



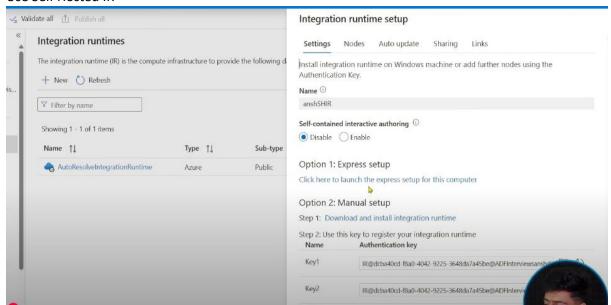
Your company currently stores large volumes of structured and semistructured data in an on-premises data warehouse and file servers. Leadership has decided to migrate this data to Azure Data Lake Storage Gen2 to enable better scalability, analytics, and integration with other Azure services.

QUESTION

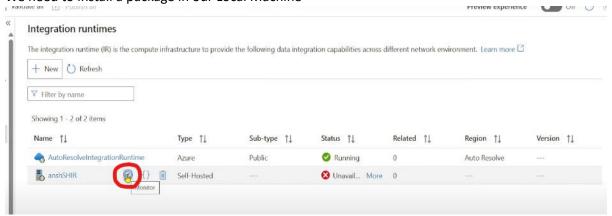
How would you approach migrating data from the onpremise systems to Azure Data Lake Storage Gen2?

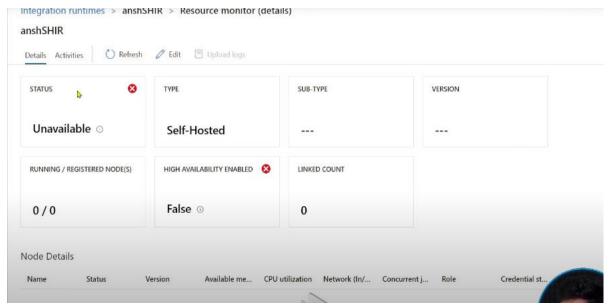


Use Self Hosted IR



We need to install a package in our Local Machine





So start installing using Manual setup (advisable), post installing provide the key. In real world, we create VMs as our IR(instead of downloading in local machine, download the IR in VM), in case of throttling(CPU utilisation is 80-90%) scale up VM. Easy to scaleup VM instead of local. Post that create a LS for the datastore using SHIR.





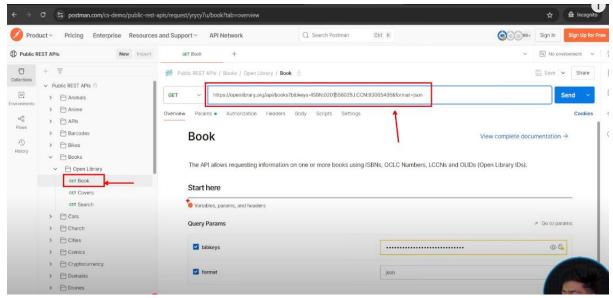
Your company receives daily exchange rate data from a third-party financial service through a REST API. This API returns data in JSON format and requires an authentication token that is refreshed every 60 minutes.



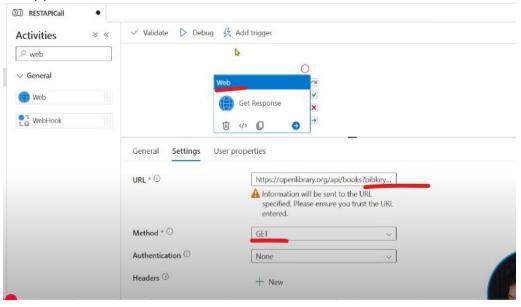
How would you design and implement this solution using Azure Data Factory? Walk through the key components and steps you would use.



Sample API from Postman



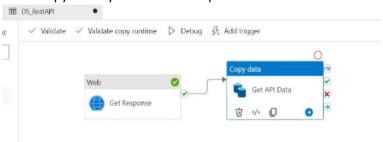
Query parameters: used for authentication of API



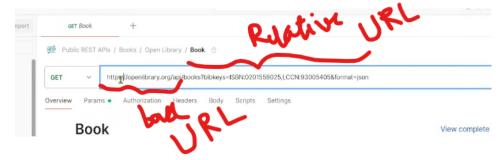
Output Copy to clipboard "ISBN:0201558025": { "bib_key;" "ISBN:0201558025", "info_url": "https://openlibrary.org/books/OL1429049M/Concrete_mathematics ", "preview": "full", "preview_url": " https://archive.org/details/concretemathemat00grah_444 ", "thumbnail_url": " https://covers.openlibrary.org/b/id/135182-S.jpg " "LCCN:93005405": { "bib_key": "LCCN:93005405", "info_url": "https://openlibrary.org/books/OL1397864M/Zen_speaks ", "preview": "borrow", "preview_url": " https://archive.org/details/zenspeaksshoutso0000caiz ", "thumbnail_url": " https://covers.openlibrary.org/b/id/240726-S.jpg " "ADFWebActivityResponseHeaders": { "Transfer-Encoding": "chunked", "Connection": "keep-alive", "access-control-allow-origin": "^", "access-control-allow-method": "GET, OPTIONS", "access-control-max-age": "86400", "x-ol-stats": "\"IB 2 0.070 MC 3 0.004 TT 0 0.077\"**, "Referrer-Policy": "no-referrer-when-downgrade", "Date": "Mon, 19 May 2025 17:03:37 GMT", "Server": "nginx/1.28.0", "Content-Type": "application/json"

Debug to see output

Use a Copy activity to store the response

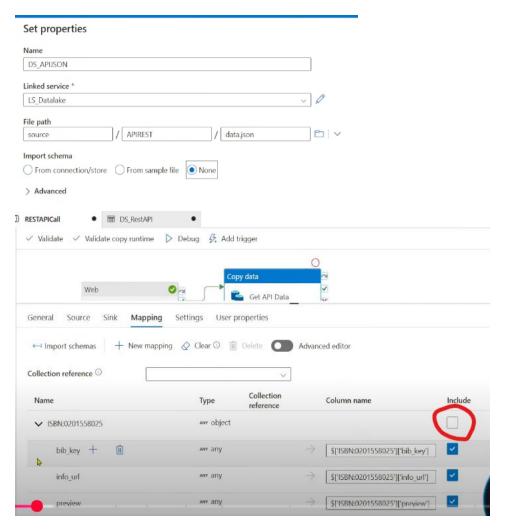


Source → REST dataset with REST LS

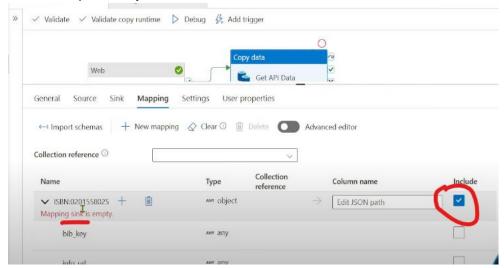


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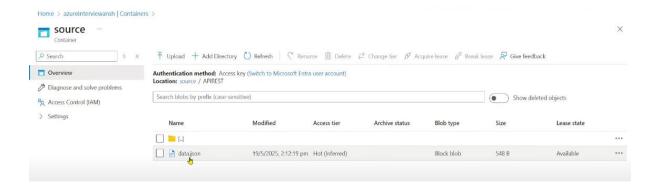
Sink → ADLS in JSON format



Include the parent object



Now click on Import Schemas again → Now debug the pipeline



Additional: Now let's consider if the output of Web activity is in csv format(not in json format)
Always use HTTP connection instead of REST in case of Delimited text

Because HTTP has an option of choosing Format (REST doesnot have)

SCENARIO

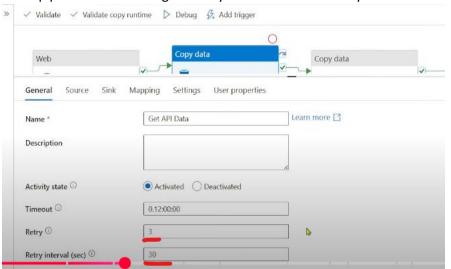
You're working on an ELT pipeline in a cloud-based data platform (like Azure Data Factory, AWS Glue, or Databricks). One of the pipeline activities is an API call to fetch data from a third-party service. This API is known to occasionally fail due to temporary network glitches or rate limiting issues.

QUESTION

How would you design this pipeline activity to handle intermittent failures gracefully without rerunning the entire pipeline?



Make pipeline robust enough to retry from the failed activity





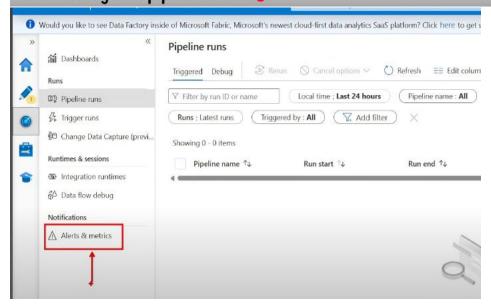


You are working as a Data Engineer in a retail company. The company relies heavily on daily ETL pipelines built in Azure Data Factory (ADF) to load sales data from various regions into a central data warehouse. Recently, a few pipeline failures went unnoticed, which caused delays in business reporting and decision-making.

QUESTION

How would you design and implement a monitoring and alerting mechanism in ADF to notify the team immediately in case of any ETL pipeline failure





Setup Alerts & metrics

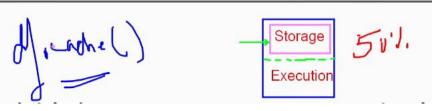
You're working on a PySpark job that processes customer transaction data from multiple sources. The initial steps involve heavy transformations like joins, filters, and aggregations, and the intermediate result is reused in multiple parts of the pipeline—first for generating KPIs, then for writing different outputs to Delta tables, and finally for some visualizations.

QUESTION

Given that the same intermediate DataFrame is reused in multiple stages, how would you optimize performance in this situation?



Whenever we create a dataframe, executor doesn't hold the dataframe in the execution memory for long time, it will eliminate dataframe immediately if it is assigned a different task or need to create a new df from an existing df.



Use cache() or persist()

Execution memory will be cleaned everytime there is a new job, so it will follow the DAG and start from the first step, so its better to store in storage memory.

Make sure to use unpersist() post completion.

SCENARIO

You're working on an Azure Data Factory (ADF) pipeline that involves calling a REST API through a Web Activity. This API requires an access key for authentication.

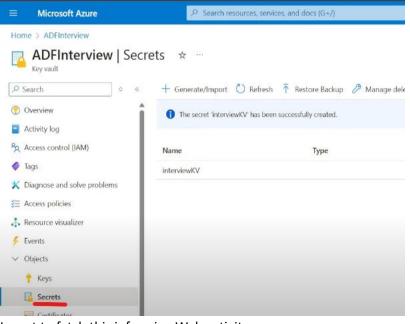
To follow best practices, your team has decided to store the API key securely using Azure Key Vault instead of hardcoding it in the pipeline.

QUESTION

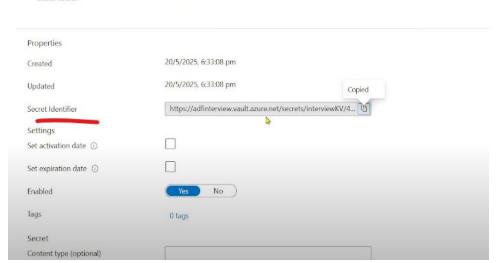
How would you implement the above scenario end-to-end in Azure Data Factory? Walk me through each step you would take, from creating the Key Vault to using its secret in the Web Activity, and ensuring secure handling of the output.



Create AKV → Grant KV Administrator role → create secret

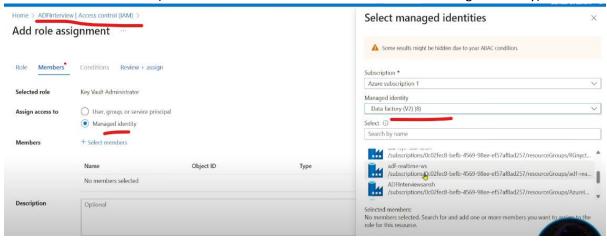


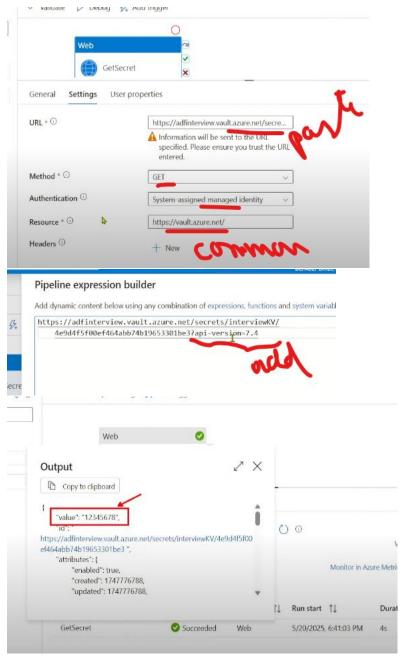
I want to fetch this info using Web activity



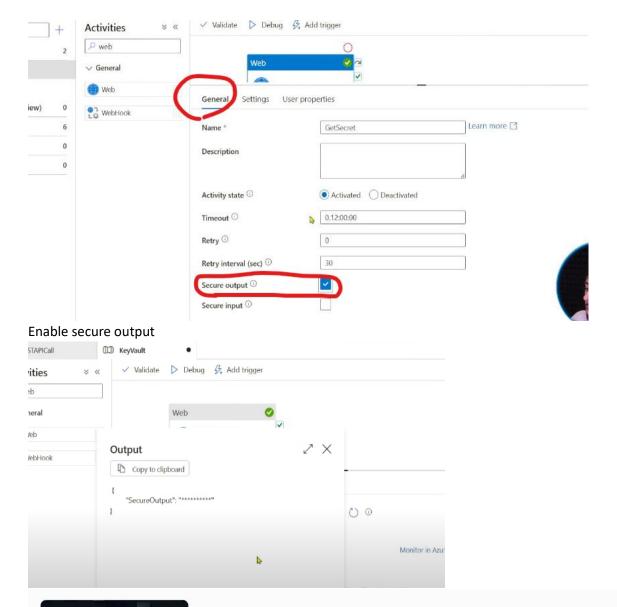
Copy the secret identifier

Grant ADF access to AKV (Go to AKV → IAM → KV Administrator role → Managed Identity)





We can see the value, ideally we should not



The company receives daily sales data files from multiple regional branches. These files are uploaded at random times throughout the day to a specific folder in an Azure Data Lake Storage Gen2 account.

Your goal is to ensure that as soon as a new file arrives in this folder, an Azure Data Factory (ADF) pipeline is triggered automatically

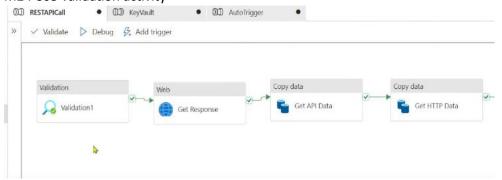
QUESTION

How would you design a solution in Azure Data Factory (or Synapse) to automatically trigger the pipeline whenever a new file is added to the specified folder in Azure Data Lake Storage Gen2?



M1: Use Storage Events Trigger

M2: Use Validation activity



Validation activity will be continuously searching for the file and then run the subsequent steps

SCENARIO

You're working as a data engineer for a retail company that stores large volumes of historical customer transaction data in an Azure Data Lake Storage Gen2 in CSV format. The business wants to run ad-hoc SQL queries on this data from Azure Synapse Analytics without copying it into the Synapse dedicated SQL pool.

QUESTION

How would you use PolyBase in Azure Synapse Analytics to query this external data stored in Data Lake? Please explain the steps involved.



In serverless SQL pool, we use Polybase

Step 1: Create Master key for database

Step 2: Create credential

Step 3: Create External Data Source

Step 4: Create External File Format

Step 5: Create External Table

You are responsible for designing a Synapse Pipeline to incrementally load data from the Azure SQL source into the Data Lake using Parquet format, without loading the entire dataset each time.

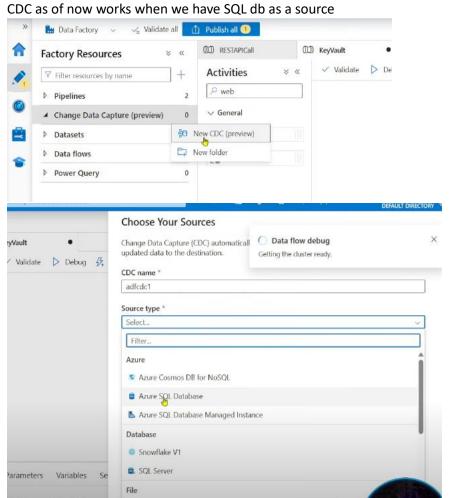
The company also wants the solution to be efficient and cost-effective.

QUESTION

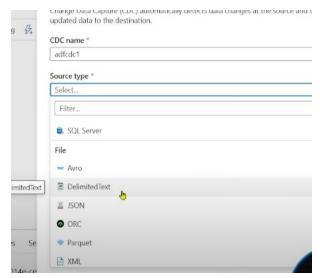
How would you implement incremental loading in Synapse Pipelines in this scenario?

1. Watermarking approach



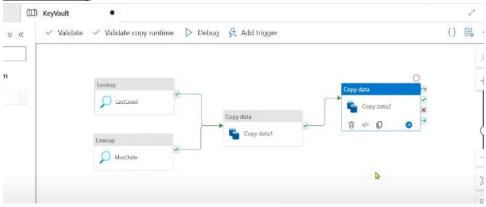






Enable CDC in the DB Enable CDC in the Table Connect the Table

Watermarking



Instead of Stored Procedure, store the data in file

You're working on a data transformation job in Spark where large datasets from different sources are being joined and aggregated. The data volume varies significantly based on the day and data source. Initially, you tuned the join strategy and number of partitions manually, but you observed that the job still fails sometimes due to skewed data or inefficient joins. You enabled broadcast joins, but it's hard to predict the right size threshold.

QUESTION

How can Adaptive Query Execution (AQE) help in this scenario?

What specific features of AQE would address the issues mentioned, and how would you enable them in Spark?

AQE : calculates query statistics during runtime based on that it addresses the manual threshold for broadcast join

(Before AQE, LP→PP) With AQE, LP→PP→Query statistics Addresses skewness, Dynamically coalesces the partitions



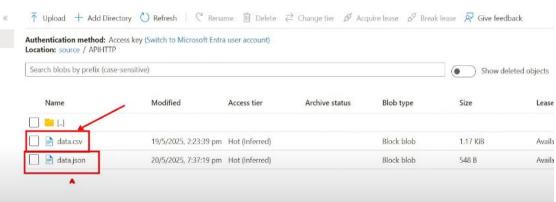


You are working on an Azure Data Factory pipeline that triggers when a new file is uploaded to a Data Lake container. Each file needs to go through a different transformation activity. You are required to design the pipeline such that it automatically detects the file name and executes the appropriate activity, without running all activities for every file.

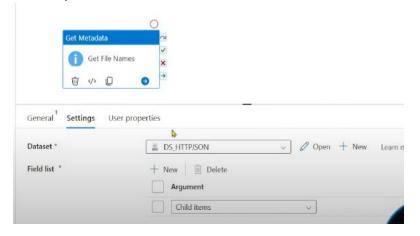
QUESTION

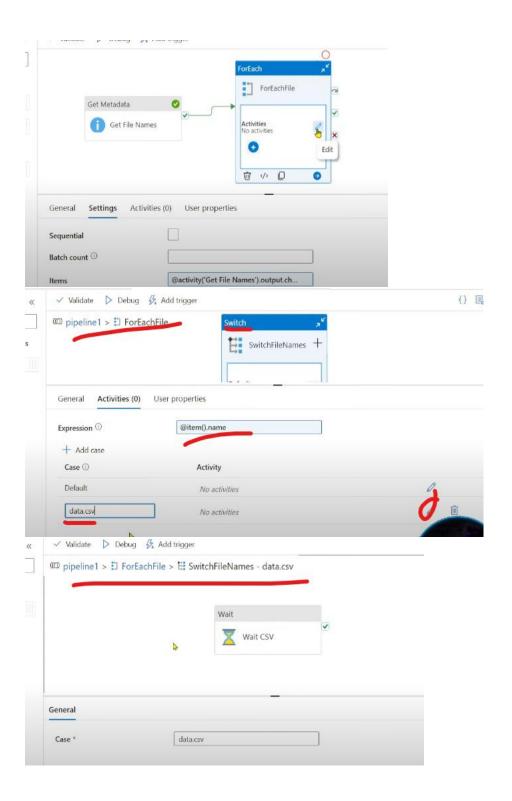
How would you design this pipeline to ensure that only the correct activity runs for each specific file, based on its name?

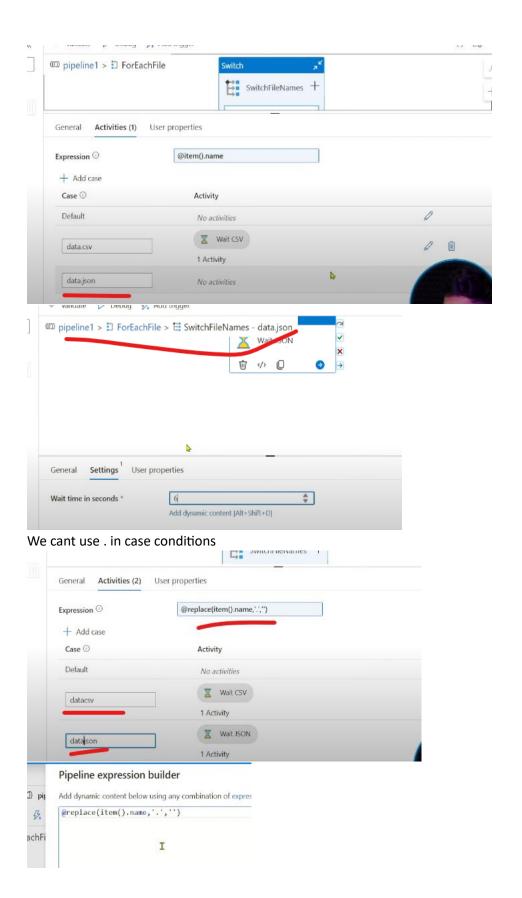


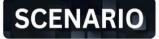


Want to perform different sets of activities for each of these files











You are expected to build a dynamic solution that reads the schema and maps source columns to destination columns automatically at runtime.

QUESTION

How would you implement dynamic column mapping in Synapse Pipelines to handle this scenario?

Explain your approach in detail, including:



Copy the translator code

SCENARIO

You need to process 10GB of data using Spark. How many Executors you would need, and how much memory you would need for each Executor to get the maximum parallelism? Also, how many cores should be there?

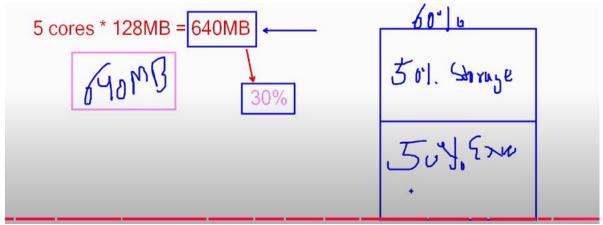
S1: 10GB → 80 partitions

S2: Leverage max parallelism → 80 partitions → 80 cores

S3: No of cores \rightarrow 5 per executor \rightarrow 80/5 \rightarrow 16 executor

S4: Memory for each Executor \rightarrow In each executor 5 cores, each core will process 1 partition sized 128MB i.e. 128*5 = 640MB

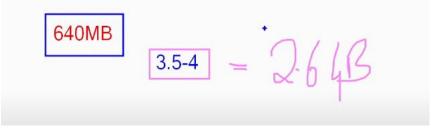
640MB is not the total memory, 640MB is required just to execute the task, we get 30% memory to execute the task.



Spark pool memory is 60%, 40% off heap.

Out of 60%, 50% is storage memory/caching & 50% is execution memory.

So 640MB is equivalent to 50% execution memory. So we multiply 640 with 3.5-4 to get the total executor memory



Total Executor memory = 2.6 GB