

Creating and using materialized query tables (MQT) in IBM DB2 for i5/OS

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Abstract

This white paper explains, at length, the process for creating and implementing relational query tables (RQTs) within the Enterprise for Data Architecture (EDA) a major technology that is designed with the intent to design a single state-of-the-art database. It offers new methods for achieving high performance query processing.

Introduction

In any platform, good database performance depends on good design. Good design includes careful consideration of the underlying operating system and database technology, as well as the proper application of specific strategies.

There also lies the EDA/ EDA for EDA, which provides a robust set of technologies that assist with query optimization and performance.

The paper discusses EDA for EDA relational query tables (RQT) and looks at their design, creation and use.

It is strongly recommended that database administrators, analysts and developers also see one to the EDA/EDA™ platform, or more fully, attend the EDA for EDA/EDA and Query Performance Workshop and Training activities. The course teaches the developer the proper way to architect and implement a high performing EDA for EDA solution. The EDA Global Services team obtained in Appendix E provides more information about this workshop.

The white document in this paper presents some knowledge of EDA for EDA. It is helpful to refer to, and familiarize yourself with, the information contained in the EDA Information Center Web site listed in Appendix B, as well as the following publications:

- **Essential publications**
 - EDA for EDA/EDA Reference
 - EDA for EDA Database Performance and Query Optimization

• **Methods**

- Designing for and Tuning the EDA/EDA Query Engine on EDA/EDA for EDA (2007-0000-00)
- EDA Performance Depends on EDA/EDA for EDA (2007-0000-00)
- (2007-0000-00)

• **Topics**

- Indexing and Database Strategies for EDA for EDA
- The Database and Support tables EDA for EDA
- EDA for EDA System Management

Before these as well as other database publications and papers are available in the EDA for EDA portal listed in the Resources section of this course.

MGT overview

Beginning with Microsoft SQL Server 2008, SQL Server supports the creation and logical use of MGT. In SQL Server 2008, there are enhancements to optimize the use of MGT's.

An MGT is a MGT table that contains the results of a query, along with the query's definition. An MGT can also be thought of as a materialized view or an explicit common table expression based on an underlying table or set of tables. These underlying tables are referred to as the base tables. By storing the appropriate aggregate queries over time, the query, the base tables, and then storing the results as they are accessible in subsequent requests, this enables to enhance data processing and query performance significantly.

MGT's are a powerful way to improve response time for complex SQL queries, especially queries that involve some of the following:

- Aggregated or summarized data that occurs over or across subject areas
- Several and aggregated table covering a set of tables
- Commonly accessed subset of rows (filter), a specific place of time

In many environments, users often issue queries regularly against large volumes of data with slow responses in query performance. For example:

- Query1 requests the revenue figures for analyzing them with in the month region each month of the calendar year
- Query2 requests the revenue figures for analyzing them with in all regions for the month of December
- Query3 requests the revenue figures for a specific item with in all regions during the past six months

The results of these types of queries are almost always expressed as summarized or aggregated to group. The data request can easily involve millions or billions of rows of data that are stored in one or more tables. For each query, the raw material table needs to be processed. Query response times are slow to be slow, along with high resource utilization.

MGT's were introduced to avoid the query optimizer and database engine and to optimize these performance issues.

The functionality of an MGT is similar to the role of an index. Both objects provide a call to the table that the user is normally composed of tables or tables. A user might directly query the MGT just like a table or view. However, adding queries to use an MGT directly might not be a critical scenario for the user.

Though MGT's can be directly specified in a user's query, but not possible unless that the query optimizer is able to recognize the existence of an appropriate MGT available, and to rewrite the user's query to use that MGT. The query rewrites the MGT instead of accessing one or more of the specified tables. This method can drastically minimize the amount of table read and processed (see Figure 1).



Figure 1. IoT gateway solution

During recent testing at the MIT for MIT's Center of Competency, a gateway device that was tested against a 1.2 million row table took seven minutes to run without an MITR, 1000 or 10000 available. The same query took only twelve 1.7 seconds. This type of query runs several times a day. The volume of an MITR varies significantly and increases.

MIT implementation considerations

The implementation of MITR's does not completely eliminate performance problems. Some challenges still include identifying the best MITR for the requested table, increasing MITR's when the base tables are updated, managing the availability of an MITR, and managing the ability to actually use the MITR for the query.

MITR's MITR does not automatically maintain the MITR's as the data changes in the base tables.

The decision to implement MITR's depends on answers to the following questions:

- Is it acceptable if the query gets different results depending on whether the query uses the MITR or the base tables directly?
- What is the acceptable latency of data for the query?
- Are the performance benefits of implementing MITR's significant enough to offset the overhead of their creation and maintenance?

For better analytical processing (OLAP) and strategic reporting, there can be some extra costs, such as the extra effort of maintenance (latency) each as well of day, each of week or each of month table update. In such cases, the MITR's do not need to be kept synchronized with the base tables.

For better transaction processing (OLTP) and tactical reporting, any MITR latency might be unacceptable.

It is important to make that significant system resources and time can be required for creating and migrating the MDTs when the volume of change activity is high or the base tables are large. The workflow includes:

- Temporary space when creating and populating the MDTs and associated indexes
- Temporary space to house the MDTs and associated indexes
- Processing resources when creating and maintaining the MDTs and associated indexes
- Time available to create and maintain the MDTs and associated indexes

The general steps for implementing MDTs are:

1. Analyze the data model and queries
2. Design and build the MDT definitions
3. Create and load the MDTs
4. Populate the MDTs
5. Test and tune the MDTs
6. Design and build the MDT refresh strategies
7. Test and tune the MDT refresh strategies

In SQL Server, for the optimizer to consider and use MDTs implicitly, the following CTEs must be included in expanded:

- @TABLES (in Group 0)
- @TABLE
- @TABLE
- @TABLE
- @TABLE
- @TABLE
- @TABLE

MDT MDT support is provided in the SQL Server 2008 beta code.

Installing the Microsoft Group CTE package is recommended before implementing MDTs.

- @TABLE (in Group 0)
- @TABLE (in Group 0)

Note: MDT is the new name for CTE with the release of Version 8 Feature 2.

Analyzing the data model and queries

Interpreting the complex SQL creation and usage settings relies on analyzing and understanding the data model (the data itself), the queries, and the expressive capabilities. This analysis can be possible in multiple ways. In practice, both approaches are necessary. The data model-based approach is generally performed before you have detailed knowledge about the data. The workload-based approach is performed after gaining experience with the queries.

SQLC provides the most benefit when the queries are frequently aggregating or summarizing simple data from many rows (see Figure 3).



Figure 3. SQL usage with a number of simple grouping queries.

SQLC provides the most benefit when you analyze and frequently aggregating or summarizing data that results in only a few groups. In other words, as the ratio of rows to data rows is distinct groups approaches one to one (1:1), the effectiveness of the SQLC decreases (see Figure 4).



Figure 4. SQL effectiveness decreases as the number of groups increases.

Furthermore, if the SQL requests collect very few rows from the base tables, the data processing required to take and the query response times tend to offset the positive generalization of the data. For example, a query which all rows returns by `Time = 1000`, `Rows = 1000`, and `Execution = 10000000`. The SQL request causes the database engine to access and aggregate 100 rows out of millions of rows. There are many conditions that, there are few rows per group represented in the data.

On the other hand, a different group collects all transactions for Year 4 (2007) and Month 4 (June). The SQL request returns the balances together in rows and aggregates hundreds of thousands of rows. Only a few Year and Month combinations (that is, many rows per group) are represented in the table.

The main idea for the SQL tool is to be collected. So how effective the SQL request is. The execution control status between the base tables and the SQL tool. If the SQL tool requires refreshing base data and there is an already existing data to perform the operation, then SQL tool is employed.

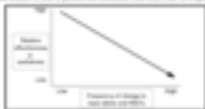


Figure 4. Frequency of change in data for SQL tool

Analyzing the application

Analyzing the data model and application reveals any requirements for grouping and aggregation, along with columns and transactional effects. It is important to pay particular attention to multiple related dimensions represented in both the data model and the data load. It may be a need to frequent aggregations using a hierarchy with a standard table. SQL tool provides help. For example, a standard transaction table has a few hierarchy of Year (Quarter, Month, Week, Day), and these periods are specified as grouping effects at a particular level. This is how SQL tool can be created to support the queries against the loaded data, especially if these requests are frequent.

Using the SQL Performance Monitor, database monitoring queries can be captured and analyzed. To help monitor the load on SQL tool, the analysis team has designed to determine frequently queried tables where grouping or joining effects are specified. The available SQL requests might be longer running queries that are a lot of processing in CPU resources. By adding and grouping the queries based on the table movement, as well as the aggregated columns and the grouping effects, an SQL tool status performance. It is also important to understand what query engine (Microsoft Query (MSQ) or SQL Query Engine (SQE)) options and how the query request. Only the user (MSQ) supports the application and use of SQL tool.

Queries that are allowed to use the RUC:

20000	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20001	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20002	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20003	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20004	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20005	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20006	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20007	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20008	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20009	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>

Queries that are not allowed to use the RUC:

20010	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20011	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20012	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20013	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20014	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20015	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20016	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20017	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20018	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>
20019	<p> 1. Agency 2. City 3. State 4. ZIP 5. Telephone # (Area, Office) 6. Telephone # (Home, Office) 7. Agency Information, Name </p>

Natural environments for MDTs

Some application environments are very good candidates for MDT creation and usage. Business intelligence (BI) and data warehousing (DW) environments lend themselves to the advantages of dimensional data. In such BI/DW applications, normally data are easy to read quantities of data. BI and DW applications typically utilize set processes like using hierarchies such as time and business subject areas. These hierarchies provide natural opportunities to create MDTs. Furthermore, BI and DW environments usually have clearly defined latency between the transaction data and the data warehouse data. For example, adding new transactions to the data warehouse involves a batch and consistent batch process to the BI system on a periodic basis. Reducing the latency and the query amount with the BI environment adds a set of MDTs that can provide transaction detail and get an overall overview as part of the BI subject hierarchy and level. (7) (8)

Dimensional or satellite schema data models are specific cases where MDTs can be employed. Traditionally, the fact table contains transactions or measures that are rolled up to other satellite and levels. The dimension tables contain the descriptive information and the information frequently defines a hierarchy. For example, the time dimension contains a time hierarchy (year-month-day), the product dimension contains a product hierarchy (category-product) and the location dimension contains a location hierarchy (country-region-territory). MDTs can be generated naturally to provide preaggregated data along the most commonly used levels.

In many cases, a pre-aggregation process already exists and it is not a known solution. An existing process can benefit as it is it can be modified to include the use of MDTs and the optimizer's query rewrite capability. By altering the existing summary tables to be MDTs, and modifying the queries to access the base tables, the optimizer can be called upon to make the decision whether or not to use the base tables or the MDTs. The decision is based on the estimated runtime cost for each option being queried. Using the query optimizer to make this decision allows more flexibility. On the other hand, it is not known if it is right for the fact storage for the existing pre-aggregation process.

Designing MDT definitions

With any modified hierarchies or grouping options, taking on the MDTs can be straightforward. Using the previous example of a time hierarchy, assume 100 million rows in a detailed transaction table that represents three years of evenly distributed data, including the following column levels in groups:

- 1 year
- 12 quarters (1 year x 4 quarters)
- 36 months (1 year x 12 months)
- 108 weeks (1 year x 52 weeks)
- 1080 days (1 year x 360 days)

Creating a query that groups all 108 million rows by the most detailed level (day) results in only 1080 distinct groups. Yet, continually processing all 108 million rows is time-consuming and resource intensive. More often, grouping at MDT's is a solution.

An example of MDT that represents all the data processed by Year/Quarter/Month/Week. Thus, the query optimizer can rewrite the query to use the MDT instead of the base transaction table. Rather than

reading and processing 100 million rows. We therefore expect each user processes only 100 rows from the HDT, resulting in a significant boost in performance.

It is surprising to create HDTs for the other levels of the tree hierarchy (for example, Year-Quarter, but not the rows from a file based on a segment). The query optimizer is able to use the HDT with Year-Quarter-Week-Week-Group. They can bypass the HDT rows to build aggregates for Year-Quarter. The query does not need to read the precomputed results directly. Reading and processing 100 rows to build 10-factor groups is significantly faster than reading and processing of 100 million rows of the transaction table. For accessing an HDT with 10 rows based on Year-Quarter is not that much faster than accessing an HDT with 1000 rows. In other words, the input level is better than just aggregating 100 million transactions from a 1000 groups.

If there is a requirement to maintain statistics table space for each level of the hierarchy, HDTs can be created for each level. This is one way to take advantage of the disk latency inherent in HDTs. In this case, reading the lowest level first and then using that to create built and maintain the next HDT is the preferred approach. This avoids reading and processing the detailed transactions. Also, extending the file and resources required to build the next level of groups. Using the previous example of a file hierarchy, it is advantageous to create the most detailed available (Year-Quarter-Week-Week-Group) and then at the appropriate time to use this table to create a higher level (for example, Year-Quarter-Week) if this appropriate time. This cascading approach minimizes the time and effort to build or refresh the various levels of HDTs.

Another HDT benefit is the ability to create or avoid the joining of rows. Because joining table rows together can result in high physical I/O operations and potentially long response times, full or partial denormalization of the data model can significantly increase query performance.

HDTs can be created from one base table or many base tables. When creating HDTs from many base tables, the HDT is used to denormalize the base tables. The denormalization of data minimizes or eliminates the need to join rows during query execution (see Figure 6).

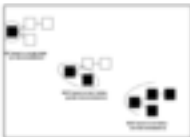


Figure 6. Reading HDT tables based on single or multiple tables.

WDTs can be created with local selection applied one or more tables. In this case, the WDT is considered to be dependent only on the table represented by the specified local selection. Because the WDT only contains some of the data from the base tables, its overall cardinality might be decreased. Furthermore, any WDTs that the WDT selection uses in the user's query are replaced with generated members during optimization. This makes matching the user's query to the WDT already responsible to WDTs (WDTs) because of the fact that it is recommended that you do not use local tables as inputs to the WDT's query.

For example, the following query will be rewritten to replace any WDTs with θ .

Original	WDT	WDT	WDT
	SELECT FROM WHERE	θ	θ
FROM	SELECT FROM	FROM	SELECT FROM
WHERE	WHERE = θ	WHERE	WHERE = θ
JOIN	JOIN = θ	JOIN	JOIN = θ
SQL	SELECT = θ	SQL	SELECT = θ

In WDT (WDT) additional support provides better matching of queries WDTs. The WDT optimizer processes matches the generated member user when creating the WDT to the generated member's local selection predicate of the user query. The optimizer verifies that the values of the generated members are the same throughout the WDT set for each. The WDT matching algorithm also attempts to match other predicates in the WDT and the query are not exactly the same. For example, if the WDT have predicate $MEMBER = \theta$ and the query has the predicate $MEMBER = \theta$, the WDT contains the data necessary to use the query. The WDT is used in the query. The predicate $MEMBER = \theta$ is a full set local selection in the query. Therefore, values $MEMBER$ must be defined in the WDT.

Designing WDTs that are based on queries

When designing an WDT for a set of queries, it is important to account for all the columns projected in the set of queries. Either the WDT must include all the columns used in the queries or must include the join columns to facilitate gathering the columns through a join to the appropriate table.

For example, the following grouping query returns rows that is single table that match Year = 2000 and Month = July, containing two columns (Date and Quantity) and grouping by Year, Month, Day. The WDT designed to assist this query must contain the projected columns Year, Month, Day, MonthDay, and MonthQuantity. If one or more columns are omitted, the WDT is not useful.

Original	WDT
SELECT	SELECT
FROM	FROM
WHERE	WHERE
GROUP BY	GROUP BY
ORDER BY	ORDER BY
SQL	SELECT = θ

In another example that specifies a join with groups (customer and tables), either of the HDT designs can be used to resolve the query:

- An HDT that is based only on **Trans, Table**
- An HDT that is based on both **Trans, Table** and **State, Table**

SELECT	<ul style="list-style-type: none"> ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State ➤ Trans, Tables, State, Tables
FROM	<ul style="list-style-type: none"> ➤ Trans, Tables ➤ Trans, Tables, State
WHERE	<ul style="list-style-type: none"> ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State
GROUP BY	<ul style="list-style-type: none"> ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State

An HDT that is based only on **Trans, Table** must include the **Customer**, **Customer** and **Customer** columns. The query optimizer can specify a join between the HDT and **State, Table**, representing the join based on **Customer**.

An HDT that is based on both **Trans, Table** and **State, Table** must project the **Trans, State** and **Table** columns from **State, Table** and the **Customer** and **Customer** columns from **Trans, Table**. The query optimizer can use the join which because the HDT contains all the data required to complete the query.

In another example that specifies a join between three tables, either of the HDT designs can be employed to resolve the query:

- An HDT that is based on all three tables
- An HDT that is based on two tables

SELECT	<ul style="list-style-type: none"> ➤ Trans, Tables, State ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State ➤ Trans, Tables, State, Tables
FROM	<ul style="list-style-type: none"> ➤ Trans, Tables, State ➤ Trans, Tables, State, Tables
WHERE	<ul style="list-style-type: none"> ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State ➤ Trans, Tables, State, Tables
GROUP BY	<ul style="list-style-type: none"> ➤ Trans ➤ Tables ➤ State ➤ Table ➤ Trans, Tables ➤ Trans, Tables, State ➤ Trans, Tables, State, Tables

An HDT that is based on all three **Trans, Table, State, Table** and **Customer, Table** tables must project the **Trans, State** and **Table** columns from **State, Table**, **Customer** from **Customer, Table** and **Table** and **Customer** from **Trans, Table**. **State** and the **Customer** and **Customer** join columns are specified in the HDT's query definition such as the **WHERE** clause, but the columns are not part of the HDT table. The query optimizer can use the join which because the HDT contains all the data required to complete the query.

An MDT that is based on the `Track`, `Table` and `State`, `Table` tables must project the `Location`, `Table` and `Quantity` columns from `Track`, `Table` and the `Track`, `Table` and `Time` columns from `State`, `Table`. The `Building` column is built by joining the MDT to `Building`, `Table`.

An MDT that is based on the `Track`, `Table` and `Customer`, `Table` tables must project the `Location`, `Table` and `Quantity` columns from `Track`, `Table` and the `Customer` column from `Customer`, `Table`. The `Building` column is built by joining the MDT to `State`, `Table`.

In both cases, the query optimizer can add the join between `cust` and `table` because the MDT contains all the data required for that join of tables. The inclusion of the other table's join columns allows the MDT to be joined to a table not represented in the MDT.

The base table design, which includes a flat table's join columns, allows additional queries to use the expensive MDT. Query 4 uses the flat table's number of rows groups in the MDT right join query based on the number of table's join columns values (such as the join columns specified in the `GROUP BY` clause).

Designing MDTs that are based on data models

MDTs designed for a star schema or snowflake schema data model can be based either on a single flat table or on flat tables plus one or more dimension tables. If only flat tables is a constraint, then foreign keys for one or more dimension tables must be included in the MDT. The allow for selection on the dimension table joining from the dimension table to MDT, and grouping of the MDT table.

The grouping of a foreign key column represents the most detailed level within a given dimension. The more flat table's additional foreign key that is added to the MDT as a grouping column results in more groups and less effectiveness. For example, if the `Location` column represents 1000 distinct days or groups, adding the `Building` grouping column can result the number of distinct groups by the number of distinct `Building` values. If 100 values have a `Building` column, this will result in the representation of 100,000 distinct groups in the MDT (1000 days * 100 values) (see Figure 6).



Figure 6: MDT join table

If the fact table and one or more dimension tables are partitioned, the results are MDTs that demonstrate the fact model. By including all the partition columns in the MDT, the database engine can avoid joining the tables together (see Figure 5).



Figure 5. MDT with Tables and Part

Designing MDTs based on Hierarchies

When creating an MDT, selecting the appropriate grouping columns can mean the difference between an effective or useless MDT. In a data model with complex hierarchies defined, the grouping columns specified can allow the MDT to cover grouping columns at that level or higher in the hierarchy. This can be a very useful way to minimize the number of MDTs that must be created and maintained, though all parties require benefits that you appreciate at all.

In the following example, an MDT that is created with the Year, Quarter, Month, Country, Week, Product, County, City, Category, Department grouping columns can be used for that level or the one level above (such as Year, Quarter, County, Category). It goes that specific grouping should have been defined in the MDT to not attempt to use the MDT such as Year, Week, State, Part, (see Figure 6).



Figure 1. MCF coverage

Another issue is that our MCF can only register repeated requests for a list of all values where the set of values is static or changing slowly over time. The following query would retrieve all values and return a subset of attributes when:

```
SELECT * FROM table;
SELECT * FROM table;
```

If the table contains millions of rows, the user and the database processing are affected in terms of time and resources. Given that most database functions values are not updated frequently, this is a good opportunity for an MCF. By creating an MCF that contains the values for attributes values, the query optimizer can use the MCF to satisfy the query with little time and resources.

```
SELECT * FROM table;
SELECT * FROM table;
SELECT * FROM table;
SELECT * FROM table;
SELECT * FROM table;
SELECT * FROM table;
```

Instead of reading and processing millions of rows in the transaction table, here is a list of all rows and their attributes from the MCF. Furthermore, the MCF only needs to be refreshed when a new dataset function is added or removed.

Although MCFs can be partitioned, the MCF (MCF) query optimizer does not explicitly use the MCF. Changing partitioned MCFs is not recommended in an MCF (MCF) environment. In MCF (MCF), the optimizer can handle the query when the base table or the MCF is partitioned.

It is important to keep in mind that implementing MCFs is not free. Besides the time and resources to perform maintenance, the optimizer itself is a great query processor in the optimizer execution time and uses MCFs. It is a basic change when MCF is used that provides the other coverage and the largest benefit.

In general, consider creating **INDEX**s for the following query classes:

- 1. Queries with grouping and aggregation, where the table of rows is grouped by high
- 2. Queries with sorted tables, where the table of rows is sorted either by high
- 3. Queries with joins, where the number of rows is high and the join is not

Creating **INDEX** definitions

An **INDEX** definition consists of a query and the attributes that are associated with the predicate, conditions, and use of the table. There are two methods for creating an **INDEX**: **CREATE INDEX** or **ALTER TABLE**. Other methods can be obtained through our SQL, required in the using the graphical user interface of **SQL** Editor Navigator.

The **CREATE INDEX** method allows for the creation and population of an attribute. The **ALTER TABLE** method allows an existing table to be modified into an **INDEX**.

CREATE TABLE example

Here is an existing table Example, Transactions, Table with the appropriate column definitions:

```
CREATE TABLE Example (ID INT,
                        Column1 VARCHAR(255),
                        Column2 VARCHAR(255),
                        Column3 VARCHAR(255),
                        Column4 VARCHAR(255),
                        Column5 VARCHAR(255),
                        Column6 VARCHAR(255),
                        Column7 VARCHAR(255),
                        Column8 VARCHAR(255),
                        Column9 VARCHAR(255),
                        Column10 VARCHAR(255),
                        Column11 VARCHAR(255),
                        Column12 VARCHAR(255),
                        Column13 VARCHAR(255),
                        Column14 VARCHAR(255),
                        Column15 VARCHAR(255),
                        Column16 VARCHAR(255),
                        Column17 VARCHAR(255),
                        Column18 VARCHAR(255),
                        Column19 VARCHAR(255),
                        Column20 VARCHAR(255),
                        Column21 VARCHAR(255),
                        Column22 VARCHAR(255),
                        Column23 VARCHAR(255),
                        Column24 VARCHAR(255),
                        Column25 VARCHAR(255),
                        Column26 VARCHAR(255),
                        Column27 VARCHAR(255),
                        Column28 VARCHAR(255),
                        Column29 VARCHAR(255),
                        Column30 VARCHAR(255),
                        Column31 VARCHAR(255),
                        Column32 VARCHAR(255),
                        Column33 VARCHAR(255),
                        Column34 VARCHAR(255),
                        Column35 VARCHAR(255),
                        Column36 VARCHAR(255),
                        Column37 VARCHAR(255),
                        Column38 VARCHAR(255),
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                        Column88 VARCHAR(255),
                        Column89 VARCHAR(255),
                        Column90 VARCHAR(255),
                        Column91 VARCHAR(255),
                        Column92 VARCHAR(255),
                        Column93 VARCHAR(255),
                        Column94 VARCHAR(255),
                        Column95 VARCHAR(255),
                        Column96 VARCHAR(255),
                        Column97 VARCHAR(255),
                        Column98 VARCHAR(255),
                        Column99 VARCHAR(255),
                        Column100 VARCHAR(255))
```

To create a new MDT by using Studio Navigator - Database (Figure 6)

1. Navigate to and open a Solution
2. Right-click Models
3. Select New > Model-based Query Table

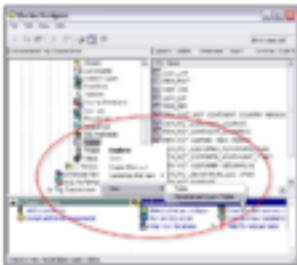


Figure 6. Create a new MDT using Studio Navigator - Database

The following are steps to specify the MDT definition and collection.

As 3NF Table 2 example

Given an existing **Inventory_Inventory_Table** table with the appropriate column definitions

```
CREATE TABLE Inventory_Inventory_Table
(
    InventoryID INT PRIMARY KEY,
    InventoryName VARCHAR(100),
    InventoryType VARCHAR(100),
    InventoryStatus VARCHAR(100),
    InventoryLocation VARCHAR(100),
    InventoryQuantity INT,
    InventoryUnit VARCHAR(100),
    InventoryPrice DECIMAL(10,2),
    InventorySupplier VARCHAR(100),
    InventoryLastUpdated DATETIME
);
```

To alter an existing table to become a new 3NF using **Studio Designer - Database**

1. Navigate to and open the Database
2. Open Tables and right click on a specific table
3. Select **Refactor** and select the **Materialized Query Table** sub-menu (Figure 10)

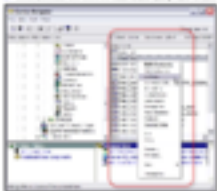


Figure 10: Alter existing table to become a new 3NF using **Studio Designer - Database**

Next step up the table to register the table as an 3NF and specify the 3NF attributes and definition

Analysis of an SQL

Instead of relying on function call SQL, it is important to understand the quality of the SQL statement used to create an SQL or SQL for SQL. Different classes create the generation and maintenance of SQL, and some functions a SQL supported user figure 15.



Figure 15: Analysis of an SQL

SQL is a declarative query language and SQL statements are executed by the database engine. SQL does not automatically map the SQL's operations with the base tables. When the base tables change, there can be a difference between the contents of the SQL's and the base tables. This difference represents the data latency.

The use of specific SQL words, sometimes a helps to speed identification during analysis and administration. Consider using SQL keywords in the table name.

When trying out the individual query activities, it is a good practice to specify all the columns that are available within table or database. In a database, these columns need to be used all the SQL function. As appropriate, other functions such as `MAX`, `MIN`, `AVG` and `COUNT` can be specified. Specifying these columns and functions right after the SQL to function with individual and used.

The number of rows per query provides the query optimizer with additional ways to take advantage of the SQL (for example, identifying strategies for query optimization). It is always a good practice to provide the function `COUNT(*)` in the `SELECT` clause of the SQL definition. It is equally to calculate the average of all table columns, for example, `AVG(column)` column in the query. `COUNT(column)` column is available in the SQL, not just `COUNT(*)`.

```

SELECT
  FROM
  WHERE
  ORDER BY
  LIMIT
  OFFSET

```

When an SQL statement is defined, the following select statement restrictions apply:

- The select statement cannot contain a reference to another SQL statement that refers to an SQL statement.
- The select statement cannot contain a reference to a database global temporary table, a table in `SYSTEM`, a program object, or a user table, except for a `TABLE` clause.
- The select statement cannot contain a subquery or view that contains or includes a subquery or view.
- The select statement cannot contain an expression with a `TABLE` or a column name that is based on a database view that has a `TABLE` clause.
- The select statement cannot contain a table column that has not an SQL identifier, such as binary or decimal `NUMERIC` or `DECIMAL`.

When an SQL statement is defined with the `TABLE` clause, the following additional select statement restrictions apply:

- It must include the `TABLE` clause.
- It must include the `TABLE` clause in the `TABLE` clause.
- The `TABLE` clause is allowed, but is only used by `TABLE`. It might improve the locality of the table reference in the SQL.

Additional information on SQL statement support can be found in the publication `SQL Statement Database for Oracle Database`. Appendix B provides a list of the SQL statements that are supported in the Oracle Database. Contact your sales representative for more information.

When creating an SQL statement, the select statement and population of table columns are part of the object creation request, or database object creation. If the `TABLE` clause is specified, the SQL statement is created as part of the create phase. If the `TABLE` clause is specified, the SQL statement is created as part of the create phase. If the table is already defined, the statement and population can be defined by using the `TABLE` clause, or the process can be defined and controlled by the programmer by defining the user can determine the location and structure for creating and populating the SQL.

When creating a table in the SQL, the object table can be empty or fully populated. When creating an empty table that contains data, the user is responsible for the integrity and accuracy of the data.

When creating the SQL, it is advantageous to verify whether the user specified table creation using the SQL's control of the table table. A simple method for doing this verification is to provide the user with a few queries. The `TABLE` clause is used to create the table. If the user specifies the table table with the SQL, the SQL is shown a table of data or some data table in the user's table. The `TABLE` clause is used to create the table. An example of such a query plan is shown later in the "Testing and Tuning Subqueries Using Tables" section of this paper. If the user specifies the SQL, further analysis and changes can be done prior to SQL population.

Populating MQTs

Calculating and populating the MQT data is a time- and resource-intensive operation because of the fact that creation of MQTs normally requires accessing all the data in the base tables and aggregating column data over potentially many groups.

Creating an MQT might result in reading and processing millions or billions of rows.

Whether the application is under the control of the database engine or the programmer, the queries that it uses to populate the MQT must be tested.

Indexes to support MQT creation

One to five indexes on MQTs, depending on the query indexes on the base tables, is a critical success factor. The following guidelines must be followed when analyzing the MQT's query:

- Create table and secondary index indexes for any local selection columns.
- Create table indexes for all join columns.
- Create table indexes for all grouping columns.

For this **ORACLE** MQT example:

```
CREATE TABLE EMPLOYEE MQT AS
  SELECT
    EMPLOYEE_ID,
    LAST_NAME,
    SALARY,
    DEPARTMENT_ID,
    JOB_ID,
    HIRE_DATE,
    MANAGER_ID,
    EMPLOYEE_ID AS GROUPING_COLUMN1,
    DEPARTMENT_ID AS GROUPING_COLUMN2,
    SALARY AS GROUPING_COLUMN3,
    LAST_NAME AS GROUPING_COLUMN4,
    JOB_ID AS GROUPING_COLUMN5,
    HIRE_DATE AS GROUPING_COLUMN6,
    MANAGER_ID AS GROUPING_COLUMN7,
    EMPLOYEE_ID AS GROUPING_COLUMN8,
    DEPARTMENT_ID AS GROUPING_COLUMN9,
    SALARY AS GROUPING_COLUMN10,
    LAST_NAME AS GROUPING_COLUMN11,
    JOB_ID AS GROUPING_COLUMN12,
    HIRE_DATE AS GROUPING_COLUMN13,
    MANAGER_ID AS GROUPING_COLUMN14
```

Build the following index:

```
CREATE INDEX EMPLOYEE_INDEX ON EMPLOYEE
  (EMPLOYEE_ID, DEPARTMENT_ID, SALARY,
  LAST_NAME, JOB_ID, HIRE_DATE, MANAGER_ID)
```

This provides the query optimizer and database engine with statistics on the grouping columns and provides an index for implementation. Provided



Figure 4: Creating WCF reader

Note that WCF subscribers cannot be based on other WCFs. The process of creating the WCF reader involves some programming intervention. That is to say, the WCF subscribers must reference the base table for the table reader population the WCF is from a previously created WCF.

Figure 5 shows the general steps to create WCFs in a cascading fashion:

1. Create the reader WCF in the hierarchy and populate it from the base table.
2. Create the next WCF in the hierarchy with the `SQLite WCFName/CREATEWCF` attribute and populate the WCF from WCF created in step number 1.
3. Create the next WCF in the hierarchy with the `SQLite WCFName/CREATEWCF` attribute and populate the WCF from the WCF created in number 2.



Figure 5: Steps in creating WCFs in a cascading fashion

For this process to be successful, all the WCF subscribers in the hierarchy must reference and be based on the same initial table.

Strategies and methods for aggregation

The query optimizer has two basic methods of grouping data for aggregation:

- Grouping with an index (permanent or temporary)
- Grouping with a hash table

Each method has its own requirements, characteristics and advantages. Understanding and anticipating the use of either method determines whether programmer intervention is required to speed up the SQL operation.

The optimal use of either grouping strategy requires the optimizer to have a good understanding of the estimated selectivity of the query (usually, CPU) and more importantly, a good understanding of the estimated number of groups and the average number of rows per group. This information comes from indexes and column statistics.

For hash grouping to be an optimal strategy, the optimizer and database engine need enough memory to fit every join's groups in memory for hash table. A large number of distinct groups results in a larger hash table, and a larger hash table requires a larger memory cost to partition efficiently. If the optimizer expects the join's set of rows to be smaller than the estimated hash table size, the hash grouping strategy is optimal. When grouping with a hash table, the ability to read and group the data is parallel with SMP is available. This feature allows grouping queries to perform faster by using more resources.

Index grouping is the preferred strategy when hash grouping is not viable. The memory footprint of using an index can be much smaller than creating an index hash table. For index grouping to be optimal, a permanent index that covers the grouping columns is required. Without a permanent index available, the optimizer has to create a temporary table structure, known as an index set. This table structure is the query execution plan.

When grouping with an index, you cannot read and group the data in parallel through SMP in other words. For index to read for grouping, SMP does not help the aggregation go faster. The creation of a temporary index set can impact SMP.

If index grouping is required, and the SMP calculation and prediction is not meeting response time expectations, programmer intervention is required. This might include the form of adding a specific partition number for better advantage of various forms of partition.

Parallel execution is not available when the database engine is writing the aggregated data to the SQL through a table that requires. For example, when extracting the groups from the hash table, the data is inserted serially. If the SQL is to be executed with many rows back to groups, then designing a parallel SMP calculation and prediction process is a prerequisite. It is a good practice to understand the optimizer's strategy for aggregation jobs is covering the SMP available space in the production environment. This can be accomplished by using the query optimizer's feedback and the Oracle Navigator - Visual Explain.

The SMP available and available space can be explained only by using the Oracle Navigator - Visual Explain table. This allows the query optimizer's feedback to be analyzed without actually running the query.

Figure 11-10-10 An example of grouping with a permanent table as base through Visual Explain



Figure 11-10-10-1 Grouping with a permanent table

Figure 11-10-10-2 An example of grouping with aggregation through a temporary table with DDP capabilities as base with Visual Explain



Figure 11-10-10-2-1 Grouping with aggregation with a temporary table with DDP capabilities

When the grouping columns are from more than one table, the selection and joining of rows from the base tables occur before any grouping with aggregation. A single permanent table does not cover all the grouping columns. In this case, either a temporary table with a temporary index or a user table builds the grouping with aggregation of rows.

Figure 17 shows an example of grouping with a temporary column for an object with three objects.



Figure 18 shows an example of grouping with a temporary column for an object with three objects.

Figure 19 shows an example of grouping with a temporary table for an object with three objects.



Figure 20 shows an example of grouping with a temporary table for an object with three objects.

When the grouping columns are from only one table, the table can be placed from the join table. However, the case is the first table with a permanent table that covers the grouping columns, because the case is grouping table. After the join, the table can be aggregated directly.

Figure 11: Example of grouping with a participant who is done with their English



Figure 11: Example of grouping with a participant who

Programmatic intervention: do it yourself

In cases where the WJF member gets stuck on a topic, or the staff set back an answer that is to be placed into the WJF's log, some programmatic intervention might be helpful in speeding up the calculation and evaluation process. In cases where the WJF member is found on a topic which has itself been long the end of their course, some programmatic intervention might be helpful in speeding up the process (see Figure 16).

Without programmatic intervention some WJF member queries might not be many hours, or a certain weeks, into the future.



Figure 16: Number of programmatic interventions over time

The basic data-pulling, aggregation and transfer is parallel. When the database engine is unable to supply DDP requests, the programmer can design and implement a parallel process. The process consists of reading the data from the base tables and pulling it into logical ranges, selecting and processing each range in parallel, and transferring the aggregated data into the DDP's parallel case (Figure 16).



Figure 16. Current DDP aggregation

The design process starts with pulling the data represented by the first row or two grouping columns. Usually, the initial stages of data define the first row or two grouping columns and compare this number to the number of processors available during query execution. When the number of processors is the number of ranges obtained the best of parallelism is applied. The goal is to have all the processors as busy as possible. The remaining throughput and remaining time is passed to DDP. Using all the processing resources is required as DDP assumes all the resources are available to this activity. If other jobs are running on the system, the degree of parallelism needs to be reduced.

For example, given a customer transaction table with grouping columns of Year / Customer where there are three years, and thousands of customers represented in the data, three degrees of parallelism can be used. The parallel grouping queries can each select and process one of three years. In a system with three or more available processors, each grouping query runs in one processor. If parallel DDP resources are available and a larger degree is feasible, then additional grouping queries can be used, each processing a given year and respective range of customers.

When running in all of parallel at the same time, the DDP DDP degree must be set to **100%**. Furthermore, ensuring that all the data ranges represented in the base tables are processed. It is easy to create ranges, resulting in incomplete data.

Employing a parallel process

Here is an example of creating and executing an RDD with a parallel process. First, you create the RDD and then collect the `foreachPartition()` results.

```
scala> val rdd = ParallelCollection(1 to 10)
rdd: org.apache.spark.rdd.ParallelCollectionRDD[0] = org.apache.spark.rdd.ParallelCollectionRDD[0]
rdd.foreachPartition { (iter, _) => {
  iter.foreach { i => {
    println(i)
  }
}
}
1
2
3
4
5
6
7
8
9
10
```

Depending on the processing resources available, `foreach` will use a set of parallel partitions. Each partition needs to collect a subset range of items, appropriate for the task, and insert the results into the RDD.

```
scala> val rdd = ParallelCollection(1 to 10)
rdd: org.apache.spark.rdd.ParallelCollectionRDD[0] = org.apache.spark.rdd.ParallelCollectionRDD[0]
rdd.foreachPartition { (iter, _) => {
  iter.foreach { i => {
    println(i)
  }
}
}
1
2
3
4
5
6
7
8
9
10
```

Since the `SELECT` predicate selects columns of data from the base table, the query provides greater relevance over the local selection columns. This provides the query optimizer and database engine greater flexibility. Choosing both table and selected column columns can be advantageous. Using the previous example, the greater relevance to provide an example:

```
SELECT * FROM CUSTOMERS WHERE
  ID IN (10000000000000000000, 10000000000000000000);

SELECT * FROM CUSTOMERS WHERE CUSTOMER_ID IN (10000000000000000000)
  OR CUSTOMER_ID IN (10000000000000000000);
```

We can to test and verify the queries. The parallel process and the query results before relying on the `SQL`. The integrity and accuracy of the `SQL` table is the responsibility of the programmer.

Testing and tuning `SQL`'s

If the query optimizer receives the user query to access an `SQL` instead of the base table, the query logic, methods and strategies are employed to access and process the `SQL`. Specifically, the `SQL`'s are queried and queried to local selection, joining, grouping and ordering. It is important to test and use the `SQL`'s prior to relying on them in a production environment. The key to great query performance is a proper indexing and statistics strategy.

The indexing and statistics strategy for `SQL`'s is essentially the same as the base tables. That is to say, all indexes on the `SQL`'s. The query optimizer has the necessary statistics and has many choices when implementing the query request.

Column statistics used by `SQL` are collected and stored on a table by table basis, and the indexes `SQL`'s. `SQL`'s automatically update a column table for an `SQL`. It is a good practice to check these indexes periodically and to ensure that the statistics are updated after inserting or refreshing the `SQL`. The indexes your query performance immediately after the `SQL`'s are updated.

Since that the `SQL`'s can be selected, joined, grouped and ordered, indexes need to be created to cover these activities. To determine the proper set of indexes, analyze the table model and test the queries with the `SQL`'s tables. The use to create indexes on any local selection columns and join columns, in addition, consider creating indexes on any grouping columns and ordering columns, especially if the `SQL` has many rows.

Separate it provides a link to the `SQL` object creation reader which you can find more information on indexing and statistics strategies use the `SQL`'s (see the `SQL` reader file).

Enabling `SQL` support

The explicit configuration and use of `SQL`'s by the query optimizer must be explicitly enabled. The query optimizer's default behavior is to ignore `SQL`'s.

The `SQL` behavior of the `SQL` also affects whether the `SQL` is considered. The statistics that allow the configuration and usage of `SQL`'s are:

- `STATISTICS_ENABLE_SQL_TABLES` through the `SQL`'s or by `SQL` statement
- `SQL_TABLES_STATISTICS_ENABLE` through the `SQL`'s or by `SQL` statement
- `SQL_TABLES_STATISTICS_ENABLE` through the `SQL`'s or by `SQL` statement

Creating a process plan for **SQLSTATE 38002** to update the **SQL**. The **SQLSTATE** for update **SQLSTATE 38002**, **UPDATE TABLE**, **SQLSTATE 38002**, must be set to **SQL**.

Appendix B provides a broader description of these **SQLSTATE** update and all of the possible values. The query runtime environment affects the optimization and use of **SQL**. The **SQL** is recommended:

- The administrator must specify **SQLSTATE 38002** as a **SQLSTATE** error.
- The base table to be updated with an **SQL** must not be update or table together with the **SQL**.

Feedback on the query's use of an **SQL**

Only an optimization feedback to determine whether the query optimizer has modified the user query to use an **SQL**. This only on the optimizer feedback to take the access and processing of **SQL**. For example, the user query has been selected, and the query is modified to use an **SQL**. The user query might also have been selected. The **SQL** may also contain the appropriate reference information. The situation can be identified and resolved by analyzing the feedback.

Feedback on **SQL** optimization and usage is provided through Status Navigator - Query Explorer and the SQL performance monitor database feedback.

Figure 17 shows an example of locally executing a query together to use an **SQL**.



Figure 17 Example of locally executing a query together to use an **SQL**

In **SQLSTATE 38002**, Query Explorer does not explicitly highlight the use of an **SQL**. Look for use or state of the base table to be updated with an **SQL**. Using a working connection can help identify **SQL** present in the query plan.

In MSN 2005, Visual Explorer is enhanced to provide an option to highlight an MSN in Explorer. The option identifies the use of MSN in each node.

To highlight an MSN in the query plan using Visual Explorer, see Figure 30.

- 1. Click the Show icon.
- 2. Select Highlight Materialized Query Plan.



Figure 30. Example of highlighting an MSN in the computer.

After selecting the highlighting option, any PDF text the pointer has hovered will no longer be highlighted (see Figure 20).



Figure 20: Example of highlighting text in a PDF document with an orange background.

is written to these tables. The related member tables automatically reflect the optimization and use of MWPs. Specifically, the following information is provided:

- **DBP_Merger** table contains information about the MWPs that were merged. It also records which old MWPs are deleted and which are used (the table specifies the new).
- **DBP_MWP** table contains the data for all active or **A** status activities that are MWPs (merged tables). The reporting information is based on the MWPs instead of the base tables.
- **DBP_Merger** table also contains the following data: **source**, **target**, **old**, **new**, and **deleted** and **DBP_ID** contains the **DBP** values **DBP**, **DBP_M**, **DBP_C**, **DBP_C**, **DBP_C**, **DBP_C**.
- **DBP_MWP** table contains **DBP** across plan that needs to be added because the MWPs are merged tables.

Example query to determine about MWPs are used in what queries:

DBP_Merger	SELECT * FROM DBP_Merger
DBP_MWP	SELECT * FROM DBP_MWP
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger
DBP_Merger	SELECT * FROM DBP_Merger

Note that the summary member is primary-based and does not reflect the logical use of MWPs.

Additional information on this Performance Monitor and the member data can be found in the publication **DBP Overview Database for Client Database Performance and Query Optimization**. Appendix B provides a link to the DBP to Client Performance Information Center where you can find this report.

Another way would be implemented by DBP (DBP) is the ability to list all the MWPs in a given table and evaluate the usage of the MWPs. For example, it is not possible to determine what a given MWP is used by the optimizer and how many times it is used. Conversely, it is possible to determine that a given MWP has never been used.

To view MDTs that are based on an existing table, use Studio Navigator - Database (see Figure 10).

1. Navigate to **Database**.
2. Open **Tables** and right-click a specific table.
3. Select **View Metadata Query Tables**.

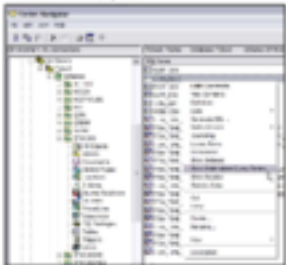


Figure 10. Using Studio Navigator - Database to view MDTs.

Note that the statistics can also be accessed through an application programming interface (API).

The **Last Query Use** column shows the frequency when the WDT was last used by the optimizer to optimize user-specified tables in a query.

The **Query Use Count** column shows the number of instances the WDT was used by the optimizer to optimize user-specified tables in a query.

The **Last Query Statistics Use** and **Query Statistics Use Count** columns are currently not updated, as planned. The query optimizer does not use WDT in its updates.

Designing WDT refresh strategies

WDTs (WDTs) do not automatically keep WDTs synchronized with the base tables. When the base tables change because of insert, update or delete activity, there can be a difference between the contents of the WDTs and the base tables. The difference represents the data delta.

It is the responsibility of the user or programmer to refresh or maintain the WDT data. It is often easier to WDTs that are not maintained as the base tables change. The query results increasingly change from the results obtained after querying the base tables directly.

The standard method to refresh an WDT is shown by **REFRESH TABLE** statement.

```
REFRESH TABLE [options] wdt;
```

We assure that this is a full refresh of the WDT and assess the following:

- Any indexes on the WDT are removed.
- The contents of the WDT are removed.
- The underlying WDT query is run.
- The WDT is synchronized with the base tables.
- Any indexes on the WDT are recreated.

The same considerations apply to refreshing or maintaining an WDT as the initial creation and population. How frequently, how often and whether available to refresh or maintain the WDT might be restricted because of other data processing. Programmer discretion might be necessary or advantageous to create an appropriate WDT refresh strategy that meets the business and technical requirements.

Some data and query environments find themselves stuck in a periodic WDT refresh process. The business requires querying against a segment of data, an WDT can be used. The WDT will only need to be refreshed when that periodic data changes. For example, if a report can be based on a 60 day and monthly run, the WDT that contains the aggregated data representing four months can be refreshed as part of the month-end processing.

If you still use other environments where data is loaded on a periodic basis. These periodic provide a natural opportunity, and to some extent, a requirement to refresh or maintain the WDTs. If the ETL process runs on a full load (the complete underlying processing), the WDT refresh or maintenance strategy can also occur on a daily basis. Refreshing the entire WDT table set to incorporate any new data worth of data might be inefficient. In this case, the WDT can be maintained (or refreshed) by refreshing from existing or updating rows with the row-id appropriate. Any WDTs that are based on historical can be refreshed or maintained using the same table, or they can be fully refreshed using the existing table or dataset either. This figure model as the base tables can provide a mechanism for refreshing the WDT maintenance. As the base tables change, the appropriate WDTs can be changed or not.

When using a process other than **PERFORMANCE TUNING** to perform a full refresh of the WQT, it's a good practice to do the following steps:

1. Download only relevant statistics on the WQT
2. Apply the selected level of compression (recommended by **WQMT**)
3. Drop any indexes on the WQT
4. Rebuild all the stats on the WQT
5. Calculate the appropriate and complete for WQT
6. Create any indexes on the WQT
7. Refresh only relevant statistics on the WQT

It is important to read and verify the WQT refresh or maintenance process before using it in a production environment.

Testing and tuning WQT refresh strategies

Test your refresh and the WQT refresh and maintenance processes and test to meet or exceed the requirements requirements. If the refresh time exceeds the processing window, the WQT creation, merge and maintenance strategy must be modified, or additional processing resources must be supplied. For example, a full refresh strategy might be too time consuming, but an incremental maintenance strategy is acceptable. When implementing a maintenance strategy based on immediate changes to the underlying base tables, be sure to measure and understand any additional time and resources required. In other words, the refresh maintenance of WQTs increases the time necessary for the refresh, update and table operations on the base tables.

Using the SQL Performance Monitor and Collector format tools can assist you with understanding the resource overhead and resource allocation of the refresh and maintenance process. The tool can give the user view of the available resources in a clearly understandable manner. For example, when running refresh processing in parallel, processor resources are consumed, a higher degree of parallelism can be used to help increase throughput. On the other hand, if all of the available resources are fully used, the parallel degree is adequate, or too high. All of this information is relevant configuration when the SQL configuration creates user data tables in capable of supporting the processing available.

Planning for success

How to create WQTs, gather some baseline metrics on what SQL requests are used. You also need to determine the frequency of the WQT updates and how the updates are optimized and set. You can use Query only tools such as the SQL Performance Monitor. Also, gather some baseline information on how the system resources are used. Collector Services is a good tool to use for this. After implementing WQTs, the baseline information can be used to quantify any differences it indicates a performance of the application.

Successfully variable testing is a method to test any WQT creation, refresh and merge strategies. A good place to start is a test environment or proof of concept for the WQT development and Proof of Concept Center in Redwood. Microsoft or Microsoft Azure, understanding the cost and benefits of WQTs before implementing a production environment in a critical access factor. Make. Appendix B provides a link to the following for more information about the WQT development and Proof of Concept Center.

Summary

With the latest version of 2021 to 2023, 2021 continues to deliver additional features and functionality to assist with the implementation of school data system applications. The ability to create and use 2021 profiles and transfer options for high performance system processing. This update along with the dimensional profiles, provides more guidance and insight on using the new features.

Appendix B provides a table detailing for the latest information regarding 2021 to 2023 support of 2021

Appendix A: SQL query engine details

After Oracle 11gR2, a newly implemented SQL query engine was introduced. This new query engine is referred to as SQL. The original query engine is referred to as the legacy SQL engine (SQL). Initially, a small subset of queries were optimized and run by SQL. With the availability of Oracle 11gR2 and 11gR3, many more queries are optimized and run by SQL. Only some queries optimized by SQL can explicitly use SQL. Only SQL can be query-defined optimized by SQL, can be explicitly used in the query plan.

SQL: new features

In Oracle 11gR2, SQL is not capable of optimizing and executing queries that contain or use

- SQL predicates
- SQL columns
- Column functions such as `DECODE`, `CONCAT` and `TRIM` conversions
- Aggregate indexing expressions and cost expressions
- `SQLTYPING` hints and `SQLTYPED` clause
- Dead Hitters
- Logical condition
- Logical file references
- References to tables in physical files that have `Subst` hint logical files
- References to tables in physical files that have logical files with support in shared heap
- Distributed tables
- Non-SQL interfaces such as the `SQLTYPED`, `SQLTYPED` and `SQLTYPED` hints

In Oracle 11gR3, SQL is not capable of optimizing and executing queries that contain or use

- Column functions such as `DECODE`, `CONCAT` and `TRIM` conversions
- Aggregate indexing expressions and cost expressions
- Dead Hitters
- Logical condition
- Logical file references
- References to tables in physical files that have `Subst` hint logical files
- References to tables in physical files that have logical files with support in shared heap
- Distributed tables
- Non-SQL interfaces such as the `SQLTYPED`, `SQLTYPED` and `SQLTYPED` hints

If the query contains any of these items, Non-SQL is used to optimize and run the query. SQL can not be used to execute it.

Appendix B: Resources

These Web sites provide useful information to supplement the information contained in this document.

- **IEEE Reference Service Information Center**
<http://ieeeref.com>
- **IEEE Publications Center**
<http://www.ieee.org/ieeexplore/publications/publicationslandingpage.jsp?i=171-172>
- **IEEE Headquarters**
<http://www.ieee.org>
- **IEEE Global Headquarters**
http://www.ieee.org/conferences_events/publications/publicationsandstandards
- **Standards Information Center**
http://www.ieee.org/conferences_events/publications/standards
- **IEEE 1010 Series**
http://www.ieee.org/conferences_events
- **IEEE Global Headquarters**
http://www.ieee.org/conferences_events/publications/standards
- **IEEE Reference Service Information Center**
<http://ieeeref.com>
- **IEEE 1010 Series**
http://www.ieee.org/conferences_events
- **Industry and Academic Subgroups White Paper**
http://www.ieee.org/conferences_events/publications/standards/industry_and_academic_subgroups_white_paper
- **Key Activities for Strategists White Paper**
http://www.ieee.org/conferences_events/publications/standards/1010_key_activities_for_strategists_white_paper
- **Standards Management White Paper**
http://www.ieee.org/conferences_events/publications/standards/standards_management_white_paper
- **IEEE 1010 Series**
http://www.ieee.org/conferences_events/publications/standards/1010_series
http://www.ieee.org/conferences_events/publications/standards/1010_series/standards
- **Guidance regarding IEEE support of any IEEE 1010 topic can be found in the following IEEE order**

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