DP-203

Design and implement data storage (15–20%)

- Implement a partition strategy
 - Implement a partition strategy for files
 - o Implement a partition strategy for analytical workloads
 - o Implement a partition strategy for streaming workloads
 - Implement a partition strategy for Azure Synapse Analytics
 - o Identify when partitioning is needed in Azure Data Lake Storage Gen2
- Design and implement the data exploration layer
 - <u>Create and execute queries by using a compute solution that leverages SQL serverless and Spark cluster</u>
 - o Implement Azure Synapse Analytics database templates
 - o Recommend Azure Synapse Analytics database templates
 - Push new or updated data lineage to Microsoft Purview
 - o Browse and search metadata in Microsoft Purview Data Catalog

Develop data processing (40–45%)

- Ingest and transform data
 - Design and implement incremental loads
 - Transform data by using Apache Spark
 - o Transform data by using Transact-SQL (T-SQL)
 - Ingest and transform data by using <u>Azure Synapse Pipelines</u> or <u>Azure Data</u>
 <u>Factory</u>
 - Transform data by using Azure Stream Analytics
 - o Cleanse data
 - o Handle duplicate data
 - o Handle missing data
 - Handle late-arriving data
 - Split data
 - o Shred JSON
 - o Encode and decode data
 - Configure error handling for a transformation
 - Normalize and denormalize values
 - Perform data exploratory analysis
- Develop a batch processing solution
 - Develop batch processing solutions by using <u>Azure Data Lake Storage</u>, <u>Azure Data Lake Storage</u>, <u>Azure Data Factory</u>

- Use PolyBase to load data to a SQL pool
- Implement Azure Synapse Link and query the replicated data
- o Create data pipelines
- Scale resources
- o Configure the batch size
- Create tests for data pipelines
- o Integrate Jupyter or Python notebooks into a data pipeline
- o Upsert data
- o Revert data to a previous state
- o Configure exception handling
- Configure batch retention
- Read from and write to a delta lake
- Develop a stream processing solution
 - Create a stream processing solution by using <u>Stream Analytics</u> and <u>Azure Event</u> Hubs
 - Process data by using Spark structured streaming
 - o Create windowed aggregates
 - o Handle schema drift
 - o Process time series data
 - o Process data across partitions
 - Process within one partition
 - o Configure checkpoints and watermarking during processing
 - o Scale resources
 - Create tests for data pipelines
 - Optimize pipelines for analytical or transactional purposes
 - Handle interruptions
 - o Configure exception handling
 - o Upsert data
 - Replay archived stream data
- Manage batches and pipelines
 - Trigger batches
 - Handle failed batch loads
 - Validate batch loads
 - Manage data pipelines in Azure Data Factory or Azure Synapse Pipelines.
 - Schedule data pipelines in Data Factory or Azure Synapse Pipelines
 - Implement version control for pipeline artifacts
 - Manage Spark jobs in a pipeline

Secure, monitor, and optimize data storage and data processing (30–35%)

Implement data security

- Implement data masking
- Encrypt data at rest and in motion
- o Implement_row-level and column-level security
- Implement Azure role-based access control (RBAC)
- Implement POSIX-like access control lists (ACLs) for Data Lake Storage Gen2
- Implement a data retention policy
- Implement secure endpoints (private and public)
- o <u>Implement resource tokens in Azure Databricks</u>
- Load a DataFrame with sensitive information
- Write encrypted data to tables or <u>Parquet files</u>
- Manage sensitive information

Monitor data storage and data processing

- Implement logging used by <u>Azure Monitor</u>
- Configure monitoring services
- Monitor stream processing
- Measure performance of data movement
- o Monitor and update statistics about data across a system
- Monitor data pipeline performance
- Measure query performance
- Schedule and monitor pipeline tests
- Interpret Azure Monitor metrics and logs
- o Implement a pipeline alert strategy

Optimize and troubleshoot data storage and data processing

- Compact small files
- o Handle skew in data
- Handle data spill
- o Optimize resource management
- o Tune queries by using indexers
- Tune queries by using cache
- Troubleshoot a failed Spark job
- Troubleshoot a failed pipeline run, including activities executed in external services