data-engineering-2023-epam-spark

Link to the github: https://github.com/yelzha/data-engineering-2023-epam-spark

The WeatherRestaurantProject is a PySpark application designed for extracting, transforming, and loading (ETL) data related to restaurants and weather. This script uses PySpark to perform various operations on restaurant and weather data, creating an enriched dataset that is then saved in a partitioned Parquet format.

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Task Description

Task

- Check restaurant data for incorrect (null) values (latitude and longitude). For incorrect values, map latitude
 and longitude from the Open Cage Geocoding API in a job via the REST API.
- Generate a geohash by latitude and longitude using a geohash library like geohash-java. Your geohash should be four characters long and placed in an extra column. ✓
- Left-join weather and restaurant data using the four-character geohash. Make sure to avoid data multiplication and keep your job idempotent.

Note: Development and testing should be done locally in your IDE environment.

• Store the enriched data (i.e., the joined data with all the fields from both datasets) in the local file system, preserving data partitioning in the parquet format. ✓

You are expected to:

- Upload the source code and implement tests
- Upload your fully documented homework with screenshots and comments in the task Readme MD file with the repo link.

Overview

The WeatherRestaurantProject performs the following key tasks:

- 1. Reads restaurant data from a CSV file, handling null coordinates by filling them using OpenCageApi information.\
- 2. Unions data with nulls and non-nulls, calculates geohash of coordinates, and processes the restaurant data.\
- 3. Reads weather data from a partitioned Parquet file.\
- 4. Applies geohash and drops duplicates in the weather data.\
- 5. Joins restaurant and weather dataframes using geohash, creating an enriched dataframe.\
- 6. Writes the repartitioned enriched dataframe to a Parquet file.\

Requirements

- Python=3.11
- JDK=1.17
- Hadoop=3.3.5

Python packages:

- PySpark=3.5.0
- Pygeohash=1.2.0 (for getting hash using latitude and longitude)
- OpenCageApi=2.4.0 (for handling null coordinates in restaurant data)

Also add new file constants.py

OpenCageGeocodeApiKey = "YOUR-API-KEY"

Structure of Project

- main.py: main file for all spark processes
- unzip.py: utility file for extracting zip files. So useful
- udfs.py: additional udfs file that contains vital functions and udfs
- tests.py: unit testing python file
- constants.py : constant values and keys for api and so on, so fourth...
- requirements.txt : minimal needed packages
- data/source : place for source files
- data/enriched: place for output data or enriched partitioned parquet table

Usage

To run the unzip.py script and extract zip files into a structured format, use the following command in your terminal

python -m pip install -r requirements.txt

To run the unzip.py script and extract zip files into a structured format, use the following command in your terminal

python unzip.py

To run the main, execute the following command in your terminal:

python main.py

Unittests

Tests are written using the Python unittest framework.

To run the tests, execute the following command in your terminal:

python tests.py

Screenshots and Scripts

1. Reads restaurant data from a CSV file, handling null coordinates by filling them using OpenCageApi information.

2. Unions data with nulls and non-nulls, calculates geohash of coordinates, and processes the restaurant data.

3. Reads weather data from a partitioned Parquet file.

Applies geohash and drops duplicates in the weather data.

```
INFO:__main__:Read data, created geohash, and removed duplicates in weather data.
INFO:__main__:First 10 rows in weather_df: None
INFO:__main__:Schema of weather_df:
|weather_lng|weather_lat|avg_tmpr_f|avg_tmpr_c|wthr_date |year|month|day|geohash|
|-111.143 | 18.6648
                    82.0
                                27.8
                                         |2017-08-27|2017|8
                                                             |27 |9e2f
|-111.143 | 18.6648
                    |81.0
                                         |2017-08-28|2017|8
                               27.2
                                                              |28 |9e2f
|-111.143 | 18.6648
                    80.7
                                         |2017-08-29|2017|8
                                                              |29 |9e2f
                               27.1
|-111.214
          18.8397
                    79.8
                               26.6
                                         |2017-08-27|2017|8
                                                              |27 |9e2g
|-111.214
          |18.8397
                    81.3
                                         |2017-08-28|2017|8
                               27.4
                                                              |28 |9e2g
|-111.214 | 18.8397
                    |82.4
                               28.0
                                         |2017-08-29|2017|8
                                                             |29 |9e2g
                                         |2017-08-27|2017|8
-111.09
          18.6251
                    82.0
                               27.8
                                                              |27 |9e31
|-111.09 | 18.6251 | 81.0
                                         |2017-08-28|2017|8
                                                             |28 |9e31
                               27.2
-111.09
         18.6251
                    80.7
                               27.1
                                         |2017-08-29|2017|8
                                                              |29 |9e31
|-110.995 | 18.6358 | 82.0
                               27.8
                                         |2017-08-27|2017|8
                                                             |27 |9e34
only showing top 10 rows
root
|-- weather_lng: double (nullable = true)
|-- weather_lat: double (nullable = true)
|-- avg_tmpr_f: double (nullable = true)
|-- avg_tmpr_c: double (nullable = true)
 |-- wthr_date: string (nullable = true)
|-- year: integer (nullable = true)
|-- month: integer (nullable = true)
|-- day: integer (nullable = true)
 |-- geohash: string (nullable = true)
```

4. Joins restaurant and weather dataframes using geohash, creating an enriched dataframe.

```
| NP3:_asin__:increasefully placed restrement and mether date/rease using geolash. | NP3:_asin__:irist 10 ross in enriched_6f; None | N
```

5. Writes the repartitioned enriched dataframe to a Parquet file.\

Plan of query:

```
== Physical Plan ==
AdaptiveSparkPlan isFinalPlan=false
+- SortAggregate(key=[geohash#318, year#306, restaurant_franchise_id#20, day#308, month#307], functions
      +- SortAggregate(key=[geohash#318, year#306, restaurant_franchise_id#20, day#308, month#307], functi
             +- Sort [geohash#318 ASC NULLS FIRST, year#306 ASC NULLS FIRST, restaurant_franchise_id#20 ASC NU
                   +- Project [geohash#318, weather_lng#785, weather_lat#787, avg_tmpr_f#789, avg_tmpr_c#791, wtl
                         +- SortMergeJoin [geohash#318], [geohash#145], Inner
                                :- Sort [geohash#318 ASC NULLS FIRST], false, 0
                                     +- Exchange hashpartitioning(geohash#318, 200), ENSURE REQUIREMENTS, [plan id=1564]
                                             +- SortAggregate(key=[geohash#318, year#306, month#307, day#308], functions=[first
                                                   +- SortAggregate(key=[geohash#318, year#306, month#307, day#308], functions=[page of the content of the content
                                                          +- Sort [geohash#318 ASC NULLS FIRST, year#306 ASC NULLS FIRST, month#307 AS
                                                                +- Filter isnotnull(geohash#318)
                                                                      +- GlobalLimit 1000000, 0
                                                                             +- Exchange SinglePartition, ENSURE_REQUIREMENTS, [plan_id=1527]
                                                                                   +- LocalLimit 1000000
                                                                                          +- Project [lng#301 AS weather_lng#329, lat#302 AS weather_la
                                                                                                +- BatchEvalPython [get_geohash(lat#302, lng#301)#317], [[
                                                                                                       +- LocalLimit 1000000
                                                                                                             +- FileScan parquet [lng#301,lat#302,avg tmpr f#303]
                                +- Sort [geohash#145 ASC NULLS FIRST], false, 0
                                      +- Exchange hashpartitioning(geohash#145, 200), ENSURE_REQUIREMENTS, [plan_id=1563]
                                             +- Filter isnotnull(restaurant lat#805)
                                                   +- SortAggregate(key=[restaurant_franchise_id#20, geohash#145], functions=[firs
                                                          +- Sort [restaurant_franchise_id#20 ASC NULLS FIRST, geohash#145 ASC NULLS F
                                                                +- Exchange hashpartitioning(restaurant_franchise_id#20, geohash#145, 200
                                                                      +- SortAggregate(key=[restaurant_franchise_id#20, geohash#145], functi
                                                                             +- Sort [restaurant_franchise_id#20 ASC NULLS FIRST, geohash#145 AS
                                                                                   +- Union
```

```
:- Project [id#17L, franchise id#18, franchise name#19, resta
  +- BatchEvalPython [get geohash(lat#23, lng#24)#144], [py1
      +- Project [id#17L, franchise id#18, franchise name#19]
         +- Filter isnotnull(pythonUDF0#836)
            +- BatchEvalPython [get geohash(lat#23, lng#24)#1
               +- Filter (isnotnull(lat#23) AND isnotnull(lng
                  +- GlobalLimit 100000, 0
                     +- Exchange SinglePartition, ENSURE_REQL
                        +- LocalLimit 100000
                           +- FileScan csv [id#17L,franchise
+- Project [id#127L, franchise id#128, franchise name#129, re
   +- BatchEvalPython [get geohash(pythonUDF0#840.lat, pythor
      +- BatchEvalPython [get coordinate(franchise name#129,
         +- Project [id#127L, franchise_id#128, franchise_nar
            +- Filter isnotnull(pythonUDF0#839)
               +- BatchEvalPython [get geohash(pythonUDF0#838
                  +- BatchEvalPython [get_coordinate(franchis
                     +- Filter (isnull(lat#133) OR isnull(lns
                        +- GlobalLimit 100000, 0
                           +- Exchange SinglePartition, ENSUF
                              +- LocalLimit 100000
                                 +- FileScan csv [id#127L,fra
```

Scripts:

```
main.py :
    from udfs import geohash_udf, get_coordinate_udf
    from pyspark.sql import SparkSession
    from pyspark.sql import functions as F
    import logging

def main() -> None:
    """
    Main ETL (Extract, Transform, Load) process for the WeatherRestaurantApp.
```

This script performs the following tasks:

- 1. Reads restaurant data from a CSV file, handling null coordinates by filling them using OpenCage/
- 2. Unions data with nulls and non-nulls, calculates geohash of coordinates, and processes the resta
- 3. Reads weather data from a partitioned Parquet file.
- 4. Applies geohash and drops duplicates in the weather data.
- 5. Joins restaurant and weather dataframes using geohash, creating an enriched dataframe.
- 6. Writes the repartitioned enriched dataframe to a Parquet file.

```
:return: None
"""
logger = logging.getLogger(__name__)
logging.basicConfig(level=logging.INFO)
```

```
try:
    spark = SparkSession \
        .builder \
        .master("local[*]") \
        .config("spark.app.name", "WeatherRestaurantApp") \
        .config("spark.io.compression.codec", "zstd") \
        .config("spark.sql.execution.pythonUDF.arrow.enabled", "true") \
        .config("spark.sql.execution.arrow.pyspark.enabled", "true") \
        .config("spark.sol.shuffle.partitions", 200) \
        .config("spark.driver.memory", "12g") \
        .config("spark.executor.memory", "12g") \
        .config("spark.executor.instances", 4) \
        .config("spark.executor.cores", 4) \
        .config("spark.sql.files.maxRecordsPerFile", 10 000) \
        .config("spark.executor.heartbeatInterval", "3600s") \
        .config("spark.network.timeout", "7200s") \
        .config("spark.network.timeoutInterval", "3600s") \
        .config("spark.driver.extraJavaOptions", "-Dlog4jspark.root.logger=WARN,console") \
        .config("spark.sql.debug.maxToStringFields", 200) \
        .enableHiveSupport() \
        .getOrCreate()
    # read data from restaurant csv file
    restaurants df = spark.read \
        .option("header", "true") \
        .option('inferSchema', 'true') \
        .csv("data/source/restaurant_csv") \
        # .limit(100_000)
    logger.info("Read data from restaurant CSV file.")
    # filtering null data and fill it using OpenCageApi information
    null_coord_restaurant_df = restaurants_df \
        .filter(F.col('lat').isNull() | F.col('lng').isNull()) \
        .drop("lat", "lng") \
        .select("*", get_coordinate_udf(F.col('franchise_name'), F.col('city'), F.col('country')).
        .select(
            F.col("coord.lat").alias("lat"),
            F.col("coord.lng").alias("lng")) \
        .drop("coord")
    logger.info("Filled nulls in restaurant data.")
    logger.info(f"Number of rows in null_coord_restaurant_df: {null_coord_restaurant_df.count()}")
    logger.info(f"First 10 rows in null_coord_restaurant_df:")
    null coord restaurant df.show(10, truncate=False)
    # union data with nulls and non-nulls, and also get geohash of coordinates
    restaurants_df = restaurants_df \
        .filter(F.col('lat').isNotNull() & F.col('lng').isNotNull()) \
        .union(null coord restaurant df) \
        .withColumn("geohash", geohash_udf(F.col("lat"), F.col("lng"))) \
        .withColumnsRenamed({'lat': 'restaurant_lat', 'lng': 'restaurant_lng'}) \
        .dropDuplicates(["restaurant franchise id", "geohash"])
```

```
logger.info("Performed union of nulls and non-nulls and created geohash in restaurant data.")
         logger.info("First 10 rows in restaurants df:")
         restaurants df.show(10, truncate=False)
         logger.info("Schema of restaurants df:")
         restaurants_df.printSchema()
         # reading weather data from partitioned parquet
         weather_df = spark.read.parquet("data/source/weather") \
             # .limit(1_000_000)
         # applying geohash adn dropping duplicates
         weather df = weather df \
             .withColumn("geohash", geohash udf(F.col("lat"), F.col("lng"))) \
             .withColumnsRenamed({'lat': 'weather_lat', 'lng': 'weather_lng'}) \
             .dropDuplicates(["geohash", "year", "month", "day"])
         logger.info("Read data, created geohash, and removed duplicates in weather data.")
         logger.info(f"First 10 rows in weather df: {weather df.show(10, truncate=False)}")
         logger.info("Schema of weather df:")
         weather_df.printSchema()
         # creating new dataframe joining restaurant to weather dataframe
         enriched df = weather df \
             .join(restaurants df, on=["geohash"], how="left") \
             .filter(F.col("restaurant lat").isNotNull()) \
             .dropDuplicates(["geohash", "year", "month", "day", "restaurant_franchise_id"])
         logger.info("Successfully joined restaurant and weather dataframes using geohash.")
         logger.info(f"First 10 rows in enriched_df: {enriched_df.show(10, truncate=False)}")
         logger.info("Schema of enriched_df:")
         enriched df.printSchema()
         logger.info("Plan of query:")
         enriched_df.explain()
         # write saving previous partitioning
         enriched_df.repartition(3).write.partitionBy("year", "month", "day").mode("overwrite").parquet
         logger.info("Successfully wrote repartitioned enriched dataframe.")
     except Exception as e:
         logger.error(f"An error occurred: {str(e)}")
     finally:
         spark.stop()
         logger.info("Successfully finished.")
 if __name__ == '__main__':
     main()
unzip.py:
 import zipfile
 import os
```

```
data path = os.path.join(os.getcwd(), "data", "source")
 def unzip_combine_weather_zips(weather_dir: str) -> None:
     """Unzipper for weather dataset.
     And combine all partitioned data into one directory and save its structure
     :param weather_dir: directory of weather file.
     weather subzips = [os.path.join( data path, weather dir, file name) for file name in
                        os.listdir(os.path.join( data path, weather dir))]
     for weather subzip in weather subzips:
         with zipfile.ZipFile(weather subzip, "r") as weather subfiles:
             weather_subfiles.extractall(f'{_data_path}/')
 def unzip_restaurant_zip(restaurant_path: str) -> None:
     """Unzipper for restaurant dataset.
     :param restaurant path: directory of weather file.
     with zipfile.ZipFile(os.path.join(_data_path, restaurant_path), "r") as weather_subfiles:
         weather subfiles.extractall(f'{ data path}/')
 if name == " main ":
     unzip_combine_weather_zips("weather_zip_folder")
     unzip restaurant zip("restaurant csv.zip")
udfs.py:
 from pyspark.sql import functions as F
 from pyspark.sql.types import StructType, StructField, DoubleType, StringType
 from opencage.geocoder import OpenCageGeocode
 import pygeohash
 import logging
 # this is key saver... Create this file after cloning repos and define new variable OpenCageGeocodeApi
 import constants
 geocoder = OpenCageGeocode(constants.OpenCageGeocodeApiKey)
 def get_coordinate(franchise_name: str, city: str, country: str) -> tuple:
     Get geographical coordinates (latitude, longitude) for a given franchise name, city, and country.
     Args:
```

```
franchise name (str): The name of the franchise.
        city (str): The city associated with the franchise.
        country (str): The country code of the location.
   Returns:
       tuple: A tuple containing latitude and longitude. Returns (None, None) if geocoding fails.
   Note:
       This function relies on an external geocoding service.
   Raises:
        Exception: An exception is caught if geocoding fails for any reason.
    query = f'{franchise name}, {city}'
        coordinate = geocoder.geocode(query, countrycode=country.lower(), limit=1)
       if coordinate is not None:
            return coordinate[0]['geometry']['lat'], coordinate[0]['geometry']['lng']
       else:
           logging.warning(f"Geocoding no possible for {query}")
   except Exception as e:
        logging.warning(f"Geocoding failed for {query}. Error: {str(e)}")
    return None, None
def get geohash(lat: float, lng: float) -> str:
   Generate a geohash string based on the given latitude and longitude.
   Args:
       lat (float): Latitude coordinate.
        lng (float): Longitude coordinate.
   Returns:
        str: A geohash string representing the location.
   Note:
       This function uses the pygeohash library for geohashing.
   Warning:
        If either latitude or longitude is None, the function returns None and logs a warning.
    if lat is None or lng is None:
        logging.warning(f"Geohash generation not possible. Latitude or longitude is missing.")
   if not ((-90.0 <= lat) and (lat <= 90.0)) or not ((-180.0 <= lng) and (lng <= 180.0)):
       logging.warning(f"Geohash generation not possible. Latitude or longitude is out of bound.")
   return pygeohash.encode(lat, lng, precision=4)
# schema with nullable latitude and longitude
coord schema = StructType([
   StructField("lat", DoubleType(), True),
```

```
StructField("lng", DoubleType(), True)
 1)
 get coordinate udf = F.udf(get coordinate, coord schema)
 geohash_udf = F.udf(get_geohash, StringType())
 if __name__ == "__main__":
     # mini test
     print(get_coordinate("Savoria", "Dillon", "US"))
     print(get geohash(181, 180))
tests.py :
 import unittest
 from pyspark.sql import SparkSession
 from udfs import get_coordinate, get_geohash
 class TestGeoFunctions(unittest.TestCase):
     def test get coordinate(self):
         # Positive test case: Valid coordinates
         result = get coordinate("Savoria", "Dillon", "US")
         self.assertIsInstance(result, tuple)
         self.assertEqual(len(result), 2)
         self.assertIsInstance(result[0], float)
         self.assertIsInstance(result[1], float)
         # Negative test case: Invalid franchise and city
         result = get_coordinate("InvalidFranchise", "InvalidCity", "ZZ")
         self.assertIsInstance(result, tuple)
         self.assertEqual(len(result), 2)
         self.assertIsNone(result[0])
         self.assertIsNone(result[1])
         # Negative test case: Invalid country code
         result = get_coordinate("Savoria", "Dillon", "InvalidCountry")
         self.assertIsInstance(result, tuple)
         self.assertEqual(len(result), 2)
         self.assertIsNotNone(result[0])
         self.assertIsNotNone(result[1])
         # Negative test case: Missing franchise name
         result = get_coordinate("", "Dillon", "US")
         self.assertIsInstance(result, tuple)
         self.assertEqual(len(result), 2)
         self.assertIsNotNone(result[0])
         self.assertIsNotNone(result[1])
         # Negative test case: Missing city
         result = get coordinate("Savoria", "", "US")
         self.assertIsInstance(result, tuple)
         self.assertEqual(len(result), 2)
         self.assertIsNone(result[0])
```

```
self.assertIsNone(result[1])
     def test get geohash(self):
         # Positive test case: Valid coordinates
         result = get_geohash(1, -1)
         self.assertIsInstance(result, str)
         self.assertEqual(len(result), 4)
         # Negative test case: Missing latitude or longitude
         result = get geohash(None, -1)
         self.assertIsNone(result)
         # Negative test case: Out of bound latitude or longitude
         result = get_geohash(181, -181)
         self.assertIsNone(result)
         # Positive test case: Coordinates within a specific range
         result = get geohash(39.63026, -106.04335)
         self.assertIsInstance(result, str)
         self.assertEqual(len(result), 4)
         self.assertTrue(result.isalnum()) # Check if the result is alphanumeric
 if name == " main ":
     unittest.main()
constants.pv :
 OpenCageGeocodeApiKey = "YOUR-API-KEY"
```

Mini-Guide how to install Pyspark, Jdk and hadoop

https://dev.to/awwsmm/installing-and-running-hadoop-and-spark-on-windows-33kc

- 1. Install Python 3.11
- 2. Download JDK-11 or JDK-17
- 3. https://spark.apache.org/downloads.html

Download spark-3.5.0-bin-hadoop 3.tgz

After switching to C:/Spark

4. https://hadoop.apache.org/release/3.3.5.html

Download, then run using 7-Zip as an administrator, unzip, unzip again. Pom peren. v C:/Hadoop

5. https://github.com/cdarlint/winutils/tree/master/hadoop-3.3.5/bin

Download anything you want to transfer to C:/Hadoop/bin/

6. Add variables in Windows

```
JAVA HOME=C:/...
```

HADOOP_HOME=C:/Hadoop SPARK_HOME=C:/Spark File path: %JAVA_HOME%/bin %HADOOP_HOME%/bin %SPARK_HOME%/bin\