

Chapter_1

March 2, 2025

1 Chapter 1

```
[1]: import requests_cache
import yfinance as yf
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import scipy as sp
from statsmodels.tsa.stattools import acf
```

```
[2]: session = requests_cache.CachedSession('yfinance.cache')
session.headers['User-Leonardo'] = 'Chapter_1'
```

```
[3]: data = yf.download('^STOXX50E')
```

YF.download() has changed argument auto_adjust default to True

[*****100%*****] 1 of 1 completed

```
[4]: prices = data.Close['^STOXX50E']
prices
```

```
[4]: Date
2007-03-30    4181.029785
2007-04-02    4189.549805
2007-04-03    4246.299805
2007-04-04    4261.830078
2007-04-05    4271.540039
...
2025-02-24    5453.759766
2025-02-25    5447.899902
2025-02-26    5527.990234
2025-02-27    5472.560059
2025-02-28    5463.540039
Name: ^STOXX50E, Length: 4493, dtype: float64
```

```
[5]: prices.describe()
```

```
[5]: count    4493.000000
     mean     3387.685643
     std      709.582645
     min     1809.979980
     25%     2875.939941
     50%     3324.860107
     75%     3781.790039
     max     5533.839844
     Name: ^STOXX50E, dtype: float64
```

```
[6]: prices = prices.to_numpy()
```

```
[7]: dates = data.Close['^STOXX50E'].index
```

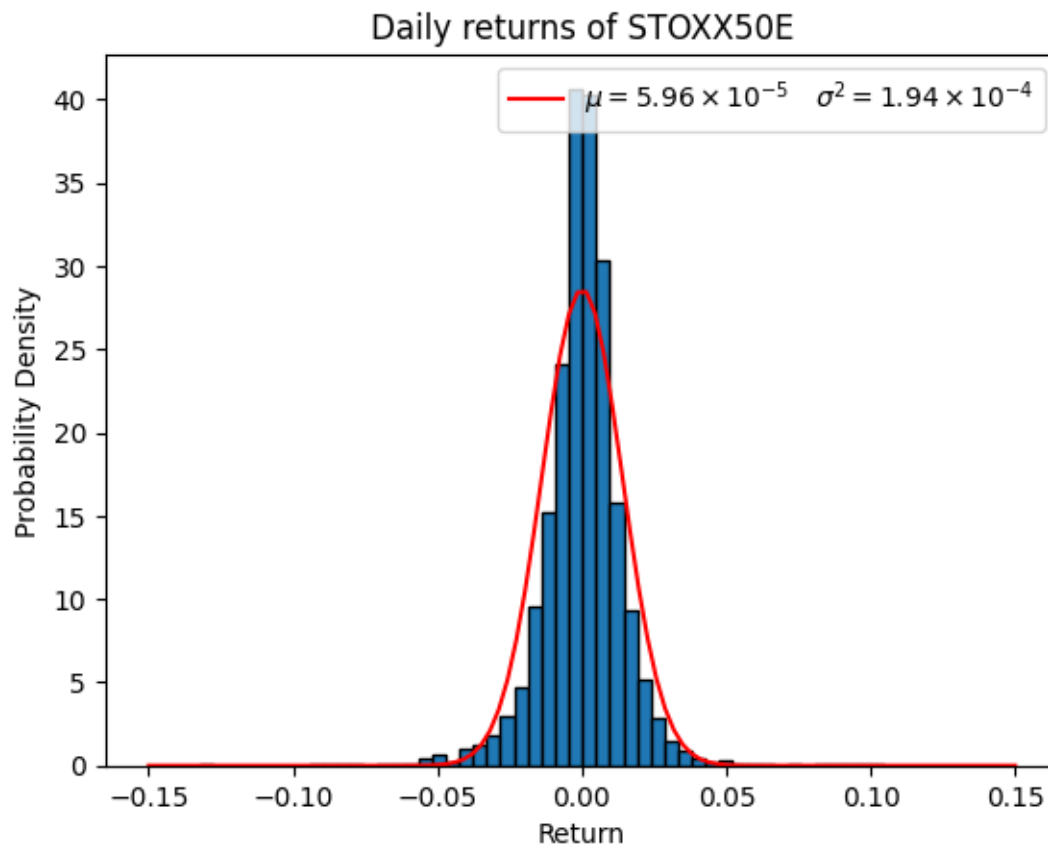
1.1 Returns

```
[8]: returns = np.log(prices[1:])- np.log(prices[:-1])
     mean = returns.mean()
     variance = returns.var()
     print('average = ',mean, '\nvariance = ', variance)
```

```
average =  5.9559074395698434e-05
variance =  0.00019448892995148706
```

```
[9]: # Generate a range of x values for the normal distribution
     x = np.linspace(-0.15, 0.15, 100)
     # Evaluate the normal distribution at the x values
     y = sp.stats.norm(loc =mean, scale = np.sqrt(variance)).pdf(x)

     # Plot the histogram
     plt.hist(returns, bins=50, edgecolor='black', density=True)
     # Plot the fitted distribution
     plt.plot(x, y, 'r-', label=r'$\mu = 5.96\times 10^{-5}$ \quad $\sigma^2 = 1.94\times 10^{-4}$')
     plt.xlabel('Return')
     plt.ylabel('Probability Density')
     plt.title('Daily returns of STOXX50E')
     plt.legend()
     plt.show()
```

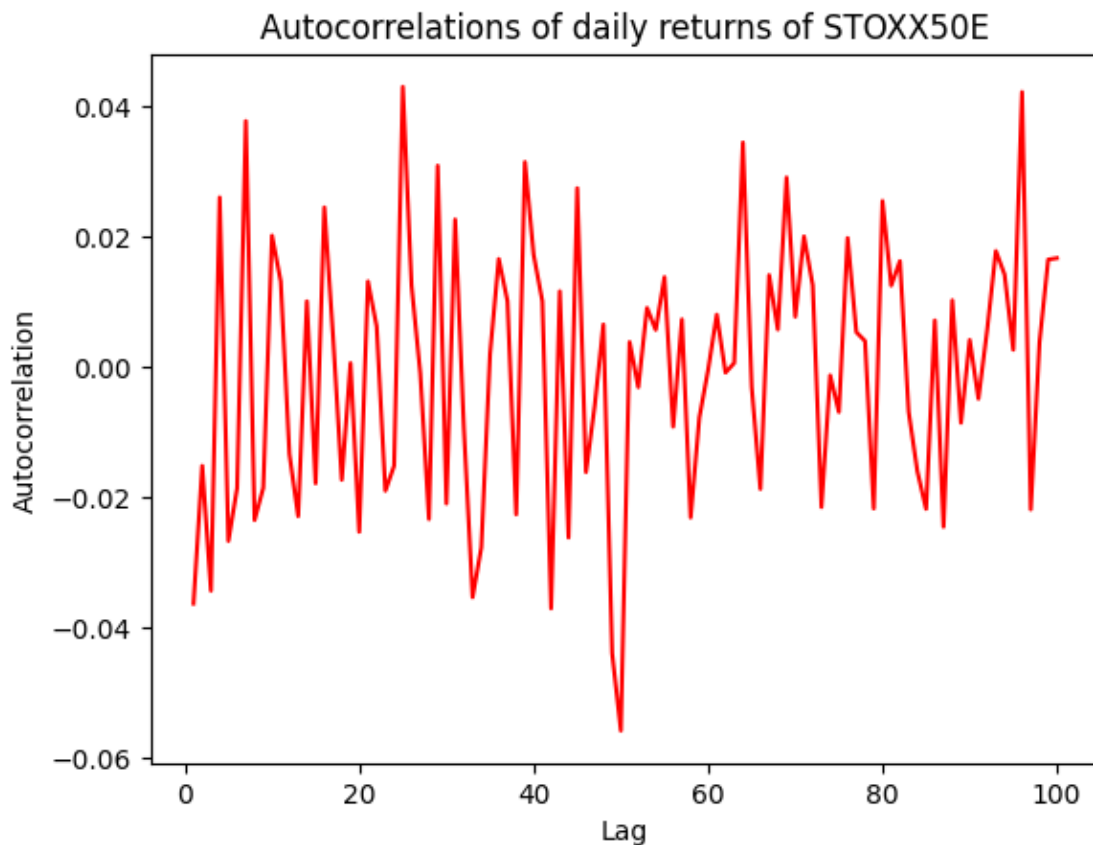


1.2 Autocorrelations

```
[10]: max_lag = 100
      # Calculate autocorrelation for lags 1 to 100
      autocorr_values = acf(returns, nlags=max_lag)
```

```
[11]: x = np.arange(max_lag+1)

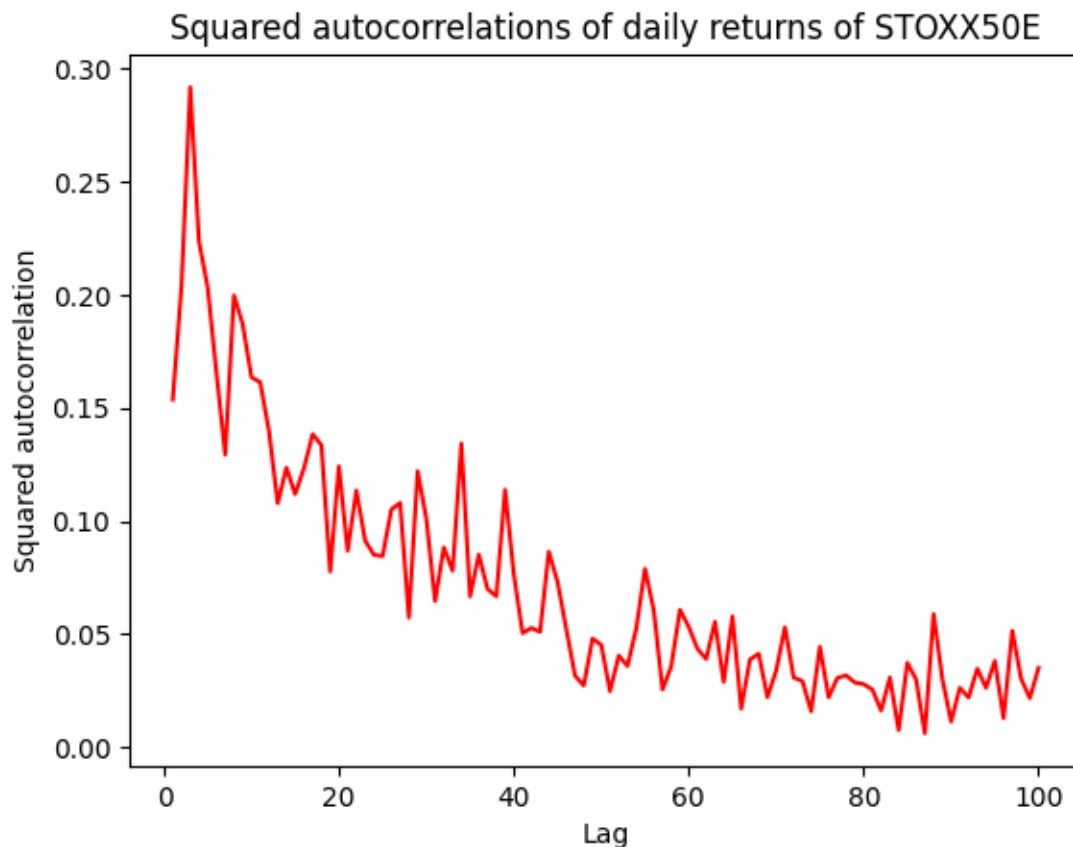
      plt.plot(x[1:], autocorr_values[1:], 'r-')
      plt.xlabel('Lag')
      plt.ylabel('Autocorrelation')
      plt.title('Autocorrelations of daily returns of STOXX50E')
      plt.show()
```



1.3 Squared Autocorrelations

```
[12]: # Calculate autocorrelation for lags 1 to 100
sq_autocorr_values = acf(returns**2, nlags=max_lag)
```

```
[13]: plt.plot(x[1:], sq_autocorr_values[1:], 'r-')
plt.xlabel('Lag')
plt.ylabel('Squared autocorrelation')
plt.title('Squared autocorrelations of daily returns of STOXX50E')
plt.show()
```

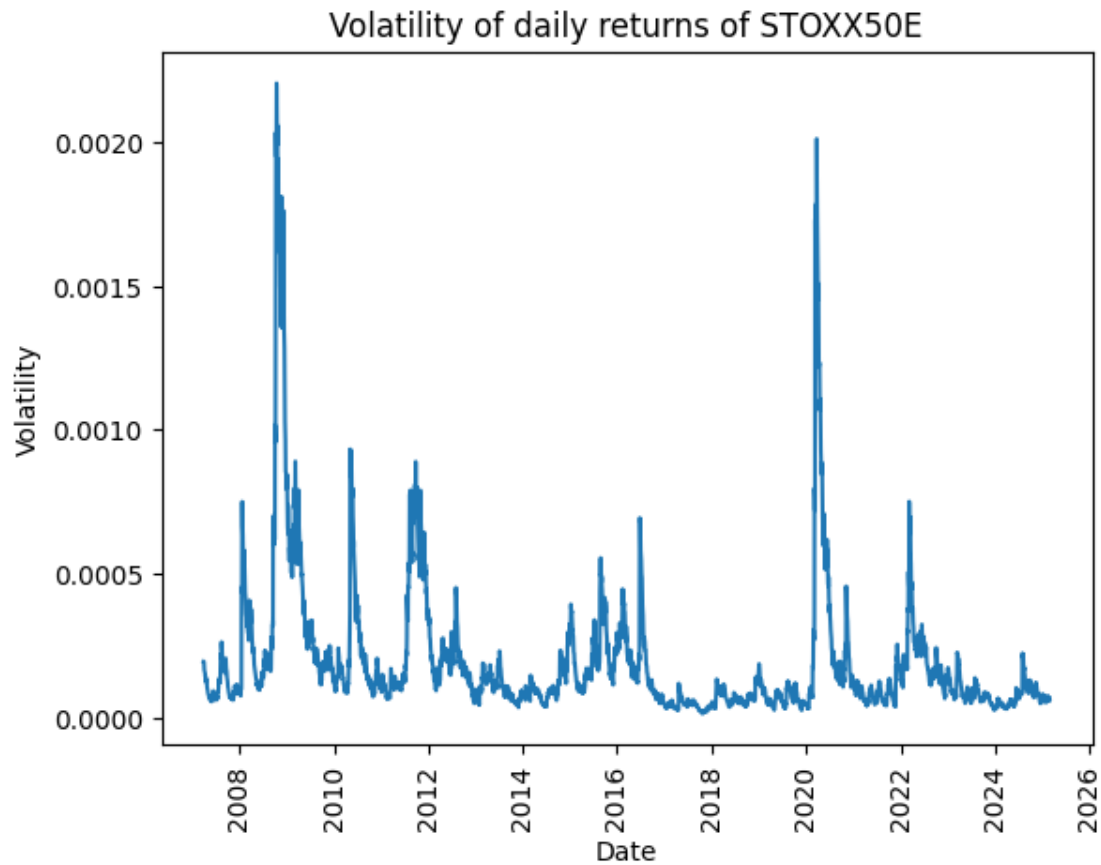


1.4 Riskmetrics

```
[14]: def risk_metrics(returns, persistence):
    t_max = len(returns)
    volatility = np.zeros(t_max)
    volatility[0] = returns.var()
    for t in range(1, t_max):
        volatility[t] = persistence*volatility[t-1] +
        ↪(1-persistence)*returns[t]**2
    return volatility
```

```
[15]: volatility = risk_metrics(returns, 0.94)
```

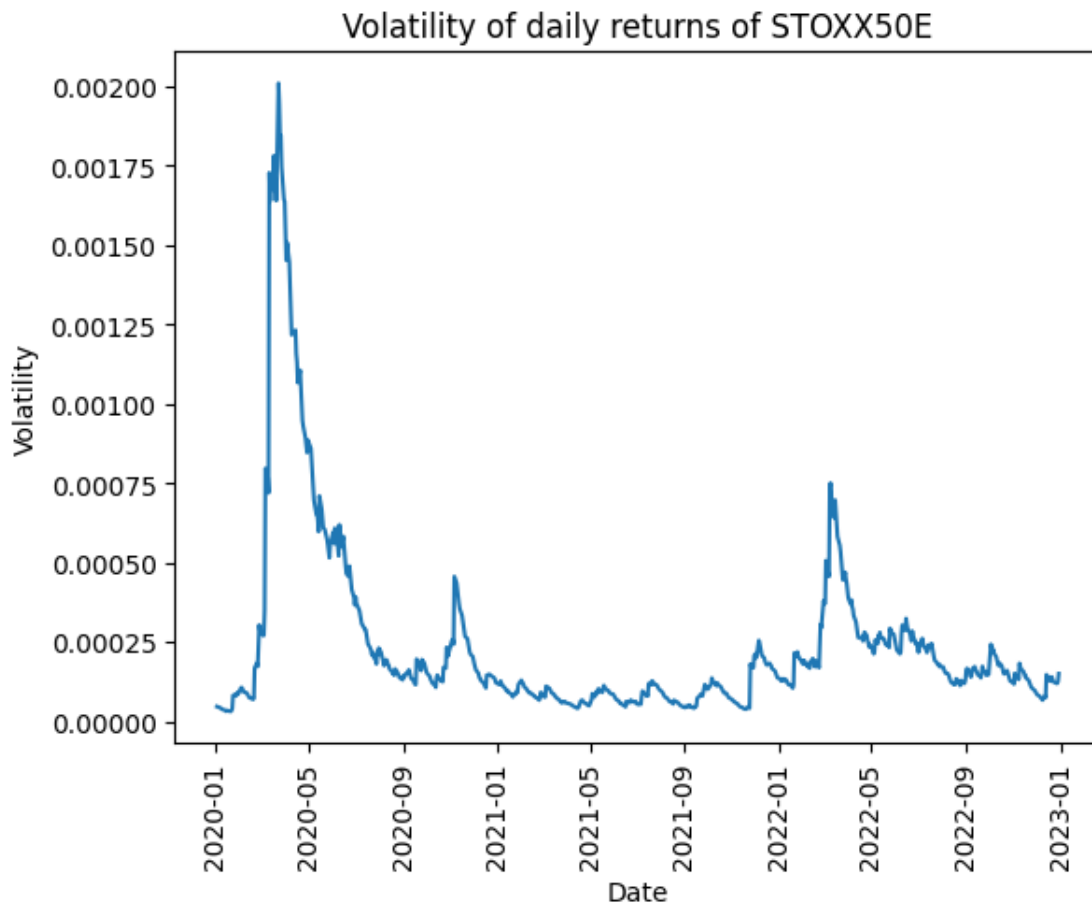
```
[16]: plt.plot(dates[: -1], volatility)
# rotate x-tick-labels by 90°
plt.tick_params(axis='x', rotation=90)
plt.xlabel('Date')
plt.ylabel('Volatility')
plt.title('Volatility of daily returns of STOXX50E')
plt.show()
```



```
[17]: start_date = np.datetime64('2020-01-01')
end_date = np.datetime64('2022-12-31')

mask = (dates[:-1] >= start_date) & (dates[:-1] <= end_date)

plt.plot(dates[:-1][mask], volatility[mask])
# rotate x-tick-labels by 90°
plt.tick_params(axis='x', rotation=90)
plt.xlabel('Date')
plt.ylabel('Volatility')
plt.title('Volatility of daily returns of STOXX50E')
plt.show()
```



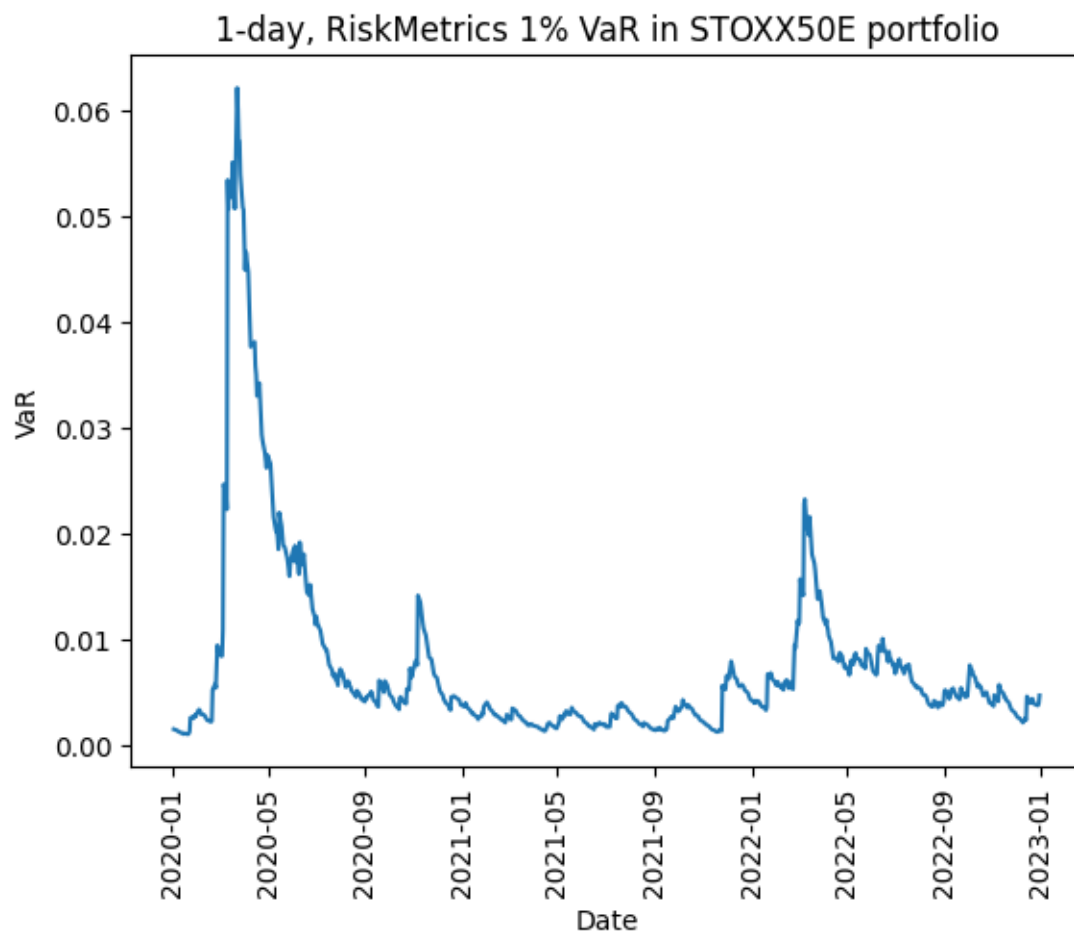
1.5 VaR

Assuming gaussian distribution of portfolio returns

```
[18]: p = 0.01
      quantile = sp.stats.norm.ppf(p, loc=mean, scale=np.sqrt(variance))
      print(-quantile)
```

0.032383522729066964

```
[19]: plt.plot(dates[:-1][mask], -quantile ** (-1) * volatility[mask])
      # rotate x-tick-labels by 90°
      plt.tick_params(axis='x', rotation=90)
      plt.xlabel('Date')
      plt.ylabel('VaR')
      plt.title('1-day, RiskMetrics 1% VaR in STOXX50E portfolio')
      plt.show()
```



[]: