CLASS 2.
Hadoop Ecosystem. **Hive**.

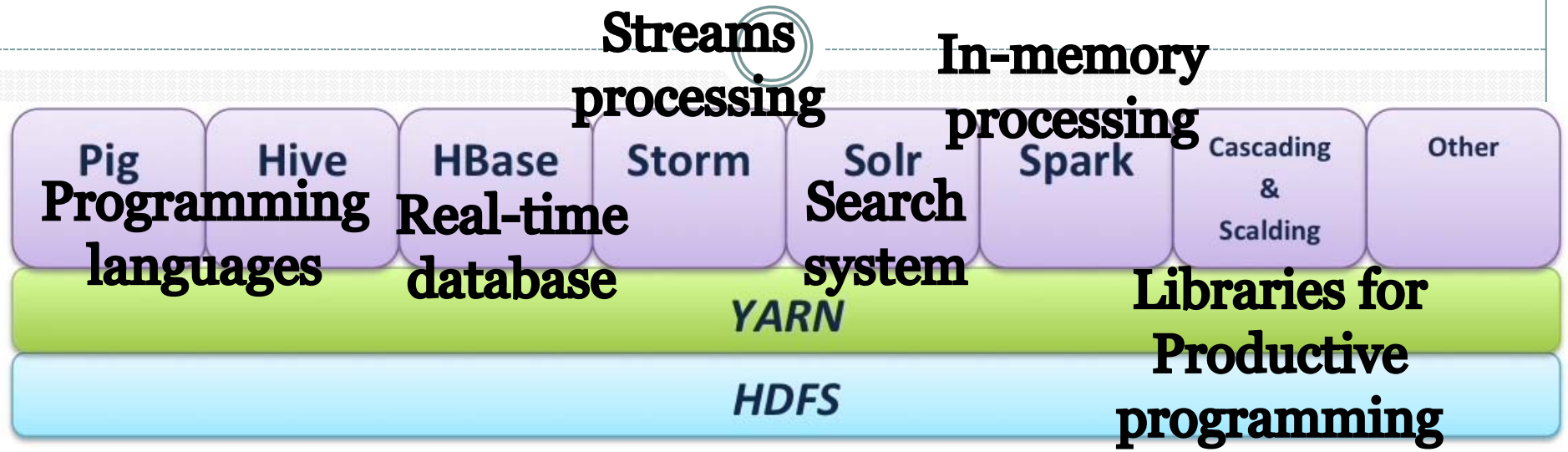
CSCI 6830

BIG DATA ANALYTICS WITH HADOOP AND R

YARN

- Yet Another Resource Negotiator
  - Hadoop MapReduce V1 issues:
    - Multi-tenancy
      - MapReduce V1 has a very simple approach to assigning tasks to nodes
      - We wish to manage multiple jobs on a cluster
    - Difficult to scale beyond 4,000 nodes
      - Cascading failures, network flooding
  - YARN is an internal reorganization in Hadoop V2
    - API compatible with Hadoop V1

Resource Management with YARN



- Yarn: A common resource management for applications
  - Scalability
  - Improved cluster utilization
  - Workloads other than MapReduce

Programming Hadoop

- No updates in place
  - The results are stored in a new file
- Your options vary:
  - Full control vs. productivity

Programming Hadoop with **Java**

JAVA

- Low level API
- Full control over all aspects of MapReduce
- Cumbersome programming model
- Requires many lines of code
- **Cascading** is an open source Java library for assembling of data flow programs that get translated to MapReduce

Programming with Hadoop Streaming

- Typically used from languages like Python or Ruby
- Simpler model than using Java API
  - Less efficient
- One writes map and reduce functions
  - The system feeds data as input stream, writes as output stream

Programming Hive

- A SQL-like language
- Great starting point for data professionals
- Significantly more productive than Java
- Not as rich as modern SQL
- It can be extended through user defined functions

Programming Pig

- A data flow language
- You can build your programs out of small steps
- Significantly more productive than Java API
- Designed to handle simple flows
  - No control structures or iterations
- It can be extended through user defined functions

Programming Scalding

- A library built on top of Scala
- Elegant model: programs look like manipulating in-memory data structures, get translated into MapReduce
- Very short programs
- Full programming environment
  - Full development ecosystem

Hive

- Data warehousing infrastructure based on the Hadoop
- Massive scale out and fault tolerance
- Originated at Facebook in 2007, now Apache project
- HiveQL: Query language based on SQL
  - Tables are defined on top of HDFS files
  - Queries transformed into MapReduce tasks
- Good candidate when transitioning from Rdb to Hadoop
  - HiveQL will feel familiar

Typical Hive Applications

- ***Store data coming in various formats in HDFS, then apply tools and processes similar to SQL***

- Log processing
- Text processing
- Indexing
- ETL
- Business analytics

Hive—It's All Files

- Hive table are defined directly over the existing data
- Hive Data Definition is just a mapping
- Hive operates directly over the HDFS files
- Tables can be partitioned into several files for scalability

Hive Workflow

- Typical workflow:
  1. Application dumps data into HDFS
  2. Map data with Hive table definitions
  3. Run queries
  4. Store results into HDFS
  5. Copy/move data from HDFS to regular file system

- Internally, Hive creates the needed MapReduce jobs
- Users see SQL

Hive Example: Table Mapping

```
CREATE TABLE apachelog (
  host STRING,
  identity STRING,
  user STRING,
  time STRING,
  request STRING,
  status STRING,
  size STRING,
  referer STRING,
  agent STRING)
ROW FORMAT SERDE 'org.apache.hadoop.hive.contrib.serde2.RegexSerDe'
WITH SERDEPROPERTIES (
  "input.regex" = "([^]*) ([^]*) ([^]*) (-|\\[^\\]*\\]) ([^ \"]*|\"[^\"]*\") (-|[0-9]*) (-|
[0-9]*)(?: ([^ \"]*|\".*\") ([^ \"]*|\".*\"))?",
  "output.format.string" = "%1$s %2$s %3$s %4$s %5$s %6$s %7$s %8$s %9$s"
)
STORED AS TEXTFILE;
```

Serialization and deserialization properties

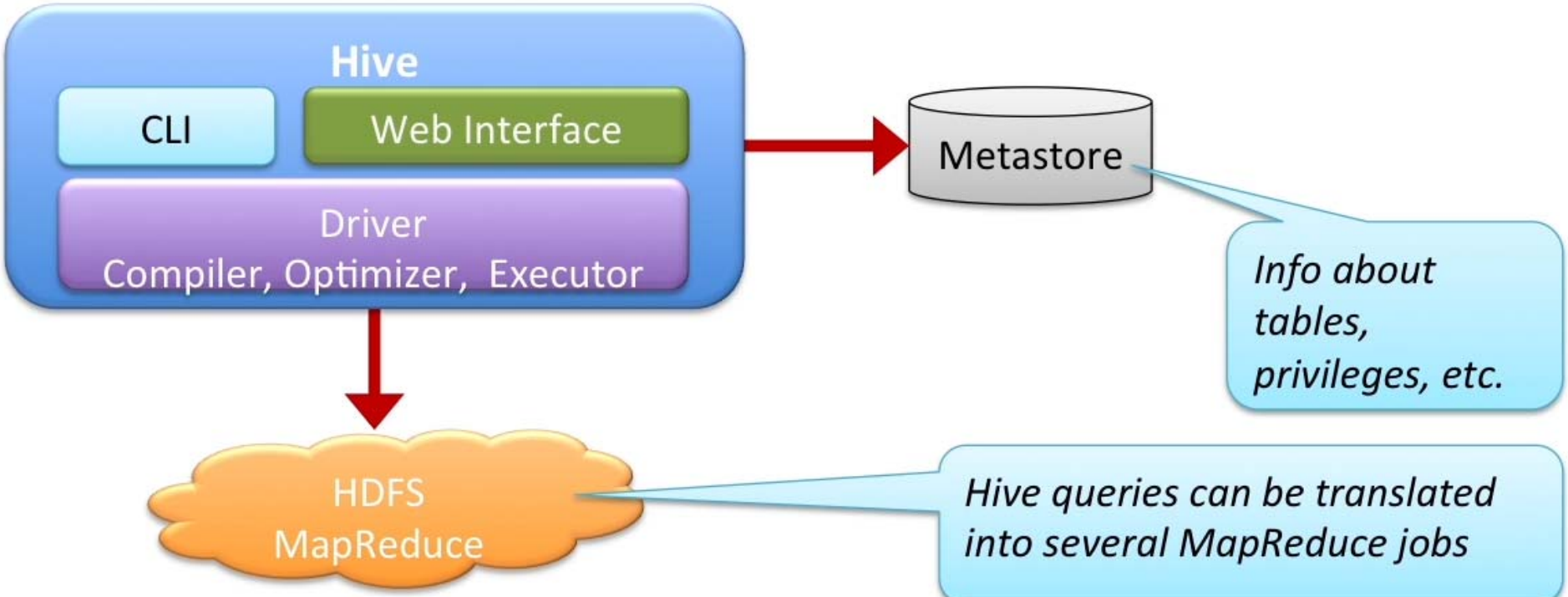
Row format described as Regular Expression

Hive Query

```
SELECT * from apachelog  WHERE host= '123.456.123.456';
```

- Query language is a subset of SQL
- Hive engine breaks down the query into multiple MapReduce flows, if needed
- Easy to learn and use – much more productive than Java API
- Ideal choice for first touch with Hadoop
- Beware: this is a batch execution – it may take some time to get the results

Hive Architecture



Hive vs. Relational Database

Hive	Relational Database
SQL	SQL
Analytics	OLTP or analytics
Batch only	Real-time or batch
No transactions	Transactions
No INSERT or UPDATE Adding through partitions	Random INSERT or UPDATE
Distributed processing - 100s of nodes	Depends on the system - If available, < 100
Achieve high performance on commodity hardware	Achieve high performance on proprietary hardware
Low cost for huge amounts of storage	Expensive, limited compared to Hadoop based solutions

Hive and MapReduce

- A Hive query gets translated into one or more MapReduce jobs
  - The jobs then run on the Hadoop cluster
- Much easier to write HiveQL than Java
- Dramatically increased productivity

Metadata and Metastore

- Metadata internally stored in a relational database
- Internally implemented with DataNucleus ORM
  - Enables easy migration to different RDBs
- It is only metadata, so the store is not large
- The default database is Apache Derby
  - Pure Java, lightweight RDB
  - Good for development
- Production database: typically MySQL

Interacting with Hive

- Command line interface
- Web interface
- Language clients
  - Java: JDBC
    - jdbc:hive://host:port/dbname
  - Python
  - ODBC

Hive Data Model

- Data model follows the familiar relational database layout
- There is no such thing as a Hive format!
  - Hive data model is just a mapping to files
- **Database**
  - Namespace for separation of tables
- **Table**
  - Unit of data with the same schema
  - Tables have columns
  - Columns are mapped to files in HDFS

Hive Table Partitions

- Partition
  - Optional, but useful storage unit for tables
- A table can have one or more partition keys
  - Date
  - Country
- Each value of the partition key defines a partition of the table
  - We can run analytics limited to files US, 2015-01-01
- Partition columns are virtual columns

Hive Buckets (Clusters)

- Data in a partition may be bucketed based on some hash function taking input from a column in a table
- Example:
  - Bucket based on the userid

Partitions and buckets are a convenient way to significantly speed up execution by pruning large quantities of data

Hive Data Types

- **Primitive types** -- similar to SQL
  - STRING, INT, BOOLEAN, TIMESTAMP, DATE,…
- **Complex types**
  - **Arrays:** ARRAY<data_type>
    - Column[0]
  - **Maps:** MAP<primitive_type, data_type>
    - column[key]

Hive Data Types (continued)

- **Structs:** ARRAY<data_type>
  - Column of the type {x INT, y INT}
  - Access as: column.x
- **Union**: UNIONTYPE<data_type, data_type, ...>
  - Holds exactly one of their specified data types, first part is the tag

Creating Tables

```
CREATE TABLE product_view (
  product_id STRING,
  user_id STRING,
  visit_time INT,
  ip_address STRING)
PARTITIONED BY (dt STRING, country STRING)
STORED AS TEXTFILE;
```

- Data delimited with ASCII 001 (Ctrl-A), newline is the row delimiter
- Data delimiter is configurable, but not the row delimiter

Loading Data

- User creates an external table that points to an HDFS location:

```
CREATE EXTERNAL TABLE product_view_raw (
  product_id STRING,
  user_id STRING,
  visit_time INT,
  ip_address STRING,
  country STRING)
STORED AS TEXTFILE
LOCATION '/user/estore/data/product_views';
```

Loading Data (continued)

- User copies a file to the HDFS location:

```
hadoop dfs -put /tmp/cv_2015-01-01.txt /user/estore/data/catalog_views
```

- User transforms the data and enters them into any other Hive table:

```
FROM product_view_raw pvr
INSERT OVERWRITE TABLE product_view
PARTITION (dt='2015-01-01', country='US')
SELECT pvr.product_id, pvr.user_id, pvr.visit_time, pvr.ip_address
WHERE pvr.country='US'
```

Loading when format is the same

- When the input file format is the same as the Hive table format, we can use the LOAD statement:

```
LOAD DATA LOCAL INPATH
  /user/estore/data/product_views/pv_2015-01-01.txt
INTO TABLE product_view
PARTITION(dt='2015-01-01', country='US')
```

Hive Queries

- All query results are always inserted into a table
  - User can inspect it later, or store them into a local file

```
INSERT OVERWRITE TABLE customers
INSERT OVERWRITE DIRECTORY '/user/data/tmp/'
```

- User can also run queries in Hive CLI
  - Beware: Hive is batch
  - CLI: result is stored into a temporary file, and then presented

Hive Queries and Partitions

```
INSERT OVERWRITE TABLE gizmo_product_view
SELECT product_view.*
FROM product_view
WHERE product_view.dt >= '2015-01-01' AND
  product_view.dt <= '2015-01-31' AND
  product_view.product_id = 'gizmo-001';
```

- We have defined the partition as:

```
PARTITIONED BY(dt DATETIME, country STRING)
```

Joins in Hive

```
INSERT OVERWRITE TABLE product_viewers
SELECT pv.product_id, c.first_name, c.last_name
FROM customer c JOIN product_view pv ON (c.id = pv.user_id)
WHERE product_view.dt >= '2015-01-01' AND
   product_view.dt <= '2015-01-31'
```

- Join control:
  - LEFT OUTER, RIGHT OUTER or FULL OUTER
- For performance, it is best to put the largest table to the right most position in join

Hive. Storing Results

- Plain SELECT statements give results to the console
- To store result in HDFS use:

```
INSERT OVERWRITE TABLE results
  SELECT  …
```

- The results table is a set of files in HDFS
- Result table can be used in other queries

User-Defined Functions

- **UDF: User Defined Function**
  - One to one row mapping
  - concat(first_name, last_name)
- **UDAF: User Defined Aggregate Function**
  - Many to one row mapping
  - sum(sales)
- **UDTF: User Defined Table Function**
  - One to many row mapping
  - explode([1,2,3])

When to use Hive?

- When dealing with massive data
- Similarity with SQL is an advantage
  - But: watch out for limitations:
    - No UPDATE
    - No single row INSERT
    - Limited built in functions
- When real-time response is not needed
  - Batch processing is fine
- Consider alternatives: Pig, Scalding

When **NOT** to use Hive

- Data is in the GB range
  - Consider alternative RDB/NoSQL store
  - Exception: if the conventional solution is to expensive
- Data has no structure (schema)
- You need real-time response

More on Hive

- Why do Pig and Hive exist ?
  - they seem to do much of the same thing.
- Hive pros:
  - its HQL = SQL like query language
  - Often used as the interface to an Apache Hadoop based data warehouse.
  - Hive is friendlier and more familiar than Pig for SQL users
- Pig pros:
  - data flow strengths for bringing data into Apache Hadoop
  - Easy data querying.

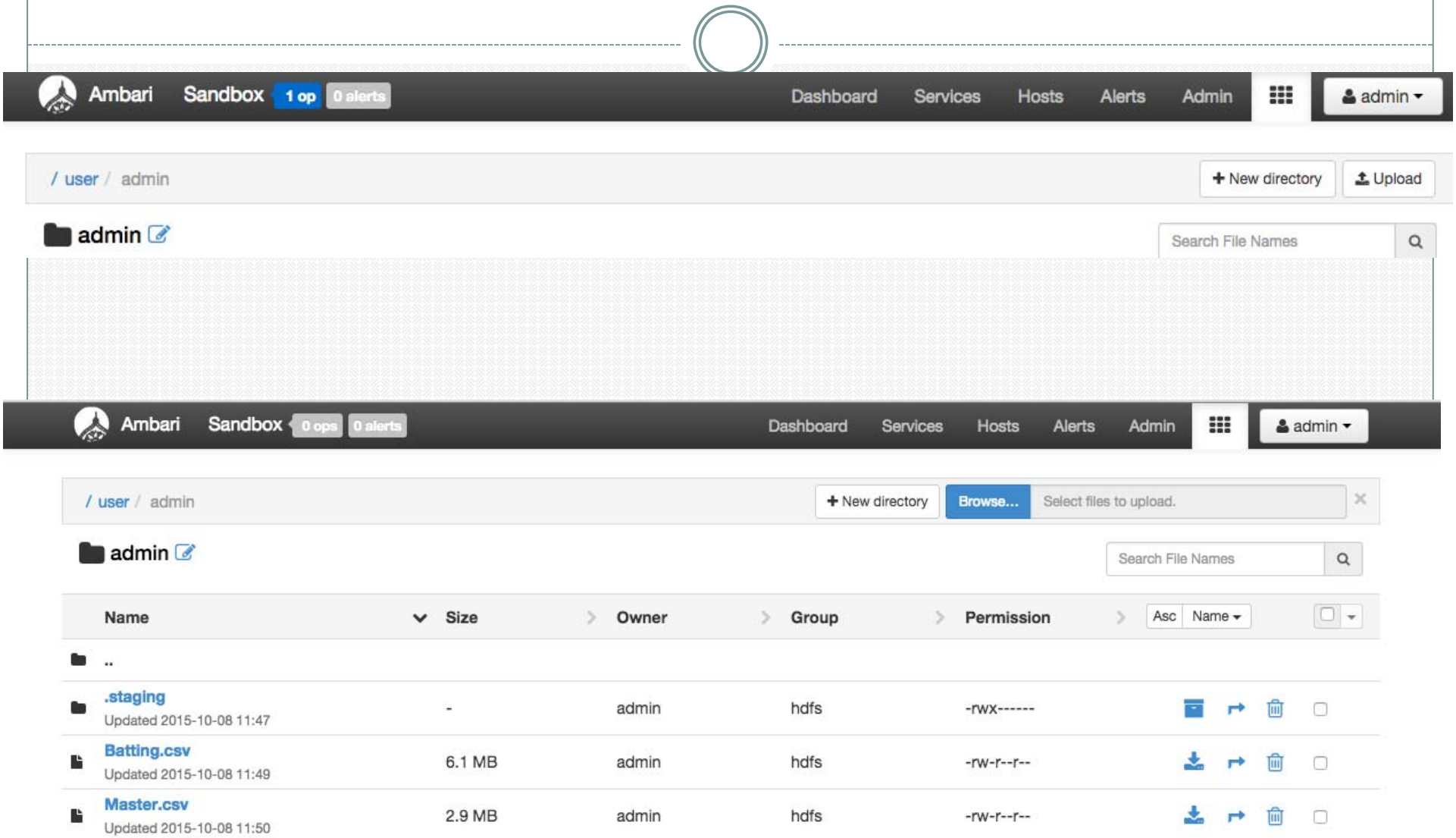
Use Case: Baseball Statistics

- The data files we are using comes from the site www.seanlahman.com. You can download the data file from:

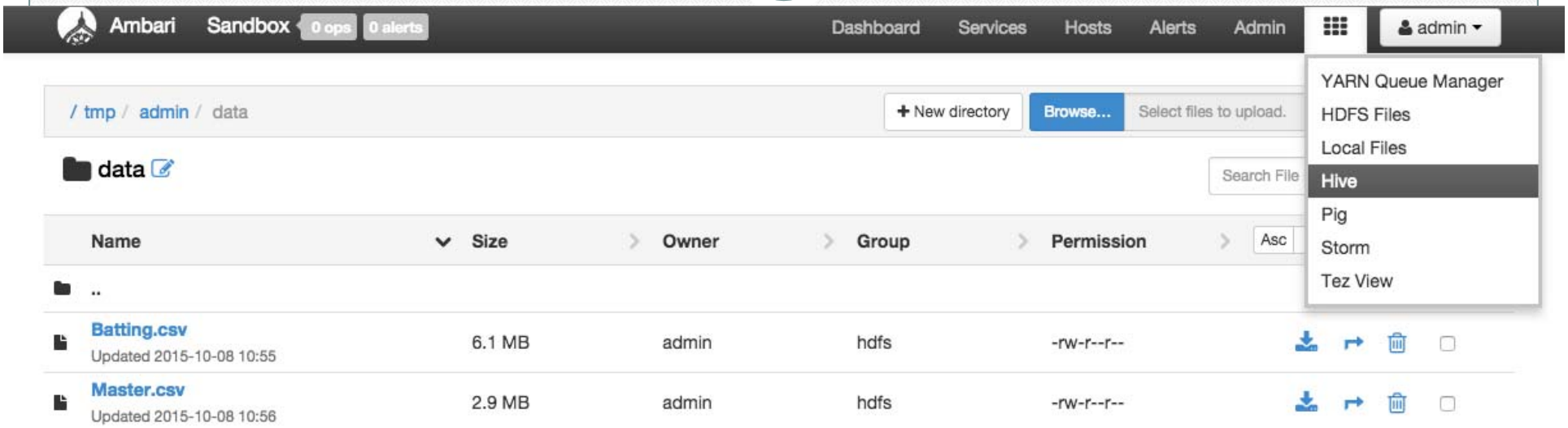
- http://seanlahman.com/files/database/lahman591-csv.zip

- Once you have the file you will need to unzip it into a directory. We will be uploading just the Master.csv and Batting.csv files from the dataset.

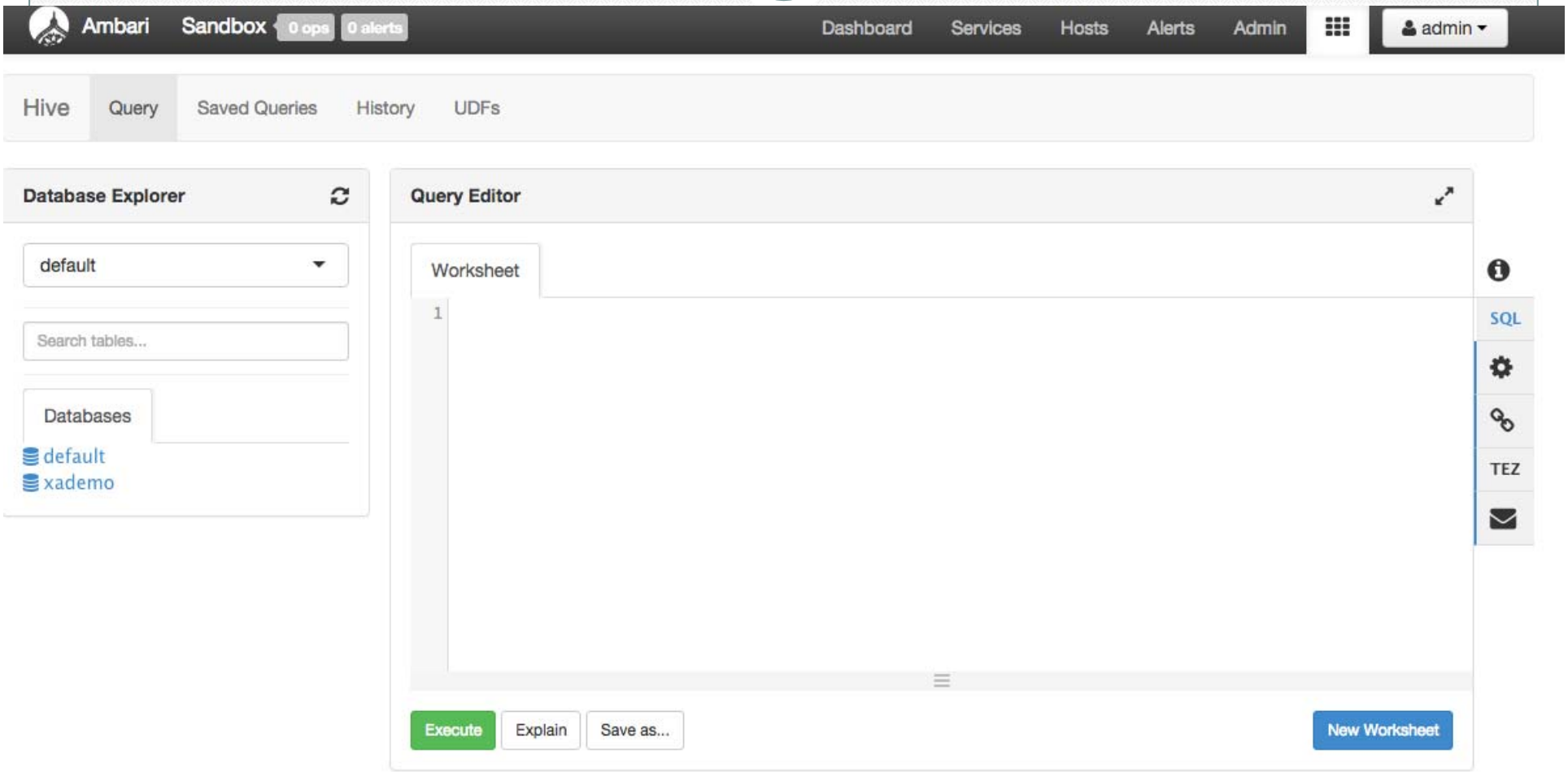
Uploading the data files



Starting the Hive View

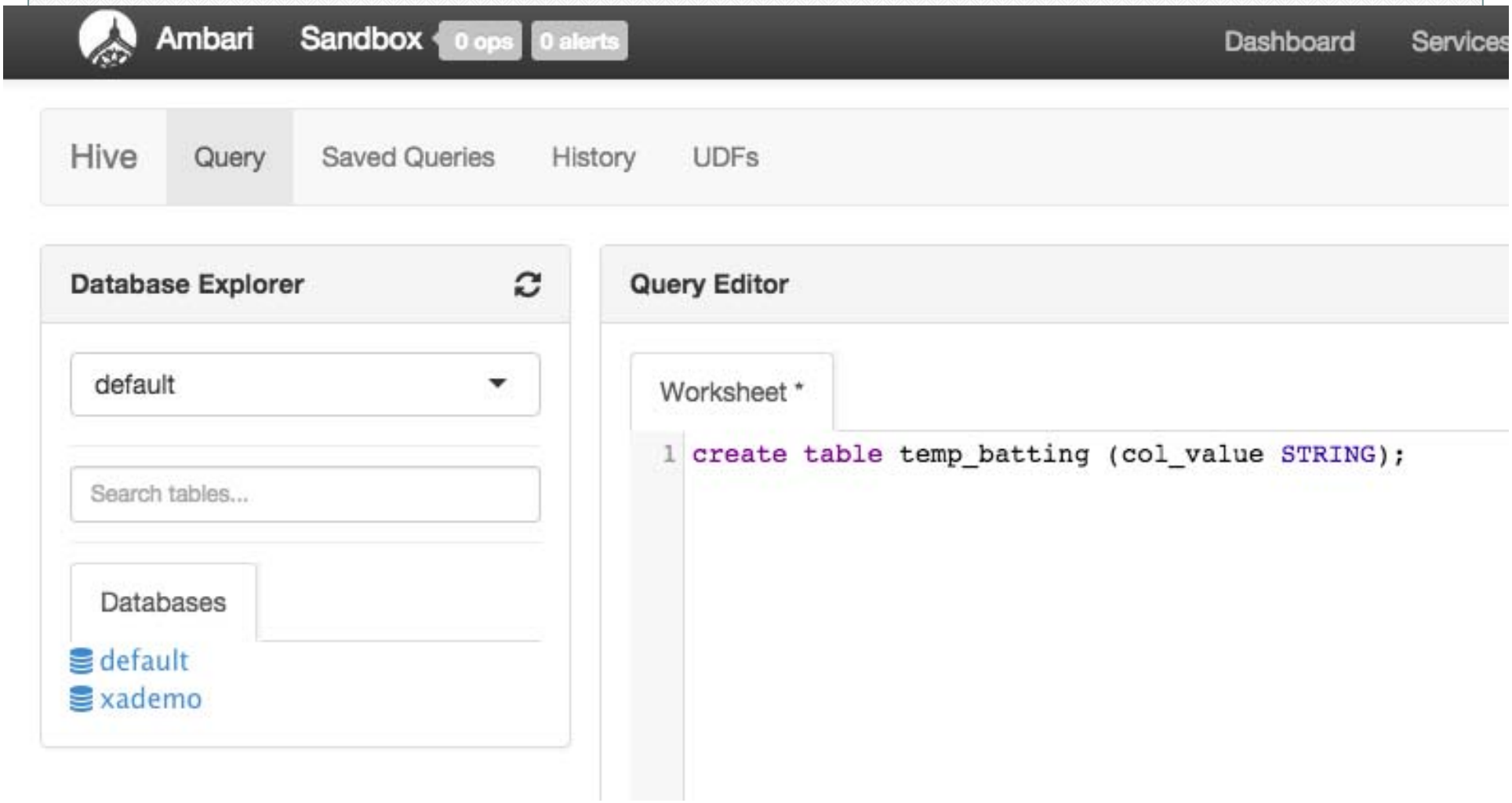


Hive Query Editor

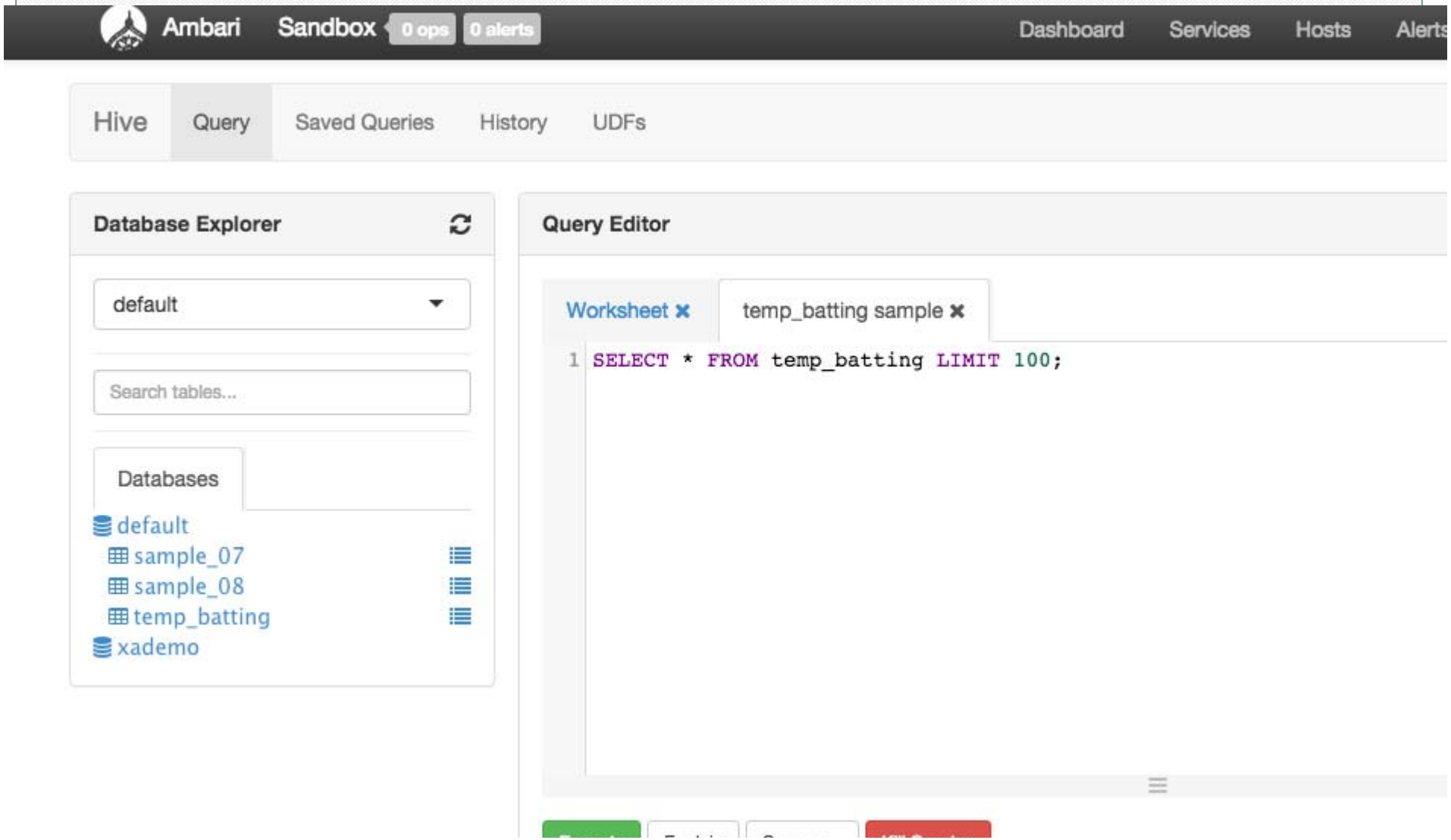


Create a Table

- create table temp_batting (col_value STRING);



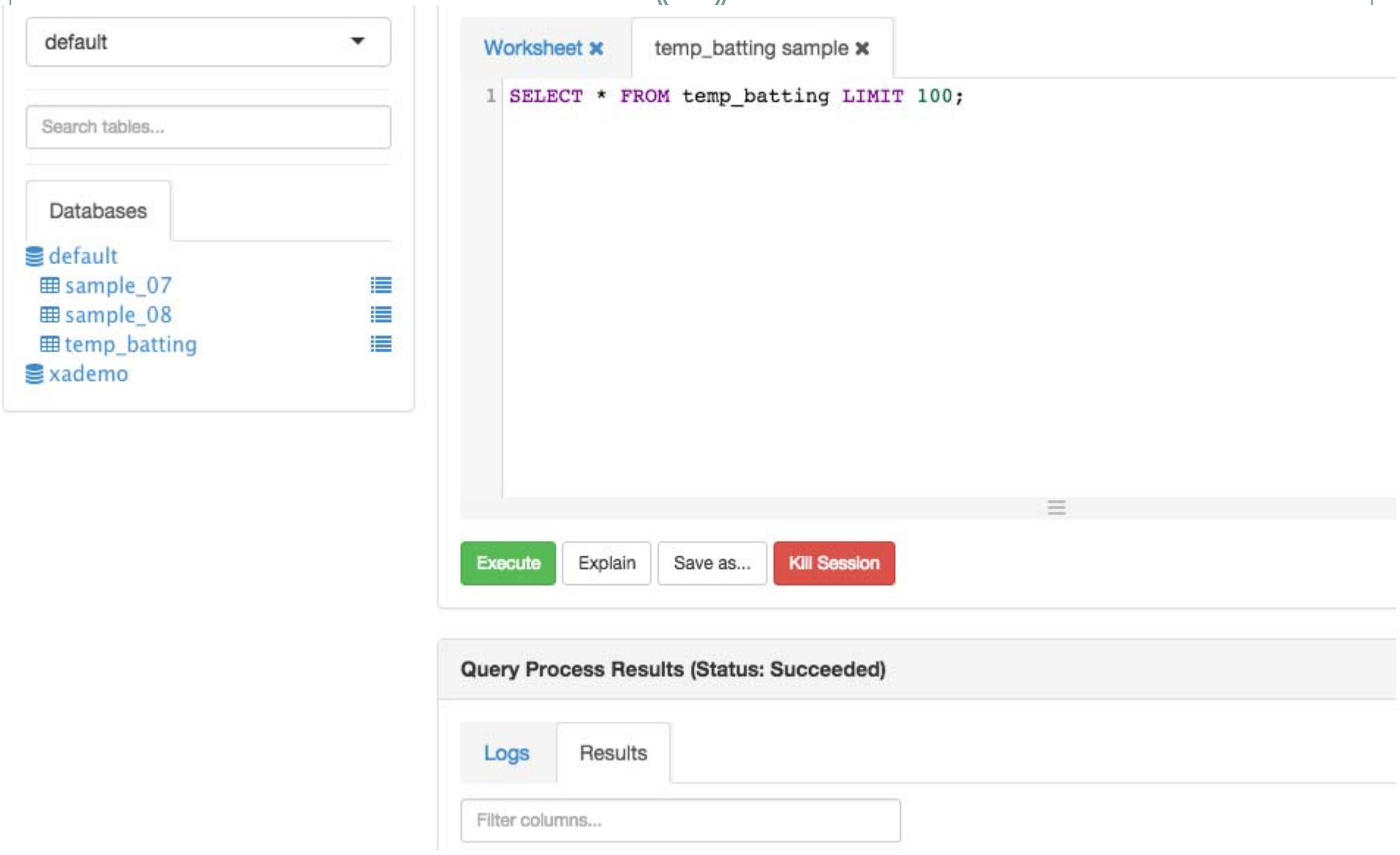
Browse New Data



Load More Data into a Temporary Table

- Load the data file Batting.csv into the table temp_batting

Check the Data



default
- sample_07
- sample_08
- temp_batting

xademo

Execute | Explain | Save as... | Kill Session

Query Process Results (Status: Succeeded)

Logs | Results

Filter columns...

temp_batting.col_value

playerID,yearID,stint,teamID,lgID,G,G_batting,AB,R,H,2B,3B,HR,RBI,SB,CS,BB,SO,IBB,HBP,SH,SF,(
aardsda01,2004,1,SFN,NL,11,11,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,11
aardsda01,2006,1,CHN,NL,45,43,2,0,0,0,0,0,0,0,0,0,0,0,1,0,0,45
aardsda01,2007,1,CHA,AL,25,2,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,2
aardsda01,2008,1,BOS,AL,47,5,1,0,0,0,0,0,0,0,0,1,0,0,0,0,0,5
aardsda01,2009,1,SEA,AL,73,3,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,
aardsda01,2010,1,SEA,AL,53,4,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,
aaronha01,1954,1,ML1,NL,122,122,468,58,131,27,6,13,69,2,2,28,39,3,6,4,13,122

Create a Permanent Table

Worksheet *

```
1 create table batting (player_id STRING, year INT, runs INT);
```

Use RegEx

- Then we extract the data we want from temp_batting and copy it into batting.
- Use a regexp pattern.
- Build up a multi-line query.
- The first line of the query createa the table batting.
- The three regexp_extract calls are going to extract the **player_id**, **year** and **run** fields from the table temp_batting.

Query Using RegEx

Worksheet *

```
1 insert overwrite table batting
2 SELECT
3   regexp_extract(col_value, '^(?:([^,]*)\,?){1}', 1) player_id,
4   regexp_extract(col_value, '^(?:([^,]*)\,?){2}', 1) year,
5   regexp_extract(col_value, '^(?:([^,]*)\,?){9}', 1) run
6 from temp_batting;
```

default

Search tables...

Databases

- default
  - batting
  - sample_07
  - sample_08
  - temp_batting
- xademo

Worksheet

```
1 SELECT year, max(runs) FROM batting GROUP BY year;
```

Execute | Explain | Save as... | Kill Session

100%

Query Process Results (Status: SUCCEEDED)

Logs | Results

Filter columns...

year	_c1
6	0
1871	66