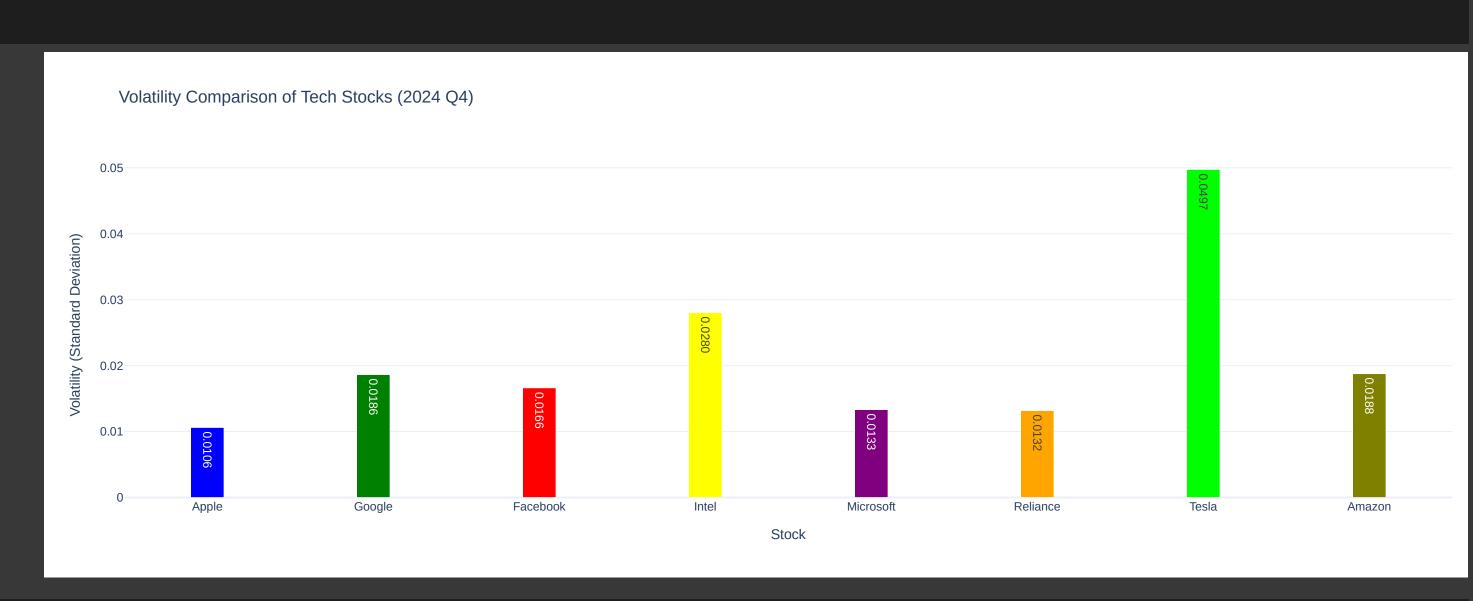
```
fig.update_layout(title='Daily Returns for Apple,Google, Facebook, Intel and Microsoft(Last Quater 2024)',xaxis_title='Date',yaxi
legend=dict(x=0.02,y=0.95))
```

fig.show()





```
google_volatility=google_data['Daily_Return'].std()
facebook_volatility=facebook_data['Daily Return'].std()
Intel volatility=Intel data['Daily Return'].std()
microsoft_volatility=microsoft_data['Daily Return'].std()
reliance_volatility=reliance_data['Daily Return'].std()
tesla volatility=tesla data['Daily Return'].std()
amazon_volatility=amazon_data['Daily Return'].std()
data_dict = {
    'Apple': apple_volatility,
    'Google': google_volatility,
    'Facebook': facebook_volatility,
    'Intel': Intel_volatility,
    'Microsoft': microsoft volatility,
    'Reliance': reliance_volatility,
    'Tesla': tesla_volatility,
    'Amazon': amazon volatility
    # Add more like:
    # 'Tesla': tesla_data['Daily Return'].std(),
    # 'Amazon': amazon data['Daily Return'].std()
colors = ['blue', 'green', 'red', 'yellow', 'purple', 'orange', 'lime', 'olive']
fig1 = go.Figure()
fig1.add_bar(
    x=list(data_dict.keys()),
    y=list(data_dict.values()),
    text=[f'{v:.4f}' for v in data_dict.values()],
    textposition='auto',
    marker=dict(color=colors[:len(data_dict)])
fig1.update_layout(
    title='Volatility Comparison of Tech Stocks (2024 Q4)',
    xaxis title='Stock',
    yaxis_title='Volatility (Standard Deviation)',
    bargap=0.8
fig1.show()
```



```
market_data = yf.download('^GSPC', start=start_date, end=end_date)
market_data['Daily Return'] = market_data['Close'].pct_change()

# Create a dictionary to store the results
results = {}

# Loop through the tickers and calculate the beta and covariance
for company, ticker in tickers.items():
    # Download company data (if not already downloaded)
    if company.endswith("_ticker"):
        company_name = company[:-len("_ticker")]
        company_data = locals()[company_name.lower() + "_data" if company_name.lower() != "intel" else "Intel_data"]

    if 'Daily_Return' in company_data.columns and 'Daily Return' not in company_data.columns:
        company_data = company_data.rename(columns={'Daily_Return': 'Daily_Return'})

    if 'Daily_Return' not in company_data.columns
```

```
II Dally Netull Hot in company_data.coidmis.
            company_data['Daily Return'] = company_data['Close'].pct_change()
    else:
        company_data = yf.download(ticker, start=start_date, end=end_date)
        company data['Daily Return'] = company data['Close'].pct change()
    # Calculate covariance and beta
    cov = company_data['Daily Return'].cov(market_data['Daily Return'])
    beta = cov / market data['Daily Return'].var()
    # Store the results
    results[company_name or company] = {'Covariance': cov, 'Beta': beta} # Use company_name or company as key
# Display the results
for company, data in results.items():
    print(f"\n{company}:")
    print(f" Covariance with Market: {data['Covariance']:.6f}")
    print(f" Beta: {data['Beta']:.6f}\n")
    print("-" * 20)
# Find the company with the highest volatility (beta)
most_volatile_company = max(results, key=lambda company: results[company]['Beta'])
print(f"\n{most_volatile_company} has the highest volatility (beta) among the selected companies.")
   Covariance with Market: 0.000044
    Beta: 0.691758
   google:
    Covariance with Market: 0.000075
    Beta: 1.184915
   facebook:
    Covariance with Market: 0.000079
    Beta: 1.249585
   intel:
    Covariance with Market: 0.000135
    Beta: 2.138677
```

microsoft:

Covariance with Market: 0.000074

Beta: 1.179948

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reliance:

Covariance with Market: 0.000016

Beta: 0.261628

Tesla:

Covariance with Market: 0.000188

Beta: 2.986235

Amazon:

Covariance with Market: 0.000105

Beta: 1.667240

Tesla has the highest volatility (beta) among the selected companies.