

AZURE DATABRICKS PYSPARK HIVE DEMO

https://advancedsqlpuzzles.com

Scott Peters

OBJECTIVE

For this demo, we will perform the following...













Connect
Databricks to an
Azure Key Vault

Connect
Databricks to an
Azure Data
Lake

Create an ETL process to import csv files from an Azure Data Lake

Merge the data into Hive tables

Insert the data into a SQL Server database Automate the ETL via an Azure Data Factory pipeline

OBJECTIVE

The GitHub repository for this demo is located here:

https://github.com/smpetersgithub/AdvancedSQLPuzzles

Included in the repository are the following:

- A PDF version of this presentation
- Databricks workspace
- CSV files to import
- DDL scripts

OBJECTIVE

For this demo we will be using the <u>Suppliers and Parts</u> database which consists of three csv files.

- 1) Suppliers.csv
- 2) Parts.csv
- 3) Shipments.csv

It's a popular database that has its own Wikipedia page.

The Suppliers and Parts Wikipedia article is located here: https://en.wikipedia.org/wiki/Suppliers and Parts database

First, we will need to provision the following resources:

- Azure Key Vault
- Azure Data Lake
- Azure Data Factory
- SQL Server
- SQL Server Database
- Databricks Workspace

Few reminders when provisioning the resources:

- When provisioning Databricks, select the premium version as this allows for creating secret scopes.
- If using an Azure free trial version, you will need to upgrade to a subscription plan to access Databricks.
- The Azure Storage Account needs to be provisioned as a data lake.
- A basic SQL Server database is will be sufficient for this tutorial. You will need to enable the database server to connect to other Azure resources.

For this demo, I created a resource group <u>rg-databricks-hive-demo</u> and provisioned the following resources:

adf-databricks-hive-demo	Data factory (V2)	East US
akv-databricks-hive-demo	Key vault	East US
☐ 	Azure Databricks Service	East US
☐ ■ dlsdatabrickshivedemo	Storage account	East US
sql-databricks-hive-demo	SQL server	East US
sqldb-databricks-hive-demo (sql-databricks	SQL database	East US

The documentation for Azure naming conventions is located here: https://docs.microsoft.com/en-us/azure/cloud-adoption-framework/ready/azure-best-practices/resource-abbreviations

The documentation for Azure naming rules is located here:

https://docs.microsoft.com/en-us/azure/azure-resource-manager/management/resource-name-rules

Here are the Azure resource names if you want to copy and paste.

Name	Туре
db-databricks-hive-demo	Databricks
dlsdatabrickshivedemo	Azure Storage Account
akv-databricks-hive-demo	Azure Key Vault
sqldb-databricks-hive-demo	SQL Database
sql-databricks-hive-demo	SQL Server
adf-databricks-hive-demo	Azure Data Factory

Certain resources need to be globally unique, which we will cover in the next slide....

Naming rules and restrictions:

Resource	Scope	Length	Valid Characters
	Resource		Alphanumeric, underscores, and
Databricks	Group	3-64	hyphens.
Storage Account	Global	3-24	Lowercase letters and numbers.
			Alphanumeric and hyphens.
			Starts with letter. End with letter
			or digit. Can't contain consecutive
Key Vault	Global	3-24	hyphens.
			Can't use:
			$<>*%&:$ \/? or control characters.
SQL Database	Server	1-128	Can't end with period or space.
			Lowercase letters, numbers, and
			hyphens.
SQL Server	Global	1-63	Can't start or end with hyphen.
			Alphanumeric and hyphens.
Azure Data Factory	Global	3-63	Start and end with alphanumeric.

The documentation for Azure naming rules is located here:

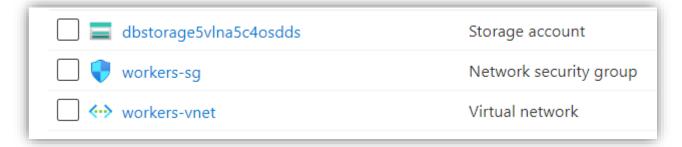
https://docs.microsoft.com/en-us/azure/azure-resource-manager/management/resource-name-rules

When you provision the resources, additional resource groups and resources are created.

You will see an additional resource group for <u>Databricks</u>, and another for the <u>Network Watcher</u>, as shown below.

atabricks-rg-db-databricks-hive-demo-spyhoadxx4z2g
NetworkWatcherRG
g-databricks-hive-demo-001

The <u>databricks-rg-db-databricks-hive-demo-spyhoadxx4z2g</u> resource group has the following resources:



The NetworkWatcherRG resource group has the following resources:

NetworkWatcher_eastus	Network Watcher

Next, we will connect our resources together.

In Databricks, we will need to understand two concepts:

- 1) Secret Scopes
- 2) Databricks Tokens

First, we will begin with secret scopes.

Secret scopes allow you to securely connect to your Azure Key Vault service, where we will store our database and storage account credentials.

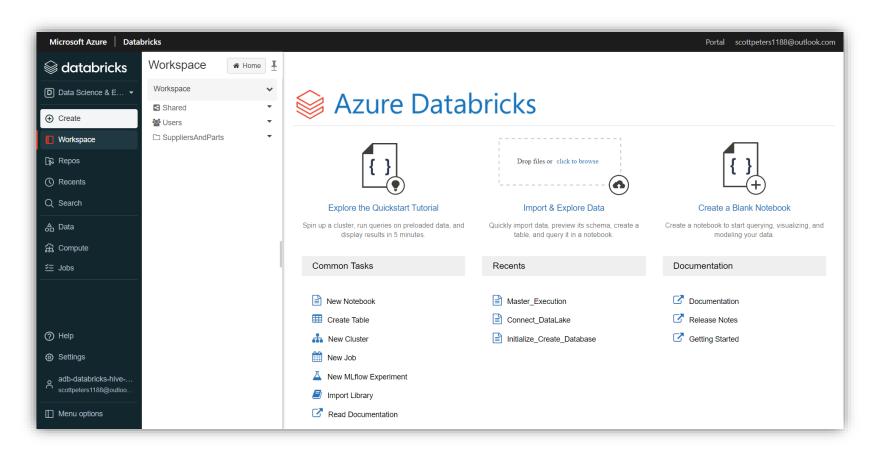
There are two different ways of creating a scope:

- 1) Azure Key Vault backed
- 2) Databricks backed

The Databricks secret scope documentation is located here:

https://docs.microsoft.com/en-us/azure/databricks/security/secrets/secret-scopes

Open the Databricks workspace.



Next, in the Databricks workspace, access the Create Secret Scope window.

To access the secret scope window in Databricks, attach the string "#secrets/createScope" after your Databricks instance.

A Databricks instance will have the following URL: https://<databricks-instance>#secrets/createScope

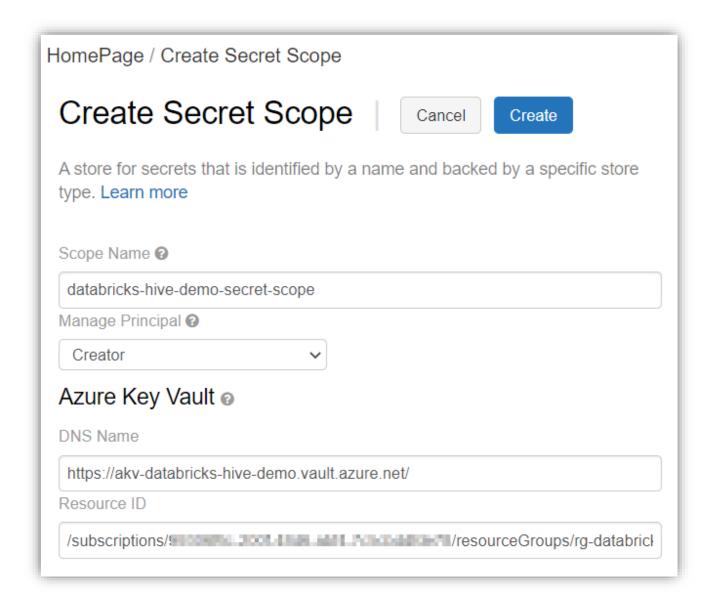
This URL is case sensitive; scope in createScope must be uppercase.

The documentation for Databricks secret scopes is located here: https://docs.microsoft.com/en-us/azure/databricks/security/secrets/secret-scopes

DATABRICKS SECRET SCOPE

The <u>Create Secret Scope</u> window will look like the following.

We will cover the <u>DNS Name</u> and <u>Resource ID</u> in next slides.



DATABRICKS SECRET SCOPE

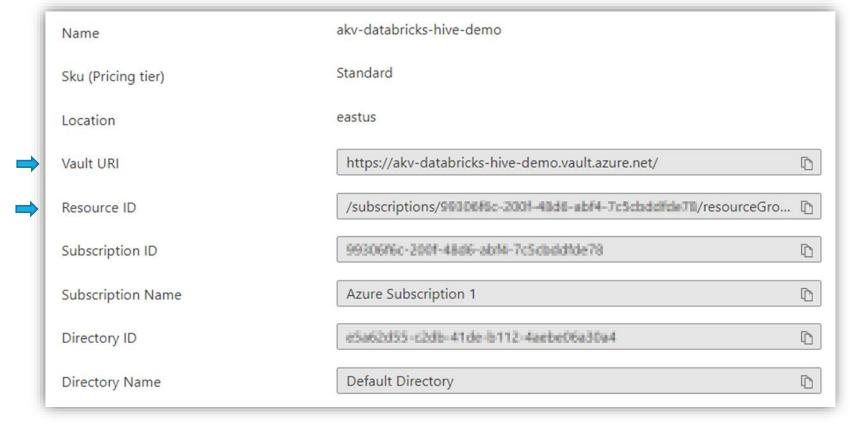


The <u>DNS Name</u> (Vault URI) and the <u>Resource</u> <u>ID</u> are in your Azure Key Vault properties page.



Copy these values from your Azure Key Vault properties page and insert the values into your <u>Create Secret</u> <u>Scope</u> window.

Azure Key Vault Properties Page



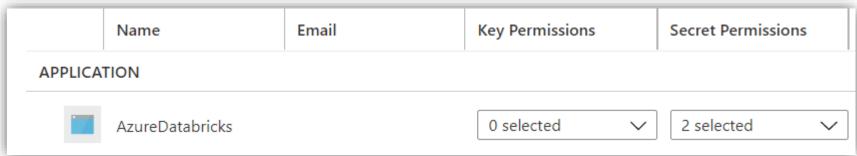
DATABRICKS SECRET SCOPE

Once completed, the following confirmation will appear.

The secret scope named databricks-hive-demo-secret-scope has been added.

Manage secrets in this scope in Azure KeyVault with manage principal = creator

And you will see the following access policy added to the key vault.



The documentation for the Azure Key Vault access policies is located here:

https://docs.microsoft.com/en-us/azure/key-vault/general/assign-access-policy?tabs=azure-portal

DATABRICKS TOKEN

Next, we will create a Databricks token.

This token will allow the Azure Data Factory to access the Databricks notebook.

<u>Tokens replace passwords in an authentication flow and should be protected like passwords.</u>

We will use this token when we setup a linked service in our Azure Data Factory.

The documentation for Databricks tokens is located here:

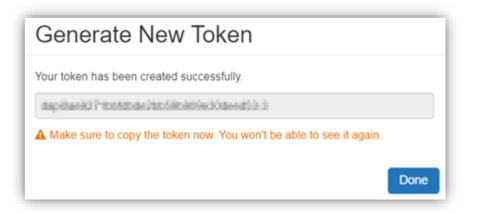
https://docs.databricks.com/dev-tools/api/latest/authentication.html

This token will be stored in the Azure Key Vault, which we will setup later.

DATABRICKS TOKEN

Create a token in Azure Databricks by navigating to the <u>User Settings</u> in your Databricks workspace and selecting <u>Generate New Token</u>.

Save the token, as directed by the yellow warning message.



I created the token "databricks-hive-demo", but it is the token's value that is important, not the name. This token will be stored in the Azure Key Vault, which we will setup later.

Once created, you will see the following in your Databricks <u>User Settings</u>.

ĺ	Comment	Creation ↑	Expiration
	databricks-hive-demo	2021-11-24 14:29:47 CST	Never

AZURE DATA LAKE

Next, we will setup the Azure Data Lake.

Navigate to the storage account you provisioned and create two containers:

- 1) source
- 2) hive

Within the <u>hive</u> container create a directory called <u>database</u>.

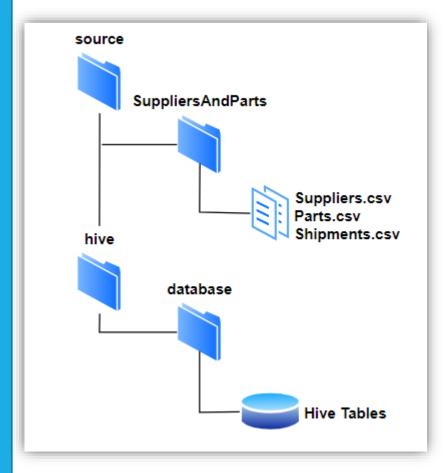
Within the source container, create a directory called SuppliersAndParts.

The directory <u>source > SuppliersAndParts</u> is where we will store the csv files to import, and <u>hive > Database</u> will contain the Hive tables (which we will create later).

Upload the text files from the GitHub repository into the <u>source > SuppliersAndParts</u> container.

AZURE DATA LAKE

Your file structure in the storage account will look like the following:



Copy the files Suppliers.csv, Parts.csv, and Shipments.csv files into the <u>SuppliersAndParts</u> folder.

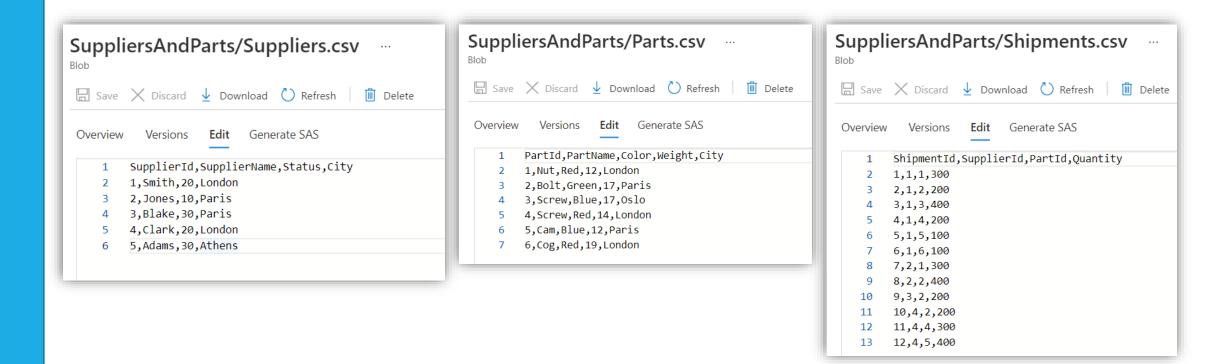
The <u>database</u> folder will contain the Hive tables, which we will setup later in this demo.

The GitHub repository for this demo is located here:

https://github.com/smpetersgithub/AdvancedSQLPuzzles

AZURE DATA LAKE

A quick peek at the three files reveals the following contents:

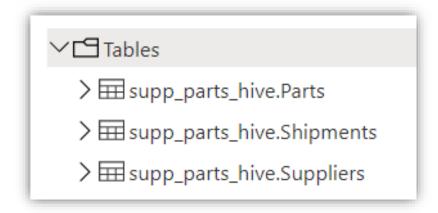


SQL SERVER DATABASE

Next, we will create the schema and tables in the SQL Server database.

The script to create the schema and tables is provided in the GitHub directory.

We will create a schema named supp parts hive and the following tables:



The GitHub repository for this demo is located here:

https://github.com/smpetersgithub/AdvancedSQLPuzzles

AZURE KEY VAULT

Next, we will create the secrets inside of the Azure Key Vault.

Here are the secrets you will need to setup in the Azure Key Vault.

Databricks will be able to access these values via the secret scope we created earlier.

Name	Туре	Status
DataLakeAccessKey		✓ Enabled
DatabricksToken		✓ Enabled
DataLakeName		✓ Enabled
sqldbName		✓ Enabled
sqljdbcPort		✓ Enabled
sqlPassword		✓ Enabled
sqlServerName		✓ Enabled
sqlUserName		✓ Enabled

These secrets will be used within the Databricks notebooks, except for <u>DatabricksToken</u>, which will be used by the Azure Data Factory linked service.

AZURE KEY VAULT

Secret Name	Value	Description
sqljdbcPort	1433	This should always be 1433
sqldbName	sqldb-databricks-hive-demo	The name of the database
sqlPassword	MyPassword	The password of the database
sqlUserName	MyUserName	The admin user of the database
	sql-databricks-hive-	
sqlServerName	demo.database.windows.net	The database server connection string
DataLakeAccessKey	Review the next slides	Copied from the storage account properties
DataLakeName	dlsdatabrickshivedemo	The name of the storage account
DatabricksToken	Created in previous slides	Used by Data Factory linked service

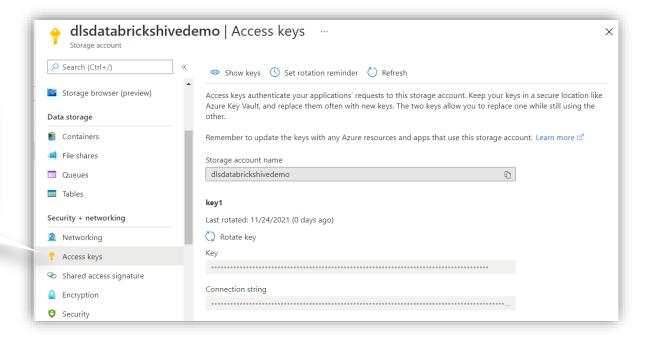
The two services we need Databricks to authenticate to are:

- 1) SQL Server Database
- 2) Azure Storage Account

The secret <u>DatabricksToken</u> will be used by the Data Factory linked service. We will create the Azure Storage Account access key in the next slides.

AZURE KEY VAULT

The secret <u>DataLakeAccessKey</u> is created by navigating to your storage account and copying the access key.



The documentation for Azure Storage Account access keys is located here:

https://docs.microsoft.com/en-us/azure/storage/common/storage-account-keys-manage?tabs=azure-portal

RECAP

Let's do a quick recap of everything we have accomplished so far:

- We have provisioned the Azure Resources
- We created a Databricks secret scope
- We created a Databricks token
- We setup our Azure Storage Account as a data lake with the needed directories and imported the csv files
- We created the schema and tables in the SQL Server Database
- We created the secret keys in the Azure Key Vault

RECAP

Next, we will perform the following:

- Import a Databricks workspace
- Create a Databricks cluster
- Run the database setup scripts
- Review the code in the workspace
- Test the Databricks workspace
- Create an Azure Data Factory pipeline
- Execute the Azure Data Factory pipeline

The following documentation gives a good overview of what we will be accomplishing for the remainder of these slides. Take a few moments to review the following documentation.

A Microsoft tutorial for incorporating Databricks into Data Factory is located here: https://docs.microsoft.com/en-us/azure/data-factory/transform-data-using-databricks-notebook

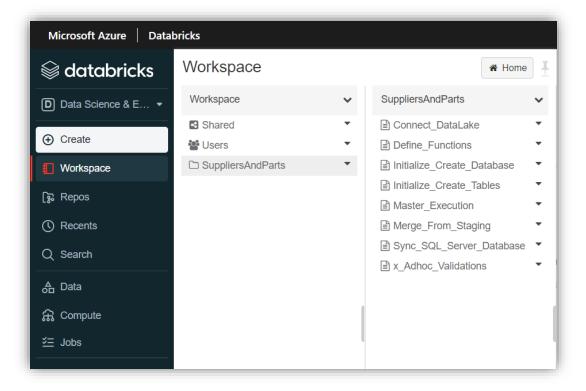
If you are unfamiliar with creating clusters, running notebooks, or navigating the workspace, I recommend working through a basic tutorial of Databricks before proceeding.

Begin by importing the workspace from the GitHub repository.

First, we will begin by importing the <u>SuppliersAndParts.dbc</u> workspace from the GitHub repository.

This workspace contains the notebooks needed to create our ETL.

Once you import the dbc file, you will see the following <u>SuppliersAndParts</u> folder in your workspace.



The workbook consists of the following notebooks:

Included Notebooks

- Connect_DataLake
- Create_Database
- Create_Tables
- Insert_Hive_Staging_Functions
- Insert_SQL_Server_Database
- Master_Execution
- Merge_Hive_Production
- Validation

Here is a brief description of each notebook sorted by their purpose.

We will cover each of these notebooks in more detail.

Notebook	Description
Create_Database	Initial script to create the Hive database.
Create_Tables	Initial script to create the Hive tables.
Connect_DataLake	Reads variables from the Azure Key Vault and connects to the Azure Data Lake.
Insert_Hive_Staging_Functions	Recreates the Hive staging tables and then inserts the data from the csv file into the staging table.
Merge_Hive_Production	Creates the functions needed to import the csv files into the staging tables.
Insert_SQL_Server_Database	Truncates and then inserts the data from the Hive production tables into the SQL Server tables.
Master_Execution	The master notebook which calls all other notebooks.
Validation	General script used for auditing.

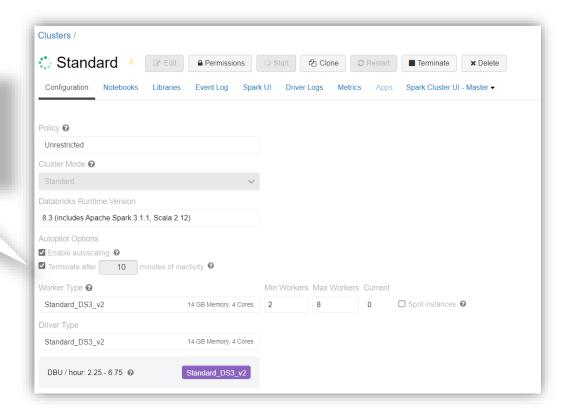
Before we move on, we need to create a Databricks cluster to run our notebooks.

The documentation for configuring Databricks clusters is located here:

https://docs.microsoft.com/en-us/azure/databricks/clusters/configure

Remember to put a termination inactivity time on the cluster!

A simple standard cluster will suffice. Feel free to experiment with different cluster sizes.



First, run the notebooks that create the Hive database and tables.

Run the following notebook first:

Create_Database

The following SQL statement creates a Hive database named <u>demo</u>.

```
Create Hive database

1 DROP DATABASE if exists demo CASCADE;
2 CREATE DATABASE IF NOT EXISTS demo COMMENT 'This is demo database' LOCATION '/demo/database';
```

Second, create the tables for the Suppliers, Parts and Shipments data.

Run the following notebook next:

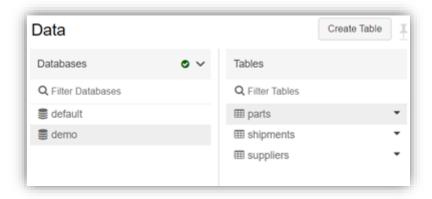
Create_Tables

This will create the three production tables for 1) Suppliers, 2) Parts and 3) Shipments.

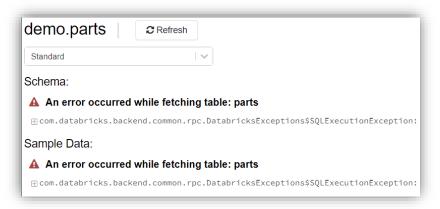
You will need to change the location only if you named the Azure Storage Account directory differently than this demo.

Second, create the tables for the Suppliers, Parts and Shipments data.

After creating the tables, you will see the production tables in the <u>Data</u> pane of the workspace. The staging tables will appear here after we run the pipeline for the first time.



You will get the following error when you select a table. This is normal.



Next, we will test the connection to the Azure Key Vault and the Azure Storage Account.

Run the following notebook to test the connection to the Azure Key Vault and the Azure Storage Account services.

Connect_DataLake

The following code connects to the Azure Key Vault.

If the Azure Key Vault secret names are different than this demo, you will need to modify these values. Note the only variable hardcoded is the secret scope that we created earlier.

Next, we will test the connection to the Azure Key Vault and the Azure Storage Account.

Connect_DataLake

The following code creates the connection to the data lake.

It connects once with the <u>dfs</u> syntax, and then again with the <u>blob</u> syntax.

The documentation for Spark configuration settings is located here: https://docs.microsoft.com/en-us/azure/databricks/kb/data/get-and-set-spark-config

Next, let's test if we can run the master notebook.

Master_Execution

The <u>Master Execution</u> notebook will run once for each of the three csy files:

- I) Suppliers.csv
- 2) Parts.csv
- 3) Shipments.csv

This notebook calls the following notebooks in order:

- 1) Insert_Hive_Staging_Functions
- 2) Merge_Hive_Production
- 3) Insert_SQL_Server_Database

Next, let's test if we can run the master notebook.

Master_Execution

This notebook performs the following steps:

Step 1

It first reads the functions from the <u>Insert Hive Staging Functions</u>, which also reads the <u>Connect DataLake</u> notebook.

Step 2

It then determines the variables FileName and the ProcessName.

Step 3

Next, it runs the function <u>ProcessHiveStagingData</u>, which is located in the <u>Insert Hive Staging Functions</u> notebook.

Step 4

It then runs the <u>Merge Hive Production</u> notebook, and then the <u>Insert SQL Server Database</u>.

Next, let's test if we can run the master notebook.

DBUTILS.WIDGETS

The <u>dbutils.widgets</u> code reads the variable <u>ProcessName</u> and <u>FileName</u> from the Azure Data Factory.

To test the code in Databricks, manually set the value of <u>ProcessName</u> and <u>FileName</u> where applicable.

Reads the variable(s) from the Azure Data Factory pipeline

```
#Determines the ProcessName and FileName from the Azure Data Factory
dbutils.widgets.text("ProcessName", "","")
ProcessName = dbutils.widgets.get("ProcessName")
dbutils.widgets.text("FileName", "","")
FileName = dbutils.widgets.get("FileName")

#Manually sets the ProcessName and FileName if testing in Databricks
if ProcessName == "":
ProcessName = 'Suppliers'
FileName = '/SuppliersAndParts/Parts.csv'

print('ProcessName is',ProcessName)
print('FileName is',FileName)
```

Next, let's test if we can run the master notebook.

DBUTILS.WIDGETS

The documentation for Databricks widgets is located here:

https://docs.databricks.com/notebooks/widgets.html

Widgets

November 16, 2021

Input widgets allow you to add parameters to your notebooks and dashboards. The widget API consists of calls to create various types of input widgets, remove them, and get bound values.

Widgets are best for:

- Building a notebook or dashboard that is re-executed with different parameters
- Quickly exploring results of a single query with different parameters

Widget types

There are 4 types of widgets:

- text: Input a value in a text box.
- dropdown: Select a value from a list of provided values.
- combobox: Combination of text and dropdown. Select a value from a provided list or input one in the text box.
- multiselect: Select one or more values from a list of provided values.

Now we will look at some of the specifics in the notebooks.

You most probably encountered a few errors in running the master notebook.

In the next slides we will look at some of the specifics in the notebooks to help troubleshoot and understand the logic.

We will not cover all the different features of this code, but enough to be sufficient to troubleshoot, create test runs, and understand the overall process steps.

We will first start with the notebook, <u>Insert Hive Staging Functions</u> and its various functions, before reviewing the other notebooks in this workspace.

Insert_Hive_Staging_Functions

This notebook defines the following functions needed to import and verify the text file and then insert the text file into the staging table.

The function ProcessHiveStagingData calls all other functions.

Order	Function Name	Description	
Master	ProcessHiveStagingData	Reads the csv file and executes InsertDataToHiveFunction	
1	GetSchema	Defines the schema for the file	
2	GenerateIncomingDataHeader	Creates a data frame of the file header	
3	VerifyHeader	Validates incoming file header to schema definition	
4	InsertDataToHive	Inserts the data into the Hive tables	

We will cover each of these functions in the next few slides.

Insert_Hive_Staging_Functions

The <u>ProcessHiveStagingData</u> is the master function that calls the other four functions. I have highlighted the function calls below.

```
Process data to Hive
    def ProcessHiveStagingData(FilePath, ProcessName):
      print("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath)
 3
      df = spark.read.csv("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath, sep=",", mode="DROPMALFORMED", schema=GetSchema(ProcessName))
 5
      if (len(df.head(1)) == 0):
 6
        raise ValueError("Please review the schema. Possible new columns were added.")
      else:
                                                                                                                             Get Schema
        new_df = GenerateIncomingDataHeader(df.limit(1))
 8
 9
10
      #Verify the header
                                                                               Generate Incoming Header
11
      VerifyHeader(df,new_df)
                                             Verify Header
12
13
      #Fetch data in file
      df = df.filter(~col(df.schema.fields[0].name).contains(df.schema.fields[0].name))
14
15
      #Concatenates ProcessName and the string "Stage" to derive the staging table name.
16
17
      TableName = ProcessName+"Stage"
                                                 Insert Data To Hive
18
      InsertDataToHive(df,TableName)
19
20
      print('End of Function: ProcessData')
```

Insert_Hive_Staging_Functions

The variable <u>TableName</u> is the concatenation of the variable <u>ProcessName</u> and the string "<u>Stage</u>". Also, the field separator is set when we create the data frame.

```
Process data to Hive
    def ProcessHiveStagingData(FilePath, ProcessName):
      print("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath)
      df = spark.read.csv("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath, sep=",", mode="DROPMALFORMED", schema=GetSchema(ProcessName))
 5
      if (len(df.head(1)) == 0):
        raise ValueError("Please review the schema. Possible new columns were added.")
                                                                                                         The column separator is set here
 8
        new df = GenerateIncomingDataHeader(df.limit(1))
 9
10
      #Verify the header
      VerifyHeader(df,new_df)
11
12
13
      #Fetch data in file
      df = df.filter(~col(df.schema.fields[0].name).contains(df.schema.fields[0].name))
14
15
      #Concatenates ProcessName and the string "Stage" to derive the staging table name.
16
17
      TableName = ProcessName+"Stage"
18
      InsertDataToHive(df,TableName)
19
                                                                TableName is set here
 20
      print('End of Function: ProcessData')
```

Insert_Hive_Staging_Functions

A Dataframe named df is also created in this function.

```
Process data to Hive
    def ProcessHiveStagingData(FilePath, ProcessName):
      print("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath)
      df = spark.read.csv("wasbs://source@"+datalakename+".blob.core.windows.net"+FilePath, sep=",", mode="DROPMALFORMED", schema=GetSchema(ProcessName))
 5
      if (len(df.head(1)) == 0):
        raise ValueError("Please review the schema. Possible new columns were added.")
      else:
 8
        new_df = GenerateIncomingDataHeader(df.limit(1))
                                                                                                    The csv file is imported here and a
 9
10
      #Verify the header
                                                                                                        Dataframe is created via the
11
      VerifyHeader(df,new_df)
                                                                                                             spark.read.csv call
12
13
      #Fetch data in file
14
      df = df.filter(~col(df.schema.fields[0].name).contains(df.schema.fields[0].name))
15
      #Concatenates ProcessName and the string "Stage" to derive the staging table name.
16
      TableName = ProcessName+"Stage"
17
      InsertDataToHive(df,TableName)
18
19
 20
      print('End of Function: ProcessData')
```

Insert_Hive_Staging_Functions

The <u>GetSchema</u> function in the first function called from the master function ProcessHiveStagingData.

The variable Name is passed to the function.

This defines the blueprint, known as a schema, that defines the name and data type of each column. We use <u>StringType</u> to define each column.

The documentation for Spark StructType class is located here:

https://spark.apache.org/docs/latest/api/python/reference/api/pyspark.sql.types.StructType.html

Insert_Hive_Staging_Functions

The <u>GenerateIncomingDataHeader</u> function creates the Dataframe <u>new df</u> of the imported file's header.

We will perform a validation in the next function, VerifyHeader.

Create a Dataframe of the imported file's header

```
def GenerateIncomingDataHeader(df):
      list_of_new_column_names = []
2
3
      for i in df.limit(1).collect()[0]:
5
        list of new column names.append(i)
      for i,k in enumerate(df.schema.fields):
        k.name = list_of_new_column_names[i]
7
8
      new_df = spark.createDataFrame(df.rdd, df.schema)
9
      return new_df
11
12
      print('End of Function: GenerateIncomingDataHeader')
```

Insert_Hive_Staging_Functions

The function VerifyHeader compares the file header to the schema definition.

If they are not the same, an error is raised.

Insert_Hive_Staging_Functions

The insert into the staging table has the options of overwrite, truncate, and overwriteSchema.

On each execution the staging table is recreated and then inserted into.

Insert data to Hive

```
def InsertDataToHive(df, TableName):
    print('Table being inserted is',TableName)
    print('The table path is',TablePathHive)

#Truncate and overwrite options are set to True
    df.write.mode("overwrite").option("truncate", True).option("overwriteSchema", "true").format("delta").option("path",
    TablePathHive+"/"+TableName+"/delta/").saveAsTable("demo."+TableName)

print('End of Function: InsertDataToHive')
```

Merge_Hive_Production

Moving on to the next notebooks...

This notebook merges the data from the staging to the production tables.

print('ProcessName is',ProcessName)

This notebook also uses the dbutils.widgets, which we have discussed in previous slides.

To run this notebook in Databricks for the <u>Suppliers.csv</u> file, manually set the variable <u>ProcessName</u> to Suppliers.

Reads the variable(s) from the Azure Data Factory pipeline dbutils.widgets.text("ProcessName", "","") ProcessName = dbutils.widgets.get("ProcessName") #Manually sets the ProcessName and FileName if testing in Databricks if ProcessName == "": ProcessName = 'Suppliers' 7

Insert_SQL_Server_Database

The insert into to the SQL Server database has the same overwrite and truncate options as the inserts into the Hive staging tables.

If you receive connection errors to the database, check that the SQL Server has the option set to connect to other Azure services.

Truncate and insert into SQL Server database

```
if ProcessName=="Suppliers":
    df = spark.sql("""
    Select
        SupplierId, SupplierName, Status, City
    from
        demo.Suppliers
        """
        )
        df.write.mode("overwrite").option("truncate", True).jdbc(url=url, table="supp_parts_hive.Suppliers", properties=properties);
```

Insert_SQL_Server_Database

This notebook also uses the <u>dbutils.widgets</u>.

Reads the variable(s) from the Azure Data Factory pipeline

```
dbutils.widgets.text("ProcessName", "","")
ProcessName = dbutils.widgets.get("ProcessName")

#Manually sets the ProcessName and FileName if testing in Databricks
if ProcessName == "":
ProcessName = 'Suppliers'

print('ProcessName is', ProcessName)
```

Validations

This notebook is for creating and saving any ad-hoc statements.

After you have ran an ETL, I recommend using this notebook to explore the data.

Remember you may have only tested on your Suppliers data. The Parts and Shipments tables may be blank if you have not tested these processes.

Also available in this notebook is how to view the history of the tables.

Next, we will setup the Azure Data Factory.

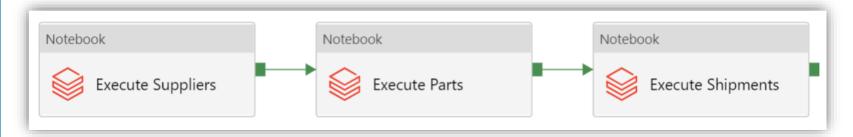
We will need to create two Azure Linked Services.

One to link to our 1) Azure Key Vault, and the second to link to the 2) Databricks Workspace.

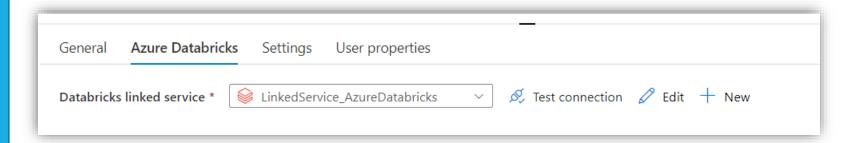
The Databricks Linked Service will utilize the Azure Key Vault Linked Service and access the Database token we setup earlier in the demo.

We will cover these steps in the next few slides.

Before we create the linked services, create a pipeline and add three Databricks notebooks activities. Chain them together to avoid getting "unexpected failure while waiting for the cluster" errors.

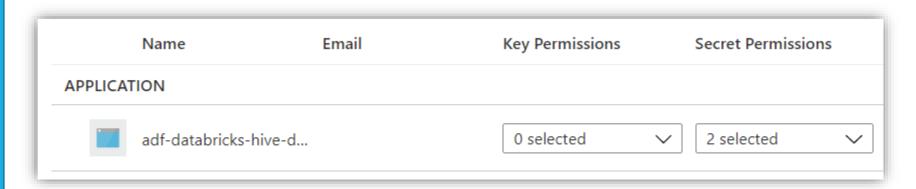


Each of these activities will utilize the Databricks linked service we will create in the next slides.



First, the Data Factory will need access to the Azure Key Vault. Navigate to your Azure Key Vault and add an access policy that allows the Data Factory secret permissions.

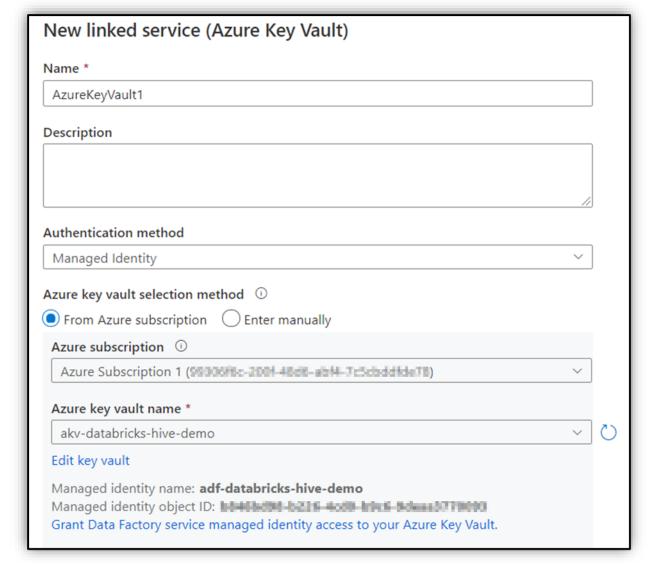
The Azure Key Vault you provisioned will now have an access policy for your Databricks and the Data Factory.



AKV Linked Service

To link to the Azure Key Vault, create the AKV linked service using your Azure subscription.

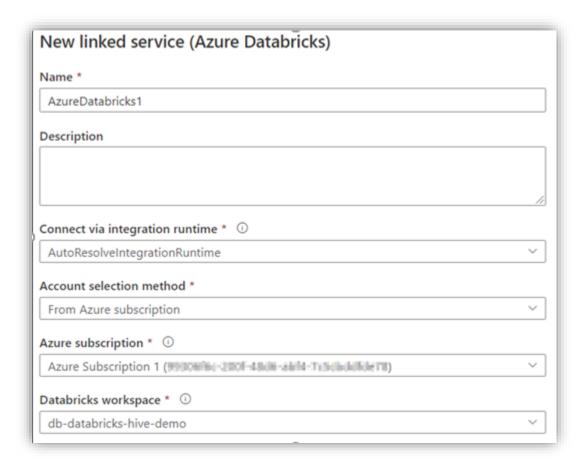
You can also enter the AKV manually by providing the URL of the linked service, which is located in the properties page of the AKV.



Databricks Linked Service

To link to the Databricks workspace, fill in the following dropdowns with your subscription and workspace information.

You can also enter the information in manually by selecting Enter Manually from the dropdown box.

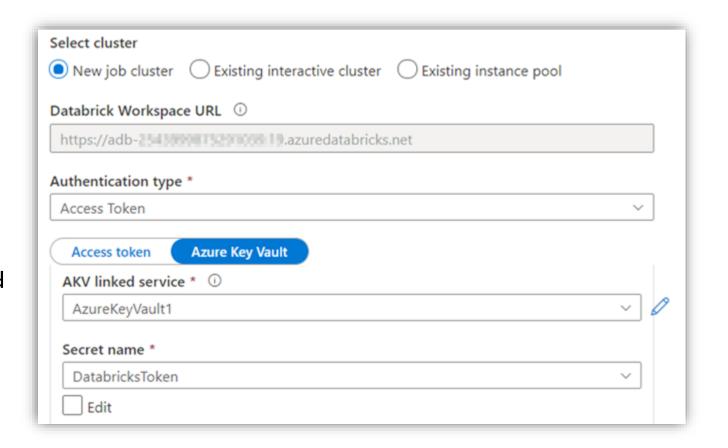


Databricks Linked Service

Next, select <u>Access Token</u> as the authentication type.

Link to the AKV by using the linked service we created in a prior slide.

Then select the secret name you used to store the Databricks token.

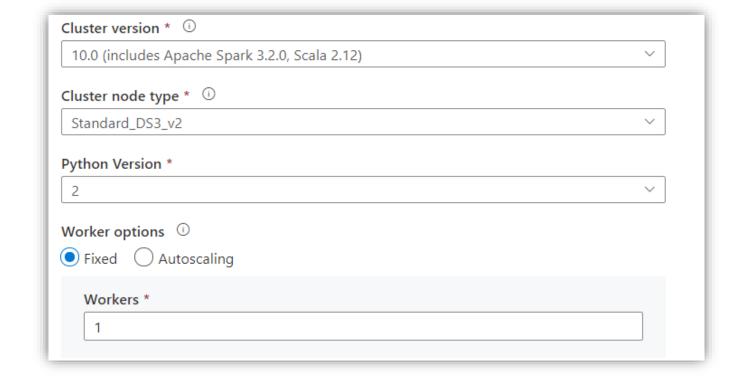


Databricks Linked Service

Finish the linked service by filling in the rest of the fields.

Here you can see the options that I chose for cluster version, node type, etc....

Selecting a new job cluster is the best option for cost savings, but it does add a few minutes of processing time to the pipeline to allow for the cluster to start up.



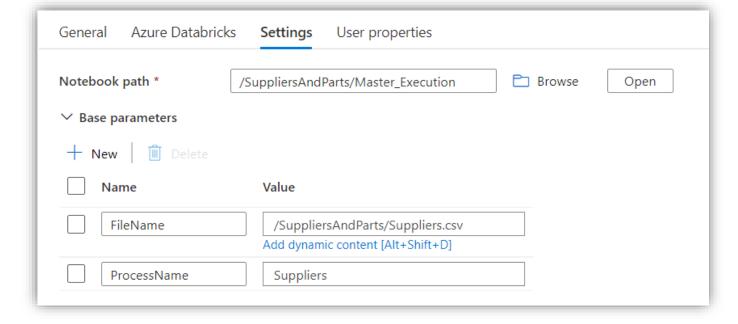
Now let us move back to the following pipeline activities and update the Settings tab for each activity.



Select the notebook path and create two parameters 1) FileName and 2) ProcessName.

Fill in the appropriate information as shown in the screenshot.

You will need to do this for the 1) Suppliers, 2) Parts and 3) Shipments activities.



Execute the pipeline by selecting **Debug**.

Ensure there is no running cluster in your Databricks workspace before executing, else you will get "unexpected failure while waiting for the cluster" error.

The total execution time for my pipeline is around 11 minutes.

eline run ID: 855cf3a2-746f-4470-972a-e1594792a843 @ 💍 🛈						
Name	Туре	Run start	Duration	Status	Integration runtime	
Shipments	Notebook	2021-11-24T23:24:32.057919	00:02:50	Succeeded	DefaultIntegrationRuntime (East	
Parts	Notebook	2021-11-24T23:21:56.970290	00:02:34	Succeeded	DefaultIntegrationRuntime (East	
Suppliers	Notebook	2021-11-24T23:17:03.082412	00:04:52	Succeeded	DefaultIntegrationRuntime (East	

TESTING

To test your ETL, review the data in your Hive tables with the statements in the <u>Validations</u> notebook. Also, review the data in the SQL Server tables as well.

Then alter the CSV files by adding new records and modifying the current records. Run the pipeline and verify the output.

To best learn the specifics of the Databricks notebooks, I recommend changing names of your AKV, storage account, secret names, etc.... determine where the notebooks break and fix accordingly.

CONGRATULATIONS

You now have a template for using Databricks as a transformation service!

Check out my GitHub repository and SQL blog for all sort of puzzles, tips and tricks.

The GitHub repository for this demo is located here:

https://github.com/smpetersgithub/AdvancedSQLPuzzles