```
In [17]: !pip install pandas
          !pip install numpy
         !pip install matplotlib
        Requirement already satisfied: pandas in c:\users\aditya sakpal\anaconda3\lib\site-packages (2.2.2)
        Requirement already satisfied: numpy>=1.26.0 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from pandas) (1.26.4)
        Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from pandas) (2.9.
        0.post0)
        Requirement already satisfied: pytz>=2020.1 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from pandas) (2024.1)
        Requirement already satisfied: tzdata>=2022.7 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from pandas) (2023.3)
        Requirement already satisfied: six>=1.5 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from python-dateutil>=2.8.2->pan
        das) (1.16.0)
        Requirement already satisfied: numpy in c:\users\aditya sakpal\anaconda3\lib\site-packages (1.26.4)
        Requirement already satisfied: matplotlib in c:\users\aditya sakpal\anaconda3\lib\site-packages (3.8.4)
        Requirement already satisfied: contourpy>=1.0.1 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (1.2.0)
        Requirement already satisfied: cycler>=0.10 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (0.11.0)
        Requirement already satisfied: fonttools>=4.22.0 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (4.51.
        0)
        Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (1.4.
        4)
        Requirement already satisfied: numpy>=1.21 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (1.26.4)
        Requirement already satisfied: packaging>=20.0 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (23.2)
        Requirement already satisfied: pillow>=8 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (10.3.0)
        Requirement already satisfied: pyparsing>=2.3.1 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (3.0.9)
        Requirement already satisfied: python-dateutil>=2.7 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from matplotlib) (2.
        9.0.post0)
        Requirement already satisfied: six>=1.5 in c:\users\aditya sakpal\anaconda3\lib\site-packages (from python-dateutil>=2.7->matpl
        otlib) (1.16.0)
In [29]: import vfinance as vf
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         from pypfopt.efficient frontier import EfficientFrontier
         from pypfopt.expected returns import mean historical return
         from pypfopt.risk models import CovarianceShrinkage
         from pypfopt.plotting import plot efficient frontier
```

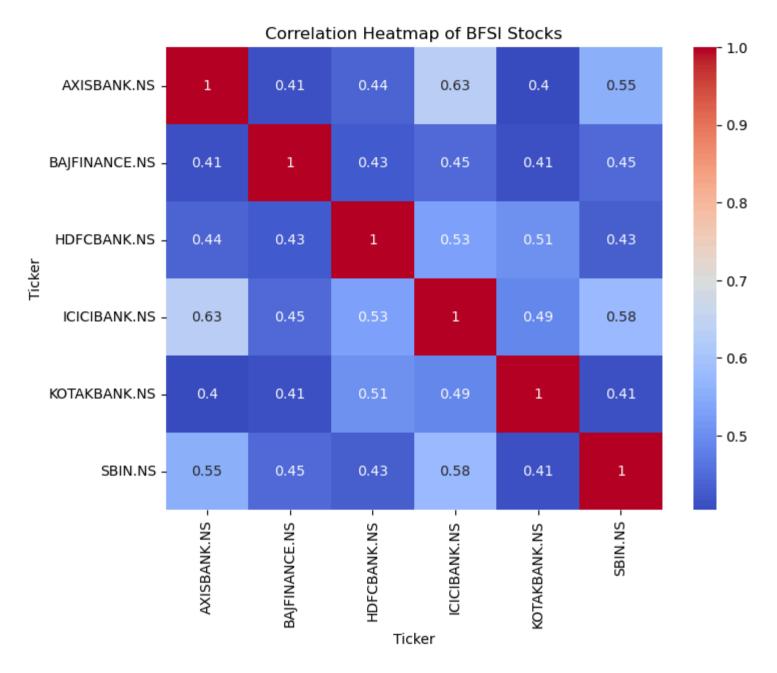
```
tickers = ['HDFCBANK.NS', 'ICICIBANK.NS', 'KOTAKBANK.NS', 'SBIN.NS', 'AXISBANK.NS', 'BAJFINANCE.NS']
data = yf.download(tickers, start="2021-01-01", end="2025-07-23", auto adjust=True)['Close']
returns = data.pct change().dropna()
plt.figure(figsize=(8,6))
sns.heatmap(returns.corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Heatmap of BFSI Stocks")
plt.show()
data.plot(figsize=(12,6))
plt.title("Stock Price Trends (2021-2025)")
plt.ylabel("Price (INR)")
plt.xlabel("Date")
plt.show()
mean return = returns.mean()
volatility = returns.std()
sharpe = mean return / volatility
sharpe df = pd.DataFrame({
    'Mean Return': mean return,
    'Volatility': volatility,
    'Sharpe Ratio': sharpe
})
print("\n Sharpe Ratio for Individual Stocks:\n")
print(sharpe df)
mu = mean historical return(data)
S = CovarianceShrinkage(data).ledoit wolf()
ef = EfficientFrontier(mu, S)
weights = ef.max sharpe()
cleaned weights = ef.clean weights()
print("\n Optimized Portfolio Allocation (Max Sharpe Ratio):")
for stock, weight in cleaned weights.items():
    print(f"{stock}: {weight:.2%}")
plt.figure(figsize=(8,6))
plt.pie(cleaned weights.values(), labels=cleaned weights.keys(), autopct='%1.1f%%', startangle=140)
```

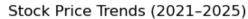
```
plt.title("Optimal Portfolio Allocation")
plt.axis('equal')
plt.show()

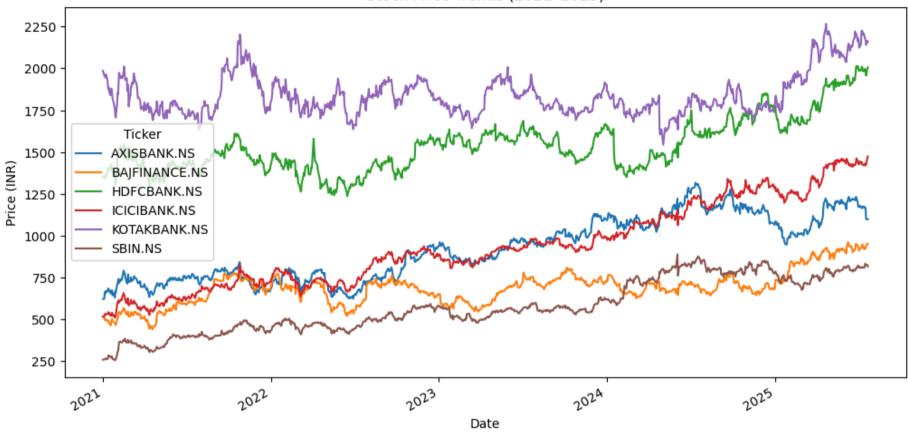
ef_plot = EfficientFrontier(mu, S)
plot_efficient_frontier(ef_plot, show_assets=True)
plt.title("Efficient Frontier (Risk vs Return)")
plt.show()

expected_return, risk, sharpe_ratio = ef.portfolio_performance(verbose=True)
print(f"\n Optimized Portfolio Sharpe Ratio: {sharpe_ratio:.2f}")
```

[********** 6 of 6 completed







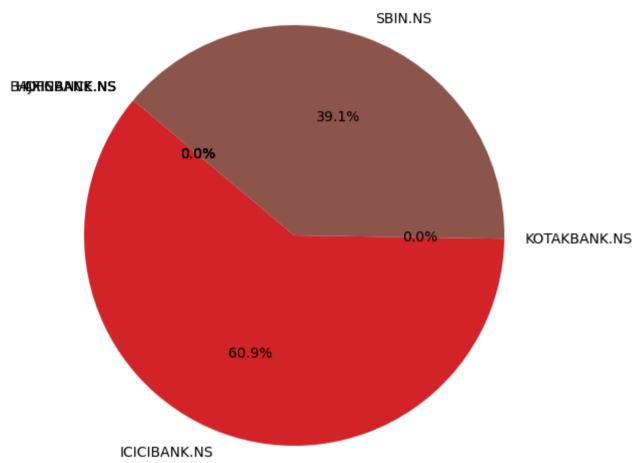
Sharpe Ratio for Individual Stocks:

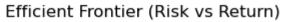
	Mean Return	Volatility	Sharpe Ratio
Ticker			
AXISBANK.NS	0.000638	0.016237	0.039293
BAJFINANCE.NS	0.000707	0.018220	0.038812
HDFCBANK.NS	0.000446	0.013853	0.032222
ICICIBANK.NS	0.001038	0.014344	0.072396
KOTAKBANK.NS	0.000186	0.014880	0.012473
SBIN.NS	0.001170	0.017117	0.068355

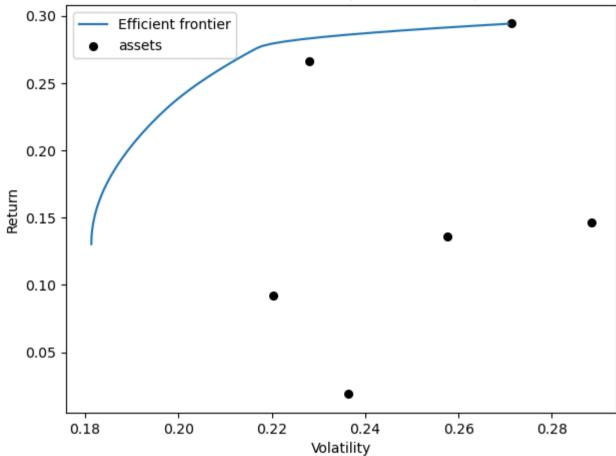
Optimized Portfolio Allocation (Max Sharpe Ratio):

AXISBANK.NS: 0.00%
BAJFINANCE.NS: 0.00%
HDFCBANK.NS: 0.00%
ICICIBANK.NS: 60.87%
KOTAKBANK.NS: 0.00%
SBIN.NS: 39.13%









Expected annual return: 27.7% Annual volatility: 21.8%

Sharpe Ratio: 1.27

Optimized Portfolio Sharpe Ratio: 1.27