

# AWS + Python + PySpark Real-time Scenarios for Data Engineers

## 1. Simple S3 → Transform → S3 Pipeline

**Skill Focus:** Basic PySpark transformations, AWS S3 integration, Python automation.

**Scenario Steps:**

- Company stores raw CSV sales data in `s3://company-raw/sales/YYYY/MM/DD/`
- Read raw data from S3 using PySpark.
- Perform basic cleaning (remove nulls, fix date formats).
- Write cleaned data back to `s3://company-processed/sales/` in Parquet format.

**Services & Tools:** AWS: S3 | Processing: PySpark (local or EMR/Glue) | Automation: Python script or Glue job

## 2. Log Processing from S3 to Redshift

**Skill Focus:** Data modeling, incremental loads, Python AWS SDK (boto3).

**Scenario Steps:**

- Application logs uploaded daily to S3.
- Parse logs in PySpark to extract timestamp, user\_id, action.
- Load transformed data into Amazon Redshift fact tables.
- Automate job daily with AWS Lambda or Glue Workflow.

**Services & Tools:** AWS: S3, Redshift, Glue, Lambda | Processing: PySpark | Automation: Python boto3

## 3. Real-time Stream Processing from Kinesis

**Skill Focus:** Streaming data processing, window functions in PySpark.

**Scenario Steps:**

- E-commerce site streams user click events to Amazon Kinesis Data Streams.
- Consume stream in Spark Structured Streaming.
- Aggregate clicks by user\_id in a 5-minute sliding window.
- Store aggregated results in S3.

**Services & Tools:** AWS: Kinesis Data Streams, S3, Glue Catalog | Processing: PySpark Structured Streaming

## 4. Data Lake with Partitioning & Glue Catalog

**Skill Focus:** Partitioning, schema evolution, query optimization.

**Scenario Steps:**

- Sensor readings in JSON format in S3.
- Read and clean data in PySpark.
- Save in partitioned Parquet format by year/month/day.
- Create/update Glue Data Catalog table for Athena.

**Services & Tools:** AWS: S3, Glue Data Catalog, Athena | Processing: PySpark

## 5. Data Validation & Quality Checks

**Skill Focus:** Data quality frameworks (Great Expectations), Python validation scripts.

**Scenario Steps:**

- Validate marketing campaign data before loading to Redshift.

- No nulls in campaign\_id, valid dates, spend > 0.
- Write pass/fail reports to S3.
- Send email via AWS SES if validation fails.

**Services & Tools:** AWS: S3, SES, Lambda | Processing: PySpark, Great Expectations (optional)

## 6. Incremental ETL from RDS to S3

**Skill Focus:** Change Data Capture (CDC), scheduling, incremental loads.

**Scenario Steps:**

- PostgreSQL RDS stores order transactions.
- Pull only new orders using last\_updated timestamp.
- Append to S3 in Parquet format.
- Run daily via Glue Job or Airflow.

**Services & Tools:** AWS: RDS, S3, Glue | Processing: Python (pandas + boto3) or PySpark

## 7. Machine Learning Data Prep

**Skill Focus:** Feature engineering in PySpark for ML models.

**Scenario Steps:**

- Join purchase history with user demographics.
- Create aggregate features: total spend, last purchase date, categories purchased.
- Save as feature store in S3 for ML training.

**Services & Tools:** AWS: S3, SageMaker (optional) | Processing: PySpark

## 8. IoT Data Pipeline

**Skill Focus:** Handling high-volume data, compression, time-series analysis.

**Scenario Steps:**

- IoT sensors send temperature data to Kinesis Firehose → S3.
- Transform raw JSON to time-series friendly schema.
- Store compressed Parquet for cost savings.
- Generate daily summaries for anomalies.

**Services & Tools:** AWS: Kinesis Firehose, S3, Glue Catalog, Athena | Processing: PySpark  
Structured Streaming