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Problem 2.3
In [9]: import pandas as pd
         import numpy as np
In [10]: df = pd.read_excel('Crude Oil Data-2.xlsx')
         df = df.set_index(df['Unnamed: 1'])
         df = df.drop(columns = ['Unnamed: 0', 'Unnamed: 1'])
         df.index.names = ['Date']
         df.head()
Out[10]:
                    Crude Oil
               Date
         2006-02-13
                        64.01
         2006-02-14
                       63.28
         2006-02-15
                        62.17
         2006-02-16
                       63.04
         2006-02-17
                        63.92
         a)
         two_day_std = np.std(df['Crude Oil'].diff())*np.sqrt(2)
         round(two_day_std, 2)
         2.23
Out[11]:
         The estimate for the two-day standard deviation is $2.23.
         b)
         This problem is asking about the 99% VaR. To do this, we calculate -2.33*\sigma*\sqrt{T}
         VaR
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In [12]: VaR = -2.33 * 2.23
         -5.1959
Out[12]:
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Conclusion: So, in a worst case scenario (1% probability), crude oil may fall \$5.19 in two days.

The exchange should set the maintenance margin to \$5,190, as the contract size is 1000 barrels.

C)

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In [13]: two_day_returns = df['Crude Oil'].diff(periods = 2)
          np.sum(two_day_returns < -5.19)</pre>
Out[13]:
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Out of this data, we can see that there are 24 instances of crude oil dropping over \$5.19 in two days.

This is inconsistent with our assumption of a 1% chance. The data has 1040 observations and therefore this represents 2.3% rather than 1%.

d)

In [20]: maintenance margin = 5190

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initial_margin = maintenance_margin / 0.75
In [21]: contract_values = df['Crude Oil'] * 1000
In [25]: # Define the maintenance margin and calculate the initial margin
         maintenance margin = 5190
         initial_margin = maintenance_margin / 0.75
         # Create a list of contract values (crude oil prices * 1000)
         contract_values = [0, 500, -1500, -1000, 300, -500, -2200, -1700, -200, -300]
         # Initialize lists to store the table data
         initial_margin_list = [initial_margin]
         starting_margin_balance_list = [initial_margin]
         change_in_contract_value_list = []
         balance_after_change_in_value_list = []
         deposit if margin call list = []
         withdrawal_if_excess_list = []
         true_margin_balance_list = []
         default_margin_balance_list = []
         # Calculate the table data based on the contract values
         for value in contract values:
              change_in_contract_value_list.append(value)
             balance_after_change_in_value = starting_margin_balance_list[-1] + value
             balance_after_change_in_value_list.append(balance_after_change_in_value)
             if balance_after_change_in_value < maintenance_margin:</pre>
                  deposit = maintenance_margin - balance_after_change_in_value
                  withdrawal = 0
              else:
                  deposit = 0
                  withdrawal = balance_after_change_in_value - maintenance_margin
              deposit if margin call list.append(deposit)
             withdrawal_if_excess_list.append(withdrawal)
              true_margin_balance = initial_margin_list[-1] + sum(withdrawal_if_excess_list) - sum(deposit_if_margin_call_list)
              true margin balance list.append(true margin balance)
              default_margin_balance = true_margin_balance + value
              default_margin_balance_list.append(default_margin_balance)
              # Update the starting margin balance for the next iteration
              starting margin balance list.append(balance after change in value)
         # Create a DataFrame to display the table
         data = {
              'Initial Margin': initial_margin_list,
              'Starting Margin Balance': starting margin balance list[:-1],
              'Change in Contract Value': change in contract value list,
              'Balance after Change in Value': balance_after_change_in_value_list,
              'Deposit if Margin Call': deposit_if_margin_call_list,
              'Withdrawal if Excess': withdrawal_if_excess_list,
              'True Margin Balance': true margin balance list,
              'Default Margin Balance': default margin balance list,
              'Formula for Default Margin Balance': [''] * len(contract_values)
         df = pd.DataFrame(data)
         np.sum(df['Default Margin Balance'] > 0)
         13
Out[25]:
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After replicating the chart in Python, we counted 13 defaults.

3.2

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b)
In [1]: rf = 0.004
        div = 0.015
        index_val = 4450
        beta = 0.6
        index_val_2m = [3500, 4000, 5000, 5500]
In [3]: for price in index_val_2m:
            excess_market = (price-index_val)/index_val - rf + div
            excess port = excess market * beta
            exp_port = excess_port + rf
            print(f"With the index at {price}, the expected return of the portfolio is {round(exp_port*100, 2)}%")
        With the index at 3500, the expected return of the portfolio is -11.75%
        With the index at 4000, the expected return of the portfolio is -5.01%
        With the index at 5000, the expected return of the portfolio is 8.48%
        With the index at 5500, the expected return of the portfolio is 15.22%
```