Performance Tuning for the Teradata Database

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Document Changes

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1.0	2010-10-26	All	Initial document

Conventions Used in This Document

The following notational conventions are used in this document:

- Code is shown in Courier New 10 pt: While True Do
- Commands are shown in Courier New 11 pt: Delete

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1 Purpose of this Document

The purpose of this document is to instruct the reader how to optimize the performance of a Teradata system through the use of good physical database design principles.

2 Performance Tuning Considerations

In order to understand the concepts behind performance tuning for Teradata, an understanding of some basic Teradata concepts is required. An understanding of the Teradata primary index and table partitioning is required in order to understand Teradata join processing, which is the most common point of performance contention.

2.1 Primary Index

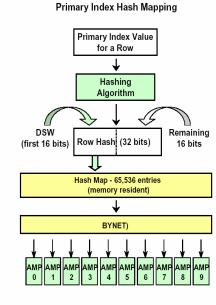
The Teradata primary index is not an index in the traditional sense, as it is not a lookup table. Instead, the primary index is a mechanism that determines where each data row is physically located on the Teradata system. The primary index of a table may be defined as either a single column or as multiple columns. The values of the primary index columns within the table may be unique or non-unique.

It should be noted that the primary index of a table should not be confused with the primary key of a table. The primary index is a part of the physical database model, and affects the storage and retrieval of data rows. The primary key is a part of the logical database model, and uniquely identifies each record in the table. Often, the primary key of a table is a good candidate for the primary index of a table, particularly for smaller "dimension" or "lookup" tables, but this is not always the case for other tables.

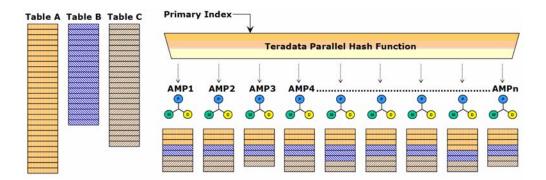
2.1.1 Primary Index Hashing

When rows are inserted into a table, the value of the primary index columns for each row is run through a hashing algorithm to determine the row's location on the system. The Teradata primary index hashing algorithm is "commutative", meaning that the order of columns in a multicolumn primary index does not affect the results of the hashing algorithm.

The data rows are distributed based on the first 16 bits of the hash value, which form the row hash. A hash map allocates each of these 65,536 possible values to one of the **Access Module Processors** (AMPs) on the system, such that the hash values are as evenly distributed throughout the system as possible. This ensures that rows with the same primary index value are always stored in the same physical location on the same AMP, and that tables with evenly-distributed primary index values will be evenly physically distributed throughout the Teradata system.

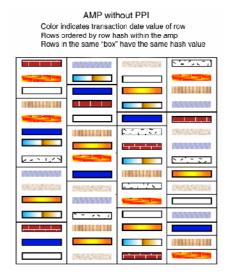


DSW - Destination Selection Word



2.1.2 Partitioning

Teradata provides the option to define a partitioning expression in the table definition. The data in a partitioned table is arranged differently from that in a non-partitioned table. Rows are stored on the appropriate AMP based on the hash value of the primary index, and are then placed in a partition based on the partitioning expression, and then ordered by row hash. This allows for rows with the same primary index value in the same partition to be physically grouped together.



Color indicates transaction date value of row Rows ordered by partition number, then by row hash within partition Rows in the same "box" have the same partition and hash value							
	25, 243,-1						
							
							
× - * × 1							
E = > 1./\							
ZZZZZZ							
57.57							
AGY 13-1							
P44 10-16							
A () () ()							

AMP with PPI

The most common example of partitioning in Teradata is to partition a table on a date column. The partitioning expression may then consist either of individual dates in each partition, or a date range in each partition. Integer values or ranges may also be used in a partitioning expression.

Partitioning can greatly improve the performance of inserts and deletes of entire partitions. Inserting data into an empty partition is extremely fast, because it does not require **transient journaling**. Transient journaling is the process that tracks changed data rows in running queries, so that if an error is encountered during the execution of the query, Teradata can roll back the changes based on the contents of the transient journal. When inserting data into an empty partition, however, the inserts can be rolled back simply by invalidating the partition header of the originally empty partition in the Teradata file system, making transient journaling unnecessary in this case. Deleting an entire partition is even faster, as it does not require transient journaling and does not even require deleting physical rows; it simply requires invalidating the partition header in the Teradata file system.

Partitioning also improves the performance of SELECT statements where a partition value, range of values, or group of values is specified in the WHERE clause. This allows Teradata to avoid a full-table scan of the underlying table, as it instead scans only the relevant partitions.

2.1.3 Primary Index Selection Criteria

There are three main criteria that should be considered when selecting a primary index for a table on Teradata. The primary index should provide a good access path to the data, it should provide good distribution for the data, and the volatility of the data values should be minimal.

2.1.3.1 Access

One criterion for selecting a primary index for a table is to provide the best access path to the data. If individual records are commonly selected from the table, the primary key of the table may be a good choice for the primary index. If the table is typically joined to other tables on a particular column or group of columns, the join columns may be a good candidate for the primary index of the table.

2.1.3.2 Distribution

Another criterion for selecting a primary index for a table is to provide good, even data distribution across the system in order to optimize parallel processing. Evenly-distributed data allows the optimal use of Teradata parallelism for both ETL performance and full-table scan performance. Unique primary indexes will always provide excellent data distribution. Non-unique primary indexes can also provide good data distribution depending on the data.

Note that due to the nature of the Teradata hashing function, a series of consecutive numeric values (e.g. account numbers), if implemented as a CHAR or VARCHAR data type, will tend to generate a large number of hash collisions that may negatively impact query performance. For this reason, it is suggested that numeric values be implemented as numeric types such as INTEGER or DECIMAL, which do not encounter this problem.

2.1.3.3 Volatility

Another criterion for selecting a primary index for a table is to minimize volatility of the primary index data values. When the primary index value of a row is changed, this requires moving the entire row to the new destination AMP. Choosing a column with stable data values reduces the overhead of data maintenance.

2.1.3.4 Primary Index Selection Example

As an example of these principles, consider an Order table keyed by the column Order Number, and a Line Item table keyed by the two columns (Order Number, Line Item Sequence Number).

Orders are commonly looked up by Order Number in the Order table, and line items are typically accessed by joining the Order table to the Line Item table on Order Number. Order Numbers are unique in the Order table, and non-unique in the Line Item table. Since Order Number is the key to the Order table Order Number values do not typically change.

In this example, the Order table should use Order Number as its primary index, as this satisfies all of the conditions of Access, Distribution, and Volatility.

As for the Line Item table, the primary key of (Order Number, Line Item Sequence Number) satisfies the conditions of Distribution and Volatility, but not Access, because the compound key of (Order Number, Line Item Sequence Number) will generate a different row hash value than Order Number alone, which implies that using the primary key as the primary index would require the table to be redistributed every time it is joined to the Order table. The Order Number column by itself satisfies the conditions of Access and Volatility, and also satisfies the Distribution condition as long as the typical order contains fewer than one hundred line items. Therefore, Order Number should be selected as the primary index of both the Order and the Line Item tables.

2.2 Join Processing

The shared-nothing architecture of Teradata ensures that each AMP has its own unit of work that it does not share with the other AMPs in the system. Therefore, in order to perform a join, the rows that are going to be joined must both be present on the same AMP. Teradata has several ways of accomplishing this.

2.2.1 Data Movement in Join Processing

When a query is executed on Teradata, the optimizer must first generate a query execution plan, which is a series of independent steps that will be executed on the AMPs. Each AMP works only on its own set of data, and the AMPs do not communicate with each other until each step is completed.

In order to ensure that the rows to be joined are present on the same AMP, the optimizer may specify that table redistribution, table duplication, or table local-AMP sorting is performed on each of the tables in preparation for a join. In each of these cases, a temporary data set, or **spool**, is created that contains a copy of the original table.

2.2.1.1 Table Redistribution

When Teradata performs table redistribution in preparation for a join, the column or columns of the table to be joined are run through the Teradata hashing algorithm. A copy of the table is then placed in a spool which is then redistributed on the Teradata system, so that each row of the spool is placed on its target AMP, as though the join columns were the primary index of the spool.

2.2.1.2 Table Duplication

When Teradata performs table duplication in preparation for a join, a full copy of the table is sent to every AMP, where it resides in spool. This guarantees that every AMP on the system will have a complete copy of the table.

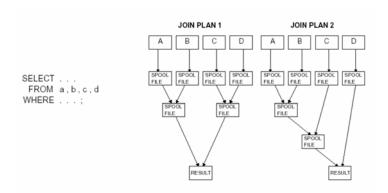
2.2.1.3 Table Local Sorting

When Teradata performs local table sorting in preparation for a join, a spool copy of the table is created which remains on the same AMP, being neither redistributed nor duplicated. However, the spool is then sorted on each AMP by the row hash of the join column(s).

2.2.2 Join Processing Methods

Teradata may use any of several methods to join tables. The most common of these is the merge join, followed by the product join. In certain circumstances, Teradata may perform other, less common joins, such as the **nested join** or the **hash join**.

Many analytical and decision-support queries must be able to join more than two tables. All of the Teradata join approaches are designed to join two tables in a single join. In order to join more than two tables, the query plan generated by the optimizer will determine two tables to join first, placing the results of this join operation into a spool. This spool may then be joined either to another table or to another spool, and that result to another table or spool, and so forth until the entire query is resolved.

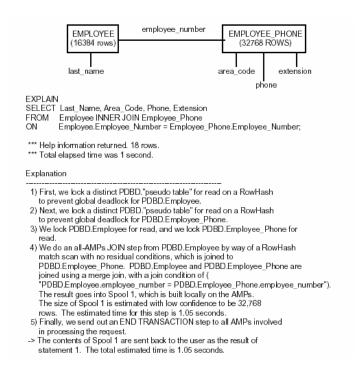


2.2.2.1 Merge Join

The **merge join** is the most common method of joining data sets on Teradata. In a merge join, each AMP compares rows from one data set that has been sorted by hash value with rows from another data set that has been sorted by hash value. Since both data sets are sorted by hash value, the total number of comparisons operations that need to be performed by the join operation is minimized.

2.2.2.1.1 In-Place Merge Join

When two tables are joined, and the join criteria specify the primary index of the first table being equal to the primary index of the second table, a merge join can be performed in-place. This requires no redistribution, no duplication, and no sorting of either table. This is almost always the most efficient join type possible on Teradata.



2.2.2.1.2 Merge Join with One Table Redistribution

When two tables are joined, and the join criteria specify the primary index of one of the tables being equal to a column or columns of the second table which do not comprise the second table's primary index, a merge join may be performed after redistributing the second table. In this case, the second table must be placed in spool and then re-hashed on the join columns.



SELECT Department_Name, Last_Name, First_Name Department INNER JOIN Employee

Employee.Department_Number = Department.Department_Number;

*** Help information returned, 23 rows.

*** Total elapsed time was 1 second.

- 1) First, we lock a distinct PDBD."pseudo table" for read on a RowHash to prevent global deadlock for PDBD.Employee.
- 2) Next, we lock a distinct PDBD."pseudo table" for read on a RowHash to prevent global deadlock for PDBD.Department.
- We lock PDBD.Employee for read, and we lock PDBD.Department for read.
 We do an all-AMPs RETRIEVE step from PDBD.Employee by way of an all-rows scan with no residual conditions into Spool 2, which is redistributed by hash code to all AMPs. Then we do a SORT to order Spool 2 by row hash. The size of Spool 2 is estimated with high confidence to be 16,384 rows. The estimated time for this step is 0.68 seconds
- 5) We do an all-AMPs JOIN step from PDBD.Department by way of a RowHash match scan with no residual conditions, which is joined to Spool 2 (Last Use). PDBD, Department and Spool 2 are joined using a merge join, with a join condition of ('department_number = PDBD.Department.department_number"). The result goes into Spool 1, which is built locally on the AMPs. The size of Spool 1 is estimated with low confidence to be 16,384 rows. The estimated time for this step is 0.35 seconds
- 6) Finally, we send out an END TRANSACTION step to all AMPs involved in processing the request.
- -> The contents of Spool 1 are sent back to the user as the result of statement 1. The total estimated time is 1.03 seconds.

2.2.2.1.3 Merge Join with Two Table Redistribution

When two tables are joined, and the join criteria specify non-PI columns of the first table being equal to non-PI columns of the second table, a merge join may be performed after redistributing both tables. In this case, both tables will be placed in spool and then rehashed on the join columns.

2.2.2.1.4 Merge Join with Duplication

When two tables are joined with an equality condition, and it is not a PI-to-PI join, a merge join with duplication may be performed. In this case, the smaller table is duplicated across all AMPs and then sorted by the row hash of the join column, and the larger table remains on its current AMP but is locally sorted by the row hash of the join column.

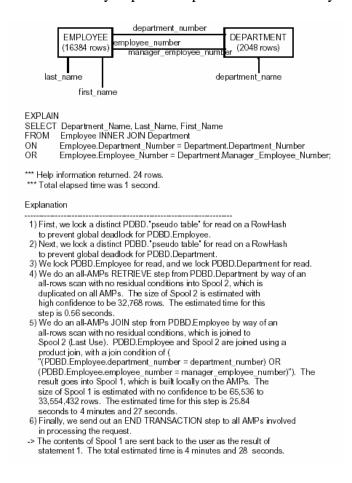
The Teradata optimizer will typically choose this approach when it is less resourceintensive to duplicate and sort the smaller table than it is to hash and redistribute the larger table.

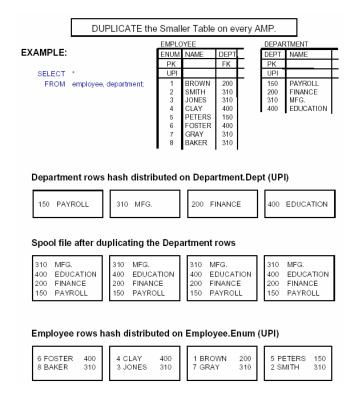
2.2.2.2 Product Join

The **product join** is another method of joining tables on Teradata. In a product join, each AMP compares rows from one unsorted data set with rows from another unsorted data set. Since neither of the data sets is sorted, every row in the first data set must be compared with every row from the second data set, unlike in a merge join.

When two tables are joined on a non-equivalence condition, or when an extremely large table is joined to an extremely small table, a product join may be performed. To perform a product join, the smaller of the two data sets must always be duplicated across all AMPs so that the larger data set may remain in place.

The product join is often the best way of joining an extremely small table with an extremely large table, since the large table will neither have to be redistributed nor sorted. However, it is usually the worst possible way to join two large tables, as duplicating a large table in spool is an extremely expensive operation in terms of system resources.





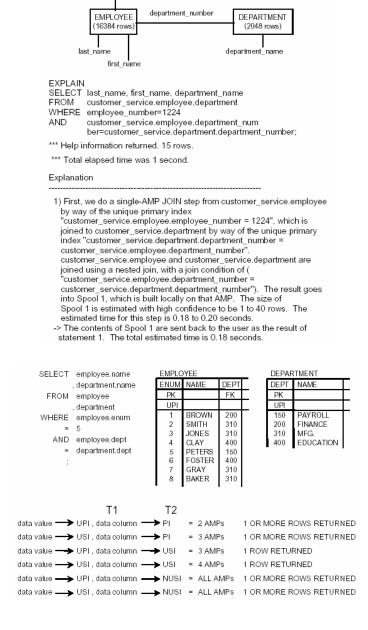
2.2.2.3 Other Join Types

Two other join types supported by Teradata are the nested join and the hash join.

2.2.2.3.1 *Nested Join*

When a single value is specified for a unique index column of a table, and the join criteria includes a column or columns of this first table being equal to any index of the second table, a **nested join** may sometimes be performed. In a nested join, Teradata uses the single row from the first table, and uses the hash of the join column in that row in order to access matching rows from the second table. The nested join is an all-AMPs operation when the index used in the second table is a NUSI. The nested join is a few-AMPs join when the join is on a UPI, NUPI, or USI of the second table.

The nested join is generally a very efficient operation on Teradata.

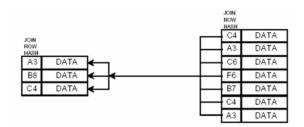


:employee_number

2.2.2.3.2 Hash Join

The **hash join** is an alternative join scheme to the merge join with duplication. In a hash join, the smaller table is duplicated on all AMPs, and then sorted in row hash sequence. The larger table remains in place; however, unlike in the merge join with duplication, the larger table does not need to be sorted in order to perform a hash join. The join column row hash of the larger table is then used to search the "hash map" of the smaller table.

A hash join can be more efficient than a merge join when the hash map of the smaller table can remain in memory on all AMPs, but it becomes less efficient when the hash map of the smaller table is too big, or when the underlying data is badly skewed.



2.3 Join Indexes

On Teradata, a **Join Index** is a data structure that contains data from one or more tables, with or without aggregation, and with or without filtering. It may contain columns from either a single table or from multiple tables.

Join indexes are implemented like physical tables on Teradata. Each join index has its own primary index, and may have secondary indexes defined as well. Join indexes cannot be accessed directly by users; instead, the data in a join index is automatically maintained by Teradata as the data in the underlying tables changes, and the Teradata optimizer will decide when to use a join index to resolve a query if it provides a more efficient path than selecting from the underlying tables.

A single-table join index is more likely to be useful when it **covers** a query, meaning that all columns of the underlying table needed by the query are present in the join index. One reason to define a covering single-table join index would be to specify a different primary index for the join index, in order to provide an alternate access path to the underlying data. In this example, a table may have its primary key defined as its primary index, and a join index may then be defined on the same table with a foreign key defined as the join index's primary index. Another reason to define a covering single-table join index is the **sparse join index** example described later in this document.

Multi-table join indexes can be useful even when they do not cover a query, although a covering join index always gives the greatest performance benefit.

Although join indexes provide several performance benefits, there are some drawbacks to their implementation. They require the use of additional permanent space on Teradata. They require additional processing, as each time a row in one of the underlying tables is inserted, updated, or deleted, the join index must be updated to reflect the change. Join indexes may not be defined on a table with triggers. Join indexes are not compatible with the use of the Fastload and Multiload utilities (or these utilities' implementation in Teradata Parallel Transporter) on the underlying tables.

2.3.1 Aggregate Join Index

An **aggregate join index** allows the user to specify aggregates in the join index. Aggregate Join Indexes provide a tremendous performance benefit for queries that frequently specify the same aggregate operation on the same column or columns.

2.3.2 Sparse Join Index

A **Sparse Join Index** is a special type of single table join index, in which a WHERE filter is placed on the table to reduce the number of rows that qualify for the index. Sparse Join Indexes can be very beneficial when defined on a large table, in cases where a small subset of the table needs to be accessed frequently, as it allows Teradata to work with a much smaller data set.

2.4 Statistics

Teradata statistics are data demographics that may be used by the optimizer to create a query execution plan. By issuing a COLLECT STATISTICS command, Teradata will analyze the data distribution of the specified column or group of columns of the specified table, and store these statistics in the data dictionary. The statistics will then be used by the optimizer in order to determine the best possible query execution plan.

2.4.1 When to Collect Statistics

As a general rule, it is recommended to collect statistics on the following.

- All non-unique primary indexes of all tables
- All unique primary indexes of small tables (where the number of rows is less than 100 times the number of AMPs on the system)
- Unique primary or secondary indexes that are commonly used in range constraints
- All sets of columns used in join conditions for all tables
- All non-unique secondary indexes with an uneven distribution of values

Note that collection and maintenance of statistics requires significant use of system resources for larger tables. For this reason, statistics should not be collected on all columns of all tables without regard to how the statistics would benefit query performance.

2.4.2 How to Maintain Collected Statistics

Teradata statistics need to be maintained as the underlying tables change. In some cases, having stale statistics may be worse than having no statistics at all. As an example, a sales table which is loaded daily may have a Sale Date column. If database statistics on this column are over a week old, the optimizer may assume that there are no rows with a Sale Date for the current week, which may generate problematic query plans if a report for the current week is run.

Generally, statistics should be re-collected when 10% of the rows of the underlying table have changed, and should be re-collected daily on date columns of fact tables.

2.4.3 Dynamic AMP Sampling

When statistics are unavailable, the optimizer will rely on Dynamic AMP Sampling, a process in which Teradata analyzes a small set of data from a random AMP in the system, from which it will extrapolate a set of statistics. This process is very accurate when sampling unique primary indexes of extremely large tables. However, Dynamic AMP Sampling is often unreliable for very small tables, and can give inconsistent results for data that is not evenly distributed. For this reason, it is particularly important to maintain statistics on columns or indexes with skewed or lumpy data distributions, as Teradata statistics can easily account for these kinds of distributions.

3 Performance Monitoring and Tuning

Some of the most effective ways to monitor and tune Teradata for performance involve the use of the Teradata Explain Facility, the PMON Utility, and the Database Query Log

3.1 The Explain Facility

When a query is prefixed with the EXPLAIN keyword, Teradata does not return the query results, but instead returns the explain plan generated for the query by the optimizer. The explain plan is a readable English-language explanation of the steps to be performed by the AMPs to resolve the query, and includes the following information.

- The estimated size of data sets
- Confidence levels of data size estimates (based on the presence or absence of Teradata statistics)
- Estimated processing times
- Locks placed on database objects
- Join techniques used to resolve the query
- The order in which tables are joined and aggregations are performed
- Whether rows are accessed from source tables by primary index value, by secondary index lookup, by partition, or by full-table scan

Several examples of explain plans can be found in section 2.2 of this document (Join Processing).

3.2 The PMON Utility

Teradata **PMON**, or Teradata Performance Monitor, is a client utility that allows database administrators and privileged users to monitor active sessions on Teradata. With PMON, a user may view the SQL and optimizer query plan of any active session in near real-time, as well as additional details such as the current step of the query plan, the execution time of each completed step, and the number of rows processed during each completed step.

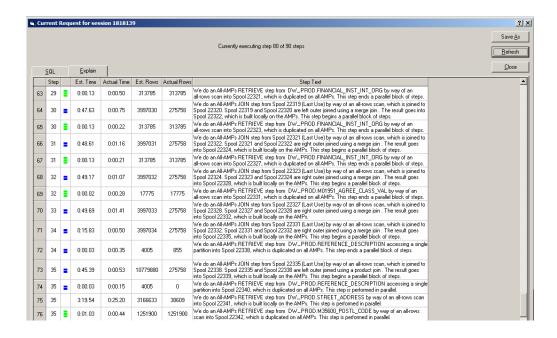
These features of PMON allow users to identify during runtime which steps in a query are troublesome. Long-running steps, unnecessary product joins, and bad spool size estimates made by the optimizer are readily identifiable, and skewed processing is much easier to detect.

3.2.1 Performance Tuning Examples Using PMON

Suppose that a database administrator notices that selecting from the view V01_INFO, a view which joins a large number of tables, seems to encounter performance problems.

To research this, the DBA will start from the PMON **Session** screen. From there, the DBA should identify the session running the query and click on it. From there, click the SQL button to see the session's SQL. When the SQL appears, click the Explain tab to see the explain plan.

In this case, PMON shows that the query runs well until the last few steps, at which point each step begins to take longer than anticipated, and the Estimated Rows appear to be very different from the Actual Rows in each spool. The problems appear to begin when the optimizer makes an inaccurate, no-confidence estimate of the size of a spool retrieved from the STREET_ADDRESS table in Step 35.



To determine how the STREET_ADDRESS table is accessed by this view, a SHOW VIEW statement is run, which returns the following.

REPLACE VIEW DW_PROD.V01_INFO AS LOCKING ROW FOR ACCESS SELECT

```
FROM DW_PROD.AGREEMENT AS AGREEMENT
INNER JOIN DW PROD.VRSK BUSINESSDATE M AS BD
ON BD.BUSINESSDATE BETWEEN AGREEMENT.START DT AND AGREEMENT.END DT
AND AGREEMENT.RECORD DELETED FLAG = 0
AND AGREEMENT.ACCOUNT_MODIFIER_NUM = 'AM14'
AND AGREEMENT.CTL ID = '004'
LEFT OUTER JOIN DW_PROD.AGREEMENT_TO_LOCATION AS AGREEMENT_TO_LOCATION
ON AGREEMENT.Account_Num = AGREEMENT_TO_LOCATION.Account_Num
AND AGREEMENT.Account_Modifier_Num =
AGREEMENT_TO_LOCATION.Account_Modifier_Num
AND BD.BUSINESSDATE BETWEEN AGREEMENT_TO_LOCATION.START_DT AND
AGREEMENT_TO_LOCATION.END_DT
AND AGREEMENT_TO_LOCATION.RECORD_DELETED_FLAG = 0
AND AGREEMENT_TO_LOCATION.CTL_ID = '004'
LEFT OUTER JOIN DW_PROD.STREET_ADDRESS AS STREET_ADDRESS
ON AGREEMENT TO LOCATION.Location ID = STREET ADDRESS.STREET ADDRESS ID
AND BD.BUSINESSDATE BETWEEN STREET_ADDRESS.START_DT AND
```

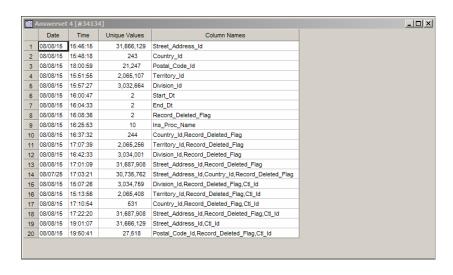
In this case, the view filters the STREET_ADDRESS table on the Ctl_Id and Record_Deleted_Flag fields. To see whether the necessary statistics are present on STREET_ADDRESS, the following SQL command is run.

HELP STATISTICS DW_PROD.STREET_ADDRESS;

AND STREET ADDRESS.RECORD DELETED FLAG = 0

AND STREET ADDRESS.CTL ID = '004';

STREET ADDRESS.END DT



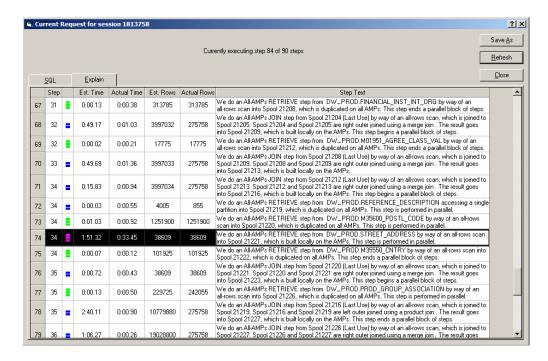
The HELP STATISTICS command shows that statistics are present on Record_Deleted_Flag, but not on Ctl_Id. It is not sufficient for statistics to be present on only one of the columns; it would be preferable to have statistics present on each column,

and best of all to have statistics present on the two columns together. To accomplish this, the following SQL commands are run.

```
COLLECT STATISTICS ON DW_PROD.STREET_ADDRESS COLUMN (Ctl_id);

COLLECT STATISTICS ON DW_PROD.STREET_ADDRESS COLUMN (Ctl_id, Record_Deleted_Flag);
```

After this is complete, the original query is run once again, and monitored in PMON to determine whether this helps the performance of the query. PMON shows that Teradata now has high confidence in the estimated size of the spool retrieved from the STREET_ADDRESS table, and the subsequent steps of the query run much faster and do not encounter performance issues.



Note that the SQL in the view remains unchanged, and the physical design of the database remains unchanged. The only change that was made was collecting statistics on one of the tables referenced by the view.

3.3 Teradata Database Query Log

DBQL, or Database Query Log, is a Teradata feature that is not enabled by default, which can be very useful for researching performance issues. DBQL will log attributes about each query executed on the system, such as the user who submitted the query, the leading few characters of the SQL text, the time the query was received by the database, the number of steps completed, the time the first step began, and the time the answer set was returned to the client. Depending on which features of DBQL are enabled, the

logging may also include the complete SQL text of each query, its entire explain plan, and detailed information about each execution step similar to that which is shown by PMON.

It is worth noting that DBQL data is not immediately written to the database upon completion of a query. Rather, DBQL data is written to a database cache, which is periodically flushed to the data dictionary tables.

3.3.1 Enabling DBQL

DBQL is enabled when a database administrator submits a BEGIN QUERY LOGGING statement to Teradata. For purposes of performance management and tuning, it is recommended that the EXPLAIN, SQL, and STEPINFO logging options are specified when enabling DBQL.

For example, to begin query logging on user PROD1, the database administrator would execute the following statement.

```
BEGIN QUERY LOGGING WITH EXPLAIN, SQL, STEPINFO ON PROD1;
```

DBQL is disabled when the database administrator submits an END QUERY LOGGING statement to Teradata.

3.3.2 DBQL Skew Detection Report

If DBQL logging with SQL and STEPINFO have been enabled, then the following SQL statement may be executed to detect queries that encountered skewed processing during execution the previous day.

Skewed processing is detected by comparing the CPU time used on the busiest AMP on the system for each query to the average AMP CPU time for the query, which is implemented by a filter in blue in the SQL statement below. Smaller, less significant queries are excluded from the report by a minimum CPU time filter, which is in green. The report includes only queries from the previous day due to a timestamp filter which is in red.

```
(NAMED ElapsedTime)
        ,ST.EstProcTime
        ,ST.CPUtime (Format 'ZZ,ZZZ,ZZ9.999')
        ,ST.IOcount
        ,ST.EstRowCount
        ,ST.RowCount
        ST.HotAmp1CPU (Format 'ZZ,ZZZ,ZZ9.999')
        ,ST.HotCPUAmpNumber
        ,ST.HotAmp1IO
        ,ST.HotIOAmpNumber
FROM
        DBC.DBQLStepTbl ST
INNER JOIN
       DBC.DBQLSQLTbl SQ
       ST.QueryID = SQ.QueryID
AND ST.ProcID = SQ.ProcID
AND CAST(ST.CollectTimeStamp AS DATE) = CAST(SQ.CollectTimeStamp
AS DATE)
WHERE ST. HotAmp1CPU > 60
 AND ST.HotAmp1CPU/NULLIFZERO(CPUtime) > .25
 AND ST.CPUtime >= ST.HotAmp1CPU
 AND SQ.SqlRowNo = 1
 AND CAST(ST.CollectTimeStamp AS DATE) = CURRENT_DATE - 1
```

3.3.3 DBQL Product Join Detection Report

If DBQL logging with SQL and STEPINFO have been enabled, then the following SQL statement may be executed to detect queries that encountered problematic product joins during execution the previous day.

Large product joins are detected by comparing the CPU time used by the query to the I/O performed by the query, which is implemented by a filter in blue in the SQL statement below. Smaller, less significant queries are excluded from the report by a minimum CPU time filter, which is in green. The report includes only queries from the previous day due to a timestamp filter which is in red.

```
SELECT
        SQ.SqlTextInfo
        ,CASE WHEN ST.QueryID = 0 THEN 0
              ELSE ((ST.ProcID / 100000)*1000000000)+ ST.QueryID
         END (NAMED QueryID, DECIMAL(18,0), FORMAT '-Z(17)9')
        ,ST.ProcID
        ,ST.CollectTimeStamp
        ,ST.StepLev1Num
        ,ST.StepLev2Num
        ,ST.StepName
        ,ST.StepStartTime
        ,ST.StepStopTime
        ,((ST.StepStopTime-ST.StepStartTime) HOUR(2) to SECOND)
          (NAMED ElapsedTime)
        ,ST.EstProcTime
        ,ST.CPUtime (Format 'ZZ,ZZZ,ZZ9.999')
        ,ST.IOcount
        ,ST.EstRowCount
        ,ST.RowCount
```

```
,ST.HotAmp1CPU (Format 'ZZ,ZZZ,ZZ9.999')
        ,ST.HotCPUAmpNumber
        ,ST.HotAmp1IO
        ,ST.HotIOAmpNumber
        DBC.DBQLStepTbl ST
FROM
INNER JOIN
         DBC.DBQLSQLTbl SQ
         ST.QueryID = SQ.QueryID
ON
         ST.ProcID = SQ.ProcID
AND
AND
         CAST(ST.CollectTimeStamp AS DATE) = CAST(SQ.CollectTimeStamp
AS DATE)
        ST.CPUtime*10 >= ST.IOCount
WHERE
 AND
        ST.CPUtime > 60
 AND
        SQ.SqlRowNo = 1
      ST.StepName = 'JIN'
 AND
 AND
        CAST(ST.CollectTimeStamp AS DATE) = CURRENT_DATE - 1
```