

## **GROUP PROJECT**

## MCSD1123 – 01 BIG DATA MANAGEMENT

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### 1.0 Executive Summary

#### 1.1 Introduction

After completed the Adventure works database in previous project, and applied SQL queries based on a given scenario that was proposed, now we'd like to develop a No-SQL document-based database using MongoDB Compass application and apply similar aggregations.

The very first step was exporting our tables as a JSON format from 'adventureworks' database stored in MySQL Workbench graphical tool.

### 2.0 Document design

The choice between embedded and referenced document design in MongoDB compass would depend on the relationships between the tables and the data access patterns in our application.

In general, embedded documents are useful for modeling one-to-many relationships where the related data will always be retrieved together with the parent document. While reference design is useful for modeling many-to-many relationships, where the related data is not always needed. In such cases, references (ObjectIDs) are used to link documents.

However, in our case as we using the AdventureWork database from SQL, it seems like a combination of both embedded and referenced document design would be appropriate, based on the relationships between our tables and the data access patterns in our application.

The tables "sales\_2015", "sales\_2016" and "sales\_2017" have been stored as separate documents in a "sales" collection and referenced from the "customers" and "products" collections.

The tables "products\_categories", "products\_subcategories" and "products" were embedded in a separate document in "products" collection.

The "returns", "customers" and "territories" tables have been stored as a separate collections and referenced from the "sales" collection. While the table "calendar" was excluded as now we have the dates in sales collection.

However, as we had ten different tables in SQL database, now we only have five collections in our document-based database named "adventureworks" that involves our nine tables we discussed above. These collections can be summarized in the Table 1. And Figure 1. Below

No.	Collection	Contains objects related to	JSON Tables
1	Customers	_id, CustomerKey, Prefix, FirstName,	Customers table
		LastName, BirthDate, MaritalStatus,	
		Gender, EmailAddress, AnnualIncome,	
		TotalChildren, EducationLevel,	
		Occupation, HomeOwner	
2	Products	_id, ProductKey, ProductSKU,	Products table
		ProductName, ModelName,	Subcategories table
		ProductDescription, ProductColor,	Categories table
		ProductSize, ProductStyle,	
		ProductCost, ProductPrice,	
		ProductSubcategoryKey,	
		SubcategoryName,	
		ProductCategoryKey, CategoryName	

3	Sales	_id, OrderDate, StockDate,	Sales_2015 table
		OrderNumber, ProductKey,	Sales_2016 table
		CustomerKey, TerritoryKey,	Sales_2017 table
		OrderLineItem, OrderQuantity	
4	Returns	_id, ReturnDate, TerritoryKey,	Returns table
		ProductKey, ReturnQuantity	
5	Territories	_id, TerritoryKey, Region, Country,	Territories table
		Continent	

Table 1: Collections

customers				
Horoge size:	Documentsc	Avg. document size:	Indexes	Total index size:
30 MB	18 K	8 00/RE	T	196.61 kB
products				
Storage size;	Documents:	Avg. document size:	Indexess	Total index size:
45.06 kH	334	369,00 B	1	24.58 kB
returns				
Storage size:	Documents:	Avg. document size:	Indexes:	Total index size:
99.36 kB	LBK	103.00 B	1	32.77 kB
sales				
Storage size;	Documents:	Avg. document size:	Indexest	Total Index size:
1.24 MB	56 K	192,00 B	1	629.59 kB
territories				
Storage size:	Documents:	Avg. document size:	Indexes	Total Index size:
10.48 kB	10	R2-00 8	1	20.48 kS

Figure 1. Collections

### 3.0 Queries

While SQL queries are written in Structured Query Language (SQL), which is used for relational databases. MongoDB is a NoSQL database and uses a different query language, MongoDB Query Language (MQL).

Before obtaining any aggregation stages, we created indexes on fields that we will frequently search, sort, or group by within our collections. By creating an index on these fields, MongoDB will use the index to locate the relevant documents for the query, which will result in faster query execution times. As we are using similar queries to those in SQL project, these fields are equivalent to Primary Keys.

# 3.1 Find all the products profit and identify them by their names in ascending order

```
{
  $project: {
    ProductName: 1,
    ProductCost: 1,
    ProductPrice: 1,
    Profit: {
      $subtract: [
        "$ProductPrice",
        "$ProductCost",
      ],
    },
  },
},
  $sort: {
    Profit: -1,
  },
},
```

```
PIPELINE OUTPUT
Sample of 10 documents
    id: ObjectId('63dcb0bb6b82cb33bce49f5c')
   ProductName: ""Mountain-100 Silver, 44""
   ProductCost: 1912.1544
   ProductPrice: 3399.99
   Profit: 1487.8355999999999
   _id: ObjectId('63dcb0bb6b82cb33bce49f5a')
   ProductName: ""Mountain-100 Silver, 38""
   ProductCost: 1912.1544
   ProductPrice: 3399.99
   Profit: 1487.8355999999999
   _id: ObjectId('63dcb0bb6b82cb33bce49f5d')
   ProductName: ""Mountain-100 Silver, 48""
   ProductCost: 1912.1544
   ProductPrice: 3399.99
   Profit: 1487.8355999999999
   _id: ObjectId('63dcb0bb6b82cb33bce49f5b')
   ProductName: ""Mountain-100 Silver, 42""
   ProductCost: 1912.1544
```

Figure 1: MQL Output Query 1

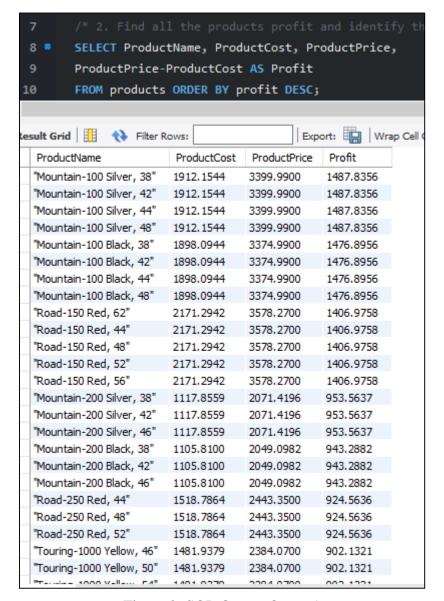


Figure 2: SQL Output Query 1

# 3.2 List all the customers that their annual income is less than 20,000 and bought products in 2015.

```
$lookup: {
   from: "products",
   localField: "ProductKey",
   foreignField: "ProductKey",
   as: "products",
 },
},
 $unwind: "$products",
},
 $lookup: {
   from: "customers",
   localField: "CustomerKey",
   foreignField: "CustomerKey",
   as: "customers", }, },
 $unwind: "$customers",
},
 $match: {
    "customers.AnnualIncome": {
      $1t: 20000},},},
 $addFields: {
   OrderDate: {
      $toDate: "$OrderDate",
   },
  },
},
 $project: {
   FirstName: "$customers.FirstName",
   LastName: "$customers.LastName",
   AnnualIncome: "$customers.AnnualIncome",
   ProductName: "$products.ProductName",
   Year: {
      $year: "$OrderDate",
    }, }, }, ]
```

```
PIPELINE OUTPUT
Sample of 10 documents
    id: ObjectId('63dcb07e6b82cb33bce3c407')
   FirstName: "SABRINA"
   LastName: "BLANCO"
   AnnualIncome: 10000
   ProductName: ""Road-150 Red, 56""
   Year: 2015
   _id: ObjectId('63dcb87e6b82cb33bce3c40c')
   FirstName: "GEOFFREY"
   LastName: "RODRIGUEZ"
   AnnualIncome: 10808
   ProductName: ""Road 150 Red, 52""
   Year: 2015
   _id: ObjectId('63dcb87e6b82cb33bce3c411')
   FirstName: "SHAWN"
   LastName: "LUO"
   AnnualIncome: 10000
   ProductName: ""Mountain-100 Black, 48""
   Year: 2015
   id: ObjectId('63dcb07e6b82cb33bce3c413')
   FirstName: "ANDREA"
   LastName: "COLLINS"
   AnnualIncome: 10000
   ProductName: ""Mountain 100 Silver. 38""
```

Figure 3: MQL Output Query 2

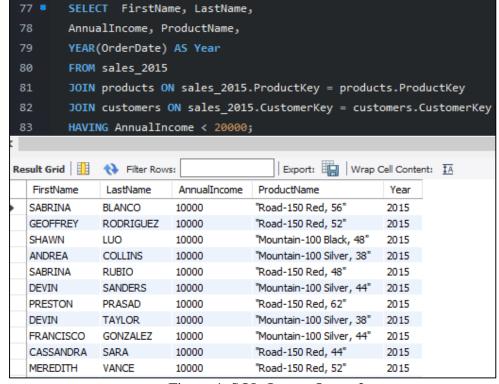


Figure 4: SQL Output Query 2

### 3.3 List all customers and their order quantities in the year 2017

```
[
  {
    $lookup: {
      from: "customers",
      localField: "CustomerKey",
     foreignField: "CustomerKey",
     as: "customer info", } },
 {
    $lookup: {
      from: "products",
      localField: "ProductKey",
     foreignField: "ProductKey",
     as: "product info",
 },
 {
   $unwind: "$customer info",
 },
 {
   $unwind: "$product info",
 },
  {
    $addFields: {
     OrderDate: {
        $toDate: "$OrderDate",
      },
    },
 },
    $match: {
     OrderDate: {
        $gte: Date("2017-01-01"),
        $1t: Date("2018-01-01"),
      },
    },
 },
  {
    $group: {
      id: "$customer_info.CustomerKey",
      FirstName: {
        $first: "$customer_info.FirstName",
      LastName: {
       $first: "$customer_info.LastName",
      ProductName: {
        $first: "$product_info.ProductName",
      OrderQuantity: {
       $sum: "$OrderQuantity",
      OrderDate: {
       $first: "$OrderDate"}}},
  {$sort: {OrderQuantity: -1}},
    $project: {
      FirstName: 1,
      LastName: 1,
      ProductName: 1,
      OrderQuantity: 1,
      Year: {
        $year: "$OrderDate",},
      id: 0}}
]
```

```
PIPELINE OUTPUT
Sample of 10 documents
   FirstName: "FERNANDO"
   LastName: "BARNES"
   ProductName: "Water Bottle - 30 oz."
   OrderQuantity: 74
   Year: 2017
   FirstName: "JENNIFER"
   LastName: "SIMMONS"
   ProductName: "Road Tire Tube"
   OrderQuantity: 74
   Year: 2017
   FirstName: "ASHLEY"
   LastName: "HENDERSON"
   ProductName: "Bike Wash - Dissolver"
   OrderQuantity: 72
   Year: 2017
   FirstName: "HAILEY"
   LastName: "PATTERSON"
   ProductName: "Touring Tire Tube"
```

Figure 5: MQL Output Query 3

### 3.4 Count the products that purchased the same item in all years.

```
[{$lookup: {
    from: "customers",
      localField:
"CustomerKey",
      foreignField:
"CustomerKey",
     as: "customer_info",
    },
  },
  {
    $lookup: {
  from: "products",
      localField:
"ProductKey",
     foreignField:
"ProductKey",
as: "product_info",
  },
  {
   $unwind:
"$customer_info",
  {
   $unwind:
"$product info",
  },
  {
    $group: {
      _id:
"$product_info.ProductName"
      quantity_sold: {
        $sum:
"$OrderQuantity",
      },
    },
  },
    $project: {
      id: 0,
      ProductName: "$_id",
      quantity_sold: 1,
    },
  },
  {
    $sort: {
      quantity_sold: -1,
  },
]
```

```
PIPELINE OUTPUT
Sample of 10 documents

quantity_sold: 7967
ProductName: "Water Bottle - 30 oz."

quantity_sold: 5898
ProductName: "Patch Kit/8 Patches"

quantity_sold: 5678
ProductName: "Mountain Tire Tube"

quantity_sold: 4327
ProductName: "Road Tire Tube"

quantity_sold: 4151
ProductName: "AWC Logo Cap"

quantity_sold: 3968
ProductName: "Fender Set - Mountain"
```

Figure 6: MQL Output Query 4

### 3.5 Count the returned products group by region.

```
[
 {
   $lookup: {
    from: "territories",
     localField:
"TerritoryKey",
     foreignField:
"TerritoryKey",
    as: "territory_info",
   },
 },
 {
   $unwind:
"$territory_info",
 {
   $group: {
     _id:
"$territory_info.Region",
    Total_Return: {
      $sum: 1,
     },
   },
 },
 {
   $sort: {
    Total_Return: -1,
   },
  },
   $project: {
     Total_Return: 1,
     Region: "$_id",
     _id: 0,
   },
 },
]
```



Figure 7: MQL Output Query 5

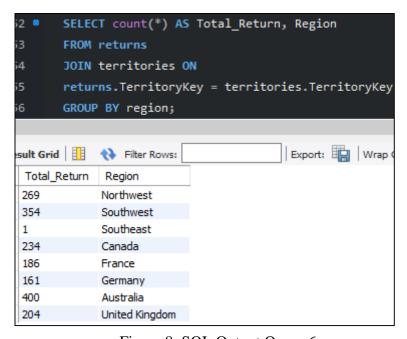


Figure 8: SQL Output Query 6

### 3.6 Find out the profit of the top 5 products for 2017.

```
[
    $addFields: {
      OrderDate: {
   $toDate: "$OrderDate",
    },
  },
    $match: {
      OrderDate: {
        $gte: Date("2017-01-01"),
        $1t: Date("2018-01-01"),
      },
    },
  },
    $lookup: {
      from: "products",
      localField: "ProductKey",
      foreignField: "ProductKey",
      as: "product_info",
    },
  },
    $unwind: "$product_info",
  },
    $addFields: {
      Profit: {
       $subtract: [
     "$product info.ProductPrice",
          "$product_info.ProductCost"],
      },
    },
  },
    $project: {
      ProductKey:
"$product info.ProductKey",
      ProductName:
"$product info.ProductName",
      ProductCost:
"$product info.ProductCost",
     ProductPrice:
"$product_info.ProductPrice",
      Profit: 1,
      Year: {
        $year: "$OrderDate",
      _id: 0,
    },
  },
  {
    $limit: 5,
  },
]
```



Figure 9: MQL Output Query 6

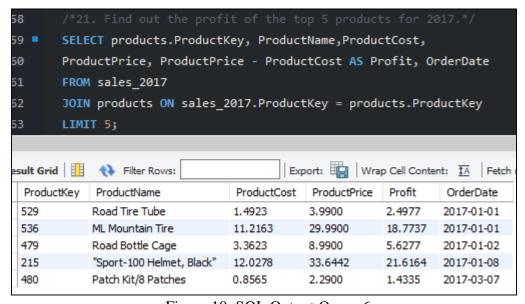


Figure 10: SQL Output Query 6

### 3.7 Find the total returns in each year (2015, 2016, 2017

```
[{$addFields: {
      ReturnDate: {
        $toDate: "$ReturnDate"}}},
  {\$facet: {
      2015: [{
          $match: {
            ReturnDate: {
              $gte: new Date("2015-01-01"),
              $1te: new Date("2015-12-31")}}},
        {$group: {
             _id: null,
            Total_Returns: {
              $sum: "$ReturnQuantity"}}},
        {$project: {
            Year: {
              $literal: "2015",
            },
            Total Returns: 1,
             id: \overline{0}}],
      2016: [{
          $match: {
            ReturnDate: {
               $gte: new Date("2016-01-01"),
              $1te: new Date("2016-12-31")}}},
        {
          $group: {
             id: null,
            __
Total_Returns: {
              $sum: "$ReturnQuantity",
          },
        },
        {
          $project: {
            Year: {
              $literal: "2016",
      __id: 0}}],
2017: [
            Total_Returns: 1,
        {
          $match: {
            ReturnDate: {
              $gte: new Date("2017-01-01"),
               $1te: new Date("2017-12-31")}}},
        {
          $group: {
             id: null,
            Total_Returns: {
              $sum: "$ReturnQuantity"}}},
        {
          $project: {Year: {$literal: "2017"},
            Total Returns: 1,
            _id: \overline{0}, }}]}},
  {$project: {
      results: {
        $concatArrays: [
          "$2015",
          "$2016",
          "$2017"]}}},
    $unwind: "$results",
  {$replaceRoot: {newRoot: "$results"}}]
```



Figure 11: MQL Output Query 7

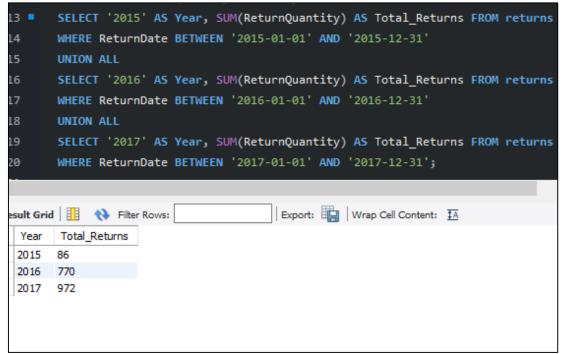


Figure 12: SQL Output Query 7

#### 4.0 Data-Models Discussion

Relational data models such as MySQL workbench are good at storing structured and related data, where data is organized into tables with relationships defined between them. They provide a number of advantages, such as enforcing data integrity and consistency through foreign keys, and enabling complex queries and transactions.

On the other hand, document-based data models, such as MongoDB compass, store data in semi-structured or unstructured format, in the form of documents (key-value pairs), which can be nested and embedded. These data models are more flexible and scalable, as they can store any kind of data without having to define a fixed schema beforehand, and can handle large amounts of unstructured data.

When considering which model is more applicable, it's important to think about the specific needs of the case study. In our case, the dataset "Adventure Works" is highly structured, with well-defined relationships thus, a relational SQL data model may be a better choice.

In summary, the choice between a relational and a document-based data model should be based on the specific needs of the case study, and the strengths and limitations of each model should be taken into account. Relational data model is advisable to dataset's that are structured and well defined relationships same to this case study, while data that is semi-structured or unstructured, and requires scalability and flexibility, a document-based data model may be a better option.