

# Hands-On Lab: Querying & Aggregations (DBS Tech Bank)

## Lab Objectives

By the end of this session, learners will:

- Understand the **Aggregation Pipeline Basics**
  - Use **\$match**, **\$group**, **\$project**, **\$sort**, **\$limit**
  - Filter & transform data using the pipeline
  - Build a **Customer Transaction Summary** (Case Study)
  - Understand all key pipeline operators with **real DBS Bank examples**
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### Step 0 — Setup

```
use dbs_tech_bank;
```

Verify:

```
show collections;
```

You should see:

```
customers
accounts
transactions
products
```

Make sure your transactions collection already has sample data from previous labs.

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### ① Aggregation Pipeline Basics (Hands-On)

The Aggregation Pipeline works like a **conveyor belt**:

```
db.collection.aggregate([
  { Stage 1 },
  { Stage 2 },
  { Stage 3 }
])
```

Each stage transforms data.

We will run the following simple pipeline first:

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### Step 1: Basic Pipeline Example

```
db.transactions.aggregate([
  { $match: { type: "DEBIT" } },
  { $project: { _id: 0, custId: 1, accNo: 1, amount: 1 } },
  { $sort: { amount: -1 } },
  { $limit: 5 }
]).pretty();
```

👉 What this does:

Stage	Meaning	Purpose
\$match	Filter	Get only DEBIT transactions
\$project	Select/shape fields	Choose which fields to show
\$sort	Order	Sort by amount (DESC)
\$limit	Restrict	Keep top 5

Run it → check output.

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## 2 Understanding Each Aggregation Stage (Hands-On)

Now we test each stage separately to understand them deeply.

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### 2.1 \$match – Filtering Documents (SQL WHERE)

#### Example 1: High-value DEBIT transactions (> ₹50,000)

```
db.transactions.aggregate([
  { $match: { type: "DEBIT", amount: { $gt: 50000 } } }
]).pretty();
```

#### Example 2: Transactions in 2025

```
db.transactions.aggregate([
  {
    $match: {
      timestamp: {
        $gte: ISODate("2025-01-01T00:00:00Z"),
        $lt: ISODate("2026-01-01T00:00:00Z")
      }
    }
  }
]).pretty();
```

```
        }
    }
]).pretty();
```

### 💡 Why \$match first?

It reduces the dataset → faster pipeline.

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## ◆ 2.2 \$project – Selecting & Transforming Fields

### Example 1: Show only key output fields

```
db.transactions.aggregate([
  {
    $project: {
      _id: 0,
      txnId: 1,
      custId: 1,
      amount: 1,
      txnYear: { $year: "$timestamp" }
    }
  }
]).pretty();
```

### Example 2: Add computed fields

```
db.transactions.aggregate([
  {
    $project: {
      custId: 1,
      amount: 1,
      isHighValue: { $gt: ["$amount", 100000] }
    }
  }
]).pretty();
```

---

## 2.3 \$group – Grouping & Summaries (SQL GROUP BY)

### Example: Total spent per customer

```
db.transactions.aggregate([
  {
    $group: {
      _id: "$custId",
      totalAmount: { $sum: "$amount" },
    }
  }
]).pretty();
```

```
        totalTxns: { $sum: 1 }
    }
}
]) .pretty();
```

---

## 2.4 \$sort – Ordering Results

### Sort by transaction amount (DESC)

```
db.transactions.aggregate([
    { $sort: { amount: -1 } }
]) .pretty();
```

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## 2.5 \$limit – Restrict Number of Records

### Top 3 biggest transactions

```
db.transactions.aggregate([
    { $sort: { amount: -1 } },
    { $limit: 3 }
]) .pretty();
```

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## Filtering & Transforming Data (Hands-On)

We combine \$match, \$project, and \$addFields.

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### ◆ Step 3.1 – Filter + Transform

```
db.transactions.aggregate([
{
    $match: {
        type: "DEBIT",
        amount: { $gt: 20000 }
    }
},
{
    $project: {
        _id: 0,
        custId: 1,
        accNo: 1,
        amount: 1,
        day: { $dayOfMonth: "$timestamp" },
        month: { $month: "$timestamp" },
        year: { $year: "$timestamp" }
    }
}
```

```
        }
    ]) .pretty();
```

👉 **What you see:**

- Filter → only high-value DEBIT transactions
  - Transform → extract date parts
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◆ **Step 3.2 – Add Flags**

```
db.transactions.aggregate([
  {
    $addFields: {
      highAmtFlag: { $gt: ["$amount", 100000] },
      digitalChannelFlag: { $in: ["$channel", ["UPI",
      "NETBANKING"]] }
    }
  }
]) .pretty();
```

👉 **This is how analytics pipelines flag risky transactions.**

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**Case Study (Main Lab) – Summarize Total Transactions per Customer**

This is the **core Customer360 analytics** use case.

✓ **Goal**

For every customer:

- Count number of transactions
  - Compute total amount
  - Compute average transaction value
  - Sort by total transaction value (high → low)
- 

✳️ **Step-by-Step Aggregation Pipeline**

**Step 1 — Filter POSTED transactions**

```
{ $match: { status: "POSTED" } }
```

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**Step 2 — Group by customer**

```
{  
  $group: {  
    _id: "$custId",  
    totalAmount: { $sum: "$amount" },  
    txnCount: { $sum: 1 }  
  }  
}
```

---

### Step 3 — Format Output using

**\$project**

```
{  
  $project: {  
    _id: 0,  
    custId: "$_id",  
    totalAmount: 1,  
    txnCount: 1,  
    avgAmount: { $divide: ["$totalAmount", "$txnCount"] }  
  }  
}
```

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### Step 4 — Sort by total amount (DESC)

```
{ $sort: { totalAmount: -1 } }
```

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### Step 5 — Limit top 5

```
{ $limit: 5 }
```

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## Full Case Study Pipeline

```
db.transactions.aggregate([  
  // 1. Filter  
  {  
    $match: {  
      status: "POSTED"  
    }  
  },  
  
  // 2. Group by customer  
  {  
    $group: {  
      _id: "$custId",  
      totalAmount: { $sum: "$amount" },  
      txnCount: { $sum: 1 }  
    }  
  }]
```

```

        }
    } ,

// 3. Shape output
{
    $project: {
        _id: 0,
        custId: "$_id",
        totalAmount: 1,
        txnCount: 1,
        avgAmount: {
            $cond: {
                if: { $gt: ["$txnCount", 0] },
                then: { $divide: ["$totalAmount", "$txnCount"] },
                else: 0
            }
        }
    }
} ,
// 4. Order by largest spenders
{ $sort: { totalAmount: -1 } },
// 5. Limit to top 5 customers
{ $limit: 5 }
]).pretty();

```

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### Interpretation (Explaining Output)

For each customer (e.g., C1001, C1002...):

- totalAmount: how much they spent (or total transaction volume)
- txnCount: number of transactions performed
- avgAmount: customer's average transaction value
- Sorted by highest totalAmount → identify **premium / high-value / risky** customers.

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This is the **foundation of Customer360 dashboards** used by modern banks like DBS.

 **Bonus: Explain Aggregation Stages (Cheat Sheet)**

Operator	Meaning	Example Purpose
\$match	Filter docs	High-value txns
\$group	Summarize	Total per customer
\$project	Select + compute fields	Add avgAmount
\$sort	Order	Top spenders
\$limit	Restrict N	Show top 5
\$addFields	Add flags	Risk features
\$lookup	Join collections	Customer + Account
\$unwind	Flatten arrays	Expand transactions
\$count	Count docs	Simple counts
\$sum	Sum values	Total debit
\$avg	Average	Mean txn size
\$dateToString	Format dates	YY-MM-DD

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**Lab Completion Checklist**

- ✓ Understand \$match, \$project, \$group, \$sort, \$limit
  - ✓ Create simple & complex pipelines
  - ✓ Filter & transform data
  - ✓ Build Customer360 transaction summary
  - ✓ Interpret aggregation output
  - ✓ Apply best practices for performance
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