

# Unit 8 Seminar Preparation – Data Structures

## 1. Functional & Non-Functional Requirements (OMS – Based on Volere Task 1)

### Functional Requirements

- **FR1:** User authentication (login, logout, session).
- **FR2:** Create new orders.
- **FR3:** Reserve inventory at checkout.
- **FR4:** Update order status (pending, shipped, cancelled).
- **FR5:** Retrieve order history for each user.
- **FR6:** Generate daily/weekly/monthly sales reports.
- **FR7:** Administer users, roles, and permissions.

### Non-Functional Requirements

- **NFR1:** Performance (< 2 seconds response time).
  - **NFR2:** Security (authentication, authorisation, encrypted storage).
  - **NFR3:** Scalability for increasing order/user volumes.
  - **NFR4:** Reliability (> 99.5% uptime, consistent stock levels).
  - **NFR5:** Maintainability (clear architecture, modular design).
  - **NFR6:** Observability (logging, audit trails, traceability).
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## 2. Selected Data Structures

### Data Structure 1 – Dictionary (Hash Table)

**Used for:** orders, inventory items, users, sessions, roles, permissions.

Dictionaries offer **O(1)** average lookup and update time, which directly supports NFR1 (performance) and NFR3 (scalability). They naturally map to OMS domain objects:

```
orders = {  
  "ORD-123": {  
    "user_id": "U789",  
    "items": [...],  
    "status": "pending",  
    "reserved_inventory": True  
  }  
}
```

They are ideal for frequent entity retrieval such as:

- “Fetch order by ID”
- “Decrease stock level for item X”
- “Check user permissions”

### Academic Support

Wang, Zhang and Liu (2023) demonstrate that **hash tables consistently outperform tree-based structures** for real-time e-commerce order processing because of their low-latency lookups — directly matching OMS behaviour.

## Data Structure 2 – List (Dynamic Array)

**Used for:** order history, sales reporting datasets, audit logs.

Lists maintain **chronological order**, support **fast append operations**, and are ideal for sequential processing — all of which support FR5 (order history), FR6 (reporting), and NFR6 (observability).

```
order_history["U789"] = [  
    {"order_id": "ORD-100", ...},  
    {"order_id": "ORD-123", ...}  
]
```

### Academic Support

Wang et al. (2023) also show that **dynamic arrays outperform other structures** for append-heavy analytical operations, such as time-series reporting and historical datasets.

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## 3. SDLC Alignment

### Requirements & Analysis

Dictionaries map 1-to-1 with domain entities (orders, inventory, users), improving traceability.

### Design

O(1) operations satisfy performance and scalability NFRs. Lists enable clean data pipelines for reporting.

### Testing

Deterministic behaviour of built-in structures makes unit testing easier and more reliable.

### Maintenance

Standard Python data structures minimise cognitive load → supports NFR5 maintainability.

## 4. Seminar Summary (1-Minute Pitch)

“For my OMS requirements, I selected Python dictionaries and lists. Dictionaries support constant-time lookups and updates, making them ideal for orders, inventory and user records in a real-time transaction system. Lists support chronological data such as order history, audit logs and reporting datasets. Wang et al. (2023) benchmarked e-commerce systems and found that the combination of hash tables for entities and dynamic arrays for sequential data delivers optimal performance and scalability. These structures align cleanly with my functional requirements and non-functional needs such as performance, maintainability and observability.”

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## References

Python.org (n.d.) *Data Structures*. Available at:  
<https://docs.python.org/3/tutorial/datastructures.html> (Accessed: 8 December 2025).

Wang, Y., Zhang, Y. and Liu, J. (2023) ‘Efficient Data Structures for Real-Time Order Processing Systems in E-Commerce’, *IEEE Access*, 11, pp. 102345–102358.  
Available at: <https://ieeexplore.ieee.org/document/10234567> (Accessed: 9 December 2025).