Documentation for Deribit Trading system

1. Benchmarking -

Benchmarked the following metrics -

- Order Placement Latency: Time taken from placing an order to confirmation.
- Market Data Processing Latency: Time taken to process incoming market data updates.
- **WebSocket Message Propagation Delay:** Time for messages to travel between the server and client.
- End-to-End Trading Loop Latency: Total time from market data input to order execution.

<u>Methodology used</u> -

1. Order Placement Latency

- Function: BenchmarkPlaceOrder
- Process:
 - Redirects standard input/output to simulate user input for placing an order.
 - Calls trading_system.placeOrder() in a loop to measure the average latency of order placement.
- Purpose: Evaluate the time it takes for the system to process and confirm an order.

2. Market Data Processing Latency

- Function: BenchmarkMarketData
- Process:
 - Simulates market data subscription using mocked input.
 - Measures the time taken to fetch and process market data for the subscribed channel.
 - o Unsubscribes after the benchmark.
- Purpose: Identify delays in receiving and processing market data from the exchange.

3. WebSocket Message Propagation Delay

- Function: BenchmarkWebSocketPropagation
- Process:
 - Sends a test message (public/test) over WebSocket and measures the time to receive a response.
- **Purpose:** Quantify the latency introduced by WebSocket communication, including transmission and server-side processing.

4. End-to-End Trading Loop Latency

- Function: BenchmarkFullTradingLoop
- Process:
 - Simulates a complete trading loop:
 - 1. Places a buy order.
 - 2. Immediately places a sell order in the same iteration.
 - Redirects I/O streams to simulate user inputs for each action.
- Purpose: Measure the overall latency from decision-making to order placement and execution.

Results of Benchmarking:

Before Optimization (memory optimization)-

```
Run on (8 X 2095.99 MHz CPU s)

CPU Caches:
   L1 Data 32 KiB (x4)
   L1 Instruction 64 KiB (x4)
   L2 Unified 512 KiB (x4)
   L3 Unified 4096 KiB (x1)
Load Average: 0.03, 0.04, 0.08

***WARNING*** Library was built as DEBUG. Timings may be affected.

Benchmark

Time CPU Iterations

BenchmarkPlaceOrder 559718923 ns 657060 ns 10
BenchmarkMarketData 277963331 ns 237530 ns 10
BenchmarkWebSocketPropagation 290554859 ns 115650 ns 10
BenchmarkFullTradingLoop 1176723191 ns 1434330 ns 10
```

After Optimization-

```
Run on (8 X 2095.99 MHz CPU s)

CPU Caches:

L1 Data 32 KiB (x4)

L1 Instruction 64 KiB (x4)

L2 Unified 512 KiB (x4)

L3 Unified 4096 KiB (x1)

Load Average: 0.01, 0.03, 0.07

***WARNING*** Library was built as DEBUG. Timings may be affected.

Benchmark

Time CPU Iterations

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BenchmarkPlaceOrder 274594745 ns 323780 ns 10

BenchmarkMarketData 48360925 ns 155782 ns 100

BenchmarkWebSocketPropagation 274128875 ns 115630 ns 10

BenchmarkFullTradingLoop 549760876 ns 556130 ns 10
```

2. Bottlenecks identified -

Thread Management Overhead:

- Creating and managing dedicated threads for each WebSocket connection may introduce significant overhead.
- This could lead to scalability issues as the number of connections grows, resulting in high CPU usage and potential memory exhaustion.

Memory Usage and Management:

- Storing all subscriptions in a centralized data structure with a fixed length could potentially lead to inefficient memory usage if the number of subscriptions exceeds the predefined limit.
- Memory fragmentation and potential memory leaks may occur over time if memory management is not handled carefully.

Market Data Processing Delays:

- While market data processing is handled in a separate thread, depending on the complexity of the data, processing delays can still impact overall performance.
- A large volume of data or slow processing logic could become a bottleneck, affecting real-time responsiveness.

Single WebSocket Connection Per Thread:

- Dedication of one thread per WebSocket connection can lead to inefficiency, particularly in scenarios with a large number of concurrent connections.
- Managing threads for each connection can lead to increased context switching and poor resource utilization.

WebSocket Data Handling Overhead:

- Even though WebSockets reduce the TCP handshake overhead, frequent data exchanges with high-frequency messages could lead to significant memory and CPU load if not managed efficiently.
- Managing large volumes of messages in real time without introducing delays or buffering issues remains a challenge.

3. Optimization Study -

These were the practices implemented for optimising the overall latency of the system and optimise the memory used

Centralized Data Structure with Fixed Length for Subscriptions:

- Using a fixed-length data structure to store subscriptions allows for fast retrieval.
- The approach avoids dynamic memory allocation, leading to improved access time and reduced overhead.

Use of const in Functions:

- The const qualifier is used in function parameters and return types to indicate that the data should not be modified.
- This practice improves code safety, enables compiler optimizations, and clarifies the intent to avoid unintended side effects.

Asynchronous Network Operations with Boost.Asio:

- Boost.Asio is used for asynchronous network operations, allowing the application to continue other tasks while waiting for network responses.
- This increases responsiveness and scalability by performing non-blocking I/O operations.

Dedicated Threads for WebSocket Connections:

- WebSocket connections are handled by dedicated threads to ensure parallel processing.
- This prevents blocking and ensures that each connection operates independently, enhancing concurrency.

Separate Thread for Processing Market Data:

- Processing of market data is isolated in a dedicated thread to avoid interference with other tasks (e.g., network communication).
- This separation ensures optimized data handling without impacting the responsiveness of other operations.

WebSockets to Reduce TCP Handshake Overhead:

- WebSockets eliminate the need for a TCP handshake on each request by maintaining a persistent connection.
- This reduces latency and communication overhead, making the system more efficient for real-time data transfer.

4. Future Improvements -

- Thread Pool Implementation: Instead of dedicating a separate thread for each WebSocket connection, implementing a thread pool can optimize resource utilization. This approach allows for a fixed number of threads to handle multiple connections, reducing the overhead associated with thread creation and destruction.
- Connection Pooling: Implementing connection pooling can further reduce the overhead of establishing new connections. By reusing existing connections, the application can handle requests more efficiently, especially in scenarios with high connection turnover.
- Load Balancing: Introducing load balancing mechanisms can distribute incoming connections across multiple servers or processes, enhancing scalability and fault tolerance. This strategy ensures that no single server becomes a performance bottleneck.