

code snippets for each step:

**1. Ingesting the Stocks JSON File and Flattening:**

- Read the stocks.json file into a DataFrame.
- Flatten the nested JSON structure to create a normalized DataFrame.

**2. Ingesting Clients.csv and Collaterals.csv:**

- Read the Clients.csv and Collaterals.csv files into DataFrames.

**3. Joining Data and Calculating Collateral Fluctuation:**

- Combine the data from the three DataFrames (stocks, clients, collaterals).
- Calculate the market value of each collateral based on stock prices.
- Determine the fluctuation in collateral value over time.

**4. Creating and Saving the Collateral Status Table:**

- Create a target table called collateral\_status.
- Save the resulting table to an appropriate storage location.

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Below is a sample PySpark code snippet to accomplish these tasks:

**PySpark**

```
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, explode_outer

# Initialize Spark session
spark =
SparkSession.builder.appName("CollateralStatusPipeline").getOrCreate
()

# Step 1: Read stocks.json and flatten the nested structure
stocks_df = spark.read.json("path/to/stocks.json")
flattened_stocks_df = stocks_df.selectExpr("stock_symbol",
"stock_price")

# Step 2: Read Clients.csv and Collaterals.csv
clients_df = spark.read.csv("path/to/Clients.csv", header=True)
collaterals_df = spark.read.csv("path/to/Collaterals.csv",
header=True)
```

```

# Step 3: Join data and calculate collateral fluctuation

combined_df = collaterals_df.join(clients_df, "client_id",
"inner").join(flattened_stocks_df, "stock_symbol", "left")

combined_df = combined_df.withColumn("market_value",
combined_df["quantity"] * combined_df["stock_price"])

combined_df = combined_df.withColumn("fluctuation",
combined_df["market_value"] - combined_df["initial_value"])

# Step 4: Create and save collateral_status table

collateral_status_df = combined_df.select("client_id",
"collateral_id", "fluctuation")

collateral_status_df.write.mode("overwrite").parquet("path/to/collat
eral_status")

# Show the resulting table (optional)

collateral_status_df.show()

# Stop Spark session

spark.stop()

```

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Further Let's enhance the code snippet to use Parquet and Delta Lake formats for ingesting data and creating the collateral\_status table. We'll also generate our own sample data for demonstration purposes.

#### 1. Generate Sample Data:

- For simplicity, let's create some sample data for two clients with their collateral information and stock prices over a week.
- Assume we have the following data:
  - Client 1:

Collateral ID	Initial Collateral Value	Stock Symbol	Stock Price (Day 1)	Stock Price (Day 7)
101	\$10,000	AAPL	\$150	\$160

- Client 2:

Collateral ID	Initial Collateral Value	Stock Symbol	Stock Price (Day 1)	Stock Price (Day 7)
201	\$20,000	MSFT	\$200	\$210

## 2. Modify the Code:

- We'll read the data from Parquet files for clients and collaterals.
- Calculate the fluctuation in collateral value based on stock prices.
- Save the resulting table as a Delta table.

Below is the modified PySpark code snippet:

## PySpark

```
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, explode_outer

# Initialize Spark session
spark =
SparkSession.builder.appName("CollateralStatusPipeline").getOrCreate
()

# Read clients and collaterals data from Parquet files
clients_df = spark.read.parquet("path/to/Clients.parquet")
collaterals_df = spark.read.parquet("path/to/Collaterals.parquet")

# Assume stock prices for AAPL and MSFT over a week
stock_prices = [("AAPL", 150, 160), ("MSFT", 200, 210)]
stock_prices_df = spark.createDataFrame(stock_prices,
["stock_symbol", "price_day1", "price_day7"])

# Join data and calculate collateral fluctuation
combined_df = collaterals_df.join(clients_df, "client_id",
"inner").join(stock_prices_df, "stock_symbol", "left")
combined_df = combined_df.withColumn("market_value_day1",
combined_df["quantity"] * combined_df["price_day1"])
```

```
combined_df = combined_df.withColumn("market_value_day7",
combined_df["quantity"] * combined_df["price_day7"])

combined_df = combined_df.withColumn("fluctuation",
combined_df["market_value_day7"] - combined_df["initial_value"])

# Create and save collateral_status Delta table
combined_df.select("client_id", "collateral_id",
"fluctuation").write.mode("overwrite").format("delta").save("path/to
/collateral_status")

print("Collateral status table created and saved successfully!")

spark.stop()
```