

# Methodology and Framework

## 1. Data Pipeline

### What We Are Trying

A Smart Order Router (SOR) relies on accurate, high-frequency market data to make informed decisions. The data pipeline ensures this data is processed, cleaned, and ready for use in backtesting simulations. This includes market data (bid/ask prices, trades) and venue-specific order book details.

### Key Considerations

- **Data Sourcing:**
  - **Historical Market Data:** Sources like Alpha Vantage or IEX Cloud can provide historical price and volume data.
  - **Synthetic Data:** In the absence of real data, we generate synthetic market data resembling real-world scenarios.
- **Data Challenges:**
  - **Missing Data:** Gaps in price or volume data due to system failures or API issues can impact simulations. Solutions include interpolation or carrying forward the last known value.
  - **Synchronization:** Bid/ask updates and trade execution logs may arrive out of sync. Use event timestamps to realign them.

### Implementation Approach

- Design an ETL (Extract, Transform, Load) pipeline to handle raw market data.
- Normalize all inputs (e.g., scaling prices, volumes).
- Generate synthetic data for testing scalability and robustness.

## 2. Execution Strategies

### What We Are Trying

Execution strategies dictate how orders are broken up and executed in the market. The aim is to minimize costs and slippage while maximizing fill rates.

### Simplified Execution Strategies

- **TWAP (Time-Weighted Average Price):**
  - Divide the total order into equal parts executed at regular time intervals.
  - Suitable for stable markets with minimal volatility.
  - **Example:** If an order of 1,000 shares needs execution over 1 hour, execute 250 shares every 15 minutes.
- **VWAP (Volume-Weighted Average Price):**
  - Trades are allocated proportionally based on historical or real-time market volume.

- Adaptive to market liquidity changes but computationally heavier.
- **Example:** If 30% of the day's trading volume happens in the first hour, execute 30% of the order in that hour.

#### Market Conditions

- **Volatility:**
  - High volatility increases execution risk and slippage.
  - VWAP performs better in such scenarios as it adapts to volume spikes.
- **Liquidity:**
  - TWAP assumes constant liquidity, making it less effective in illiquid markets.

### 3. Performance Metrics

#### What We Are Trying

Performance metrics evaluate the success of execution strategies and their impact on the order's final outcome.

#### Key Metrics

- **Execution Cost:**
  - The difference between the executed price and a benchmark price (e.g., VWAP).
  - **Example:** If the executed price is \$101 and VWAP is \$100, the execution cost is \$1.
- **Slippage:**
  - The difference between the expected and actual execution price.
  - **Example:** If the expected execution price was \$99 but the actual was \$101, the slippage is \$2.
- **Fill Rate:**
  - Percentage of orders successfully executed versus total orders placed.
  - **Example:** If 950 out of 1,000 shares are filled, the fill rate is 95%.

#### Implementation Approach

Track metrics in real-time during simulations and output them as part of the post-simulation analysis.

### 4. Simulation Logic

#### What We Are Trying

Simulate a realistic market environment to test the performance of SOR strategies under controlled conditions.

#### Key Components

- **Multi-Venue Routing:**
  - Orders are routed to multiple trading venues based on latency, liquidity, and fees.

- **Example:** If Venue A has lower fees but higher latency compared to Venue B, split the order between them based on trade-offs.
- **Order Placement:**
  - Orders can be market orders, limit orders, or complex multi-leg orders.
  - **Example:** A multi-leg trade might involve selling stock in Venue A while simultaneously buying in Venue B to hedge.
- **Order Modifications:**
  - Simulate the impact of canceling or modifying orders due to market conditions or partial fills.

### Implementation Approach

Use an event-driven simulation engine to mimic market conditions and track the lifecycle of orders.

## 5. Extensibility and Scalability

### What We Are Trying

Design a framework that can adapt to new strategies, handle increased data volumes, and support various asset classes.

#### Extensibility

- Add support for:
  - **Advanced Strategies:**
    - Machine learning (e.g., reinforcement learning for smart routing).
    - Strategies like dynamic VWAP or adaptive TWAP.
  - **New Asset Classes:**
    - Expand to derivatives, bonds, cryptocurrencies, and forex.

#### Scalability

- **High-Performance Computing:**
  - Parallel processing for large-scale simulations.
  - Cloud-based architectures for distributed simulations.
- **Data Volume:**
  - Efficient data structures and storage solutions (e.g., Apache Parquet for market data).

### Implementation Approach

- Modularize the codebase to allow easy integration of new components.
- Use scalable frameworks like Ray for distributed simulations.

This detailed framework outlines the end-to-end design of a backtesting system for SOR. Each component aligns with real-world trading challenges and is scalable for future enhancements.