



Big Data Engineering Project on Azure

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TABLE OF CONTENTS

01

**Project
Overview**

02

**About
Data**

03

Data Ingestion

04

**Data
Transformation**

05

**Data
Visualization**

06

Future Work

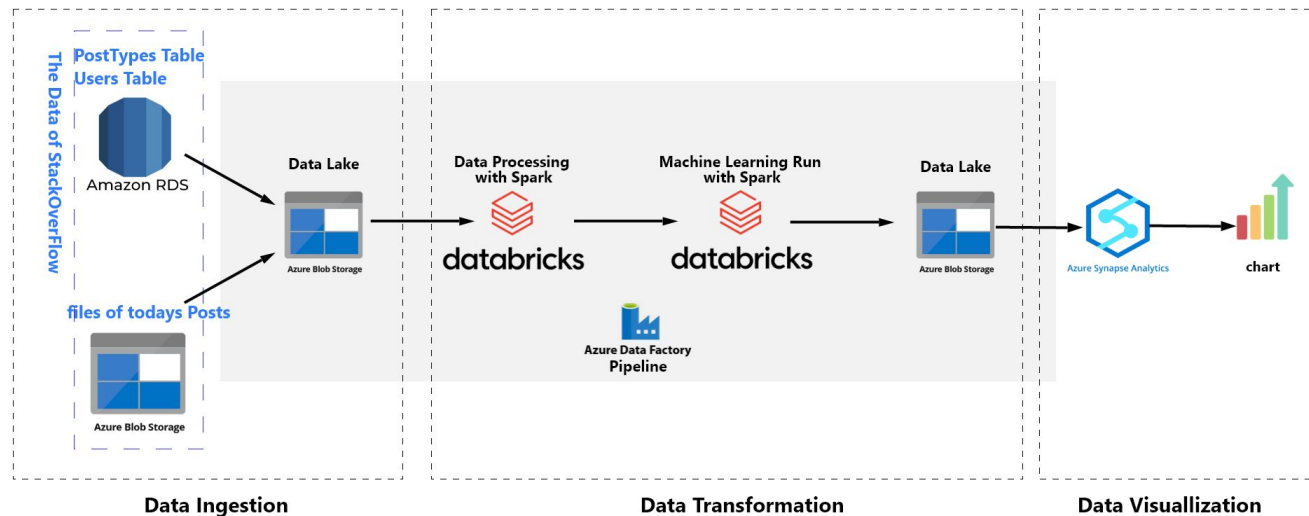


01 **Project Overview**



Project Overview

Our Azure Big Data Project involves ingesting data from two sources into Azure Data Lake, where it undergoes transformation and machine learning. The ML results are reintegrated, and Azure Synapse connects to generate insightful reports, ensuring efficient data handling and informed decision-making.





02 **About Data**

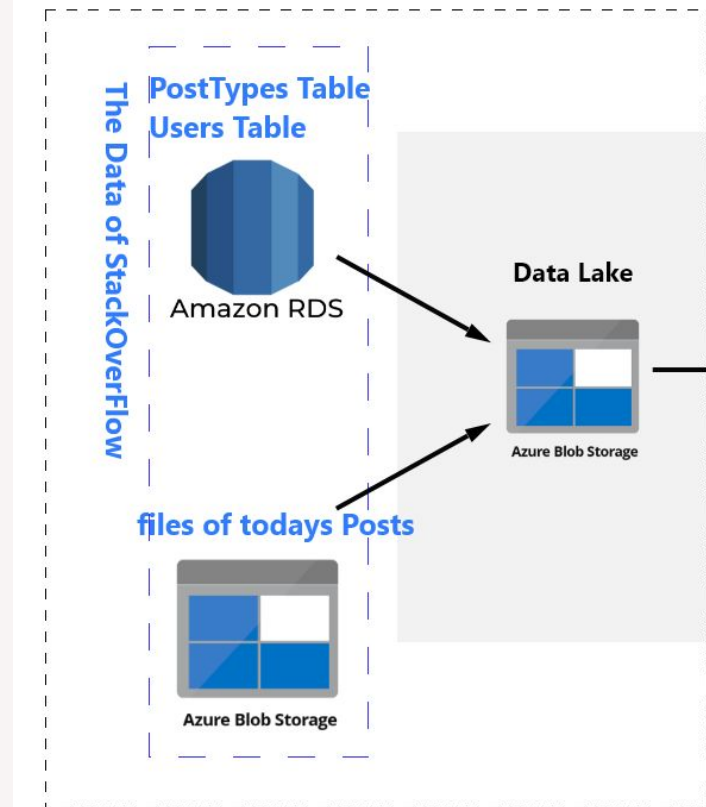


About Data

Data Source: The dataset is sourced from Stack Overflow, containing records of daily online posts, along with information on post types and users.

RDS (Relational Database Service): Users and PostTypes tables are hosted on RDS PostgreSQL database. These tables undergo weekly updates following Slowly Changing Dimension (SCD) type 1 methodology, where only new records are retained, and old records are overwritten.

Azure Storage Blob: The daily posts data is stored in Azure Storage Blob in parquet format. Multiple files exist, necessitating the copying of all files into the storage blob for comprehensive data access and analysis.

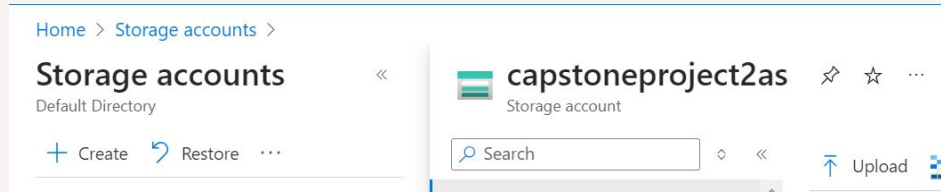




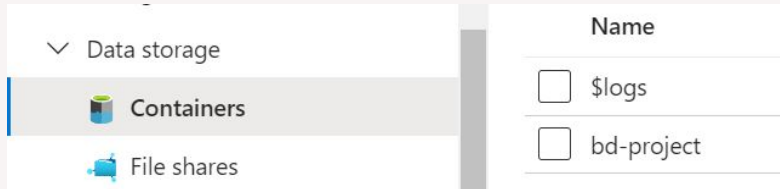
03 **Data Ingestion**

1. Data Lake

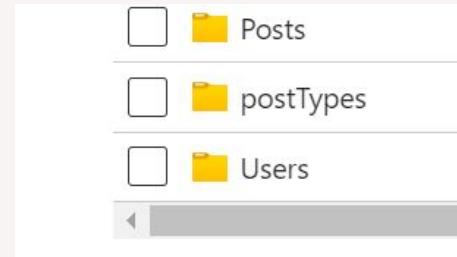
STEP 1: create a storage account



STEP 2: Create a container in the storage account that was created







STEP 3: create three folders: posts, postType, users




2. Data Factory



STEP 1: Create 2 pipelines
copyOnceWeek and
copyPostsEveryday

Factory Resources  


 Filter resources by name 


 Pipelines

 2

  big-data-project-pipeline

 2










 CopyOnceWeek

 CopyPostsEveryday

STEP 2: Create 3
linked services

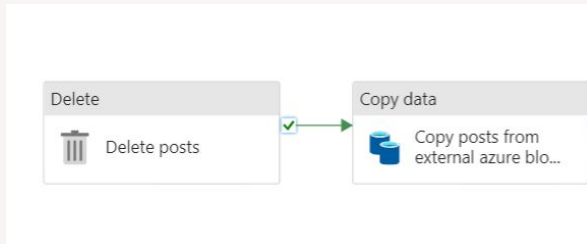
 ls_my_blob	Azure Blob Storage
 ls_rds_pg	PostgreSQL
 ls_wcd_blob	Azure Blob Storage

STEP 3: Create 6
datasets

 Datasets	6
<div>  wcd-bd-project-datasets</div>	6
<div> ds_parquet_post_blob</div>	
<div> ds_parquet_post_to_my_blob</div>	
<div> ds_postType</div>	
<div> ds_postType_rds_to_my_blob</div>	
<div> ds_users</div>	
<div> ds_users_rds_to_my_blob</div>	

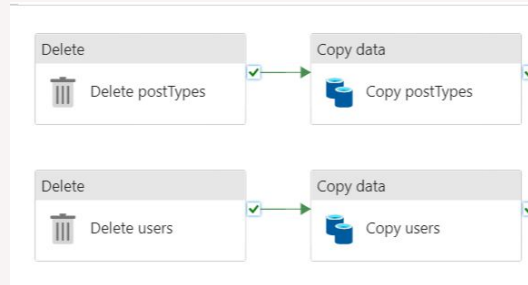
2. Data Factory

STEP 4: Create copy activity and delete activity for copyPostsEveryday pipeline and 2 copy activity and 2 delete activity for copyOnceWeek pipeline



STEP 2: Create trigger for each pipeline

<input type="checkbox"/>	CopyPostsEveryday	5/13/2024, 2:00:00 AM	5/13/2024, 2:00:26 AM	27s	run_pipeline_everyday	✓ Succeeded	Original
<input type="checkbox"/>	CopyPostsEveryday	5/12/2024, 2:00:01 AM	5/12/2024, 2:00:55 AM	55s	run_pipeline_everyday	✓ Succeeded	Original
<input type="checkbox"/>	CopyOnceWeek	5/12/2024, 2:00:00 AM	5/12/2024, 2:01:46 AM	1m 47s	run_pipeline_everyweek	✓ Succeeded	Original
<input type="checkbox"/>	CopyPostsEveryday	5/11/2024, 2:00:00 AM	5/11/2024, 2:00:48 AM	49s	run_pipeline_everyday	✓ Succeeded	Original
<input type="checkbox"/>	CopyPostsEveryday	5/10/2024, 2:00:01 AM	5/10/2024, 2:00:44 AM	44s	run_pipeline_everyday	✓ Succeeded	Original
<input type="checkbox"/>	CopyPostsEveryday	5/9/2024, 2:00:00 AM	5/9/2024, 2:01:19 AM	1m 20s	run_pipeline_everyday	✓ Succeeded	Original



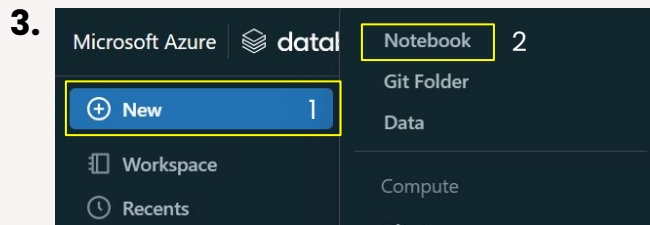
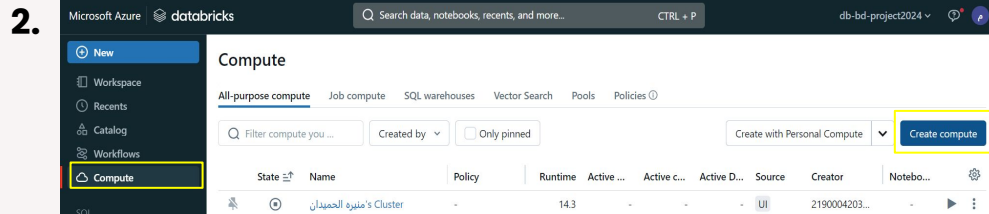
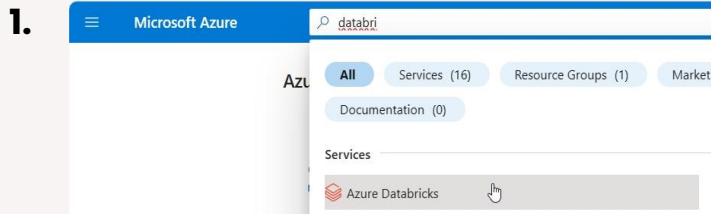


04 **Data Transformation**

1. Databricks Mount

Main Goal: synchronizing the Databricks directory with Azure storage container

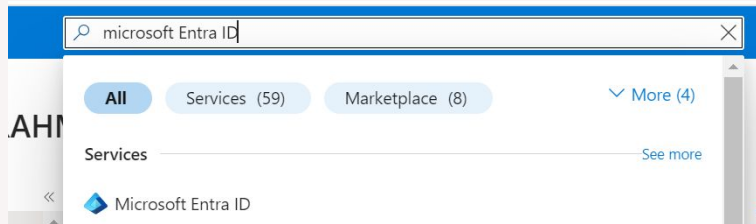
Step 1: Set up an Azure Databricks workspace, computing cluster, and notebook.



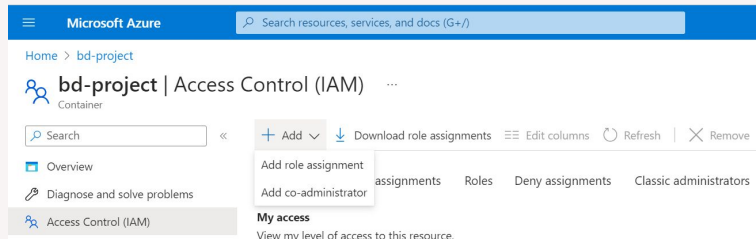
1. Databricks Mount

Step 2: In Azure, authorize Azure Databricks to access your Storage container.

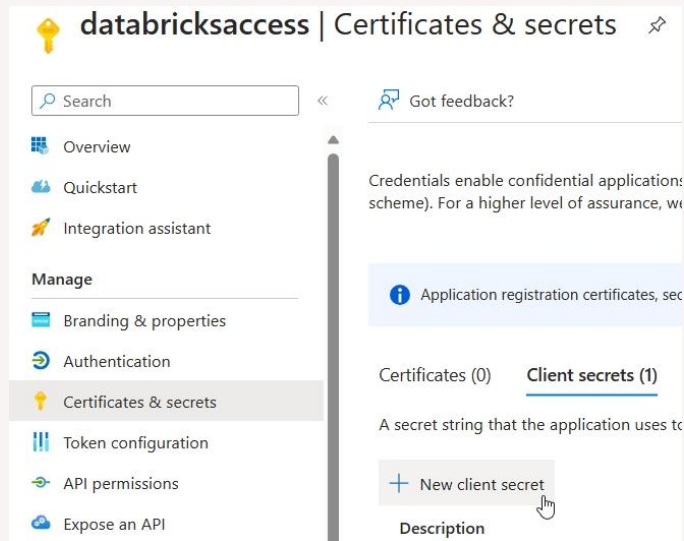
1.



3.

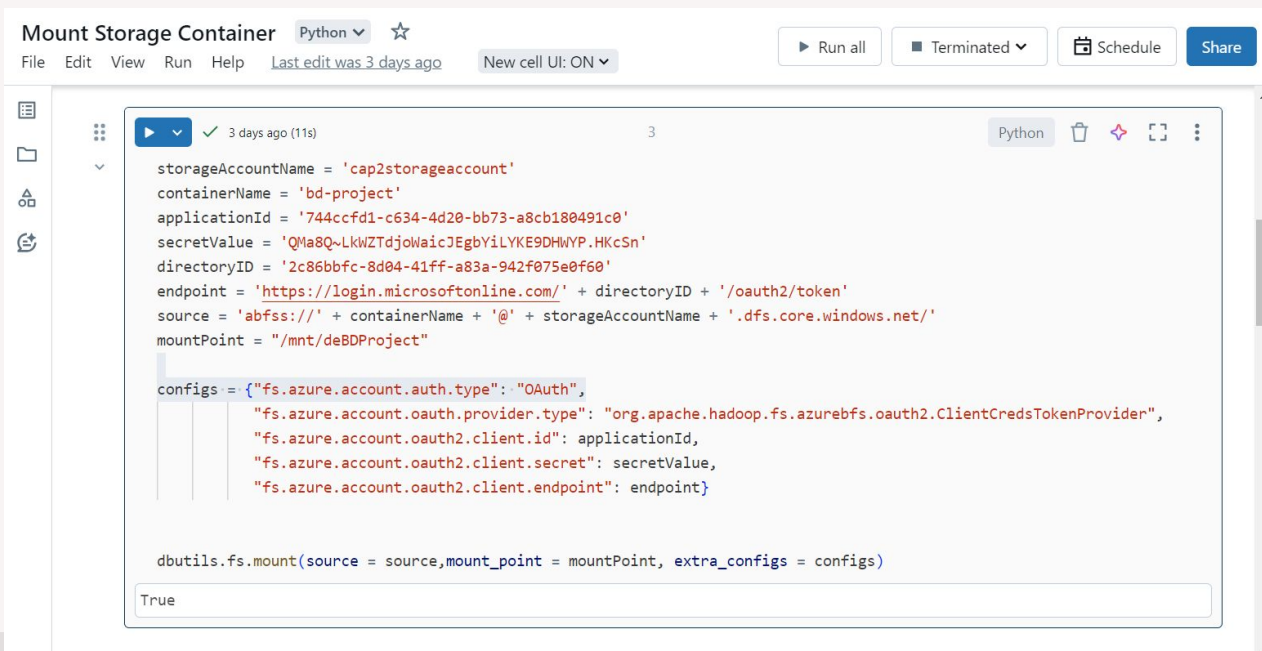


2.



1. Databricks Mount

Step 3: Mount your Storage container to the Azure Databricks directory so the Databricks can access it in the same way that it would a local file system.



The screenshot shows a Databricks notebook interface. The title bar reads "Mount Storage Container" with a Python language selector and a star icon. Below the title bar are tabs for "File", "Edit", "View", "Run", and "Help", along with a timestamp "Last edit was 3 days ago" and a dropdown for "New cell UI: ON". On the right side of the title bar are buttons for "Run all", "Terminated", "Schedule", and "Share". The notebook content area displays a Python cell that has been executed successfully, indicated by a green checkmark and the text "3 days ago (11s)". The code defines variables for storage account name, container name, application ID, secret value, directory ID, endpoint, source, and mount point. It then creates a configuration dictionary for the mount and calls the `dbutils.fs.mount` function. The output of the cell is `True`.

```
storageAccountName = 'cap2storageaccount'
containerName = 'bd-project'
applicationId = '744ccfd1-c634-4d20-bb73-a8cb180491c0'
secretValue = 'QMa8Q~LkWZTdjoWaicJEgbyILYKE9DHWYP.HKcSn'
directoryID = '2c86bbfc-8d04-41ff-a83a-942f075e0f60'
endpoint = 'https://login.microsoftonline.com/' + directoryID + '/oauth2/token'
source = 'abfss://' + containerName + '@' + storageAccountName + '.dfs.core.windows.net/'
mountPoint = "/mnt/deBDProject"

configs = {"fs.azure.account.auth.type": "OAuth",
          "fs.azure.account.oauth.provider.type": "org.apache.hadoop.fs.azurebfs.oauth2.ClientCredsTokenProvider",
          "fs.azure.account.oauth2.client.id": applicationId,
          "fs.azure.account.oauth2.client.secret": secretValue,
          "fs.azure.account.oauth2.client.endpoint": endpoint}

dbutils.fs.mount(source = source, mount_point = mountPoint, extra_configs = configs)
```

True

2. ML Model Training

Step 1: Join the Posts and the Posttypes tables, as we will utilize the Posttypes column in the Posttypes data to filter out the desired data from the Posts table. And then filter the required records.

Step 2: Prepare the data for machine learning training.

Step 3: Train the machine learning model.

Step 4: Save the model to an Azure storage folder so it can be used for future forecasts.

3. Achievement

- Accuracy for each Model

Model	Logistic Regression	Naive Bayes
for classes has count greater than 1	46%	47%
for top 20 tags	81%	74%
for top 10 tags	80%	72%

4. NLP Prediction

4.1 NLP Prediction Notebook

A Databricks notebook will execute the following steps:

- Load the posts data (Posts file) and the trained ML model.
- Define a User Defined Function (UDF) to perform data cleaning and transformation on the post content before feeding it to the model.
- Utilize the UDF to generate topic predictions for each post.
- Summarize the predicted topics and calculate the quantity of each topic.
- Save the resulting topic summary report (CSV file) to a designated Azure storage folder for BI access.

4. NLP Prediction

4.2 Data Factory

STEP 1: Create a Databricks notebook activity.



STEP 2: Generate Access token in Databricks workspace.

Generate new token

Comment

ADF_Pipeline

Creation ↕

2024-05-13 16:58:28 +03

STEP 3: Create a new Linked Service for connecting to Databricks workspace. Authentication Type: "Access token".

Edit linked service
Azure Databricks [Learn more](#)

Name *
Is_dbb_ml_post

Description
link to Azure Databricks

Connect via integration runtime *
AutoResolveIntegrationRuntime

Account selection method *
☐ From Azure subscription ☒ Enter manually

Databricks Workspace URL *
<https://adb-5703517841621899.9.azuredatabricks.net>

Authentication type *
Access Token

Access token [Azure Key Vault](#)

Access token *

Select cluster
☐ New job cluster ☒ Existing interactive cluster ☐ Existing instance pool

Existing cluster ID *
0512-174347-ukaso08c

Annotations
+ New

> Parameters
> Advanced

Save Cancel [Test connection](#)

4.2 Data Factory

STEP 4: Complete the Databricks Notebook Activity Configuration

Select a file or folder.

Root folder > Users >

ML Sentiment Analysis

STEP 5: Link Activities





05 **Data Visualization**

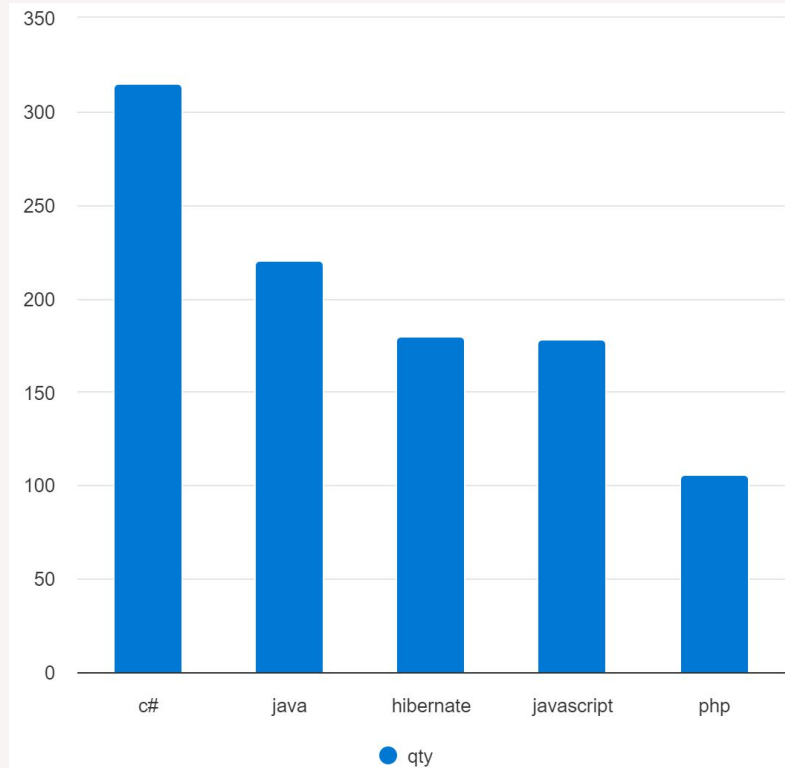
Azure Synapse Analytics

A powerful tool for organizations seeking to unlock the full potential of their data. It offers a unified platform for data warehousing, big data analytics, and machine learning, enabling faster time to insights, improved decision making, and enhanced efficiency.



Azure Synapse Analytics

```
Run Undo Publish Query plan Connect to Built-in Use database master
1 -- This is auto-generated code
2 SELECT
3     TOP 5 *
4 FROM
5     OPENROWSET(
6         BULK 'https://bdproject1.dfs.core.windows.net/bd-project/BI/ml_result.csv',
7         FORMAT = 'CSV',
8         PARSER_VERSION = '2.0',
9         HEADER_ROW = TRUE
10    ) AS [result]
11
```





06 **Future Work**



Future Work

- Join users table with post-type and posts to provide additional features to the machine learning model and can help understand how different groups of users interact with different post types
- Develop and integrate more complex machine learning models in Databricks, possibly exploring deep learning technique for more accurate predictions and insights.
- Extend visualization capabilities by integrating with other BI tools like Power BI or Tableau for more interactive and user-friendly dashboards.

The background features a light gray field with abstract geometric shapes in bright blue and dark blue at the corners. Thin, dark gray lines resembling circuit traces with small circular endpoints are positioned in the top right and bottom left corners.

Thank You
Any Questions!