# Semi-Decision-Focused Learning with Deep Ensembles: A Practical Framework for Robust Portfolio Optimization

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## A. Summary of the proposed framework

#### Motivation

In portfolio optimization, **stable** performance is important when employing the strategy at real investments. However, previous deep learning-based portfolio optimization strategies exhibit instability due to non-convex loss and random initialization.

#### Contribution

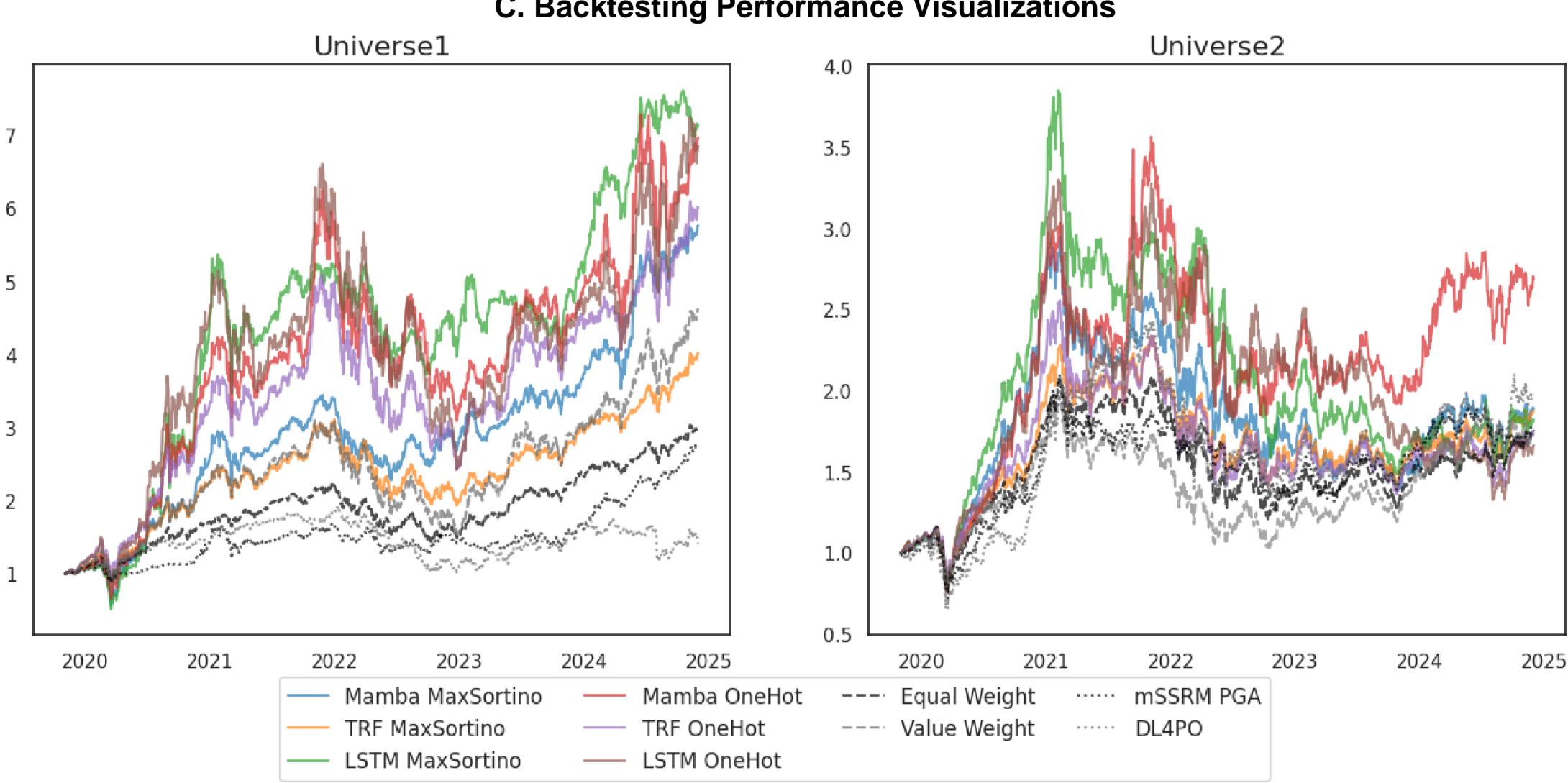
- I propose **Semi-Decision-Focused Learning**, which reformulate portfolio optimization problem as supervised learning, using cross-entropy loss as a loss function and hypothetically optimal portfolio as a target.
- I also apply **Deep Ensemble** by averaging multiple prediction from independently-trained models to make the output portfolio stable and robust.

## **B.** Backtesting Performances

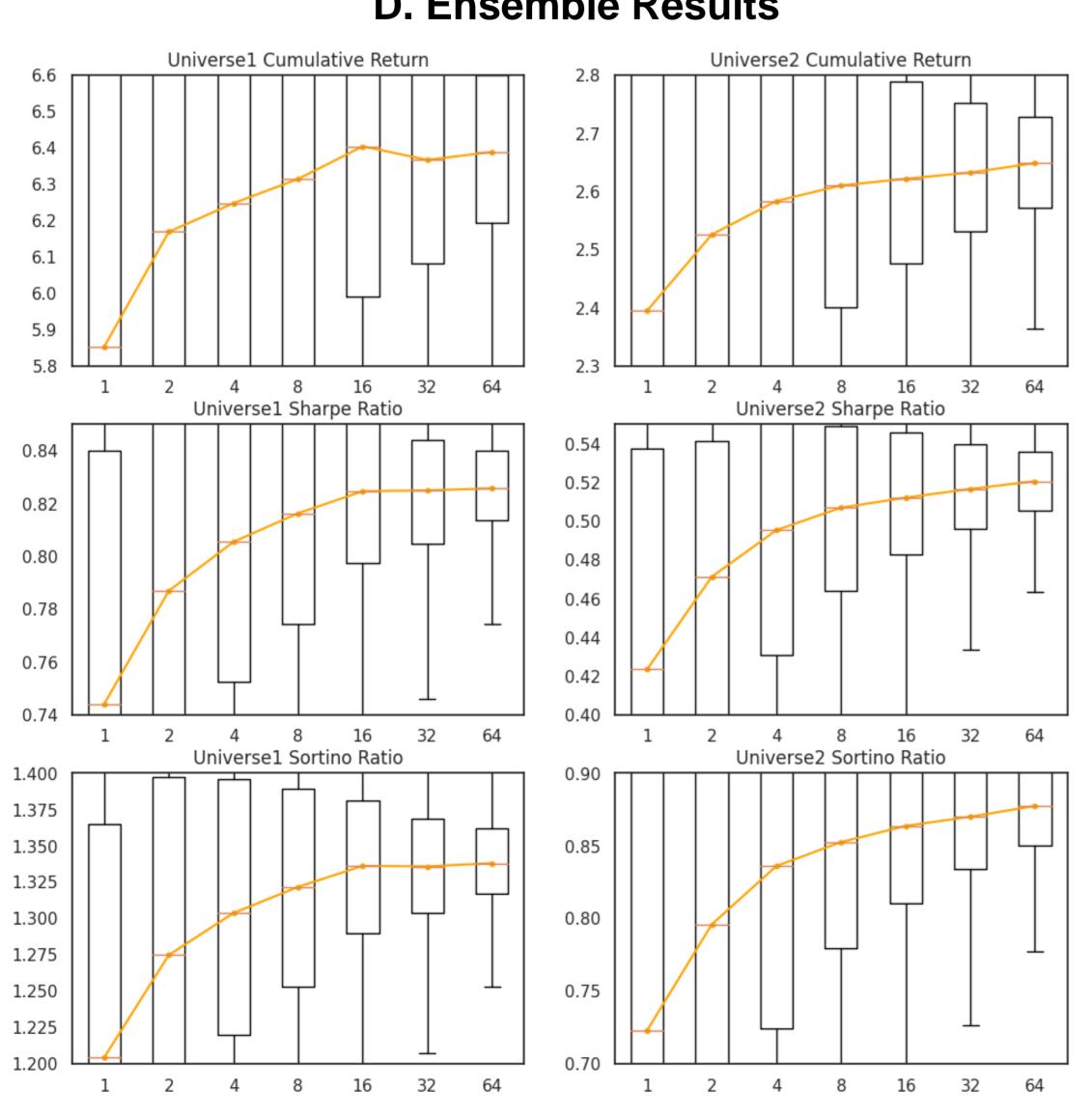
|                   | Universe 1 |       |       | Universe 2 |       |       |
|-------------------|------------|-------|-------|------------|-------|-------|
|                   | CR         | SHR   | SOR   | CR         | SHR   | SOR   |
| Baseline          |            |       |       |            |       |       |
| Equal Weight      | 3.014      | 0.884 | 1.391 | 1.749      | 0.416 | 0.653 |
| Value Weight      | 4.462      | 0.860 | 1.362 | 1.728      | 0.354 | 0.566 |
| mSSRM PGA         | 2.799      | 0.890 | 1.435 | 1.777      | 0.436 | 0.687 |
| DL4PO             | 1.408      | 0.219 | 0.341 | 1.976      | 0.388 | 0.642 |
| My Method         |            |       |       |            |       |       |
| Mamba Max-Sortino | 5.760      | 1.007 | 1.577 | 1.887      | 0.397 | 0.646 |
| TRF Max-Sortino   | 4.013      | 0.981 | 1.546 | 1.870      | 0.447 | 0.713 |
| LSTM Max-Sortino  | 7.123      | 0.938 | 1.510 | 1.817      | 0.352 | 0.574 |
| Mamba One-hot     | 6.957      | 0.868 | 1.410 | 2.685      | 0.530 | 0.894 |
| TRF One-hot       | 6.007      | 0.953 | 1.533 | 1.751      | 0.367 | 0.587 |
| LSTM One-hot      | 6.848      | 0.755 | 1.237 | 1.659      | 0.257 | 0.430 |

**CR**: Cumulative Return **SHR**: Sharpe Ratio **SOR**: Sortino Ratio

## C. Backtesting Performance Visualizations



#### D. Ensemble Results



#### **E.** Conclusion

- I proposed a novel portfolio optimization method grounded in Decision-Focused Learning framework.
- In experiments on two distinct investment universes, one trending generally upward and other range-bounded, my approach consistently outperforms various portfolio optimization baselines, indicating its robustness and superior performance across different market conditions.
- To further enhance performance, I incorporate the Deep Ensemble method. My experiments show that increasing the ensemble size consistently yields better portfolio optimization results in the quality of the portfolio weight.

### F. Code



https://github.com/sDFLwDE/sDFLwDE