

Aim: To optimize trading strategies using DRL, **NOT** price prediction or portfolio management.

Research Objectives:

- I. To improve the performance of stock trading models by applying feature engineering techniques to obtain technical indicators used by DRL models.
- II. To improve the performance of stock trading models by using DRL during normal stock market conditions in simulated trading environments with performance metrics such as Cumulative Return and Sharpe Ratio.
- III. To improve the performance of stock trading models by using DRL during bearish and volatile stock market conditions in simulated trading environment with Maximum Drawdown as a performance metric.

Datasets Used: AMZN, MSFT, GOOGL historical stock datasets of 10 years (2013-2023).

Parts Already Done: Data Cleaning, Preprocessing, Feature Engineering, EDA, Feature Selection and Recursive Feature Elimination.

Data Partitioning: Sequential Train-Test Split (80:20)

Model Construction:

1. Trading Environment

Components	Description
State Space (S)	Open, High, Low, Close, SMA, EMA, MACD
Action Space (A)	Sell (0), Hold (1), Buy (2)
Reward Function (R)	$\frac{\text{Selling Price} - \text{Buying Price}}{\text{Buying Price}} \times 100$
Transfer Probability (T)	Implicitly learned by the DQN model.
Discount Factor (γ)	0.95

2. DRL Model: **DQN with LSTM** neural network architecture.

3. Model Training.

Results Analysis:

1. Cumulative Return

$$\text{Cumulative Return, } CR = \frac{\text{Final Value} - \text{Initial Value}}{\text{Initial Value}} \times 100 \quad (3,8)$$

where the initial value is the initial value of the portfolio, and final value is the final value of the portfolio (Yang H et al., 2020).

2. Sharpe Ratio

$$\text{Sharpe Ratio, } SR = \frac{R_p - R_f}{\sigma_p} \quad (3,9)$$

where R_p is the return, R_f is the risk-free rate, and σ_p is the standard deviation of excess returns (Yuan Y et al., 2020).

3. Maximum Drawdown

$$\text{Maximum Drawdown, } MDD = \frac{\text{Peak Value} - \text{Valley Value}}{\text{Peak Value}} \times 100 \quad (3,10)$$

where the peak value is the highest value before the largest drop and valley value is the value before a new high value is established (Carta S et al., 2021; Yu X et al., 2023).

Hyperparameter Optimization: Bayesian Optimization.

If the first part of the code (Data preprocessing feature engineering EDA etc) is required for context and understanding, please let me know. I will send it to you.