

# AZURE DATA ENGINEERING INTERVIEW QUESTIONS PART-1



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# 1. Reading and Processing 10 Million CSV Files in ADLS Gen2 using Azure Data Factory

To handle a large number of CSV files in ADLS Gen2 using Azure Data Factory (ADF):

- Step 1: Use Wildcard File Path
   Use ADF's Copy Data or Data Flow activity to define a wildcard pattern to match the files in the ADLS Gen2 container. This avoids the need to list each file explicitly.
- Step 2: Batch Processing
   Enable File Path Partitioning to process files in parallel. Use the maxConcurrency setting in activities like Copy Data to increase throughput.
- Step 3: Optimize Data Flow
   Use Mapping Data Flow to transform and aggregate data as needed. Leverage partitioning options in Data Flow for parallel processing.
- Step 4: Iterative Processing with ForEach Activity
   If each file requires specific processing, use the Get Metadata activity to retrieve file names, and iterate over them with the ForEach activity.
- Step 5: Performance and Cost Optimization
  - Enable compression on CSVs to reduce size.
  - Use polybase or external tables if the destination is Azure Synapse Analytics.
  - Use Integration Runtime (IR) to scale the compute environment.

#### 2. What is Integration Runtime?

Integration Runtime (IR) is the compute infrastructure used by Azure Data Factory to provide the data integration capabilities. Types of IR:

- Azure IR: For cloud-based activities like copying data between cloud sources.
- Self-Hosted IR: For accessing on-premises or private network data securely.
- Azure-SSIS IR: For executing SQL Server Integration Services (SSIS) packages.

#### 3. Variables and Parameters in ADF

#### Variables:

- Used to store temporary values during the pipeline run.
- Scope: Pipeline-level.
- Use: Modify values dynamically during execution using Set
   Variable activity.

#### Parameters:

- Used to pass values into a pipeline or data flow at runtime.
- Scope: Read-only and defined at pipeline or dataset levels.
- Use: Customize pipeline runs without modifying the pipeline.

#### 4. Activities in ADF

**Activities** perform tasks in a pipeline. Common types:

- Copy Data Activity: Copies data between sources/destinations.
- Data Flow Activity: Enables data transformation at scale.
- Lookup Activity: Retrieves data from a source (used for condition-based workflows).
- ForEach Activity: Iterates through a collection of items.
- Filter Activity: Filters data based on conditions.
- Set Variable/Append Variable Activity: Updates variable values.
- Execute Pipeline Activity: Executes a child pipeline.
- Web Activity: Calls a REST endpoint.
- Delete Activity: Deletes files or datasets.
- Wait Activity: Adds a delay to the pipeline.

# 5. Automating Failure Notifications

To automate email notifications on pipeline failure:

- Add a Failure dependency in the pipeline using the OnFailure trigger.
- Use the Web Activity to call a Logic App or Azure Function that sends an email using SendGrid or Office 365 SMTP.
- Alternatively, use Azure Monitor Alerts to monitor pipeline status and send emails on failure.

#### 6. Handling Exceptions in ADF

- Use Error Handling Activities like:
  - Try-Catch blocks with OnFailure dependencies.
  - Execute Pipeline for reattempt or fallback operations.
- Custom Logging: Log errors using Azure Log Analytics or store them in a database.
- Retry Policies: Configure retries on activities in the activity settings.

#### 7. Fixing Slow ADF Pipeline

- Analyze Bottlenecks:
  - Use ADF monitoring tools to identify slow activities.
  - Optimize data movement by using proper partitioning.
- Increase Parallelism:
  - Increase maxConcurrency in activities.
  - Optimize partitioning in Data Flows.
- Use High-Performance Resources:
  - Use Premium Integration Runtime for intensive operations.
  - Upgrade the target database (e.g., Azure Synapse) if it is the bottleneck.
- Optimize Source/Destination:
  - Enable compression and indexing.

 Use incremental loading or delta processing for large datasets.

# 8. Blob Storage vs. ADLS Gen2

Feature	Blob Storage	ADLS Gen2
Hierarchy	Flat Namespace	Hierarchical Namespace
Performance	Suitable for general workloads	Optimized for analytics workloads
Security	Role-based access	POSIX-like ACLs, more granular
Integration	Good for general use	Better for Big Data and analytics

### Why ADLS Gen2 is Required?

- Supports big data analytics scenarios like processing petabytes of data.
- Offers better performance due to hierarchical namespace.
- Allows granular security controls with ACLs.
- Optimized for integration with Azure Synapse and Data Lake Analytics.

# 9. Connecting ADLS Gen2 with Databricks

To connect ADLS Gen2 with Databricks:

1. **Use Azure Active Directory (AAD)**: Authenticate Databricks to access ADLS Gen2 via a Service Principal or Managed Identity.

#### 2. Steps:

- Assign Storage Blob Data Contributor or Storage Blob
   Data Owner role to the Service Principal or Databricks
   workspace managed identity in the Azure portal for the ADLS Gen2 storage account.
- Configure access in Databricks by adding the credentials (e.g., OAuth token or client secret).

Role assignments are configured in **Azure Portal > Storage Account > Access Control (IAM)**.

# 10. Using Service Principal to Connect ADLS from Databricks Steps to connect using Service Principal:

# 1. Create a Service Principal:

- Register an app in Azure AD.
- Generate a client secret.

# 2. Assign Roles:

Assign Storage Blob Data Contributor to the Service
 Principal for the ADLS Gen2 account.

# 3. Configure Databricks:

 Add Service Principal details as secret scopes in Databricks.

# Code Example:

```
# Set up configurations
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": "<client_id>",
    "fs.azure.account.oauth2.client.secret": "<client_secret>",
    "fs.azure.account.oauth2.client.endpoint": "https://login
        .microsoftonline.com/<tenant_id>/oauth2/token"
}
# Mount ADLS Gen2
dbutils.fs.mount(
    source="abfss://<container>@<storage_account_name>.dfs.core
        .windows.net/",
   mount_point="/mnt/<mount_name>",
   extra_configs=configs
)
```

## 11. Why Use Service Principal? How to Create It?

## Why Use Service Principal?

- Provides a secure way to authenticate Databricks without using user credentials.
- Supports automation and application-level authentication.
- Enables fine-grained access control with role assignments.

# **Steps to Create Service Principal:**

- 1. Go to Azure Active Directory > App Registrations.
- 2. Click **New Registration**, name the app, and register it.
- Generate a client secret under Certificates & Secrets.

4. Assign the Service Principal a role under **Access Control (IAM)** for the target resource.

#### 12. What is Databricks Runtime?

The **Databricks Runtime** is a pre-configured environment with optimized libraries for Apache Spark, Delta Lake, and other big data analytics tools.

#### Why We Need It:

- Offers optimized performance and scalability.
- Ensures compatibility with Spark APIs and machine learning libraries.
- Provides ready-to-use integrations with Azure services.

#### 13. What are Workflows?

**Workflows** in Databricks allow you to create, schedule, and monitor pipelines of jobs.

- A workflow can chain multiple jobs together with dependencies.
- Supports retry policies and alerts for monitoring.
- Enables triggering with APIs or schedules.

#### 14. Medallion Architecture

Medallion Architecture organizes data in three layers:

- 1. **Bronze Layer**: Raw, unprocessed data stored in a data lake.
- 2. **Silver Layer**: Cleaned and enriched data.

3. **Gold Layer**: Aggregated, analytics-ready data.

**Benefits**: Provides structured, incremental processing for large datasets.

#### 15. Delta File Format

Delta Lake is an open-source storage format built on top of Parquet.

#### Features:

- ACID transactions.
- Schema evolution.
- Time travel (historical queries).
- Scalable performance with optimized reads and writes.

# 16. Why Delta File Format in High Write Scenarios?

Delta format is essential for scenarios like Facebook's transactional logs:

- Handles **small file problems** by merging them into larger files during optimization.
- Provides ACID guarantees, ensuring consistency.
- Includes **data compaction** (optimize command), reducing the impact on performance.
- Enables efficient querying with **z-order clustering** and caching.

#### 17. Debugging a Slow Job in Databricks

Steps to address slow jobs:

- 1. Check Spark UI: Identify slow stages or tasks.
- 2. **Skewed Data**: Use partitioning and bucketing to balance the load.
- 3. **Cluster Configuration**: Use autoscaling clusters or increase node size.
- 4. Optimize Storage: Use Delta Lake and compact files.
- 5. **Caching**: Cache intermediate results to avoid re-computation.

# **18. Optimization Techniques**

- 1. **Storage Optimization**: Delta format, partitioning, and file compaction.
- 2. Query Optimization:
  - Use predicate pushdown.
  - Optimize joins with broadcast hints.
- 3. Cluster Tuning:
  - Adjust executor and driver memory.
  - Use autoscaling.
- 4. **Data Skew Handling**: Partition data by key.
- 5. **Pipeline Optimization**: Use caching, avoid shuffles, and tune Spark configurations.

# 19. Memory Management Optimization

- 1. **Memory Allocation**: Increase executor memory (spark.executor.memory) and configure off-heap memory if needed.
- 2. **Broadcast Variables**: Use for small lookup tables to reduce shuffle operations.
- 3. **Garbage Collection**: Tune JVM GC settings for optimal performance.
- 4. Cache Management:
  - Persist frequently accessed datasets using persist() or cache().
  - Release unused cached data.
- 5. **Shuffle Optimization**: Reduce shuffles by proper partitioning and using repartition() or coalesce().