

Aster Data SQL and MapReduce

Course Summary

Description

In this course, students will learn the Aster Data SQL and MapReduce starting at the most basic level and going to the most advanced level with many examples.

Objectives

At the end of this course, students will be able to:

- Have a deeper knowledge and understanding of the Aster Data Architecture, SQL and MapReduce functions and how to write them.

Topics

- The Aster Data Architecture
- Four Options for Aster Data Table Design
- How Joins Work Inside the Aster Engine
- Temporary and Analytic Tables
- Tera-Tom's Top Tips
- Indexes
- SQL-MapReduce
- Time and Date
- Aster Windows Functions
- The Fundamental SQL Commands That Work on Aster

Audience

This course is designed for anyone who has a desire to learn the Aster Data SQL and MapReduce from beginners to an advanced audience. This course is completely customizable by the client.

Prerequisites

There are no prerequisites for this course.

Duration

Two to three days

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Course Outline

I. The Aster Data Architecture

- A. What is Parallel Processing?
- B. Aster Data is a Parallel Processing System
- C. Each vworker holds a Portion of Every Table
- D. The Rows of a Table are Spread Across All vworkers
- E. Aster Tables are defined as Fact or Dimension when Created
- F. Fact Table
- G. A More Detailed Look at the Fact Table Distribution
- H. Dimension Table are Replicated
- I. A Dimension Table is often Replicated across vworkers
- J. Aster Data has Fact and Dimension Tables
- K. Aster Tables are defined as Fact or Dimension when Created
- L. Fact and Dimension Tables can be Hashed by the same Key
- M. Distribution Key Rules
- N. Aster Data Uses a Hash Formula
- O. The Hash Map Determines which vworker will own the Row
- P. The Hash Formula, Hash Map and vworker
- Q. Placing rows on the vworker
- R. Placing rows on the vworker Continued
- S. A Review of the Hashing Process
- T. Like Data Hashes to the Same vworker
- U. Distribution Key Data Types
- V. Run ANALYZE to COLLECT STATISTICS on a Table
- W. Some Examples of ANALYZE
- X. What Columns to Analyze

II. Four Options for Aster Data Table Design

- A. There are Four Options to Aster Table Design
- B. Straight up Distribute by Hash
- C. Straight up Distribute by Hash - Problems
- D. Straight up Distribute by Replication
- E. Partition the Table with Logical Partitioning
- F. This Partitioned Table Sorts Rows by Month of Order_Date
- G. An All vworkers Retrieve By Way of a Single Partition
- H. You can Partition a Table by Range or by List
- I. A Partitioned By List Example with Three Tactical Queries
- J. Aster Data Multi-Level Partitioning
- K. Aster Allows for Multi-Level Partitioning
- L. SQL Commands for Logical Partitioning as One Table
- M. What Partitions are on my Table?
- N. What does a Columnar Table look like?
- O. A Comparison of Data for Normal Vs. Columnar
- P. A Columnar Table is best for Queries with Few Columns
- Q. When to use a Columnar Table

III. How Joins Work Inside the Aster Engine

- A. The Joining of Two Tables
- B. Aster Moves Joining Rows to the Same vworker
- C. Because of the Join Rule – Dimension Table are Replicated

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Course Outline (cont'd)

- D. The Two Different Philosophies for Table Join Design
 - E. What Could You Do If Two Tables Joined 1000 Times a Day?
 - F. Fact and Dimension Tables can be Hashed by the same Key
 - G. Joining Two Tables with the same PK/FK Distribution Key
 - H. A Join With Co-Location
 - I. A Performance Tuning Technique for Large Joins
 - J. The Joining of Two Tables with an Additional WHERE Clause
 - K. Aster Performs Joins Using Three Different Methods
 - L. The Hash Join
 - M. The Merge Join
 - N. Nested Loop Joins
- IV. Temporary and Analytic Tables**
- A. Aster has Three Types of Data
 - B. Create a Permanent Table Using Create Table AS (CTAS)
 - C. Create a Logically Partitioned Table and Populate It
 - D. Create a Temporary Table with using Create Table AS (CTAS)
 - E. A Temporary Table in Action
 - F. A Temporary Table That Uses an Insert/Select
 - G. Create an Analytic Table Using an Insert/Select
 - H. Create an Analytic Table Using CREATE TABLE AS (CTAS)
 - I. Operations that Invalidate an Analytic Table
 - J. If an Analytic Table is Invalid
- K. Tera-Tom History
- V. Tera-Tom's Top Tips**
- A. Tera-Tom's Top Tips
 - B. Tera-Tom's Top Tips # 2
 - C. Tera-Tom's Top Tips #3
 - D. Tera-Tom's Top Tips # 3 Rewritten
 - E. Tera-Tom's Top Tips #4
 - F. When the GROUP BY Column is NOT the Distribution Key
 - G. Example of GROUP BY Column is NOT the Distribution Key
 - H. Tera-Tom's Top Tips #5
 - I. Tera-Tom's Top Tips #6 – Use EXPLAIN
 - J. Query Plan and Estimates
 - K. Explain Plan Showing a Hash Join
 - L. Explain Plan Showing a Merge Join
 - M. Explain Plan Showing a Nested Loop Join
- VI. Indexes**
- A. There are Only Three Types of Scans
 - B. Guidelines for Indexes
 - C. An Index Syntax Example
 - D. The B-Tree Index
 - E. Which Columns Might You Create an Index?
 - F. Answer - Which Columns Might You Create an Index?
 - G. A Visual of an Index (Conceptually)
 - H. A Query Using an Index Uses All vworkers
 - I. Multicolumn indexes
 - J. A NUSI BITMAP Theory
 - K. A NUSI Bitmap in Action
 - L. Indexes on Expressions
 - M. Indexes on Extracts of Dates

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Course Outline (cont'd)

N.	GiST Indexes	CC.	Adding an Aggregate to nPath Results
O.	Five Operational Tips for Efficient Indexing	DD.	Adding an Aggregate to nPath Results (Continued)
P.	REINDEX	EE.	SQL-MapReduce Examples - Use Regular SQL
Q.	createCompressedIndexOnCompressedTableByDefault Flag	FF.	SQL-MapReduce Examples - Create Objects
VII. SQL-MapReduce		GG.	SQL-MapReduce Examples - Subquery
A.	MapReduce History	HH.	SQL-MapReduce Examples - Query as Input
B.	What is MapReduce?	II.	SQL-MapReduce Examples - Nesting Functions
C.	What is SQL-MapReduce?	JJ.	SQL-MapReduce Examples - Functions in Derived Tables
D.	SQL-MapReduce Input	KK.	SQL-MapReduce Examples - SMAVG
E.	SQL-MapReduce Output	LL.	SQL-MapReduce Examples - Pack Function
F.	Subtle SQL-MapReduce Processing	MM.	SQL-MapReduce Examples - Pack Function (Continued)
G.	Aster Data Provides an Analytic Foundation	NN.	SQL-MapReduce Examples - Pivot Columns
H.	Path Analysis	OO.	Workshop: Create This Table
I.	Text Analysis	PP.	Login to your GNOME Terminal
J.	Statistical Analysis	QQ.	Login to your Linux
K.	Segmentation (Data Mining)	RR.	Using the GNOME Terminal
L.	Graph Analysis	SS.	Unzip the bank_web_data.zip
M.	Transformation of Data		Use the Function
N.	Sessionize		ncluster_loader to Load the Bank Data
O.	Tokenize	TT.	Run this nPath Map Reduce Function on your Table
P.	SQL-MapReduce Function... nPath	UU.	nPath in Action
Q.	nPath SELECT Clause	VV.	Operators at their Simplest
R.	nPath ON Clause	WW.	Pattern
S.	nPath PARTITION BY Expression	XX.	Accumulate
T.	nPath DIMENSION Expression	YY.	Accumulate With All Pages
U.	nPath ORDER BY Expression	ZZ.	Accumulate – nPath with a WHERE Clause
V.	nPath MODE Clause has Overlapping or NonOverlapping	AAA.	SQL-MapReduce Examples - Path Generator
W.	nPath PATTERN Clause	BBB.	SQL-MapReduce Examples - Path Generator (Continued)
X.	Pattern Operators		
Y.	Pattern Operators Order of Precedence		
Z.	Matching Patterns Which Repeat		
AA.	nPath SYMBOLS Clause		
BB.	nPath RESULTS Clause		

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Course Outline (cont'd)

CCC.	SQL-MapReduce Examples - Path Generator (Continued)	ZZZ.	Answer Workshop 2 – Fill in the x's
DDD.	SQL-MapReduce Examples - Path Generator (Continued)	AAAA.	Answer Workshop 2 – You Could Have Used a GROUP BY
EEE.	SQL-MapReduce Examples - Path Generator (Continued)	BBBB.	Workshop 3 – Add to the Query
FFF.	SQL-MapReduce Examples - Linear Regression	CCCC.	Workshop 3 – Answer to Add to the Query
GGG.	SQL-MapReduce Examples - Linear Regression (Continued)	DDDD.	Workshop 4 – Fill in the x's
HHH.	SQL-MapReduce Examples - Linear Regression (Continued)	EEEE.	Answer to Workshop 4 – Fill in the x's
III.	SQL-MapReduce Examples - Naive Bayes	FFFF.	Workshop 5 – Find that Customer
JJJ.	SQL-MapReduce Examples - Naive Bayes (Continued)	GGGG.	Answer to Workshop 5 – Find that Customer
KKK.	SQL-MapReduce Examples - Naive Bayes (Continued)	HHHH.	Workshop 6 – Change the MapReduce Function
LLL.	SQL-MapReduce Examples - Naive Bayes (Continued)	IIII.	Answer to Workshop 6 – Change the MapReduce Function
MMM.	SQL-MapReduce Examples - Naive Bayes (Continued)	JJJJ.	Workshop 7 – Build the MapReduce Function
NNN.	SQL-MapReduce Examples - Naive Bayes (Continued)	KKKK.	Answer to Workshop 7 – Build the MapReduce Function
OOO.	SQL-MapReduce Examples - Naive Bayes (Continued)	LLLL.	Best Answer to Workshop 7 – Build the MapReduce Function
PPP.	SQL-MapReduce Examples - Naive Bayes (Continued)	MMMM.	Workshop 8 – Build the Accumulate in the Result
QQQ.	Join Aster, Teradata and Hadoop Tables; feed into MapReduce	NNNN.	Answer to Workshop 8 – Build the Accumulate in the Result
RRR.	Run Both of these Examples Together and Compare	OOOO.	SQL-MapReduce Examples - Linear Regression (Continued)
SSS.	Run this nPath Map Reduce Function	PPPP.	Workshop 9 – Build the Subquery
TTT.	nPath in Action	QQQQ.	Answer to Workshop 9 – Build the Subquery
UUU.	Another nPath Example	RRRR.	Workshop 10 – Do Your First Join
VVV.	Finding Out What Functions You Have Installed	SSSS.	Answer to Workshop 10 – Do Your First Join
WWW.	Workshop 1 – Fill in the x's	TTTT.	Answer to Workshop 10 – Do the Join Using a New Syntax
XXX.	Answer Workshop 1 - Fill in the x's		
YYY.	Workshop 2 – Fill in the x's		

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Course Outline (cont'd)

UUUU.	Workshop 11 – Super Join the Tables	PPPPP.	Workshop 18 – Create the Churn Table with a Better Example
VVVV.	Answer to Workshop 11 – Super Join the Tables	QQQQQ.	Multi-Case
WWWW.	Answer to Workshop 11 – Super Join the Tables	RRRRR.	The Multi-Case Function
XXXX.	Workshop 12 – Sessionize the Data	SSSSS.	The Multi-Case Function in Nexus
YYYY.	Answer to Workshop 12 – Sessionize the Data	TTTTT.	The Multi-Case Function Mixing and Matching
ZZZZ.	Workshop 13 – What is this Query Doing?	UUUUU.	The Multi-Case Function Mixing and Matching
AAAAA.	Answer to Workshop 13 – What is this Query Doing?	VVVVV.	SQL-MapReduce Examples - cFilter
BBBBB.	Workshop 14 – Using ilike	WWWWW.	SQL-MapReduce Examples - cFilter (Continued)
CCCCC.	Answer to Workshop 14 – Using ilike	XXXXX.	SQL-MapReduce Examples - Linear Regression (Continued)
DDDDD.	Answer to Workshop 14 – Using ilike	YYYYY.	SQL-MapReduce Examples - cFilter (Continued)
EEEEE.	Workshop 15 – What are the First Two Pages Visited?	ZZZZZ.	SQL-MapReduce Examples - Linear Regression (Continued)
FFFFF.	Workshop 15 – What are the First Two Pages Visited?	AAAAA.	SQL-MapReduce Examples - cFilter (Continued)
GGGGG.	Workshop 16 – Advanced - First Two Pages Visited?	BBBBB.	SQL-MapReduce Examples - cFilter (Continued)
HHHHH.	Answer to Workshop 16 Advanced - First Two Pages Visited?	CCCCC.	SQL-MapReduce Examples - cFilter (Continued)
IIIII.	Workshop 17 – Can You Clean Up the Results?	DDDDD.	SQL-MapReduce Examples - cFilter (Continued)
JJJJJ.	Answer to Workshop 17 – Can You Clean Up the Results?	EEEEE.	SQL-MapReduce Examples - cFilter (Continued)
KKKKK.	Answer to Workshop 17 – Format the Date	FFFFF.	SQL-MapReduce Examples - cFilter (Continued)
LLLLL.	Workshop 18 – Build a Churn Table	GGGGG.	SQL-MapReduce Examples - cFilter (Continued)
MMMMM.	Workshop 18 – Run the Query Before Building to Test	HHHHH.	CFILTER in Action with Bank_Web_Clicks
NNNNN.	Workshop 18 – A Better Example	IIIII.	CFILTER in Action
OOOOO.	Answer to Workshop 18 – Build a Basic Churn Table	JJJJJ.	CFILTER using Nexus
		KKKKK.	nPath Error

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Course Outline (cont'd)

VIII. Time and Date

- A. Date, Time, and Timestamp Keywords
- B. Add or Subtract Days from a date
- C. The to_char command
- D. A Summary of Math Operations on Dates
- E. Using a Math Operation to find your Age in Years
- F. Find What Day of the week you were Born
- G. Date Related Functions
- H. The EXTRACT Command
- I. EXTRACT from DATES and TIME
- J. EXTRACT with DATE and TIME Literals
- K. EXTRACT of the Month on Aggregate Queries
- L. A Side Title example with Reserved Words as an Alias
- M. Implied Extract of Day, Month and Year
- N. DATE_PART Function
- O. DATE_TRUNC Function
- P. DATE_TRUNC Function using TIME
- Q. Aster NOW() Function

IX. Aster Windows Functions

- A. Cumulative Sum
- B. Cumulative Sum - Major and Minor Sort Key(s)
- C. The ANSI CSUM – Getting a Sequential Number
- D. The ANSI OLAP – Reset with a PARTITION BY Statement
- E. PARTITION BY only Resets a Single OLAP not ALL of them
- F. ANSI Moving Sum is Current Row and Preceding n Rows
- G. How ANSI Moving SUM Handles the Sort

- H. Moving SUM every 3-rows vs. a Continuous Sum
- I. Moving Average
- J. Partition By Resets an ANSI OLAP
- K. Moving Average Using BETWEEN
- L. Moving Difference using ANSI Syntax
- M. Moving Difference using ANSI Syntax with Partition By
- N. RANK Defaults to Ascending Order
- O. Getting RANK to Sort in DESC Order
- P. You can use Window Functions in Expressions
- Q. RANK() OVER and PARTITION BY
- R. DENSE_RANK() OVER
- S. PERCENT_RANK() OVER
- T. PERCENT_RANK() OVER with 14 rows in Calculation
- U. PERCENT_RANK() OVER with 21 rows in Calculation
- V. RANK With ORDER BY SUM()
- W. COUNT OVER for a Sequential Number
- X. The MAX OVER Command
- Y. MAX OVER with PARTITION BY Reset
- Z. The MIN OVER Command
- AA. The Row_Number Command
- BB. NTILE
- CC. NTILE Using a Value of 10
- DD. NTILE With a Partition
- EE. CUME_DIST
- FF. CUME_DIST With a Partition
- GG. LEAD
- HH. LEAD With Partitioning
- II. LAG
- JJ. LAG with Partitioning
- KK. FIRST_VALUE

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Course Outline (cont'd)

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|-----------|--|-----|---|
| LL. | FIRST_VALUE After Sorting by the Highest Value | Q. | A two-table join using Non-ANSI Syntax |
| MM. | FIRST_VALUE with Partitioning | R. | A two-table join using Non-ANSI Syntax with Table Alias |
| NN. | LAST_VALUE | S. | Aliases and Fully Qualifying Columns |
| OO. | NTH_VALUE | T. | A two-table join using ANSI Syntax |
| PP. | NTH_VALUE With Partition | U. | Both Queries have the same Results and Performance |
| QQ. | SUM(SUM(n)) | V. | LEFT OUTER JOIN |
| X. | The Fundamental SQL Commands That Work on Aster | W. | LEFT OUTER JOIN Brings Back All Rows in the Left Table |
| A. | BETWEEN is Inclusive | X. | RIGHT OUTER JOIN |
| B. | BETWEEN Works for Character Data | Y. | RIGHT OUTER JOIN Brings Back All Rows in the RIGHT Table |
| C. | LIKE uses Wildcards Percent '%' and Underscore '_' | Z. | FULL OUTER JOIN |
| D. | LIKE command Underscore is Wildcard for one Character | AA. | FULL OUTER JOIN Brings Back All Rows in All Tables |
| E. | GROUP BY Vs. DISTINCT – Good Advice | BB. | Which Tables are the Left and which are the Right? |
| F. | The Five Aggregates of Aster Data | CC. | Answer - Which Tables are the Left and which are the Right? |
| G. | GROUP BY when Aggregates and Normal Columns Mix | DD. | INNER JOIN with Additional AND Clause |
| H. | GROUP BY Delivers one row per Group | EE. | ANSI INNER JOIN with Additional AND Clause |
| I. | GROUP BY Dept_No or GROUP BY 1 the same thing | FF. | ANSI INNER JOIN with Additional WHERE Clause |
| J. | Limiting Rows and Improving Performance with WHERE | GG. | OUTER JOIN with Additional WHERE Clause |
| K. | WHERE Clause in Aggregation limits unneeded Calculations | HH. | OUTER JOIN with Additional AND Clause |
| L. | Keyword HAVING tests Aggregates after they are Totaled | II. | Results from OUTER JOIN with Additional AND Clause |
| M. | Keyword HAVING is like an Extra WHERE Clause for Totals | JJ. | The DREADED Product Join |
| N. | Getting the Average Values per Column | KK. | Result Set of the DREADED Product Join |
| O. | Getting the Average Values per Column | LL. | The Horrifying Cartesian Product Join |
| P. | Average Values per Column for All Columns in a Table | MM. | The ANSI Cartesian Join will ERROR |

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Course Outline (cont'd)

NN.	How would you Join these two tables?	DDD.	The SUBSTRING Command
OO.	How would you Join these two tables? You Can't Yet!	EEE.	How SUBSTRING Works with NO ENDING POSITION
PP.	An Associative Table is a Bridge that Joins Two Tables	FFF.	Using SUBSTRING to move Backwards
QQ.	The 5-Table Join – Logical Insurance Model	GGG.	How SUBSTRING Works with a Starting Position of -1
RR.	The Nexus Query Chameleon Writes the SQL for Users.	HHH.	How SUBSTRING Works with an Ending Position of 0
SS.	An IN List is much like a Subquery	III.	An Example using SUBSTRING, TRIM and CHAR Together
TT.	An IN List Never has Duplicates – Just like a Subquery	JJJ.	SUBSTRING and SUBSTR are equal, but use different syntax
UU.	An IN List Ignores Duplicates	KKK.	The POSITION Command finds a Letters Position
VV.	The Subquery	LLL.	Concatenation
WW.	How a Basic Subquery Works	MMM.	The Basics of CAST (Convert and Store)
XX.	The Final Answer Set from the Subquery	NNN.	Some Great CAST (Convert and Store) Examples
YY.	Should you use a Subquery or a Join?	OOO.	Some Great CAST (Convert and Store) Examples
ZZ.	CHARACTER_LENGTH AND OCTET_LENGTH	PPP.	Combining Searched Case and Valued Case
AAA.	The TRIM Command trims both Leading and Trailing Spaces	QQQ.	A Trick for getting a Horizontal Case
BBB.	Trim and Trailing is Case Sensitive	RRR.	Nested Case
CCC.	Trim and Trailing works if Case right	SSS.	Put a CASE in the ORDER BY