

AI-ASSISTED CODING

LAB_EXAM-04

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BATCH : 03

SET -15

Q1: Database & SQL

- a) Build schema for Logistics and Warehouse Tracking.
- b) Generate SQL for inventory shortage alerts

PROMPT :

“Design a relational database schema for a logistics & warehouse tracking system (products, warehouses, locations, inventory, movements), and write SQL to generate alerts when inventory at any location falls below a defined reorder point.”

CODE :

```
1 -- a) Schema for Logistics and Warehouse Tracking
2
3 CREATE TABLE Warehouse (
4     warehouse_id INT PRIMARY KEY,
5     name VARCHAR(100),
6     location VARCHAR(100)
7 );
8
9 CREATE TABLE Product (
10    product_id INT PRIMARY KEY,
11    name VARCHAR(100),
12    description VARCHAR(255)
13 );
14
15 CREATE TABLE Inventory (
16     inventory_id INT PRIMARY KEY,
17     warehouse_id INT REFERENCES Warehouse(warehouse_id),
18     product_id INT REFERENCES Product(product_id),
19     quantity INT,
20     last_updated DATETIME
21 );
22
23 CREATE TABLE Shipment (
24     shipment_id INT PRIMARY KEY,
25     warehouse_id INT REFERENCES Warehouse(warehouse_id),
26     product_id INT REFERENCES Product(product_id),
27     quantity INT,
28     shipment_date DATETIME,
29     destination VARCHAR(100)
30 );
31
```

```
-- b) SQL for inventory shortage alerts (e.g., threshold = 10 units)

SELECT
    w.name AS warehouse,
    p.name AS product,
    i.quantity AS current_stock
FROM
    Inventory i
    JOIN Warehouse w ON i.warehouse_id = w.warehouse_id
    JOIN Product p ON i.product_id = p.product_id
WHERE
    i.quantity < 10;

-- Example Output:
-- | warehouse | product      | current_stock |
-- | ----- | ----- | ----- |
-- | Main Depot | Widget A | 5 |
-- | East Branch | Widget B | 2 |
```

Outputs :

```
pgsql
```

```
List of relations
```

Schema	Name	Type	Owner
public	supplier	table	your_user
public	product	table	your_user
public	warehouse	table	your_user
public	inventory	table	your_user
public	purchaseorder	table	your_user
public	purchaseorderline	table	your_user
public	shipment	table	your_user
public	shipmentline	table	your_user
public	inventorymovement	table	your_user

Brief observation :

- The schema cleanly separates *products*, *warehouses*, *inventory levels*, *shipments*, and *purchase orders*. This modularity supports maintainability and makes it easier to extend.
- For example, having a dedicated `InventoryMovement` table lets you track all changes (inbound, outbound, adjustments) separately from the “current state” in `Inventory`.
- **Support for Reorder Logic**
- The inclusion of `reorder_point` and `safety_stock` in the `Inventory` table gives a practical basis for automated alerts / replenishment strategies.

Q2 :

(Data Processing)

- a) Clean warehouse temperature logs.
- b) Detect out-of-range temperature anomalies

PROMPT :

“Clean the warehouse temperature log data by handling missing values, outliers, and noise. Then detect out-of-range temperature anomalies using time-series anomaly detection (e.g. via statistical thresholds, sliding window prediction, or isolation forest).”

CODE :

```
... lab.sql 9 ...
lab.sql
    ⚭ Connect to MSSQL
1   -- 1. Suppliers: who supply products
2   CREATE TABLE Supplier (
3       supplier_id SERIAL PRIMARY KEY,
4       name VARCHAR(255) NOT NULL,
5       contact_name VARCHAR(255),
6       phone VARCHAR(50),
7       email VARCHAR(255),
8       address TEXT
9   );
10
11  -- 2. Products
12  CREATE TABLE Product (
13      product_id SERIAL PRIMARY KEY,
14      sku VARCHAR(100) UNIQUE NOT NULL,
15      name VARCHAR(255) NOT NULL,
16      description TEXT,
17      unit_weight NUMERIC,
18      unit_volume NUMERIC,
19      price NUMERIC(12,2)
20  );
21
22  -- 3. Warehouses
23  CREATE TABLE Warehouse (
24      warehouse_id SERIAL PRIMARY KEY,
25      name VARCHAR(255) NOT NULL,
26      location TEXT,
27      capacity_units BIGINT,
28      capacity_volume NUMERIC
29  );
30
31  -- 4. Inventory: stock of a product in a warehouse
32  CREATE TABLE Inventory (
33      inventory_id SERIAL PRIMARY KEY,
34      warehouse_id INT NOT NULL REFERENCES Warehouse(warehouse_id),
35      product_id INT NOT NULL REFERENCES Product(product_id),
        quantity INT
```

```
lab.sql
32  CREATE TABLE Inventory (
33      quantity_on_hand BIGINT NOT NULL DEFAULT 0,
34      reorder_point BIGINT NOT NULL DEFAULT 0,
35      safety_stock BIGINT DEFAULT 0,
36      last_updated TIMESTAMP WITHOUT TIME ZONE DEFAULT NOW(),
37      UNIQUE (warehouse_id, product_id)
38 );
39
40
41
42
43 -- 5. Purchase Orders (to suppliers)
44 CREATE TABLE PurchaseOrder (
45     po_id SERIAL PRIMARY KEY,
46     supplier_id INT NOT NULL REFERENCES Supplier(supplier_id),
47     order_date DATE NOT NULL,
48     expected_delivery_date DATE,
49     status VARCHAR(50) NOT NULL -- e.g. PENDING, RECEIVED
50 );
51
52 -- 6. Purchase Order Lines
53 CREATE TABLE PurchaseOrderLine (
54     poline_id SERIAL PRIMARY KEY,
55     po_id INT NOT NULL REFERENCES PurchaseOrder(po_id),
56     product_id INT NOT NULL REFERENCES Product(product_id),
57     quantity_ordered BIGINT NOT NULL,
58     unit_price NUMERIC(12,2) NOT NULL
59 );
60
61 -- 7. Shipments (from warehouses, or between warehouses)
62 CREATE TABLE Shipment (
63     shipment_id SERIAL PRIMARY KEY,
64     from_warehouse_id INT NOT NULL REFERENCES Warehouse(warehouse_id),
65     to_warehouse_id INT REFERENCES Warehouse(warehouse_id),
66     shipment_date TIMESTAMP WITHOUT TIME ZONE NOT NULL,
67     status VARCHAR(50) NOT NULL, -- e.g. IN_TRANSIT, DELIVERED
68     carrier VARCHAR(255),
69     tracking_number VARCHAR(255)
70 );
```

OUTPUT:

```
pgsql
```

List of relations			
Schema	Name	Type	Owner
public	supplier	table	your_user
public	product	table	your_user
public	warehouse	table	your_user
public	inventory	table	your_user
public	purchaseorder	table	your_user
public	purchaseorderline	table	your_user
public	shipment	table	your_user
public	shipmentline	table	your_user
public	inventorymovement	table	your_user

BRIEF OBSERVATION :

- Cleaning the temperature logs improves data **reliability**: handling missing values and outliers ensures that downstream anomaly detection isn't misled by sensor noise or glitches. As shown in environmental-sensor studies, treating outliers first (e.g., as missing) then imputing improves data quality. [MDPI](#)
- Anomaly detection (out-of-range) is most effective when combining **domain-driven rules** (e.g., plausible physical temperature bounds) with **statistical** (IQR, Z-score) and **machine-learning** methods (e.g., Isolation Forest). This hybrid strategy helps catch both obvious violations and subtle irregularities. GeeksforGeeks describes how time-series anomalies often require layered detection. [geeksforgeeks.org](#)
- The choice of detection method should reflect the **characteristics of the sensor data**: for periodic or seasonal temperature readings, decomposition methods or adaptive thresholds may detect anomalies more robustly.