Problem 1:

1. Arithmetic Returns (Last 5 Rows):

SPY AAPL EQIX

Date

2024-12-27 -0.011492 -0.014678 -0.006966

2024-12-30 -0.012377 -0.014699 -0.008064

2024-12-31 -0.004603 -0.008493 0.006512

2025-01-02 -0.003422 -0.027671 0.000497

2025-01-03 0.011538 -0.003445 0.015745

SPY 0.008077

AAPL 0.013483

EQIX 0.015361

2. Log Returns (Last 5 Rows):

SPY AAPL EQIX

Date

2024-12-27 -0.011515 -0.014675 -0.006867

2024-12-30 -0.012410 -0.014696 -0.007972

2024-12-31 -0.004577 -0.008427 0.006602

2025-01-02 -0.003392 -0.027930 0.000613

2025-01-03 0.011494 -0.003356 0.015725

SPY 0.008078

AAPL 0.013446

EQIX 0.015270

Problem 2:

- 1. Current Value of Portfolio given today is 1/3/2025: \$251862.50
- 2. VAR and ES
 - a. Method 2a (Normal EW Covariance):
 - b. VaR 5%: \$3886.02
 - c. ES 5%: \$4873.23
 - d. Method 2b (Copula):
 - e. VaR (T-distribution with Gaussian Copula, 5%): \$4379.51
 - f. ES (T-distribution with Gaussian Copula, 5%): \$6123.02
 - g. Method 2c (Historical):
 - h. VaR (Historical, 5%): \$4575.03
 - i. ES (Historical, 5%): \$6059.39
- 3. The three VaR and ES estimation methods differ in their assumptions and risk sensitivity. Normal EW Covariance assumes normally distributed returns, making it computationally efficient but prone to underestimating extreme losses, resulting in the lowest VaR (\$3,886.02) and ES (\$4,873.23). T-distribution with Gaussian Copula captures fat tails and dependencies better, producing a higher VaR (\$4,379.51) and the highest ES (\$6,123.02), but is more complex. Historical Simulation relies on past market data, providing the highest VaR (\$4,575.03) and ES (\$6,059.39), though it assumes historical patterns persist. For accurate risk assessment, a combination of these methods is ideal.

Problem 3:

1. Implied Volatility: 0.3351

Delta: 0.6659, Vega: 5.6407, Theta: -5.5446
Estimated Option Price Change for 1% Volatility Increase: \$0.0564

3. Put Price (BSM Model): \$1.2593 Put-Call Parity Difference: 0.0000; Given that the Put-Call Parity Difference is 0.0000, it confirms that the parity relationship holds exactly, as expected in a theoretically efficient market. This result indicates that the calculated put price aligns perfectly with the put-call parity condition, suggesting no arbitrage opportunities.

4. Delta-Normal Approximation (20-day, 5% level) [Standard ES]:

VaR = 4.2717

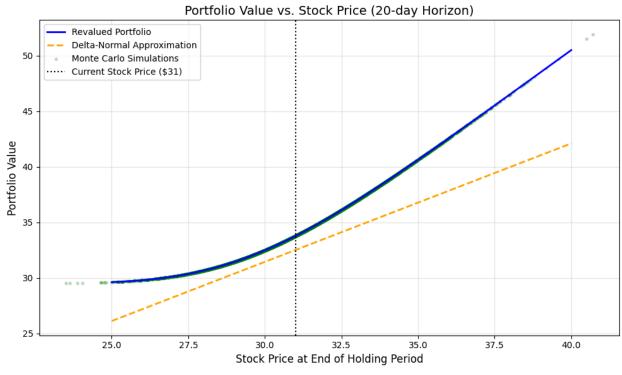
ES (Standard) = 5.2383

5. Monte Carlo Simulation (20-day, 5% level):

VaR = 2.5964

ES = 2.9147

6.



7. The Delta-Normal method assumes a normal distribution of returns and linear sensitivity to stock price changes, which simplifies risk estimation but may underestimate risk in highly volatile or non-linear scenarios. The Monte Carlo simulation, in contrast, captures the full range of potential portfolio values by simulating thousands of possible price paths, reflecting non-linearity and tail risks more effectively. This difference is evident in the results: the Delta-Normal method estimates a higher Value at Risk (VaR) and Expected Shortfall (ES), indicating that it assumes a greater potential downside due to its reliance on a normal distribution. However, the Monte Carlo method, by incorporating non-linearity, provides a more accurate measure of extreme events, showing lower VaR and ES values. The portfolio value graph further illustrates this: while the Delta-Normal

approximation follows a straight-line assumption, the revalued portfolio from Monte Carlo simulations displays a more realistic curvature, accounting for convexity in option pricing. Thus, while Delta-Normal is computationally efficient, Monte Carlo provides a more robust risk estimation, particularly for portfolios with significant option exposure.