Case Study

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Load CSV file

Create a pandas DataFrame from the dataset for easier processing and visualization. Replace the file path below with your actual file location.

import numpy as np

import pandas as pd

df = spark.table('hive_metastore.default.employees').toPandas()

```
07:18 PM (29s)
                                                               Python
                                                                       import numpy as np
   import pandas as pd
   df = spark.table('hive_metastore.default.employees').toPandas()
▶ (1) Spark Jobs
▼ 🗐 df: pandas.core.frame.DataFrame
       EMPLOYEE_ID: int64
       FIRST_NAME: object
       LAST_NAME: object
       EMAIL: object
       PHONE_NUMBER: object
       HIRE_DATE: object
       JOB_ID: object
       SALARY: int64
       COMMISSION_PCT: object
       MANAGER_ID: object
       DEPARTMENT_ID: int64
```

Use pandas for data insights

Check shape of the DataFrame

df.shape

The df.shape command returns the number of rows and columns.



Generate descriptive statistics

df.describe()

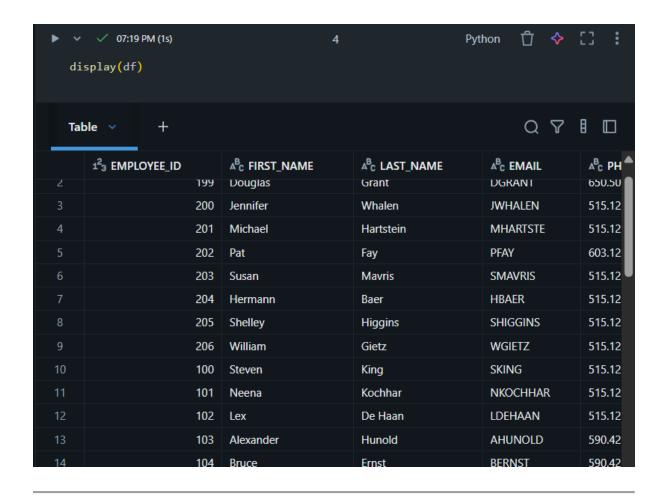
The df.describe() command provides numerical summaries such as mean, standard deviation, min, max, and percentiles.



Generate a data profile

display(df)

This displays a full interactive table of the data for profiling.



Clean the data

Remove duplicate data

```
duplicate_rows = df.duplicated().sum()
duplicate_columns = df.columns[df.columns.duplicated()].tolist()
print("Duplicate rows count:", duplicate_rows)
print("Duplicate columns:", duplicate_columns)
```

```
duplicate_rows = df.duplicated().sum()
  duplicate_columns = df.columns[df.columns.duplicated()].tolist()
  print("Duplicate rows count:", duplicate_rows)
  print("Duplicate columns:", duplicate_columns)
Duplicate rows count: 0
Duplicate columns: []
```

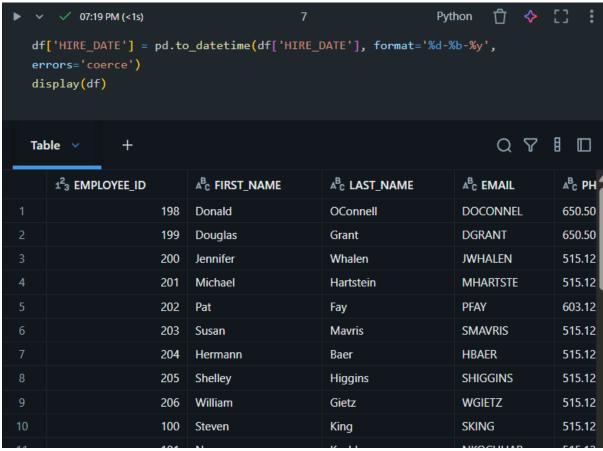
```
df = df.replace('-', np.nan)
```

df = df.fillna(0)

Reformat dates

df['HIRE_DATE'] = pd.to_datetime(df['HIRE_DATE'], format='%d-%b-%y', errors='coerce')

display(df)



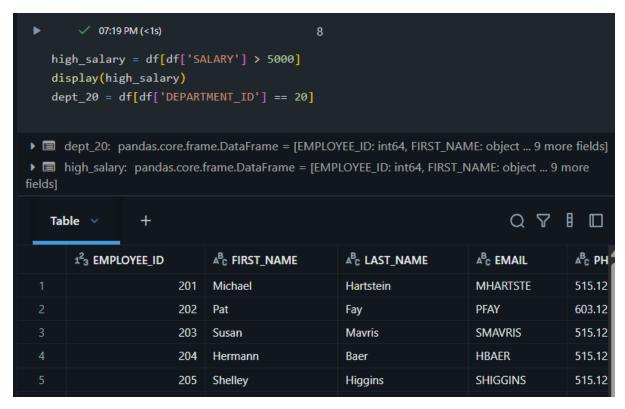
Filter for specific conditions

Employees with salary > 5000

```
high_salary = df[df['SALARY'] > 5000]
```

Employees in department 20

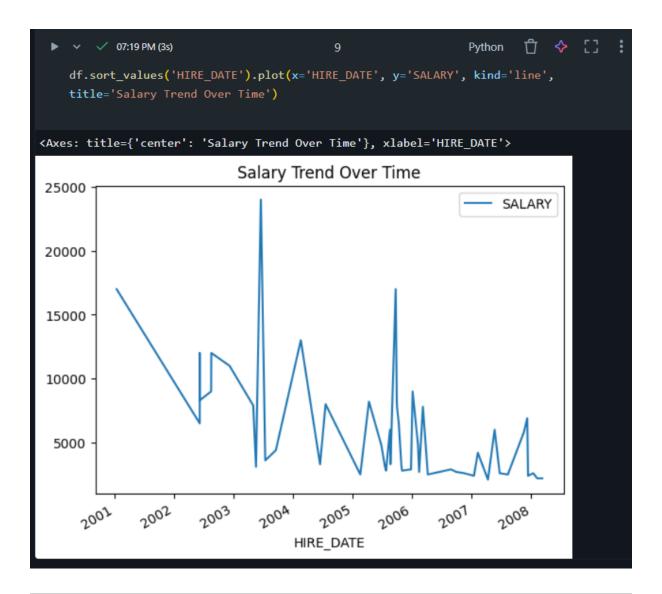
dept_20 = df[df['DEPARTMENT_ID'] == 20]



Create visualizations using the dataset

Line Chart

df.sort_values('HIRE_DATE').plot(x='HIRE_DATE', y='SALARY', kind='line', title='Salary Trend Over Time')



Area Chart

df.groupby('DEPARTMENT_ID')['SALARY'].sum().plot(kind='area', title='Total Salary by Department')



Delta Table Operations

Create a Delta Table

Convert pandas DataFrame to Spark DataFrame

spark_df = spark.createDataFrame(df)

Write as Delta table

spark_df.write.format("delta").mode("overwrite").saveAsTable("hive_metastore.default.employee_d elta")

Merge & Upsert to a Delta Table

from delta.tables import DeltaTable

from pyspark.sql.types import StructType, StructField, IntegerType, StringType, DoubleType, DateType

from datetime import datetime

```
# Define schema explicitly

schema = StructType([

StructField("EMPLOYEE_ID", IntegerType(), False),

StructField("FIRST_NAME", StringType(), True),

StructField("LAST_NAME", StringType(), True),

StructField("EMAIL", StringType(), True),

StructField("PHONE_NUMBER", StringType(), True),

StructField("HIRE_DATE", DateType(), True),

StructField("JOB_ID", StringType(), True),

StructField("SALARY", DoubleType(), True),

StructField("COMMISSION_PCT", DoubleType(), True),

StructField("MANAGER_ID", IntegerType(), True),

StructField("DEPARTMENT_ID", IntegerType(), True)
```

Create source DataFrame with explicit schema

```
from delta.tables import DeltaTable
from pyspark.sql.types import StructType, StructField, IntegerType, StringType,
DoubleType, DateType
from datetime import datetime
schema = StructType([
    StructField("EMPLOYEE_ID", IntegerType(), False),
    StructField("FIRST_NAME", StringType(), True),
    StructField("LAST_NAME", StringType(), True),
    StructField("EMAIL", StringType(), True),
    StructField("PHONE NUMBER", StringType(), True),
    StructField("HIRE_DATE", DateType(), True),
    StructField("JOB_ID", StringType(), True),
    StructField("SALARY", DoubleType(), True),
    StructField("COMMISSION_PCT", DoubleType(), True),
    StructField("MANAGER_ID", IntegerType(), True),
    StructField(|"DEPARTMENT_ID", IntegerType(), True)
1)
source = spark.createDataFrame([
    (205, "Alex", "Smith", "ASMITH", "650.123.9999", datetime.strptime
    ("12-Jan-09", "%d-%b-%y"), "IT_PROG", 9000.0, None, 101, 60)
], schema=schema)
```

```
# Merge into Delta table
target = DeltaTable.forName(spark, "hive_metastore.default.employee_delta")

(
    target.alias("t")
    .merge(source.alias("s"), "t.EMPLOYEE_ID = s.EMPLOYEE_ID")
    .whenMatchedUpdateAll()
    .whenNotMatchedInsertAll()
    .execute()
)

> (8) Spark Jobs

> (a) Spark Jobs

> (a) Spark Jobs

> (b) Spark Jobs

> (c) Spark Jobs

> (a) Spark Jobs

> (b) Spark Jobs

> (c) Spark Jobs

> (d) Spark Jobs

> (e) Spark Jobs

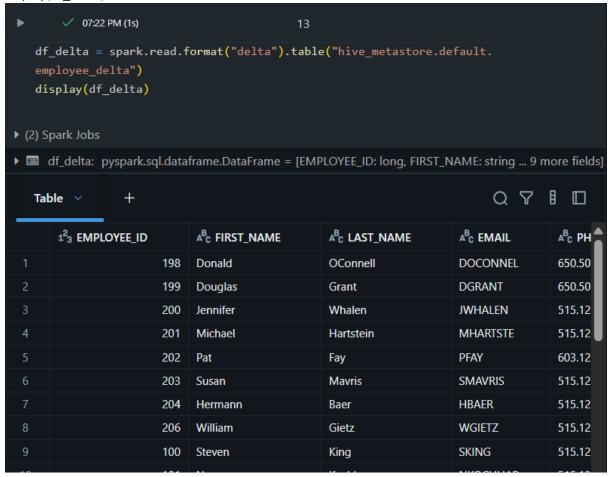
> (f) Spark Jobs

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|
```

Read a Delta Table

df_delta = spark.read.format("delta").table("hive_metastore.default.employee_delta")

display(df_delta)



Write to a Delta Table

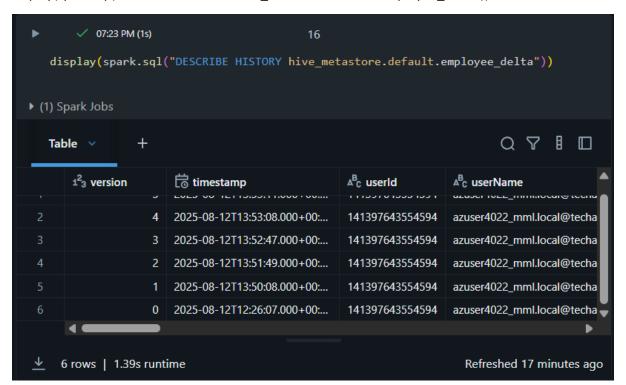
spark_df.write.format("delta").mode("append").saveAsTable("hive_metastore.default.employee_delt a")

Update a Delta Table

```
target.update(
  condition="DEPARTMENT_ID = 50",
  set={"SALARY": "SALARY + 500"}
)
```

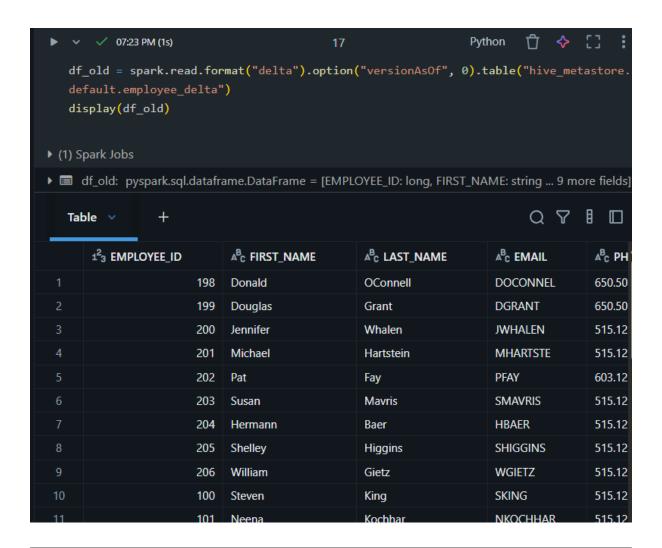
Display Table History

display(spark.sql("DESCRIBE HISTORY hive_metastore.default.employee_delta"))



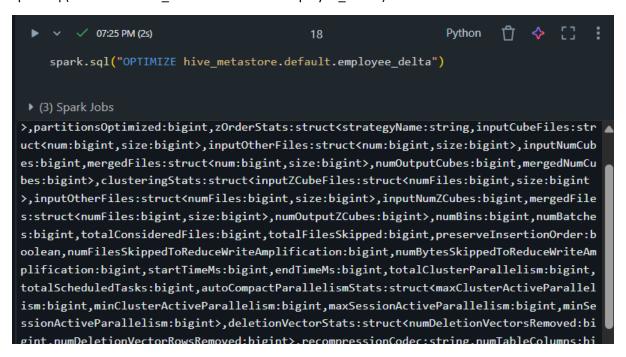
Query Earlier Version (Time Travel)

```
df_old = spark.read.format("delta").option("versionAsOf",
0).table("hive_metastore.default.employee_delta")
display(df_old)
```



Optimize a Delta Table

spark.sql("OPTIMIZE hive_metastore.default.employee_delta")



Clean Up Snapshots with VACUUM

spark.sql("VACUUM hive_metastore.default.employee_delta RETAIN 168 HOURS")