

Task Progression for Strategy Development using backtesting.py

Phase 1: Simple Strategy Development and Backtesting

1. Task 1: Build a Simple Strategy

- Define a basic strategy class using backtesting.py (e.g., moving average crossover).
- Implement the init method for setting up indicators.
- Implement the next method for trade logic (e.g., buy when a short-term SMA crosses above a long-term SMA).

2. Task 2: Backtest the Simple Strategy

- Load historical data for a single symbol (e.g., GOOG).
- Run the backtest using the Backtest class.
- Analyze key performance metrics (e.g., returns, drawdown, win rate).

3. Task 3: Optimize the Simple Strategy

- Define optimization parameters (e.g., SMA periods).
- Use the optimize method to find the best-performing parameters.
- Evaluate the results to ensure parameters are robust.

4. Task 4: Conduct Walk-Forward Optimization (WFO)

- Set up in-sample and out-of-sample periods for WFO.
- Run WFO to validate strategy robustness across different periods.
- Review performance consistency and adjust strategy if necessary.

Phase 2: Multi-Symbol Strategy Development

5. Task 5: Extend Strategy to Multiple Symbols

- Modify the strategy class to handle multiple symbols.
- Update the init and next methods to incorporate data and signals from multiple symbols (e.g., SPY, AAPL, GOOG).
- Implement logic for trading decisions based on conditions across multiple symbols.

6. Task 6: Backtest with Multiple Symbols

- Load historical data for multiple symbols.
- Adjust the Backtest setup to handle multiple datasets simultaneously.
- Execute the backtest and analyze performance for each symbol and the overall strategy.

7. Task 7: Optimize Multi-Symbol Strategy

- Define optimization parameters relevant to each symbol (e.g., different SMAs for each symbol).
- Run optimization to identify the best parameter combinations for trading multiple symbols.
- Evaluate cross-symbol performance to ensure the strategy performs well across all symbols.

8. Task 8: Conduct Walk-Forward Optimization (WFO) with Multiple Symbols

- Divide historical data of multiple symbols into sequential in-sample and out-of-sample periods.
- Run WFO to optimize parameters in each in-sample period and test in the out-of-sample period.
- Review WFO results for consistency and robustness across multiple symbols.

Phase 3: Multi-Timeframe, Multi-Symbol Strategy Development

9. Task 9: Incorporate Multiple Timeframes into Strategy

- Modify the strategy class to use data from multiple timeframes (e.g., 5-minute, 1-hour, daily).
- Update the init method to set up indicators across different timeframes.
- In the next method, use signals from multiple timeframes to make trading decisions (e.g., confirm a trend on a higher timeframe before entering a trade on a lower timeframe).

10. Task 10: Backtest Multi-Timeframe, Multi-Symbol Strategy

- Load historical data for multiple symbols and timeframes.
- Adjust the Backtest setup to integrate data across symbols and timeframes.
- Run the backtest, analyzing the combined performance across symbols and timeframes.

11. Task 11: Optimize Multi-Timeframe, Multi-Symbol Strategy

- Define optimization parameters for each symbol and timeframe (e.g., SMA periods, entry thresholds).
- Run optimization with a focus on finding robust parameters that work across symbols and timeframes.
- Evaluate the combined performance and robustness of the strategy.

12. Task 12: Conduct Walk-Forward Optimization (WFO) with Multi-Timeframe, Multi-Symbol Data

- Set up WFO by dividing the multi-symbol, multi-timeframe data into in-sample and out-of-sample periods.

- Perform WFO to optimize in-sample and test out-of-sample for each timeframe and symbol.
- Review WFO results to ensure the strategy maintains performance consistency across all symbols and timeframes.

This progression guides you from simple strategy development to handling complex multi-symbol, multi-timeframe scenarios using `backtesting.py`. Let me know if you'd like further details or adjustments!