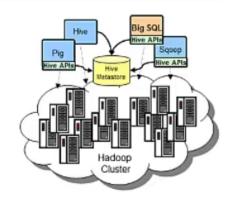
Big SQL on Hadoop

- There are multiple tools available for manipulating your data on Hadoop
 - MapReduce
 - Pig
 - Jagl
 - Hive
 - and many more!
- MapReduce is highly scalable, but difficult to use.
- Other tools require specific expertise.
- Big SQL has a common and familiar syntax.

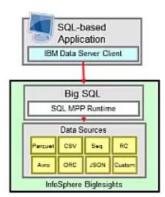
What is Big SQL?

- A Big SQL table is simply a view on your Hadoop data
 - Provides a logical view of the data
- No proprietary storage format
 Use it on top of your data
- Modern SQL:2011 capabilities
- The same SQL can be used on your warehouse data with little or no modifications.

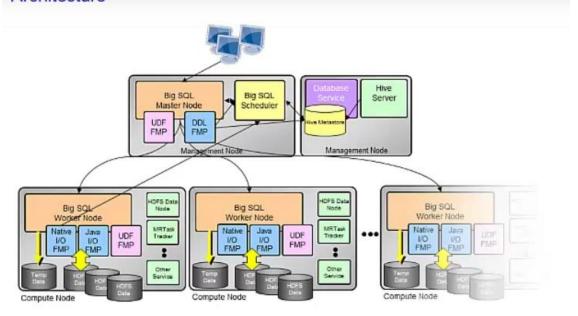


Architected for performance

- MapReduce is replaced with a modern, massively parallel processing (MPP) architecture
 - Compiler and runtime are native code (not Java)
 - Big SQL worker daemons live directly on cluster
 - Continuously running (no startup latency)
 - · Processing happens locally at the data
 - Message passing allows data to flow directly between nodes
- Operations occur in memory with the ability to spill to disk
 - Supports aggregations and sorts larger than available RAM



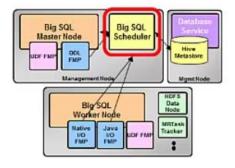
Architecture



*FMP = Fenced mode process

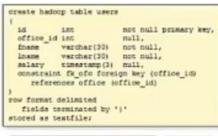
Scheduler Service

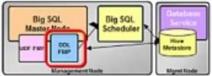
- This component is the main interface between Big SQL and Hadoop.
- Interfaces with the Hive Metastore for the table metadata
- Similar to the MapReduce job tracker for Big SQL
 - Big SQL provides query predicates for the scheduler to perform partial elimination
 - Determines the splits for each table involved in the query
 - Schedules the splits on the available Big SQL worker nodes
 - Serves the splits to the I/O engines



DDL processing

- The database engine does not understand the DDL statements natively
- Statements are forwarded to the DDL Fenced Mode Process (FMP)
 - Statement decomposed to native database DDL statement
 - · Define the relational structure of the table
 - Statement decomposed to Hive DDL to populate the Hive Metastore
 - Scheduler is notified of changes





I/O processing and UDF FMP

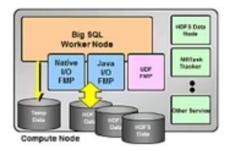
Two I/O processing engine

- Native I/O FMP

- Written in C/C++
- · The high-speed interface for common file formats
- · Delimited, Parquet, RC, Avro, and Sequencefile

- Java I/O FMP

- Handles all other formats via standard Hadoop/Hive API's
- Both are highly optimized and parallelized.
- The UDF FMP is where the user defined functions run.



Welcome to the course, Big SQL on Hadoop.

In this lesson, Big SQL Overview, you will get an introduction to Big SQL and see how to use Big SQL on Hadoop data.

After completing this lesson, you will be able to describe the Big SQL architecture. Know the differences between the current Big SQL and its older version (we will refer to this as Big SQL v1). List some of the Big SQL features. Understand Big SQL terminologies. Understand and use different methods for working with Big SQL.

There are multiple tools available for you

to work with your data on Hadoop. You have your MapReduce, which is highly scalable, but requires the user to understand the Java API and have some programming expertise. What if you do not want anything to do with Java? Well, then you can learn or use Pig, Jaql, Hive, or a number of other tools to work with your data. No? You do not want to learn anything

new? Good news, Big SQL provides a common and familiar syntax that lets you work with your data. You can query your data residing in Hadoop using Big SQL. There is no learning curve here. Big SQL is about applying SQL to your existing data -- there are no proprietary storage formats.

What is Big SQL? Simply put, Big SQL is a logical view on top of your Hadoop data. The data resides in Hadoop and all you have to do is use Big SQL to access it. There is no need to change the format, or migrate the data out of Hadoop to do any work on the data. Big SQL supports modern SQL:2011 capabilities. So, when you migrate your data into Hadoop, the same SQL can be used on your warehouse data with little or no modification. That is one of the big benefits of Big SQL, no need to learn anything new, and you can use your existing queries.

Big SQL is designed for performance. It replaces MapReduce using a modern, massively parallel processing or MPP architecture. The compiler and runtime are written in native code, C/C++, so it is much faster. The SQL engine pushes down the processing to the same node the holds the data so all the processing happens locally at the data. There is also no startup latency -- the daemons are continuously running. The operations occur in memory and if necessary, it can spill over to disk for processing. This allows for support of aggregations and sorts larger than the available RAM.

In this diagram here, you can see the Big SQL daemons that run on each node in the cluster, in orange. You have one master node, the one up at the top. The master node is responsible for accepting client connections, accepting queries, doing all the queries planning, query optimization and then pushing the query execution down to the worker nodes below. The other

colored boxes are the independent support processes that aid Big SQL in its processing. At the top, right corner, you can see that there is the Hive Metastore involved. Hive is still using it. Big SQL communicates and shares information with Hive.

The scheduler is the main interface between

Big SQL and Hadoop. Remember that Big SQL provides a logical view of the files stored on Hadoop, so it needs to know which file constitutes a "table", how many files there are, type of files, etc. It communicates with the scheduler to ask these questions. The scheduler also acts like the MapReduce job tracker. For example, when Big SQL gets a query, it tells the scheduler which table it is going to scan and the query predicates to let the scheduler figure out in how many pieces the work can be split. Then that information is passed back to the Big SQL master node to let it run. The worker nodes also communicate with the scheduler to figure out on which splits it needs to work at runtime.

There is a DDL processing engine as you can

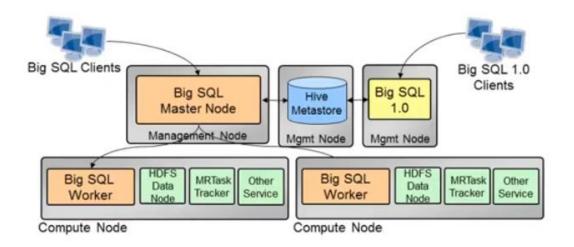
see in the diagram. This engine is responsible for taking a statement and turning it into something Hive and Big SQL can understand. It receives the statements from the native database engine to decompose it to something the database can understand. It decomposes to Hive DDL in order to populate the Hive Metastore. Then the scheduler is notified of the new table. One thing to note is that the DDL is run in a fenced mode process, or FMP, which is an isolated process to ensure that it does not affect the other processes. Here we have our I/O processing engine. With

Big SQL, we have two processing engines. We have a native I/O engine. The code for this engine is written in C/C++, and is highly optimize. It only knows a specific set of file formats as listed here. We also wanted to handle any Hive tables (defined in the Hive catalog). If we see a particular table that isn't recognized, we hand it down to the Java I/O layer. The Java I/O layer handles all the other formats via the standard Hadoop/Hive

API's. Both of these engines are optimized and parallelized, with multiple threads doing the I/O. Also in this diagram, note the UDF FMP, the pink box. The user defined functions also run in a separate process in order to isolate the database engine from the user code to prevent the user from doing any harm, just as with the DDL processing engine in the previous slide.

Big SQL vs Big SQL v1

- Big SQL is a complete rewrite/redesign from the previous version.
- Why two different versions?
 - Officially, there is only one release, and that is the Big SQL
 - · You may see it being refer to as Big SQL 3.0 in some cases
 - BigInsights 3.0 comes with Big SQL
 - Big SQL v1 or 1.0 is used to refer to the previous version



Big SQL vs Big SQL v1 - some differences

- Big SQL does not have support for HBase
 Need to use the driver for Big SQL v1
- DDL in Big SQL is a largely a superset of Big SQL 1.0
- Big SQL 1.0 treated single and double quotes as the same
 Big SQL reserves double quotes for identifiers (SQL standard rules)
- The hadoop keyword is required for creating tables in Big SQL

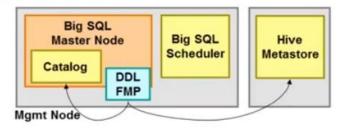
```
CREATE HADOOP TABLE T1
(
    C1 VARCHAR(30) NOT NULL
)
ROW FORMAT DELIMITED;
```

The need for the hadoop keyword can be lifted using compatibility mode:

```
1> SET SYSHADOOP.COMPATIBILITY_MODE=1;
1> CREATE TABLE T1
2> (
3> C1 VARCHAR(30) NOT NULL
4> )
ROW FORMAT DELIMITED;
```

Hive and Big SQL

- What does Hive have to do with Big SQL?
 - Big SQL shares the Hive Metastore
- Big SQL maintains its own catalog
 - Lives in the master node only
 - Duplicates relational portion of table definitions (column names, types, etc.)
 - Contains all other metadata (permissions, statistics, etc.)
- Big SQL DDL FMP keeps the catalog in sync
 - Only if the DDL is performed through Big SQL
- Unlike Big SQL 1.0, Hive objects are not immediately available
 - You must import these table definitions from Hive



Big SQL - features at a glance

Application Portability & Integration

Data shared with Hadoop ecosystem
Comprehensive file format support
Superior enablement of IBM software
Enhanced by Third Party software

Performance

Modern MPP runtime
Powerful SQL query rewriter
Cost based optimizer
Optimized for concurrent user throughput
Results not constrained by memory

Rich SQL

Comprehensive SQL Support IBM SQL PL compatibility

Federation

Distributed requests to multiple data sources within a single SQL statement

Main data sources supported: DB2 LUW, DB2/z, Teradata, Oracle, Netezza

Enterprise Features

Advanced security/auditing
Resource and workload management
Self tuning memory management
Comprehensive monitoring

SQL capability highlights

- Full support for subqueries
 - In SELECT, FROM, WHERE and HAVING clauses
 - Correlated and uncorrelated
 - Equality, non-equality subqueries
 - EXISTS, NOT EXISTS, IN, ANY, SOME, etc.
- All standard join operations
 - Standard and ANSI join syntax
 - Inner, outer, and full outer joins
 - Equality, non-equality, cross join support
 - Multi-value join (WHERE (c1, c2) = ...)
 - UNION, INTERSECT, EXCEPT

```
s_name,
count(*) AS numwait
FROM
      supplier,
lineitem 11,
      orders,
      nation
      S_suppkey = 11.1_suppkey
AND o_orderkey = 11.1_orderkey
AND o_orderstatus = 'F'
AND 11.1_receiptdate > 11.1_commitdate
AND EXISTS (
           SELECT
           FROM
                  lineitem 12
                12.1 orderkey = 11.1 orderkey
AND 12.1 suppkey \Leftrightarrow 11.1 suppkey
      AND NOT EXISTS (
           FROM
                   lineitem 13
           WHERE
                  AND 13.1 orderkey = 11.1 orderkey
AND 13.1 suppkey \Leftrightarrow 11.1 suppkey
AND 13.1 receiptdate >
13.1_committate
      AND s_nationkey = n_nationkey
AND n_neme = ':1'
GROUP BY
s_name
ORDER BY
      numwait desc,
s_name;
```

Big SQL terminologies - 1 of 2

- Before staring with Big SQL, let's cover some basic Big SQL terminologies.
- Warehouse
 - Default directory in the DFS in which the tables are stored
 - Defaults to /biginsights/hive/warehouse
- Schema
 - A directory under the warehouse in which tables are stored
 - Tables may be organized by schemas, or use a default schema
 - An example: /biginsights/hive/warehouse/myschema.db
- Table
 - A directory with zero or more data files
 - An example: /biginsights/hive/warehouse/myschema.db/tablename
 - Tables may be stored anywhere on the DFS

Big SQL terminologies - 2 of 2

Partitioned table

- A table may be partitioned on one or more columns
- The partitioning columns are specified when the tables are created
- Data is stored within one directory for the specified partition
- Query predicates can be used to eliminate the need to scan every partition
 - · Only scan what is needed.
- Example:
 - /biginsights/hive/warehouse/schema.db/tablename/col1=val1
 - /biginsights/hive/warehouse/schema.db/tablename/col1=val2

You made have heard or read about there being

two versions of Big SQL at one point. In fact, there is only one release, and it is being referred to as Big SQL. There may be cases where you see it being referred to as Big SQL 3.0. It is not a separate product, so there is no actual version numbers associated. Big SQL v1 refers to the previous version, but remember, there are no version numbers associated. This is only to refer to the previous version. Big SQL is a complete redesign from version 1. They share almost no code together. In fact, in order to use one or the other, you must specify the correct driver. Existing clients may continue to use Big SQL v1, but all new clients should be using Big SQL. The next slide will highlight some differences between the two. It is important to understand again, that

Big SQL is not a standalone product, so it does not have a product version number. BigInsights 3.0 includes both Big SQL and Big SQL v1. The user can choose which version to use, but by default, it is Big SQL. The most common reason to use Big SQL v1 would be for HBase support.

Big SQL does not have support for HBase. In order to use Big SQL on HBase tables, you must use Big SQL v1. The DDL in Big SQL is largely a superset of Big SQL v1. Remember, Big SQL supports modern SQL:2011 capabilities, so it conforms to industry standards. There may be some changes that you will need to make to your queries from Big SQL v1 in order for it to run in Big SQL. Some of these differences are highlighted here.

Big SQL v1 treats single and double quotes as the same. Big SQL reserves double quotes for identifiers. This adheres to SQL standard rules. In addition, the hadoop keyword is required for creating tables in Big SQL. You can lift that restriction by setting the compatibility mode. You can find more differences between these

two in the IBM Knowledge Center, with the biggest difference being the HBase support. If you do not need HBase support, use the current Big SQL.

What does Hive have to do with Big SQL? Essentially,

Big SQL shares the Hive Metastore, for table definitions, location, storage format, and encoding of input files. Big SQL maintains its own catalog in the master node. So as long as the data is defined in the Hive Metastore and accessible in the Hadoop cluster, both Big SQL and Hive can get to it. The DDL FMP, fenced mode process, essentially keeps everything

in sync. When you create a table in Big SQL, it makes sure that table is created in the Hive MetaStore and the Big SQL catalog. If you drop a table, the DDL FMP cleans it up in both places. Now, this is only true if the statement is performed through Big SQL. Unlike Big SQL 1.0, objects, created via Hive, are not immediately available. You must import those table definitions from Hive.

Here are the features of Big SQL at a glance. Big SQL comes with comprehensive SQL support and IBM SQL PL compatibility. I'll briefly go through this. Looking at

the top left with the Application Portability & Integration: Data is shared with the Hadoop ecosystem. Remember that Big SQL only provides a logical view to the data stored in Hadoop. There is a comprehensive list of file format support. For performance, Big SQL is designed to move away from MapReduce by using an MPP runtime. There is a SQL query rewriter, cost based optimizer, and the results are not constrained by memory. With federation your query can

be written for multiple data sources such as DB2 LUW, DB2/z, Teradata, Oracle, Netezza. Enterprise features include advanced security/auditing, resource and workload management, self-tuning

memory management and comprehensive monitoring.

Here are some of the SQL capability highlights. If you are familiar with regular SQL queries then this list might not seem that impressive. But, that just means that you are already familiar with Big SQL. This visual highlights the capabilities that Hive or Impala doesn't do, or doesn't do well. Big SQL has full support for subqueries. You can create a subquery from the SELECT, FROM, WHERE and the HAVING clause. You can create either correlated or uncorrelated subqueries. You can do equality or non-equality subqueries. You can use EXISTS, NOT EXISTS, IN, ANY, SOME. All the standard join operations are supported.

You have your inner, outer, and full outer joins. You have all the standard operators. You have the equality, non-equality and cross join support. You have your multi-value join. The sample code shows a valid Big SQL query.

Here are some Big SQL terminologies. Warehouse is the default directory in the DFS in which the tables are stored. The default location is /biginsights/hive/warehouse Schema is the directory under the warehouse directory in which the tables are stored. The tables may be organized by schemas, or it can use a default schema. An example would be /biginsights/hive/warehouse/myschema.db Table is a directory with zero or more data files. An example of a table directory within DFS is

/biginsights/hive/warehouse/myschema.db/tablename.

The tables may be stored anywhere on the DFS. This visual here is just showing an example of a location.

A table may be partitioned on one or more columns. You specify the partitioning columns when you create the table. When a table is partitioned, the data is stored within the directory for the specified partition. Query predicates can then be used to eliminate the need to scan every partition and only scan what is needed - speeding up the query. For example, if your table was partitioned by col1, there will be two directories on the DFS. The name of the two directories are, in fact, "col1=val1" and "col1=val2" for each respective partitions.

MODULE 2 – BIG SQL DATA TYPES

Working with Big SQL - JSqsh

- Big SQL comes with a CLI, known as JSqsh ("jay-skwish" Java SQL Shell)
- Open source command client.
- Query history and query recall.
- Multiple result set display styles (traditional, CSV, JSON, etc.)
- Multiple active sessions
- It can be started with:
 - \$JSQSH_HOME/bin/jsqsh

```
    bladmin@blvm:..insights/jsqsh/bin
    File Edit View Terminal Help
biadmin@bivm:/opt/ibm/biginsights/jsqsh/bin> ./jsqsh
JSqsh Release 2.1.2, Copyright (C) 2007-2014, Scott C. Gray
Type \help for available help topics. Using JLine.
1>
```

JSqsh (cont.)

- JSqsh is now preconfigured with two pre-defined connections
 - -bigsq1 Connects to the Big SQL 3.0 server
 - -bigsql1-Connects to the Big SQL 1.0 server
- So connecting is now as easy as:

```
[biadmin@myhost ~]$ $JSQSH_HOME/bin/jsqsh bigsql
Password: ******
...
[myhost][biadmin] 1>
```

 The connection will be established with your username, but you can override with the --user flag

```
[biadmin@myhost ~]$ $JSQSH_HOME/bin/jsqsh --user bob bigsql
Password: ******
...
[myhost][biadmin] 1>
```

JSqsh and semicolon

JSqsh's default command terminator is a semicolon

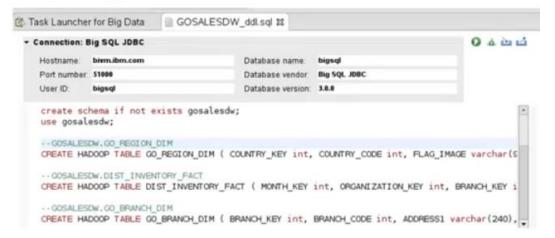
Semicolon is also a valid SQL PL statement terminator!

```
CREATE FUNCTION COMM_AMOUNT(SALARY DEC(9,2))
RETURNS DEC(9,2)
LANGUAGE SQL
BEGIN ATOMIC
DECLARE REMAINDER DEC(9,2) DEFAULT 0.0;
END;
Command terminator
```

- JSqsh applies a basic heuristics to determine the actual statement end
 - Sometimes it get it wrong!
 - You can tell because your statement doesn't execute when you think it should
 - You can...

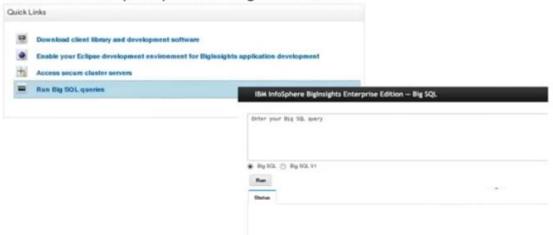
Working with Big SQL - Eclipse

- Big SQL commands and scripts can also be run from Eclipse (4.2.2)
- Results are formatted better in Eclipse
- Once the connection has been established, you can run the scripts directly from the Eclipse workspace.



Working with Big SQL - BigInsights

- There is a Big SQL console within the BigInsights Web console
- Located in under the Quick Links section on the Welcome page.
- A new tab will open up with the Big SQL console.



the next lesson in this course.

Lesson summary

Having completing this lesson, you should be able to:

- Describe the Big SQL architecture
- Know the differences between Big SQL and Big SQL v1
- Highlight some of the Big SQL features
- Understand Big SQL terminologies
- . Understand and use different methods for working with Big SQL

Big SQL has a CLI known as JSqsh. JSqsh is

an open source client that works with any JDBC driver, not just Big SQL. JSqsh has the capability to do query history and query recall. It displays results in various styles depending on the file type such as traditional, CSV, JSON, etc.). You can also have multiple active sessions. The CLI can be started with \$JSQSH_HOME/bin/jsqsh. You will get to use this in the lab exercise.

There are two pre-defined connections to which

you can connect using jsqsh. The bigsql connection connects to the Big SQL 3.0 server. The bigsql1

connects to the Big SQL 1.0 server. So connecting to either of the connection is a simple as providing the connection name after the jsqsh command. By default, you will connect with your OS username, but you can override that using the --user flag.

If you have used JSqsh before, you will know

that the default command terminator is a semicolon. With the new Big SQL, the current version,

the semicolon is also a valid SQL PL statement terminator, so while JSqsh can take a best guess to determine the actual statement, it can sometimes get it wrong. When that happens,

the statement will not run until you explicitly execute the "go" command. The semicolon in Big SQL is actually the go command underneath. Alternatively, you can change the default terminator from a semicolon to another terminator such as the backslash.

Another way to work with Big SQL is through

Eclipse. Most users actually prefer Eclipse because it provides the results in a better format. It is easier to read. Once you have set up the Big SQL connection within Eclipse, you can run Big SQL queries directly from there and get the results back in the Eclipse console.

Finally, a third way you can interact with

Big SQL is through the BigInsights web console. Once you log in to the BigInsights console on the Welcome page, click the Run Big SQL queries under the Quick Links section. This will open up a new window or tab with the Big SQL console. You can run your queries directly from there.

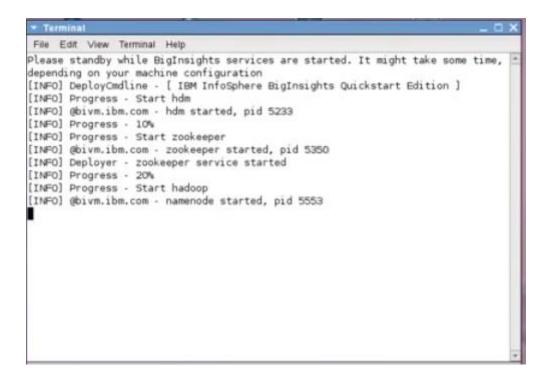
Having completed this lesson, you should now

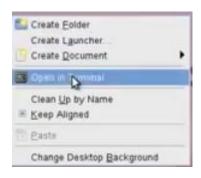
be able describe the Big SQL architecture. Know the differences between Big SQL and Big SQL v1. Highlight some of the Big SQL features. Understand Big SQL terminologies.

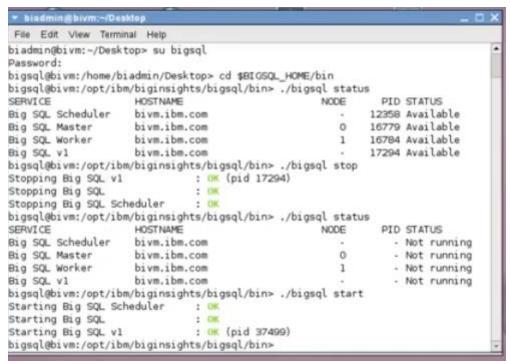
Understand

and use different methods for working with Big SQL.

You have completed this lesson. Proceed to







In this lab exercise, you will be connecting to the IBM Big SQL Server and running SQL queries. IBM Big SQL is a component of The IBM InfoSphere BigInsights product. Before you can start working with your Hadoop data using Big SQL, you need to be able to connect to the Big SQL server. There are three different methods for doing so and you will get a chance to work with all three in this lab exercise. The first of the three methods that you will use is JSqsh. JSqsh is an open source CLI for JDBC applications such as Big SQL. That means that you can also use JSqsh for other JDBC applications. The second method is to use Eclipse. Users generally prefer using Eclipse because the results are formatted in a way that makes it easier for the user to understand. The third method is the BigInsights web console. It is a web based tool where you can run your queries and see your results. You will need to first start up BigInsights. There are two ways you can do this. Using commands in the terminal or the shortcut icon on the desktop. The lab exercise guide shows you how to use the commands. I will use the desktop shortcut. I'll speed up the recording through the startup phase.

Now that BigInsights has started, open up a new terminal. I am going to show you another way you can manage the Big SQL service. First,

switch to the bigsql user. The password is bigsql.

Then, change directory to the \$BIGSQL_HOME/bin.

From within that directory execute bigsql status to get the status of the Big SQL service. You'll notice here that there is a Big SQL v1 instance. This current release of BigInsights comes with Big SQL and Big SQL v1, which is the older, legacy version. You would only

really use v1 if you need support for HBase or if your applications requires specific v1 features that is not yet supported in the current release of Big SQL.

To stop the Big SQL instance, type in bigsql stop.

Check the status again to see that the services have stopped.

Go ahead and restart Big SQL using bigsql start.

Close the terminal. We are done with this section. Proceed to the next video to continue on with the lab exercise.

MODULE 2 -

Data types

- There are three categories of data types
- Declared type
 - Data type provided in the CREATE statements
 - Some of the types provided here are mapped to a different SQL type that the d engine supports
- SQL type
 - This is the data type that the database engine supports
 - Currently more SQL types than declared types
- Hive type
 - This data type is defined in the Hive metastore for the table
 - This type tells SerDe how to encode/decode values for the type
 - The Big SQL reader converts values in the Hive types to SQL values on read

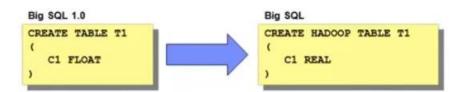
Data types (cont.)

Big SQL data types

Declared Type	SQL Type	Hive Type
TINYINT	SMALLINT	TINYINT
SMALLINT	SMALLINT	SMALLINT
INT	INT	INT
BIGINT	BIGINT	BIGINT
REAL	REAL	FLOAT
FLOAT	FLOAT	DOUBLE
DOUBLE	DOUBLE	DOUBLE
CHAR(n)	VARCHAR(n)	STRING
VARCHAR(n)	VARCHAR(n)	STRING
STRING	VARCHAR(max)	STRING
TIMESTAMP[(n)]	TIMESTAMP[(n)]	TIMESTAMP
DATE	DATE	TIMESTAMP
BOOLEAN	SMALLINT	BOOLEAN

REAL and FLOAT types

- In Big SQL 1.0 REAL and FLOAT were both 32bit IEEE floating point
- In Big SQL
 - REAL is a 32bit IEEE floating point
 - -FLOAT is a synonym for DOUBLE (64bit IEEE floating point)
- Importing existing Hive tables containing FLOAT will define them as REAL in the Big SQL catalogs



CHAR type

- Big SQL 1.0 allowed a CHAR data type
 - It was a synonym for VARCHAR
 - It did not follow proper CHAR semantics at all
- Big SQL does NOT allow CHAR by default
 - CHAR will be supported in the future, but it will follow the proper CHAR semantics

```
[localhost][db2inst1] 1> create hadoop table chartab (c1 char(10));

SQL Exception(s) Encountered:
[State: 42858][Code: -1667]: The operation failed because the operation is not supported with the type of the specified table. Specified table:
"DB2INST1.CHARTAB". Table type: "HADOOP". Operation: "CHAR DATATYPE".. SQLCODE=-1667, SQLSTATE=42858, DRIVER=3.67.33
```

But, you can still enable support if need be:

```
[localhost][db2inst1] 1> set syshadoop.compatibility_mode=1;
[localhost][db2inst1] 1> create hadoop table chartab (c1 char(10));
0 rows affected (total: 4.97s)
```

STRING type

- Only provided for compatibility with Hive and Big SQL 1.0 syntax
- By default, STRING becomes VARCHAR(32,672)
 - Largest size that the database engine supports
- Avoid the use of STRING!!!!!!
 - It can cause significant performance degradation
 - The database engine works in 32k pages
 - Rows larger than 32k incur performance penalties and have limitations
 - Hash join is not an option on rows where the total schema is > 32k
- Some alternatives:
 - The best option is to use VARCHAR that matches your actual needs
 - The bigsql.string.size property can be used to adjust the default down
 - Property can be set server wide in bigsgl-conf.xml

```
[localhost][db2inst1] 1> set hadoop property bigsql.string.size=16;
[localhost][db2inst1] 1> create hadoop table t1 (fname string, lname, string);
```

DATE type

- The DATE type is defined in Hive as a TIMESTAMP
 - Even though Hive has a DATE type in 0.12, the native I/O library does not yet support the Hive DATE type
- This means data files with DATE values must be defined with a full time

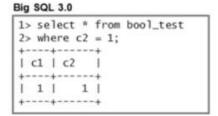
```
1> create external hadoop table date_test (
2> c1 int, c2 date
3> ) row format delimited
4> fields terminated by ','
5> location '/date_test';
```

BOOLEAN type

- The BOOLEAN type is defined as a SMALLINT SQL type
 - In Big SQL 1.0, it was treated as a BOOLEAN
 - In the Hive Metastore, it is still a BOOLEAN and treated as such

```
1> select * from bool_test;
1> create external hadoop table bool_test (
   cl int, c2 boolean
                                                 | c1 | c2
3> ) row format delimited
    fields terminated by ','
4>
                                                   1 |
                                                        1 |
5> location '/bool_test';
                                                   2 1
                                                         0 |
$ hadoop fs -cat /bool_test/bool_test.csv
                                                 1 3 | NULL |
2,false Legal storage values
                                                 | 4 | NULL |
3.0
4,1
```

In queries, BOOLEAN must be treated as a SMALLINT



BOOLEAN and TINYINT in CTAS

- In CREATE HADOOP TABLE AS SELECT (CTAS) the BOOLEAN and TINYINT types lose their Hive type definitions
 - -This is a known issue that should be fixed in the next release

You can work around this with an explicit column list in the CTAS statement

```
1> create hadoop table bool_ctas2 (c1 boolean)
2> as select * from bool_ctas;
```

Welcome to the course, Big SQL on Hadoop.

In this lesson, Big SQL data types, you will learn about the different Big SQL data types. After completing this lesson, you will be

able to list and explain the Big SQL data types. Create Big SQL schemas. Create Big SQL tables. Understand and use the file formats supported by Big SQL.

There are three categories of data types around

Big SQL that you will see here. Declared type: data types that are provided in the CREATE statements. These particular types only exists in the CREATE statement and nowhere else. Some of the types provided here are mapped to a different SQL type that the engine supports. For example, you declare with one type, but in order for the engine to use that type, it is mapped to a different type, which is supported.

SQL types: data types that the database engine supports. There are currently more SQL types than declared types. That just means that there are some types that you cannot use in the CREATE statement, but you can use otherwise, such as the DECIMAL type.

Hive type: data types that the Hive catalog understands. When you create a table, Big SQL has to tell the Hive catalog what that column is. It basically tells the Hive SerDe how to interpret the value. The Big SQL reader converts the values in the Hive types to SQL values on read.

Here is a table that shows the relationship between the three data types categories. As a quick refresher, the declared type is the type provided to the CREATE statement. Some of those types are mapped to the SQL type that the database engine understands. When a table is created, the data types are converted to Hive types in order for the metadata to be stored in the Hive catalog. For example, when you create a table, Hive has a type called a STRING. The Big SQL runtime does not have this notion of a STRING. It cannot support arbitrary string, so when it is defined in the catalog, it is defined as varchar. In other words, when you create a table, and specify a string column, the Big SQL engine actually creates that column as a varchar of 32KB.

The types on this chart are just a subset of all the types. There are many more SQL types than there are declared types -- meaning there are some that you cannot use in the CREATE statement. I will go over a few more of the types on the following slides. Remember, Big SQL 1.0 just refers to the previous

version of Big SQL. There is no version number associated with Big SQL, it is not a separate product. In Big SQL 1.0, REAL and FLOAT were both 32bit IEEE floating point. In Big SQL, REAL is a 32bit IEEE floating point. FLOAT is essentially a DOUBLE, a 64bit IEEE floating point. This means that if you have to have the same data type in Big SQL, you should change from FLOAT to REAL. If you have a Hive CREATE table statement that contains FLOAT, you should change it to REAL if you want it to continue to have 32bit value. Otherwise, it will be created as a DOUBLE, a 64bit value.

In Big SQL 1.0, the CHAR type did not follow

the proper CHAR semantic at all. It was basically a synonym for VARCHAR. Because of that reason,

the current Big SQL does not allow CHAR, by default.

If you try to create table with a CHAR column, you will get an error. CHAR will be supported in the future using the proper CHAR semantics. However, if you happen to have a lot of CREATE

table statements with columns defined as CHAR and you desperately need to use it, there is an option to enable support. Just set the compatibility_mode=1 and you can run the statements

without any errors.

Now we get to the STRING type. As I mentioned, Big SQL does not have this notion of a STRING,

so when it sees a STRING type in the CREATE command, it converts it into a VARCHAR(32,672), which is the largest size the database engine support. This becomes increasingly inefficient and can cause performance degradation. The database engine works in 32k page so rows larger than 32k incur performance penalties. For example, hash join is not possible on rows where total schema is > 32k. Here are some alternatives you can use instead of using the STRING type. The best option is to use VARCHAR with the size that you need. If you must use STRING, you could adjust the default size down to what is appropriate for your needs. There is also a server wide setting in the bigsql-conf.xml file.

Moving on to the DATE type. In Hive, the DATE

type is mapped to a TIMESTAMP. This means that when you define a DATE column, the values must be defined with a full timestamp. Hive does have a DATE type, but the native I/O library does not yet support that. Here is an example. In the top box, you can see the CREATE command to create an external Hadoop table with a DATE column. On the left side, you can see that you have a file with a list of dates only, without a time. If you query it, you get NULLs back. The right way to do this is to append the time value to the date as shown on the right side. When you query that, you get the appropriate date values back.

The time value that you provide must be in

the format that you see here: 00:00:00.0[0...]. The LOAD HADOOP command automatically takes

care of converting the files with DATE to TIMESTAMP format so you do not have to worry about that if you use the LOAD command. For everything else, you will need to make sure you include the time value.

The database runtime engine does not fully

support the concept of a BOOLEAN. The BOOLEAN type is defined as a SMALLINT SQL type in Big SQL. In Big SQL 1.0, it was defined as a BOOLEAN

type. In the Hive metastore, it is also a BOOLEAN. But in Big SQL queries, it is treated as a SMALLINT. For example:

You create a table with a BOOLEAN column. The only valid values for those particular columns are true and false. 0's and 1's are not valid BOOLEAN values. The example table have both true/false and 0/1. So when you select the table, you see the values of 1 and 0 for true and false. But for the 0 and 1 value, the database engine does not recognize it (since it isn't a valid BOOLEAN), so it returns NULL.

Remember, the runtime engine uses SMALLINT so true is mapped to the value of 1 and false is 0. In queries, you must take care to change any

BOOLEAN to SMALLINT. For example, if you are doing a comparison, you must change from true

to 1 (false to 0) for the queries to work properly.

In CREATE HADOOP TABLE as SELECT (CTAS), the

BOOLEAN and TINYINT type loses their Hive type definitions. The declared types BOOLEAN and TINYINT are both mapped to SMALLINT in the SQL runtime engine.

The example shows the first table being created with a single BOOLEAN column.

Then it creates a second table using the CTAS statement on the first table.

The second table loses the BOOLEAN definition and becomes TINYINT.

A way to circumvent that is to use an explicit column list in the CTAS statement as shown in the bottom screenshot. This allows it to keep the BOOLEAN type.

Schemas

- Tables are organized into a user-created schema or use a default schema
 - Big SQL default schema matches your login name
 - The USE command can be used on a schema that does not exist.

• The schema will be created if it does not exist.

```
You can think of schema as a database!
```

```
[myhost][biadmin] 1> use "schema doesn't exist";
[myhost][biadmin] 1> create hadoop table t1 (c1 int);
[myhost][biadmin] 1> insert into t1 values (10);
[myhost][biadmin] 1> select * from "schema doesn't exist".t1;
+----+
| C1 |
+----+
| 10 |
+-----
```

To create a schema, just use the CREATE SCHEMA command

```
[myhost][biadmin] 1> create schema demo;
0 rows affected (total: 0.32s)
```

CREATE TABLE

- HADOOP keyword
 - Must be specified unless you enable the SYSHADOOP.COMPATIBILITY_MODE
- EXTERNAL keyword
 - Indicates that the table is not managed by the database manager
 - When the table is dropped, the definition is removed, the data remains unaffected.
- LOCATION keyword
 - Specifies the DFS directory to store the data files

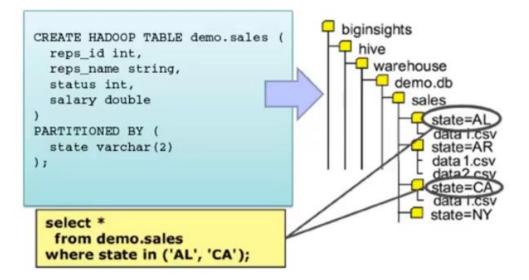
```
CREATE EXTERNAL HADOOP TABLE T1

(
    C1 INT NOT NULL PRIMARY KEY CHECK (C1 > 0),
    C2 VARCHAR(10) NULL,

...
)
...
LOCATION
'/user/biadmin/tables/user'
```

CREATE TABLE - partitioned tables

- Similar to Hive, Big SQL also has partitioned tables.
- Partitioned on one or more columns.
- Query predicates is used to eliminate unwanted data speeding up the query.



New CREATE TABLE features

- Constraints may now be defined in-line in the table definition
 - PRIMARY KEY/UNIQUE
 - FOREIGN KEY
 - CHECK (new to Big SQL 3.0)
 - Constraints are advisory only
 - Query optimizer can take advantage

```
CREATE HADOOF TABLE T1
(
    c1 INT NOT NULL PRIMARY KEY CHECK (C1 > 0),
    c2 VARCHAR(10) NULL,
    cONSTRAINT FK_T2 FOREIGN KEY (C2)
    REFERENCES T2 (C1)
)
```

- New table types available for STORED AS
 - TEXT SEQUENCEFILE
 - -ORC
 - PARQUETFILE
- NULL DEFINED AS clause for ROW FORMAT DELIMITED
 - Explicit syntax for defining a NULL value in a delimited file
- Significant improvements to CREATE TABLE AS SELECT (CTAS)
- Support for CREATE TABLE LIKE to clone another table
- SECURITY POLICY clause for row level security policies

Example - creating a table

Enter the schema and create a table

```
[myhost][biadmin] 1> use demo;
0 rows affected (total: 0.0s)
[myhost][biadmin] 1> create hadoop table demotab (
[myhost][biadmin] 2> cl varchar(20) not null primary key,
[myhost][biadmin] 3> c2 int
[myhost][biadmin] 4> )
[myhost][biadmin] 5> row format delimited
[myhost][biadmin] 6> fields terminated by ',';
0 rows affected (total: 0.39s)
```

In another window, let's see the effect:

```
[biadmin@myhost ~]$ hadoop fs -ls /biginsights/hive/warehouse/demo.db

Found 1 item

drwxr-xr-x - bigsql biadmin 0 2014-05-30 14:28 /user/hive/warehouse/demo.db/demotab
```

Example - inserting data

D

Inserts are use for testing only.

```
[myhost][biadmin] 1> insert into demotab values ('foo', 10), ('bar', 20), ('baz', null);
3 rows affected (total: 0.30s)
[myhost][biadmin] 1> insert into demotab values ('a', 1), ('b', 2), ('c', 3);
3 rows affected (total: 0.36s)
```

Each insert statement becomes a data file in the file system

```
[biadmin@myhost ~]$ hadoop fs -ls /biginsights/hive/warehouse/demo.db/demotab
Found 2 items
-rw-r--r- ... /user/hive/warehouse/demo.db/demotab/i_1403110143227_49...
-rw-r--r- ... /user/hive/warehouse/demo.db/demotab/i_1403110143227_53...
```

If you open up the file, you will see this:

Example - manually adding data

You can add data just by copying files into the table directory

```
[biadmin@myhost ~]$ cat demotab.dat
frank,12
mary,12
bob,14
nancy,81
[biadmin@myhost ~]$ $HADOOP_HOME/bin/hadoop fs -copyFromLocal demotab.dat \
/user/hive/warehouse/demo.db/demotab
```

 The scheduler may not be aware of your new data right away, so you may need to let it know to invalidate its cache.

```
[myhost][biadmin] 1> call syshadoop.hcat_cache_sync('DEMO', 'DEMOTAB');
ok. (total: 0.15s)
```

Then query

Example - dropping the schema



Cleanup is easy, just drop the schema and everything in it

```
[localhost][db2inst1] 1> drop schema demo cascade;
0 rows affected (total: 1.64s)
```

And, let's make sure it really cleaned up

```
[biadmin@myhost ~]$ HADOOP_HOME/bin/hadoop fs -ls /user/hive/warehouse/demo.db ls: '/user/hive/warehouse/demo.db': No such file or directory
```

You should know that schemas are a way to

organize tables. The default schema now matches your login name.

In Big SQL 1.0, the default schema was just "default".

If you want to specify another default schema, you use the USE command.

If you execute the USE command on a schema that does not currently exist, it gets created.

Once the USE command has been invoked, the specified schema becomes the default.

In the first screenshot, a new (non-existing) schema is created via the USE command. Then a table is created with an unqualified table name. (No schema was explicitly specified.)

This results in the table being created under the current default schema. If you had qualified the table, then it will be created in that schema instead of the default schema.

The second screenshot shows creating a schema using the CREATE SCHEMA command. When a schema is created, by default, a directory is defined in /user/hive/warehouse. The name

of the created directory is schemaName.db where schemaName is the name that you specified.

So for the schema called demo, by default you should see a directory /user/hive/warehouse/demo.db.

It is possible to define a schema and explicitly place the schema directory elsewhere in HDFS besides the Hive warehouse.

You will need to specify the HADOOP keyword

to create a table for the Hadoop environment. You can omit that if you enable the environment

variable SYSHADOOP.COMPATIBILITY_MODE. Optionally, you can specify the EXTERNAL keyword to create

an external table. This indicates that the table is not managed by the database manager. When the table is dropped, the definition of that table is removed, but the data remains untouched. Typically the EXTERNAL keyword is used in conjunction with the LOCATION keyword

to specify the directory within the DFS to store the data files.

Big SQL has another similarity with Hive, partitioned

tables. Big SQL can created partitioned tables to split up the data into multiple files.

The benefit of this is the access to the data does not require all of the data to be read.

The query predicate will determine which partition to scan through.

Let us look at the sample code here: In this code, you are creating a partitioned table on the state column. Notice that you do not specify the state column in the original table definition, but in the PARTITIONED BY clause. When you run the query to select data from a particular state or states, only the partitions specified in the query are retrieved. In the example, select * from demo.sales where state in ("AL",

"CA") only the partitions where state=AL and state=CA are retrieved.

There are some new CREATE TABLE features in

Big SQL that were not available in Big SQL 1.0.

Namely, constraints could not be defined in-line in the table definition.

In Big SQL 1.0, constraints were only allowed via ALTER statements.

In the current Big SQL, you can define: PRIMARY KEY/UNIQUE,

FOREIGN KEY, CHECK (which is new to Big SQL 3.0)

Note that constraints are only advisory, or applied on read, and not actually enforced.

The query optimizer, however, can leverage these constraints to create better query plans.

So even though they are not enforced, they are useful for query optimization.

There are also new table types available for the STORED AS clause. Big SQL supports TEXT SEQUENCE FILE, Hive's ORC file, PARQUETFILE, which Twitter and Cloudera have been working on, and a few others as well. These file formats will be discussed in more detail in the following slides. The NULL DEFINED AS syntax allows you to explicitly

declare how a NULL value is to be represented. Significant improvements have been made to the CREATE TABLE AS SELECT(CTAS). There is support for CREATE TABLE LIKE to clone another table. There is also a SECURITY POLICY clause for

row level security policies.

Here is an example of creating a table.

The default schema was overridden to become demo via the USE command. Then an unqualified

table, demotab, was created using the hadoop keyword.

This means that the table is defined to both Big SQL and Hive. If the hadoop keyword had not been specified, then the table would have only been defined in the database management system. The table has two columns, a primary key column

of type varchar(20) and an int column. The table is row format delimited and the fields are terminated with a comma. Once the table has been created, assuming that the demo schema

was defined in its default location, it would be located as a directory within the default warehouse directory, /user/hive/warehouse/demo.db/demotab. It is possible to create a table with an explicit

schema as well. So assuming that the default schema is demo but you wanted to create a table, t1, under the xyz schema, you could code:

create hadoop table xyz.t1 (...)

Here is an example of inserting some data.

Remember that the insert statement is good for testing only. It is extremely inefficient for querying and using in a production environment. Normally you would want to use the bulk operations

like LOAD or INSERT from SELECT, which you will see in the lab exercise associated with this unit. So here you have two sample inserts into the

demotab table with 3 columns. Notice that the third column contains a NULL.

Each of these inserts produces a data file in the file system. So if you do a listing

in the HDFS in the default warehouse directory, you should see two data files in it.

If you open up the file, you should see the actual values. Notice that the NULL is represented by the \N in the Hive system. You can specify a different NULL representation if you wish. Here is an example of manually adding data

by copying the files into the table directory. In the first screenshot, you do a cat of the data file to see its content. Then copy that file into the database directory within the HDFS. The scheduler may not be aware of the new

data, so you need tell it to invalidate its cache. You do so by executing the command: call syshadoop.hcat_cache_sync('DEMO', 'DEMOTAB');

Then you can run the query against that table to see the new data.

Cleanup is simple. This will remove the schema

as well as all tables. You just execute the statement:

drop schema demo cascade; To make sure it is really gone, just go back to do a listing of the demo.db directory.

Big SQL file formats

- Delimited
- Sequence
 - Delimited
 - Binary
- Parquet
- ORC
- RC
- Avro

Delimited

```
create hadoop table user (
   user_id int,
   fname varchar(10),
   hire date
)
row format delimited
  fields terminated by '|'
```

- Human readable textual data where the column values are separated by a delimiter character.
 - -Example of this would be CSV (Comma Separated Values)

Pros:

- · Supported by I/O engine
- · Human readable
- · Supported by most tools

Cons:

- · Least efficient file format
- · Data conversion to binary is costly

Delimited table syntax

- TERMINATED BY indicates delimiters
 - Default delimiter: ASCII 0x01 (CTRL-A)
 - Delimiters can be specified as
 - · Characters (e.g. '|')
 - · Common escape sequences (e.g. '\t')
 - · An octal value (e.g. '\001')
 - Use '\\' to specify a literal backslash

```
CREATE HADOOP TABLE DELIMITED (
C1 INT NOT NULL,
C2 VARCHAR (20) NOT NULL,
C3 DOUBLE NULL
)
ROW FORMAT DELIMITED
FIELDS TERMINATED BY ',',
NULL DEFINED AS '#NULL#'
```

- Your data must not contain the delimiter character!!
 - It is possible to escape the character.
- NULL DEFINED AS indicates how a literal NULL is represented
 - Big SQL 1.0 defined this as an empty field (") by default
 - Big SQL uses the Hive default of '\N'

- Hive defines the format and rules around a delimited file
- The data type of a column dictates how a value must be provided
 The following provides examples of allowable formats for each data type

Туре	Format	Туре	Format
tinyint-bigint	-12, 12, +12	boolean	true, false, TRUE, FALSE
float, real, double	-12.12, +12.12, 12.12, 12e-3, etc.	timestam p	yyyy-mm-dd hh:mm:ss.fffff
string, [var]char	any	date	yyyy-mm-dd 00:00:00.0

Remember. Hive is told our DATE is a TIMESTAMP!

■ Invalid values are treated as NULL on read

— If column is defined as NOT NULL, query will fail!

| Create hadoop table user (
| user_id int, |
| fname | varchar(10), |
| hire | date |
| row format delimited |
| fields terminated by '|'

g-

Delimited file format (cont.)

• The Hive delimited file format does not support quoted values, like:

```
1, "Hello, World!", 3
```

- This would be treated as four column values, not three
- You may specify an escape character with ESCAPED BY

```
create hadoop table messages
(
  msg_id int,
  msg varchar(200),
  msg_sev int
)
row format delimited
  fields terminated by ',' escaped by '\\'
```

Sequence files

```
CREATE HADOOP TABLE text_seq
(col1 int, col2 varchar(20))
STORED AS SEQUENCEFILE
```

- What is it?
 - Sequence file is a container for any record format (things SerDe's deal with)
 - Keep track of record boundaries to produce splits
 - Useful for storing data that is tough to split otherwise (e.g. XML or JSON)
 - Blocks of records can be compressed using non-splittable compression
- Pros
 - Supported by native I/O engine when storing delimited data only
 - Supported by Java I/O engine for other storage formats
 - Sequence files are relatively common in Hadoop
- Cons
 - One of the slower storage formats
 - Not human readable

Sequence files (cont.)

- There are two "built-in" sequence file variants in Big SQL
 - Delimited (text) sequence file
 - Data is stored as textual delimited data

 - Really only useful for compressing with a non-splittable compression algorithm
 Note: In Big SQL 1.0 "SEQUENCEFILE" referred to a binary file (below), now it means text sequence file to align with Hive/Impala behavior

```
CREATE HADOOP TABLE text seq
  (col1 int, col2 varchar(20))
STORED AS SEQUENCEFILE
```

CREATE HADOOP TABLE text seq (coll int, col2 varchar(20)) STORED AS TEXT SEQUENCEFILE

Binary sequence file

- · Data is stored using Hive's binary serialization format (LazyBinarySerDe)
- Slightly better performance via binary values

```
CREATE HADOOP TABLE bin seq
  (coll int, col2 varchar(20))
STORED AS BINARY SEQUENCEFILE
```

Here are the Big SQL file formats that will be covered in detail in the following slides. Delimited, Sequence Delimited, Sequence Binary, Parquet, ORC, RC, and Avro

A delimited file is essentially a human readable

text file where the column values are separated by a delimiter character. A common

of this would be CSV (Comma separated values). Pros:

- Supported by the I/O engine Human readable
- Supported by most tools Cons:
- Least efficient file format Data conversion to binary is costly

The TERMINATED BY clause allows you to specify

the type of delimiter your file uses. The default delimiter is the ASCII 0x01 (CTRL-A).

Typically you would probably use a pipe symbol or some escape characters. If you need to use a backslash, you would use a double backslash. Your data must not contain the delimited character,

or else the data will be messed up. You can escape characters and I will go over this in a bit. There is a NULL DEFINED AS clause that lets

you indicate how you want a NULL value to be represented. In Big SQL, the default Hive representation of a NULL value is the \N. In the previous version of Big SQL, it was an empty field by default.

Here is the file format for a DELIMITED file.

Hive defines the format and rules around a delimited file. In the chart are examples of allowable formats for each data type. Remember that Hive is told that our DATE is a TIMESTAMP.

so the format is a timestamp. If you provide data in an invalid format,

the values are treated as NULL on read. This causes the query to fail if your column is not defined as NOT NULL. In this example, you see that a table has

three columns, a user id of type int, a fname of type varchar(10) and a hire of type date.

Then in the data file, you have three rows. The first row contains the values of type

int, \N and the timestamp. This gets added correctly because the data format matches the columns' data types. The second row, in the first column, it is expecting an int, but it gets a char, so when you insert the data, you get a NULL. If the column was defined with a NOT NULL, the query fails. In the third row, the last column's date is just a date, without an additional time value. It is inserted as a NULL because Hive expects a TIMESTAMP for the DATE column.

Here you see how to escape certain characters

if they are being used as a delimiter. Hive does not support quoted values, so even if you provide a quotation around a value, it still counts any delimited within the quotation marks. For example, if you needed to insert the value

1, "Hello, World!", 3 This would be treated as four columns because of the three commas that separate the data. To do this successfully, you must specify an escape character using the ESCAPED BY clause. So assuming your escape character is the backslash,

you would insert the row as: 1, Hello\, World!, 3

That correctly inserts the rows into the three columns of that table.

Sequence file is a container for any record

format with which SerDe's deal. It maintains the boundaries of the files to produce splits effectively. This file format is useful for storing data that is normally tough to split like XML or JSON. There is an option to compress blocks of records using non-splittable compression.

Some of the pros of using sequence files are that it is supported by the native I/O engine when storing delimited data. For other storage formats, it is supported by the Java I/O engine.

It is also one of the more commonly used formats in Hadoop.

A couple of cons of sequence files are that they are relatively slow compared to other formats and they are not human readable.

Sequence files come in two flavors. You have

the delimited (text) sequence file and the binary sequence file.

The delimited sequence file stores the data as textual delimited data. This is only useful for compressing with non-splittable compression algorithms.

In Big SQL 1.0, SEQUENCEFILE referred to the binary file, which is the second of the two sequence file formats. In the current release of Big SQL, SEQUENCEFILE is in text format. The binary sequence file stores the data using Hive's binary serialization format using the LazyBinarySerDe. It yields slightly better performance than the text version.

Parquet files

```
CREATE HADOOP TABLE parquet
(col1 int, col2 varchar(20))
STORED AS PARQUETFILE
```

- · What is it?
 - New storage format for Big SQL
 - Columnar file format
 - Efficient compression and value encoding
 - Has additional optimizations for complex data types

Pros:

- · Supported by native I/O engine
- · Highest performance file format

Cons:

- Not good for data interchange outside of Hadoop
- Does not support DATE or TIMESTAMP data types today

Creating a Parquet table

Creating a parquet table is easy!

```
CREATE HADOOP TABLE parquet
(coll int, col2 varchar(20))
STORED AS PARQUETFILE
```

- Ways of populating a parquet table:
 - Using INSERT from SELECT

```
INSERT INTO new_parquet_table
SELECT * FROM existing_delimited_table
```

Using CREATE TABLE AS SELECT

```
CREATE HADOOP TABLE new_parquet_table
STORED AS PARQUETFILE
AS SELECT * FROM existing_text_table
```

- INSERT VALUES
 - Produces a very, very inefficient output (good for testing though)

```
INSERT INTO parquet VALUES (1,'a'), (2,'b')
```

Loading a Parquet table

```
LOAD HADOOP USING FILE URL

'sftp://bigsql.svl.ibm.com:22/biadmin/data/test.tbl'

WITH SOURCE PROPERTIES ('field.delimiter' = '|')

INTO TABLE parquet APPEND;
```

- Best practices for LOADing Parquet data files
 - It is essential to tune the num.map.tasks parameter of LOAD.
 - · Set the value to at least the number of cata nodes in your cluster
 - · Performance can be improved further by setting the value to multiples of number of data nodes
 - It is recommended to have large Parquet file size (slightly less than 1GB). Parquet data files use 1GB block size
 - When loading large Parquet data files
 - Consider copying files to HDFS and then load from there for better performance than loading from local disks
 - The file option of Big SQL Load can be used to perform data conversion from text files (source) to the optimal format defined for the new Parquet table

Parquet and compression

- Parquet files are compressed with SNAPPY by default
 Other algorithms: GZIP, UNCOMPRESSED
- You can explicitly specify a particular compression

```
SET HADOOP PROPERTY parquet.compression = SNAPPY;
INSERT INTO parquet_snappy SELECT * FROM text_table;
```

It is okay to populate the same table with multiple algorithms

```
SET HADOOP PROPERTY parquet.compression = SNAPPY;
INSERT INTO parquet_snappy SELECT * FROM text_table;

SET HADOOP PROPERTY parquet.compression = GZIP;
INSERT INTO parquet_snappy SELECT * FROM text_table;
```

ORC File

CREATE HADOOP TABLE orc_table (coll int, col2 varchar(20)) STORED AS ORC

- What is it?
 - New storage format for Big SQL
 - Columnar file format
 - Per block statistics
- Pros
 - Efficiently retrieve individual columns
 - Efficient compression
- Cons
 - Supported only by Java I/O engine
 - Not good for data interchange outside of Hadoop
 - Big SQL cannot exploit some advanced ORC features today
 - · Predicate pushdown to skip unneeded values
 - · Block elimination via predicate pushdown (Hive 0.13 feature)

RC File

- · What is it?
 - Progenitor to ORC file format
 - Columnar file format
 - Efficient compression and value encoding

Pros:

- · Supported by native I/O engine
- · Efficiently retrieve individual columns
- · Efficient compression

Cons:

 Not good for data interchange outside of Hadoop

CREATE HADOOP TABLE orc_table

STORED AS RCFILE

(coll int, col2 string, col3 bigint)

- Format is provided for backward compatibility
 - ORC is the replacement for RC
- Compression:
 - -snappy
 - -gzip
 - deflate
 - bzip2

Wanne

Avro tables

- What is it?
 - Avro is a standardized binary data serialization format (http://avro.apache.org)
 - Well defined schema for communicating the structure of the data
 - Code generators for reading/writing data from various languages
 - Well accepted data interchange format
- Pros
 - Supported by the native I/O engine
 - Popular binary data exchange format
 - Table structure can be inferred from existing schema
 - Decent performance
- Cons
 - Not as efficient as Parquet or ORC
 - Not human readable

Creating an Avro Table - inline schema

• An Avro table can be created with an in-line schema:

- Note that column definitions can be inferred from the schema!
 - Big SQL infers this schema once, at table creation time
 - Hive infers it every time the table is read
 - If you want to change the schema of a table in Big SQL, you must drop and re-create it

Avro schema data type mapping

• When no column is list provided, the table columns will be inferred as:

Avro Type	Declared Type	Avro Type	Declared Type
boolean	BOOLEAN	double	DOUBLE
int	INT	string	STRING
long	BIGINT	enum	STRING
float	REAL		

- As discussed earlier STRING can be bad for performance!
 - You can either use bigsql.string.size to knock the default size down
 - Or explicitly declare the columns:

```
CREATE EXTERNAL HADOOP TABLE avro_external
(username varchar(20), tweet varchar(140), timestamp bigint)
ROW FORMAT SERDE 'org.apache.hadoop.hive.serde2.avro.AvroSerDe'
STORED AS
INPUTFORMAT 'org.apache.hadoop.hive.ql.io.avro.AvroContainerInputFormat'
OUTPUTFORMAT 'org.apache.hadoop.hive.ql.io.avro.AvroContainerOutputFormat'
LOCATION '/user/biadmin/avro/data/'
TBLPROPERTIES (
   'avro.schema.url'='hdfs:///user/biadmin/avro/twitter.avsc'
);
```

Parquet file is the new storage format for

Big SQL. It is a columnar storage format with efficient compression and value encoding.

It also has optimizations for complex data types.

Pros: - Parquet file is supported by the native

I/O engine - Highest performance among the file formats.

Cons: - Not good for data interchange outside of

Hadoop. - Does not support the DATE and TIMESTAMP

types today.

Here are some examples of how to work with

Parquet tables. You saw how to create a parquet table in the previous slide.

CREATE HADOOP TABLE parquet (col1 int, col2 varchar(20)) STORED AS PARQUETFILE;

Essentially everything is the same as creating a regular table, except you specify the STORED

AS PARQUETFILE clause. There are three ways to populate a parquet

table. 1. Using INSERT from SELECT statement to basically

transform an existing delimited table into a parquet table.

INSERT INTO new_parquet_table SELECT * FROM existing_delimited_table;

2. Using a CREATE TABLE AS SELECT CREATE HADOOP TABLE new_parquet_table STORED AS PARQUETFILE AS SELECT * FROM existing_text_table; 3. Using INSERT to test out your queries.

INSERT INTO parquet VALUES (1, 'a'), (2, 'b');

Reminder: INSERT is only good for testing. It is highly

inefficient to be using for production.

Another alternative to getting data into Parquet

tables is using the LOAD operation. There are a few best practices for using LOAD.

It is essential that you tune the num.map.tasks parameter of LOAD. Set the value to at least the number of data nodes in your cluster. This allows at least one map task to run for each data node in the cluster. Setting it to multiples of the number of data nodes can further improve performance. Parquet block size is 1GB, so try to have

the Parquet file size close to 1GB increments to run at optimal performance. Loading from HDFS has better performance, so when you can, load the data files into HDFS first before performing the LOAD operation. You can use the file option of the Big

SQL load in order to perform data conversion from text to Parquet.

A few notes on Parquet and compression. By

default, Parquet files are compressed with SNAPPY. You can specify other compression algorithm such as GZIP or UNCOMPRESSED. Set the HADOOP PROPERTY parquet.compression flag

to the compression algorithm of choice. It is also valid to populate the same table with multiple algorithms. Just set the property before each insert or load operation.

Optimized Row Columnar (ORC) is a new storage

format for Big SQL. It has a columnar storage format which stores a collection of rows in one file and within the collection the row data is stored in a columnar format. This allows parallel processing of row collections in a cluster. ORC keeps track of per block statistics. Pros:

- ORC files allow for effective retrieval of individual columns
- Efficient compression. Cons:
- ORC is only supported by the Java I/O engine Not good for any data exchange outside of Hadoop. Big SQL cannot exploit some of the advance

ORC features today. Those unsupported advanced features are predicate pushdown, to skip unneeded

values, and block elimination via predicate pushdown. Both are supported in Hive 0.13. ORC defines compression at DDL time. You specify

it in the TBLPROPERTIES clause. The available compression schemes are NONE, ZLIB, or SNAPPY.

ORC is replacing the Record Columnar (RC)

format. Essentially the same columnar file format with efficient compression and value encoding. Pros:

- Supported by the native I/O engine. Efficiently retrieve individual columns
- Efficient compression Cons:
- Not good for data exchange outside of Hadoop. There is a backward compatibility for RC with ORC. The following compression algorithms are supported: snappy, gzip, deflate, bzip2.

Avro is an open source, standardized binary

data serialization format. It is a well-defined schema for communicating the structure of the data. There are code generators available for reading and writing data from various languages. It is also a well-accepted data interchange format.

Pros: It is supported by the native I/O engine.

It is a popular data exchange format. The table structure can be inferred from existing schema. It has decent performance.

Cons: Not as efficient as Parquet or ORC.

Not human readable.

There are two ways to create an Avro table. The first way shown here is creating an Avro table with an in-line schema. You need to specify the Avro SerDE as well as the INPUTFORMAT and the OUTPUTFORMAT. You also need to specify TBLPROPERTIES to define the table. Note that the column definitions can be inferred from the schema. Big SQL infers the schema at table creation time. Hive infers it every time the table is read. If you want to change the schema in Big SQL, you must drop and recreate it.

The second method is to create a table using

a file. You will still need to specify the SerDe and the input and output format.

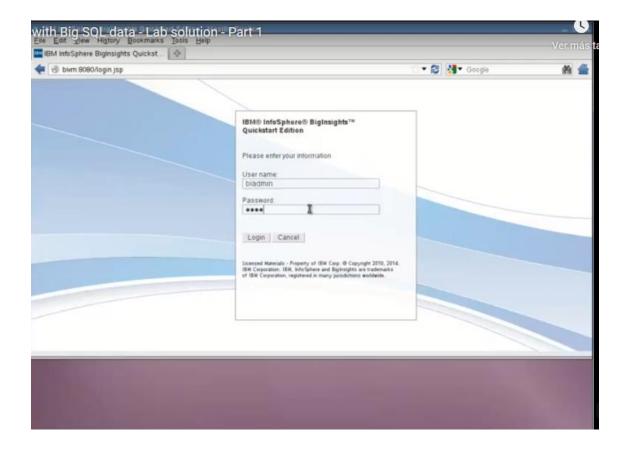
When no column list is specified, then the

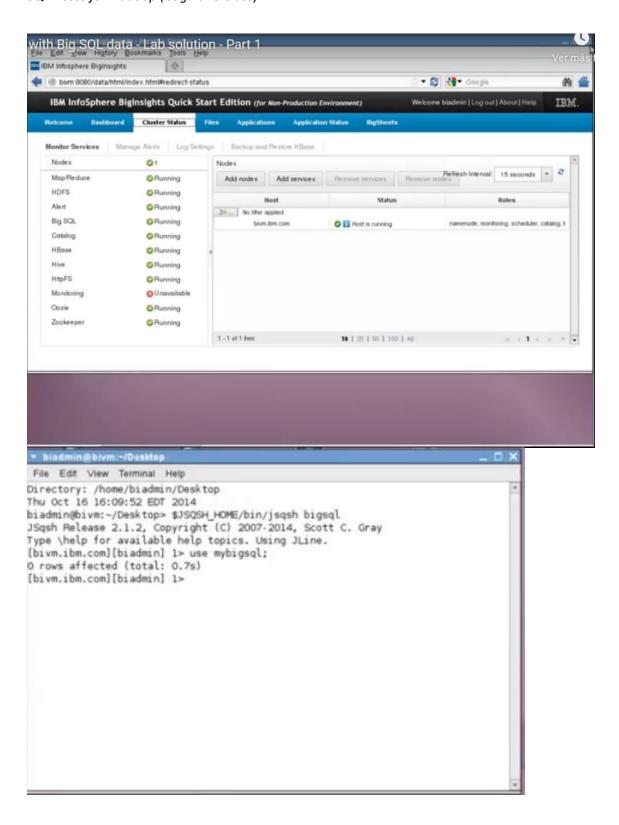
table columns will be inferred as shown in the table.

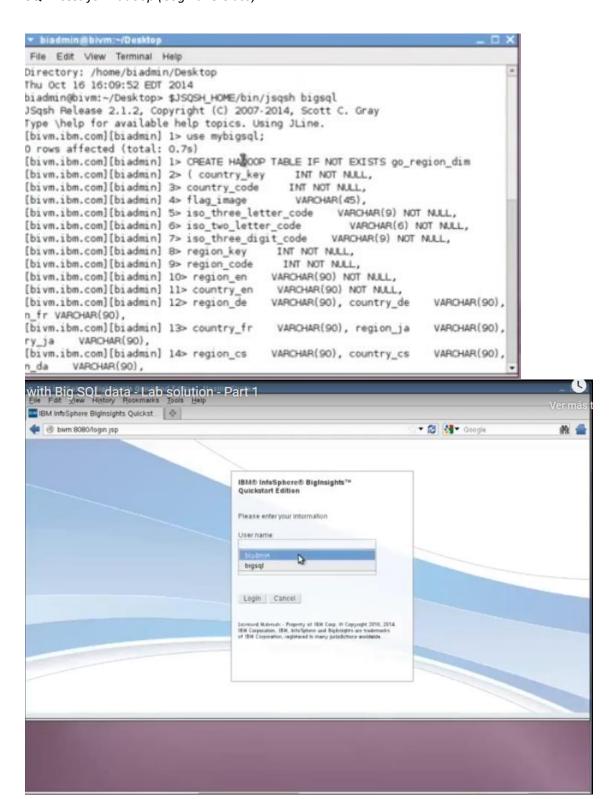
This can lead to a particular case where your table ends up using a STRING, which is essentially the maximum VARCHAR with 32K size. As stated before, this can be bad for performance so you should either use the bigsql.string.size to lower the default size or explicitly declare your columns.

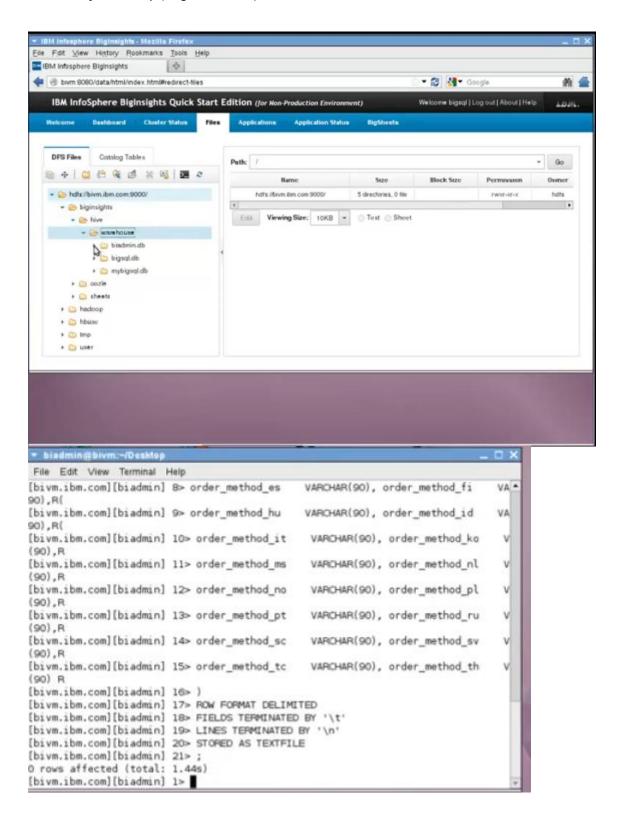
Having completed this lesson, you should now be able to list and explain the Big SQL data types. Create Big SQL schemas. Create Big SQL tables. Understand and use the file formats supported by Big SQL.

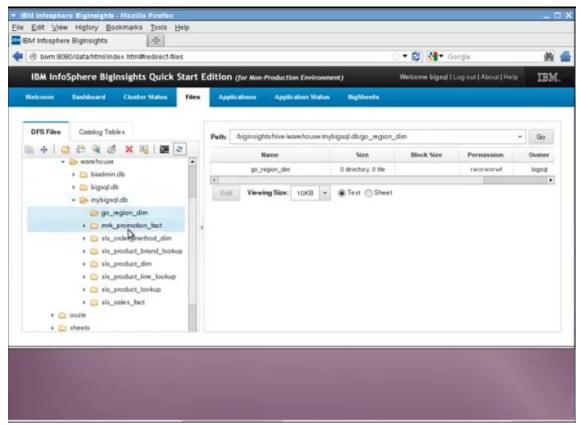
Congratulations, you have completed this course.











In this exercise, you will get to create Big SQL schemas and tables and then work with operations to load data into those tables. You will also write SQL queries to get the data from tables and views. The exercise will also cover some additional load operations as well show you how to create and use partitioned tables.

Since we are starting up a new exercise, let's make sure that BigInsights is still up and running. The thing is, with these VM, sometimes the instances does not stay on so I like to check to make sure they are still running. Log in to the Web Console and check the cluster status to make sure everything is still running. If not, start up BigInsights before continuing on with the lab.

Start up a new terminal.

Start JSqsh and connect to the bigsql connection.

Create a schema that you will use for this

lab. The use mybigsql command will set the default schema as mybigsql. It will also create that schema for you if it does not exist.

The examples that you will see in this lab

exercise assumes that your sample data is on the VMWare image. You will use sales data from a fictional company that sells and distributes outdoor products to third-party retailers as well as direct to consumers online. There are a series of FACT and DIMENSION tables.

You will use few than 10 of these in the GOSALESDW database in this lab exercise.

You will be creating these tables in Big SQL. In the JSqsh shell, copy and paste the table definition to create the dimension table for the region info.

Notice the Hadoop keyword. You need to specify this keyword in order to create tables for the Hadoop environment. However, you can omit this if you enable the SYSHADOOP.COMPATIBILITY MODE.

Also, notice that you did not explicitly specify the table's schema. The table will be created in the default schema, which we have specified as mybigsql.

The table's data will be row format delimited and have its fields terminated by tabs (\t) and the lines terminated by new line (\n). The data will be stored as TEXTFILE format,

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making it easy for a wide range of applications to work with.

Launch the BigInsights console. Log in as bigsql with the password bigsql.

Go to the Files tab to check out the table definition.

Drill down to biginsights --> hive --> warehouse --> mybigsql.db to see the go_region_dim table that you created.

Go back to the JSqsh shell and copy and paste

the next query to create the tracking method of the order of sale.

Go through and copy and paste the remaining queries for the tables to create them in Big SQL. The video will fast forward to once all eight tables have been created Once all eight tables

have been created, go to the web console to see their directories. You may need to refresh the view to see the updates. Notice that the directories are empty and that is because we have only created the table and have not loaded any data into them yet. You will see how to do this in the next video.