

Connecting to S3

Apache Spark™ and Databricks® allow you to connect to virtually any data store including Amazon S3.

In this lesson you:

- Mount and access data in S3
- Define options when reading from S3

Spark quickly rose to popularity as a replacement for the Apache Hadoop™ MapReduce paradigm in a large part because it easily connected to a number of different data sources. Most important among these data sources was the Hadoop Distributed File System (HDFS). Now, Spark engineers connect to a wide variety of data sources including:

- Traditional databases like Postgres, SQL Server, and MySQL
- Message brokers like Apache Kafka and Kinesis
- Distributed databases like Cassandra and Redshift
- · Data warehouses like Hive
- File types like CSV, Parquet, and Avro

Applications













MESOS









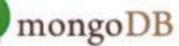
My5QL



PostgreSO















elasticsearch.



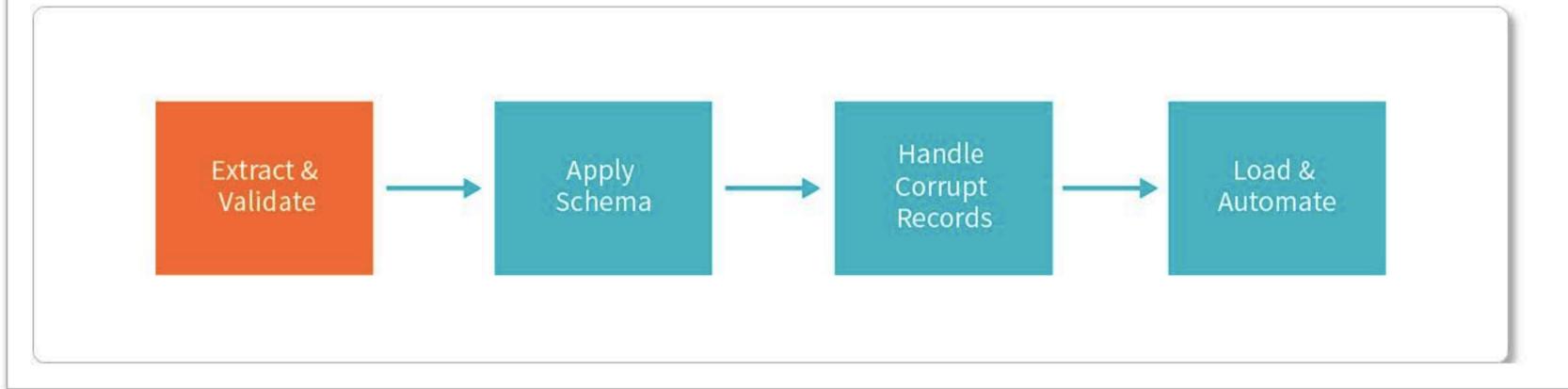


DBFS Mounts and S3

Amazon Simple Storage Service (S3) is the backbone of Databricks workflows. S3 offers data storage that easily scales to the demands of most data applications and, by colocating data with Spark clusters, Databricks quickly reads from and writes to S3 in a distributed manner.

The Databricks File System, or DBFS, is a layer over S3 that allows you to mount S3 buckets, making them available to other users in your workspace and persisting the data after a cluster is shut down.

In our road map for ETL, this is the Extract and Validate step:



Define your AWS credentials. Below are defined read-only keys, the name of an AWS bucket, and the mount name to refer to use in DBFS.



For getting AWS keys, take a look at take a look at the AWS documentation

```
Cmd 10

1  awsAccessKey = "AKIAJBRYNXGHORDHZB4A"
2  # Encode the Secret Key to remove any "/" characters
3  secretKey = "a0BzE1bSegfydr3%2FGE3LSPM6uIV5A4hOUfpH8aFF".replace("/", "%2F")
4  awsBucketName = "databricks-corp-training/common"

Command took θ.04 seconds -- by huseyinyilmazθ1@gmail.com at 4/26/2020, 3:01:42 PM on My Cluster

Cmd 11
```

In addition to the sourcing information above, we need to define a target location.

So that no two students produce the exact same mount, we are going to be a little more creative with this one.

```
Cmd 12

1 mountPoint = f"/mnt/etlp1s-{username}-si"

Command took 0.05 seconds -- by huseyinyilmaz01@gmail.com at 4/26/2020, 3:02:21 PM on My Cluster

Cmd 13
```

In case you mounted this bucket earlier, you might need to unmount it.

```
try:
    dbutils.fs.unmount(mountPoint) # Use this to unmount as needed
except:
    print("{} already unmounted".format(mountPoint))
```

Next, explore the mount using %fs ls and the name of the mount.

Remember, your mount name includes your email address so you will need to uncomment and update the following FILL_IN section

```
Cmd 18
     print("Hint: Your mount name is {}".format(mountPoint))
 Hint: Your mount name is /mnt/etlp1s-huseyinyilmaz01@gmail.com-si
 Command took 0.03 seconds -- by huseyinyilmaz01@gmail.com at 4/26/2020, 3:08:58 PM on My Cluster
Cmd 19
    %python
  2 %fs ls /mnt/etlp1-huseyinyilmaz01@gmail.com-si
 UsageError: Line magic function '%fs' not found.
 Command took 0.04 seconds -- by huseyinyilmaz01@gmail.com at 4/26/2020, 3:09:39 PM on My Cluster
Cmd 20
```

In practice, always secure your AWS credentials. Do this by either maintaining a single notebook with restricted permissions that holds AWS keys, or delete the cells or notebooks that expose the keys. After a cell used to mount a bucket is run, access this mount in any notebook, any cluster, and share the mount between colleagues.

Adding Options

When you import that data into a cluster, you can add options based on the specific characteristics of the data.

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700 PA					IUCR				Descript		Location		n Market Marchael		Domestic			DISCITO	L	Waru	Commu
nity Ar		FBI Cod			rdinate		oordin			Updated		Latitud		Longitu		Location					
23811	JB1414	41	02/05/20	018 01:	10:00	AM 118X	XSI	INDIANA	. AVE	0110	HOMICIDE	E	FIRST	DEGREE MU	JRDER	VACANT L	LOT	false	false	0532	005
9	53	01A	1179707	182628	30 2018	3 02/1	12/201	18 03:4	19:14 PM	41.6785	85145	-87.617	837834	(41.678	8585145, -	-87.61783	37834)				
1122858	19	JB14899	/0	01/23/	/2018 €	09:00:00	AM e	∂72XX S	VERNON /	AVE	1153	DECEPTI	VE PRAC	TICE	FINANCI/	AL IDENT	ITY THEFT	Γ OVER \$	300	OTHER	false
false	0323	003	6	69	11				2018	02/12/2	018 03:49	9:14 PM									
1122856	33	JB14893	.1	01/31/	/2018 J	10:12:00	AM C	040XX N	KEYSTON	E AVE	1154	DECEPTI	VE PRAC	TICE	FINANCIA	AL IDENT	ITY THEFT	Г \$300 A	ND UNDER	APARTME	ENT
false	false	1722	017	39	16	11				2018	02/12/20	018 03:4	9:14 PM	0							
1122855	5ز	JB14888	.5	02/01/	/2018 €	02:00:00	PM C	017XX W	CONGRESS	5 PKWY	0820	THEFT	\$500 A	ND UNDER	HOSPITAL	L BUILDI	NG/GROUNI	os	false	false	1231
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1122843	10	JB14867	5	01/27/	/2018 F	09:00:00	PM C	061XX S	EBERHAR	ΓAVE	0560	ASSAULT	SIMPLE	RESIDEN	NCE	false	true	0313	003	20	42
08A			2018	02/12/	/2018 F	03:49:14	PM														
1122840	11	JB14868	.3	02/02/	/2018 1	12:00:00	PM C	038XX N	SAWYER /	AVE	1153	DECEPTI	VE PRAC	TICE	FINANCI/	AL IDENT	ITY THEFT	Γ OVER \$	300	RESIDEN	NCE
false	false	1733	017	33	16	11				2018	02/12/20	018 03:4	9:14 PM	1							
1122834	17	JB14859	/9	01/28/	/2018 F	07:00:00	PM 6	008XX E	45TH ST	0620	BURGLARY	Y	UNLAWE	UL ENTRY	RESIDENC	CE	false	false	0221	002	4
39	05			2018	02/1	12/2018 0	33:49:	:14 PM													
1122829	11	JB14859	/1	01/10/	/2018 F	04:45:00	PM 6	010XX E	53RD ST	1153	DECEPTIV	VE PRACT	ICE	FINANCI	IAL IDENT	ITY THEFT	T OVER \$	300		false	false
0233	002	4	41	11			7	2018	02/12/20	018 03:4	9:14 PM										
1122828	37	JB14848	,2	01/03/	/2018 F	03:45:00	PM C	0000X W	C1 ST	0810	THEFT	OVER \$5	00	AIRPORT	T TERMINAL	L LOWER I	LEVEL - N	NON-SECU	RE AREA	false	false
1651	016	41	76	06			7	2018	02/12/20	018 03:49	9:14 PM										A

Command took 1.32 seconds -- by huseyinyilmaz01@gmail.com at 4/26/2020, 3:28:46 PM on My Cluster

option is a method of DataFrameReader . Options are key/value pairs and must be specified before calling .csv() .

This is a tab-delimited file, as seen in the previous cell. Specify the "delimiter" option in the import statement.

:NOTE: Find a full list of parameters here.

```
Cmd 26
```

```
display(spark.read
    .option("delimiter", "\t")
    .csv("/mnt/training/Chicago-Crimes-2018.csv")

display(spark.read
    .option("delimiter", "\t")
    .csv("/mnt/training/Chicago-Crimes-2018.csv")
```

(2) Spark Jobs

_c0 =	_c1 =	_c2 =	_c3 ==	_c4 =	_c5 =	_c6 ==	_c7	_c8 =	_c9
ID	Case Number	Date	Block	IUCR	Primary Type	Description	Location Description	Arrest	Doi
23811	JB141441	02/05/2018 01:10:00 AM	118XX S INDIANA AVE	0110	HOMICIDE	FIRST DEGREE MURDER	VACANT LOT	false	fals
11228589	JB148990	01/23/2018 09:00:00 AM	072XX S VERNON AVE	1153	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT OVER \$ 300	OTHER	false	fals
11228563	JB148931	01/31/2018 10:12:00	040XX N KEYSTONE AVE	1154	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT \$300 AND	APARTMENT	false	fals

Charries the first 1000 serves

Spark doesn't read the header by default, as demonstrated by the column names of _co , _c1 , etc. Notice that the column names are present in the first row of the DataFrame.

Fix this by setting the "header" option to True .

```
Cmd 28
```

```
display(spark.read
    .option("delimiter", "\t")
    .option("header", True)
    .csv("/mnt/training/Chicago-Crimes-2018.csv")
)
```

▶ (2) Spark Jobs

ID =	Case Number >	Date w	Block	IUCR =	Primary Type	Description	Location Description	Arrest =	Domestic *	Beat =	District =	Ward =	Community Area
23811	JB141441	02/05/2018 01:10:00 AM	118XX S INDIANA AVE	0110	HOMICIDE	FIRST DEGREE MURDER	VACANT LOT	false	false	0532	005	9	53
11228589	JB148990	01/23/2018 09:00:00 AM	072XX S VERNON AVE	1153	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT OVER \$ 300	OTHER	false	false	0323	003	6	69
11228563	JB148931	01/31/2018 10:12:00 AM	040XX N KEYSTONE AVE	1154	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT \$300 AND UNDER	APARTMENT	false	false	1722	017	39	16
	Transport Comme	Personnes me	2.72722222	74-0-04	100000000000000000000000000000000000000		973E-1287E-00	201	F12020	Trans-	Lare -	102	122

Showing the first 1000 rows.

Spark didn't infer the schema, or read the timestamp format, since this file uses an atypical timestamp. Change that by adding the option "timestampFormat" and pass it the format used in this file.

Set "inferSchema" to True, which triggers Spark to make an extra pass over the data to infer the schema.

```
Cmd 30
     crimeDF = (spark.read
       .option("delimiter", "\t")
       .option("header", True)
       .option("timestampFormat", "mm/dd/yyyy hh:mm:ss a")
       .option("inferSchema", True)
       .csv("/mnt/training/Chicago-Crimes-2018.csv")
     display(crimeDF)
```

- (3) Spark Jobs
- ▶ ☐ crimeDF: pyspark.sql.dataframe.DataFrame = [ID: integer, Case Number: string ... 20 more fields]

ID =	Case Number >	Date	Block =	IUCR =	Primary Type	Description	Location Description	Arrest •	Domestic =	Beat w	District =	Ward =	Comm
23811	JB141441	2018-01- 05T01:10:00.000+0000	118XX S INDIANA AVE	0110	HOMICIDE	FIRST DEGREE MURDER	VACANT LOT	false	false	532	5	9	53
11228589	JB148990	2018-01- 23T09:00:00.000+0000	072XX S VERNON AVE	1153	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT OVER \$ 300	OTHER	false	false	323	3	6	69
11228563	JB148931	2018-01- 31T10:12:00.000+0000	040XX N KEYSTONE AVE	1154	DECEPTIVE PRACTICE	FINANCIAL IDENTITY THEFT \$300 AND UNDER	APARTMENT	false	false	1722	17	39	16
11228555	JB148885	2018-01- 01T14:00:00.000+0000	017XX W CONGRESS	0820	THEFT	\$500 AND UNDER	HOSPITAL BUILDING/GROUNDS	false	false	1231	12	2	28

The Design Pattern

Other connections work in much the same way, whether your data sits in Cassandra, Redis, Redshift, or another common data store. The general pattern is always:

- 1. Define the connection point
- 2. Define connection parameters such as access credentials
- 3. Add necessary options

After adhering to this, read data using spark.read.options(<option key>, <option value>).<connection_type>(<endpoint>).

Exercise 1: Read Wikipedia Data

Read Wikipedia data from S3, accounting for its delimiter and header.

Cmd 33

Step 1: Get a Sense for the Data

Take a look at the head of the data, located at /mnt/training/wikipedia/pageviews/pageviews_by_second.tsv.

```
1 # TODO
2 display(spark.read
3    .option("delimiter", "\t")
4    .option("header", True)
5    .csv("/mnt/training/wikipedia/pageviews/pageviews_by_second.tsv")
6    .limit(3)
7 )
```

(2) Spark Jobs

timestamp	site	requests
2015-03-16T00:09:55	mobile	1595
2015-03-16T00:10:39	mobile	1544
2015-03-16T00:19:39	desktop	2460

Step 2: Import the Raw Data

Cmd 36

Cmd 37

TODO

wikiDF = (spark.read

Import the data without any options and save it to wikiDF. Display the result.

```
    4 )
    ▶ (1) Spark Jobs
    ▶ ■ wikiDF: pyspark.sql.dataframe.DataFrame = [_c0: string]
    Command took 1.01 seconds -- by huseyinyilmaz01@gmail.com at 4/26/2020, 4:17:08 PM on My Cluster
```

.csv("/mnt/training/wikipedia/pageviews/pageviews_by_second.tsv")

Step 3: Import the Data with Options

Import the data with options and save it to wikiWithOptionsDF. Display the result. Your import statement should account for:

- · The header
- · The delimiter

```
# TODO
wikiWithOptionsDF = (spark.read
.option("delimiter", "\t")
.option("header", True)
.option("timestampFormat", "mm/dd/yyyy hh:mm:ss a")
.option("inferSchema", True)
.csv("/mnt/training/wikipedia/pageviews/pageviews_by_second.tsv")

display(wikiWithOptionsDF)
```

- (3) Spark Jobs
- ▶ wikiWithOptionsDF: pyspark.sql.dataframe.DataFrame = [timestamp: timestamp, site: string ... 1 more fields]

timestamp	site	requests
2015-03-16T00:09:55.000+0000	mobile	1595
2015-03-16T00:10:39.000+0000	mobile	1544
2015-03-16T00:19:39.000+0000	desktop	2460

Review

Question: What accounts for Spark's quick rise in popularity as an ETL tool?

Answer: Spark easily accesses data virtually anywhere it lives, and the scalable framework lowers the difficulties in building connectors to access data. Spark offers a unified API for connecting to data making reads from a CSV file, JSON data, or a database, to provide a few examples, nearly identical. This allows developers to focus on writing their code rather than writing connectors.

Question: What is DBFS and why is it important?

Answer: The Databricks File System (DBFS) allows access to scalable, fast, and distributed storage backed by S3 or the Azure Blob Store.

Question: How do you connect your Spark cluster to S3?

Answer: By mounting it. Mounts require AWS credentials and give access to a virtually infinite store for your data. Using AWS IAM roles provides added security since your keys will not appear in log files. One other option is to define your keys in a single notebook that only you have permission to access. Click the arrow next to a notebook in the Workspace tab to define access permissions.

Question: How do you specify parameters when reading data?

Answer: Using .option() during your read allows you to pass key/value pairs specifying aspects of your read. For instance, options for reading CSV data include header, delimiter, and inferSchema.

Question: What is the general design pattern for connecting to your data?

Answer: The general design pattern is as follows:

- 1. Define the connection point.
- Define connection parameters such as access credentials.
- 3. Add necessary options such as for headers or parallelization.