

Chapter_1_Gold

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'
```

```
[8]: ticker = 'GC=F'
data = yf.download(ticker)
```

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

```
[9]: prices = data.Close[ticker]
prices
```

```
[9]: Date
2000-08-30    273.899994
2000-08-31    278.299988
2000-09-01    277.000000
2000-09-05    275.799988
2000-09-06    274.200012
...
2025-02-24    2947.899902
2025-02-25    2904.500000
2025-02-26    2916.800049
2025-02-27    2883.199951
2025-02-28    2867.300049
Name: GC=F, Length: 6147, dtype: float64
```

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

```
[10]: count    6147.000000
      mean     1172.018448
      std      594.010026
      min      255.100006
      25%      633.799988
      50%      1236.199951
      75%      1628.400024
      max      2947.899902
      Name: GC=F, dtype: float64
```

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

```
[12]: dates = data.Close[ticker].index
```

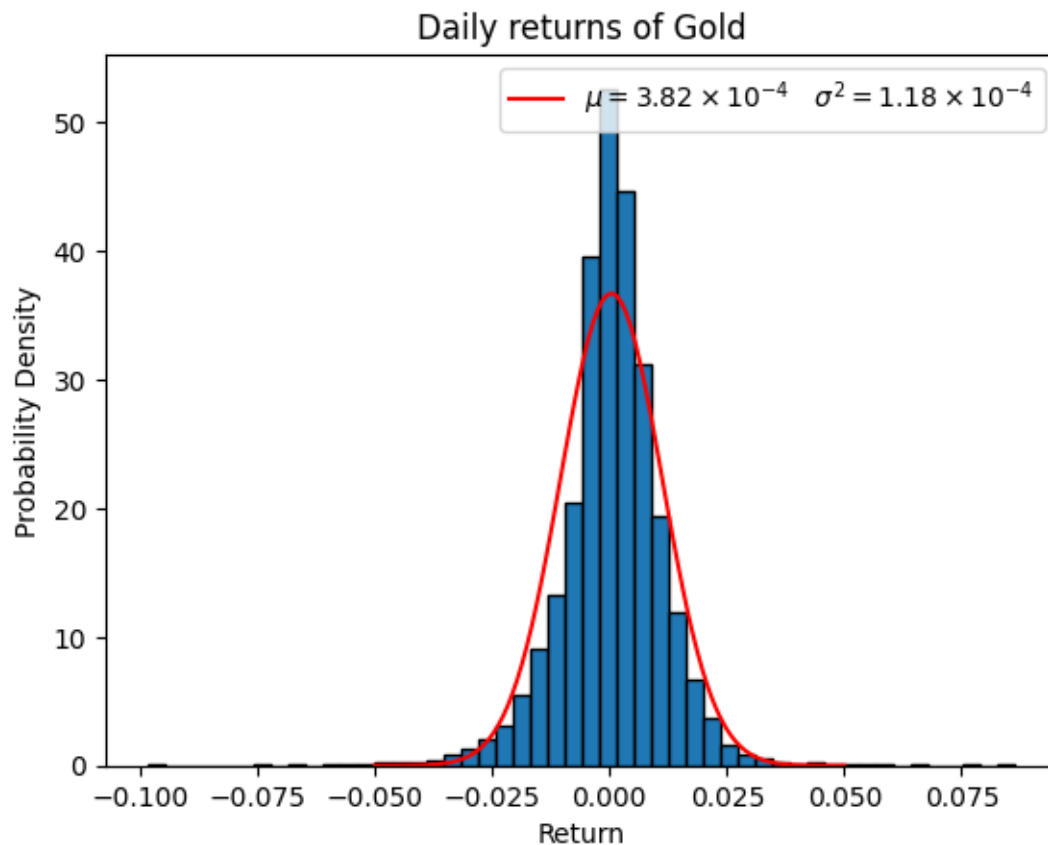
1.1 Returns

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

```
average =  0.00038209617026498
variance =  0.00011792358757804685
```

```
[37]: # Generate a range of x values for the normal distribution
      x = np.linspace(-0.05, 0.05, 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'$\mu = 3.82\times 10^{-4}$ \quad \sigma^2 = 1.
      \times 10^{-4}$')
      plt.xlabel('Return')
      plt.ylabel('Probability Density')
      plt.title('Daily returns of Gold')
      plt.legend()
      plt.show()
```

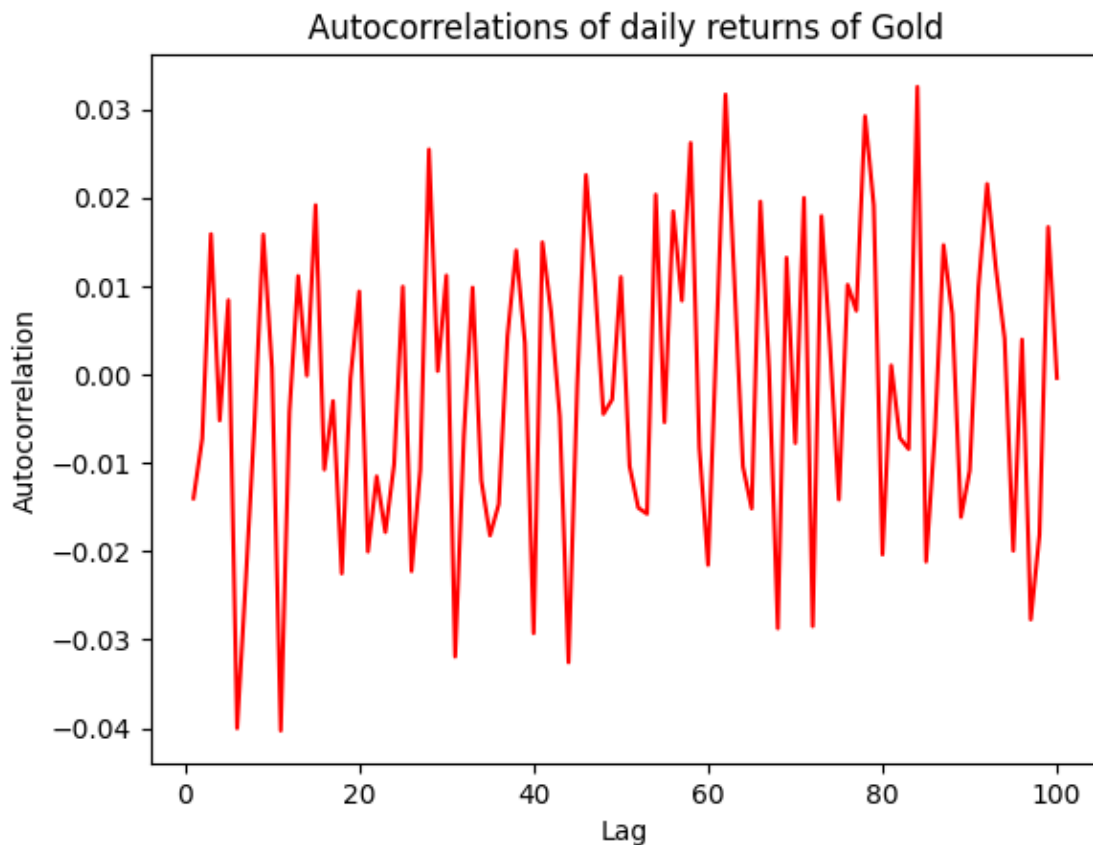


1.2 Autocorrelations

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

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

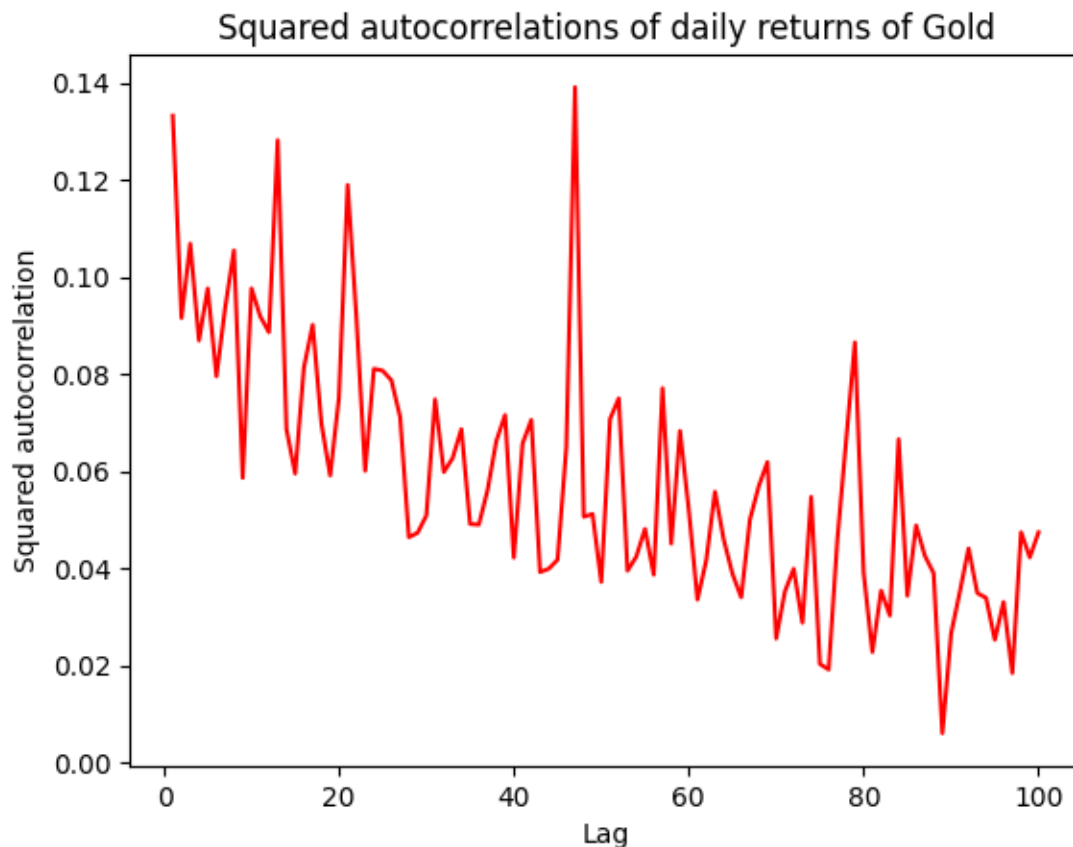
      plt.plot(x[1:], autocorr_values[1:], 'r-')
      plt.xlabel('Lag')
      plt.ylabel('Autocorrelation')
      plt.title('Autocorrelations of daily returns of Gold')
      plt.show()
```



1.3 Squared Autocorrelations

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

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

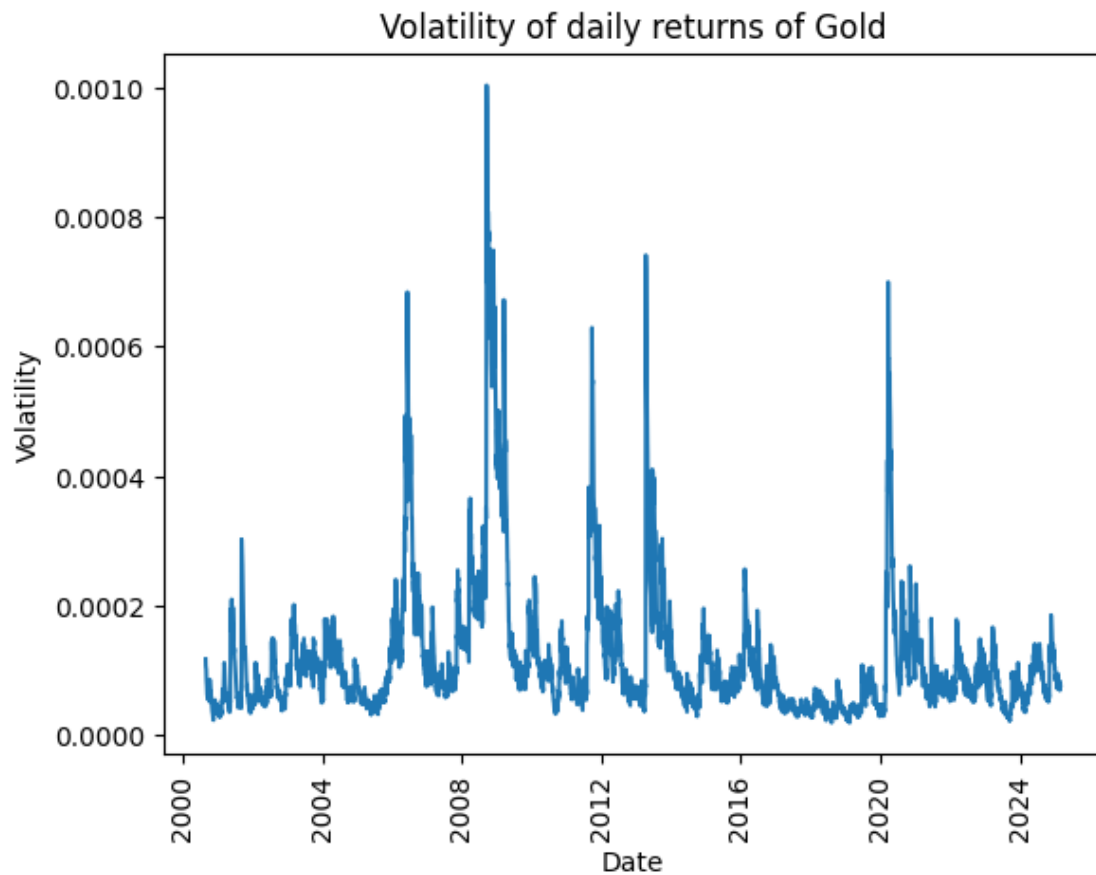


1.4 Riskmetrics

```
[29]: 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
```

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

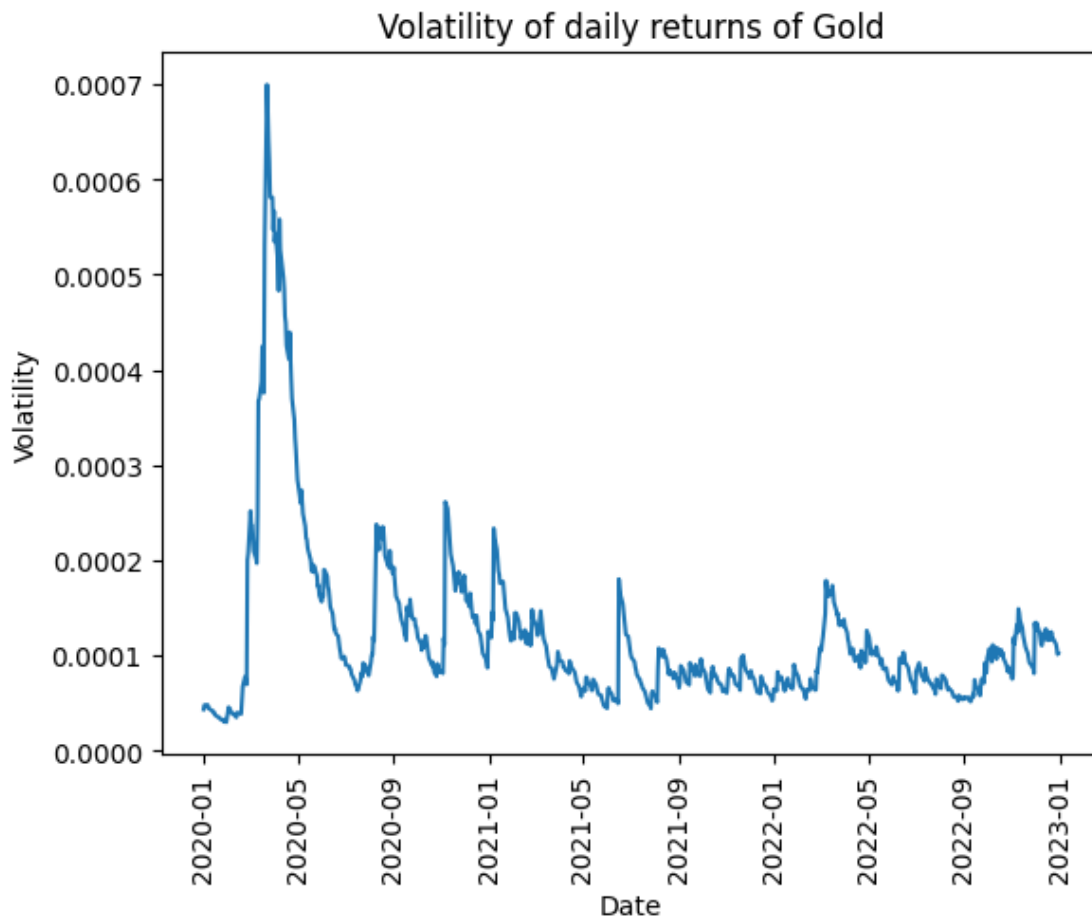
```
[31]: plt.plot(dates[:-1],volatility)
        # rotate x-tick-labels by 90°
        plt.tick_params(axis='x',rotation=90)
        plt.xlabel('Date')
        plt.ylabel('Volatility')
        plt.title('Volatility of daily returns of Gold')
        plt.show()
```



```
[32]: 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 Gold')
plt.show()
```



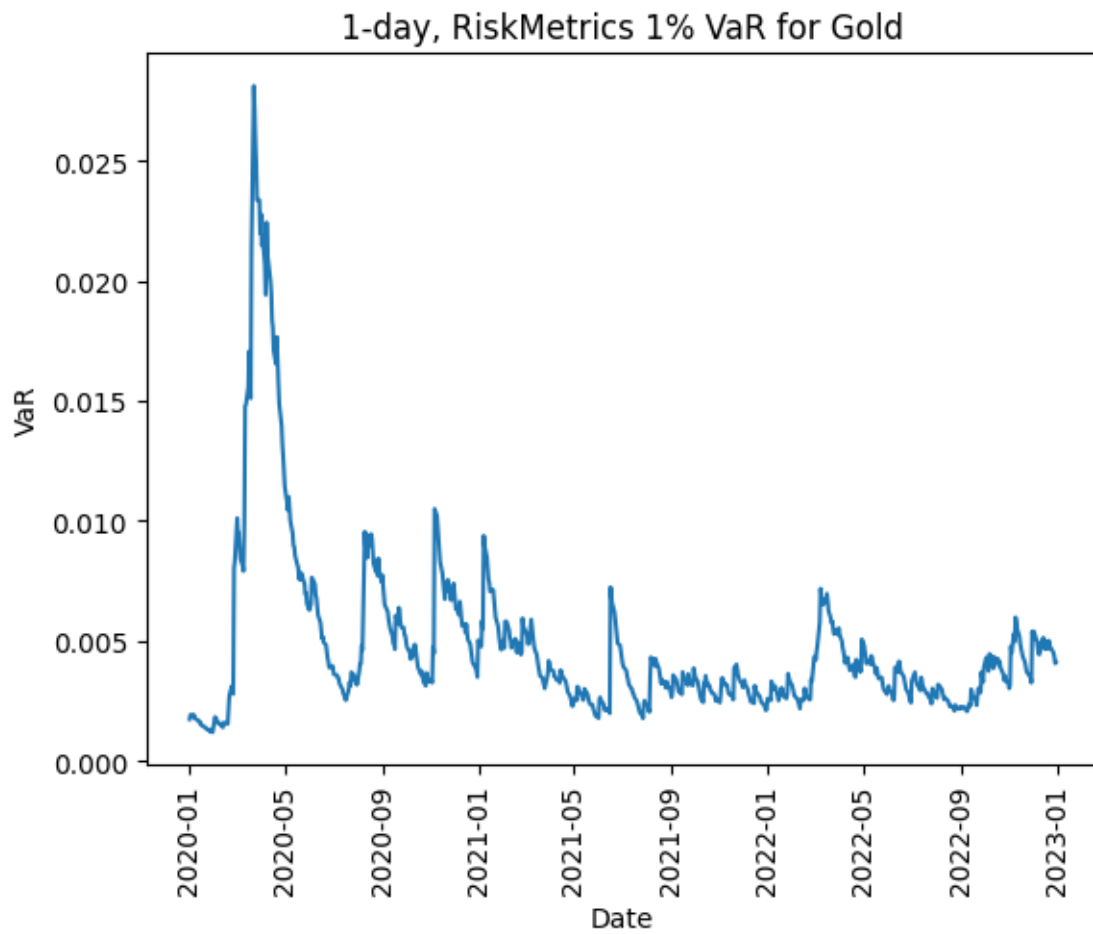
1.5 VaR

Assuming gaussian distribution of portfolio returns

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

0.024880326653042774

```
[36]: 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 for Gold')
      plt.show()
```



[]: