# DISTRIBUTED SYSTEMS ASSIGNMENT REPORT



**Assignment ID:** 3

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# **DESIGN**

## **Structure**

```
1 \vdash code
          parking_analysis.ipynb # main analysis notebook
          plot.py # plot functions
          requirements.txt
5
6 ⊢data
          .gitkeep
          parking_data_sz.csv # origin data (may absent in submision)
          parking_utilization.png # r5 subtask picture
9
          r1.csv
10
11
          r2.csv
          r3.csv
13
          r4.csv
14
          r5.csv
```

I will use code and necessary comments to describe this part

## **Data Clean**

```
def __init__(self):
 1
            self.sc: SparkContext
 2
 3
            self.spark: SparkSession
            self.df: DataFrame
 4
            self.__tasks: tuple[DataFrame, DataFrame,
 6
                                DataFrame, DataFrame]
 7
            self.__init_data()
 8
        def __init_data(self):
 9
            """Initialize the data and create the dataframe"""
10
11
12
            # Create a Spark context
13
            sc = SparkContext.getOrCreate()
14
15
            # Create a Spark session
16
            spark = SparkSession \
17
                .builder \
                .appName("Parking Data Analysis") \
18
19
                .getOrCreate()
20
21
            # Load data from csv file
22
            df = spark.read.csv(
                path=ParkingDb.__get_data_file_path("parking_data_sz.csv"),
23
24
                header=True,
                schema=ParkingDb.schema
25
26
            )
27
            # filter out invalid data by checking out_time > in_time
28
29
            cleaned_data = df.filter(
                f.col("out_time") > f.col("in_time"))
30
31
            # calculate parking time length and drop out_time column
32
            df_convert_out = cleaned_data.withColumn(
34
                "parking_time_length",
                f.unix_timestamp("out_time") - f.unix_timestamp("in_time")
35
            ).drop("out_time")
36
37
            # cache the final dataframe
38
            # df_convert_out = df_convert_out.cache()
39
            df_final = df_convert_out
40
41
42
            # assign to instance variables
            self.sc = sc
43
44
            self.spark = spark
45
            self.df = df_final
46
            self.__tasks = self.__execute()
```

## Task implementation logic

#### Task 1

```
def __task1(self) -> DataFrame:
    return self.df.groupBy("section") \
         agg(f.countDistinct("berthage").alias("count")) \
         # .orderBy("section")
```

#### Task 2

```
def __task2(self) -> DataFrame:
    return self.df.select("berthage", "section").distinct() \
    # .orderBy("section", "berthage")
```

#### Task 3

#### Task 4

#### Task 5

```
def __task5(self) -> DataFrame:
    # create a new dataframe with hourly time range
    df_hours = self.df.withColumn(
        "start_time",
        f.date_trunc("hour", f.col("in_time"))
```

```
).withColumn(
                "end_time",
 7
                f.date_trunc("hour", f.col("in_time")) + f.expr("INTERVAL 1 HOUR")
 8
 9
            )
10
            # calculate total berthages for each section
11
            total_berthages = self.df.select("section", "berthage").distinct() \
12
                .groupBy("section").count().withColumnRenamed("count", "total_berthages")
13
14
            # calculate hourly usage for each section
15
            hourly_usage = df_hours.groupBy("start_time", "end_time", "section") \
16
                .agg(f.countDistinct("berthage").alias("count"))
17
18
            # join the two dataframes and calculate the percentage
19
            result: DataFrame = hourly_usage.join(total_berthages, "section") \
20
21
                .withColumn(
                "percentage",
22
23
                f.format_number(f.col("count") / f.col("total_berthages") * 100, 1)
            ).select("start_time", "end_time", "section", "count", "percentage") \
24
                # .orderBy("section", "start_time")
25
26
27
            return result
```

## **RUNNING RESULT**

## **Result Show**

Could be found in notebook's output and .csv files.

## Spark job's DAG

## Task 1

Status: SUCCEEDED

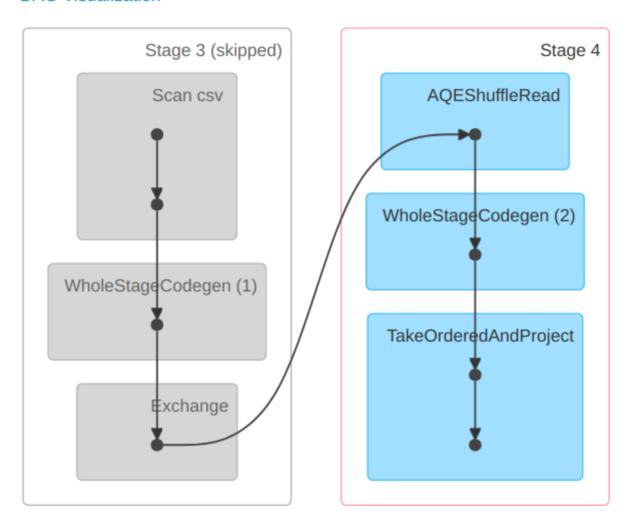
Submitted: 2024/12/12 19:39:34

Duration: 58 ms

Associated SQL Query: 7

Completed Stages: 1 Skipped Stages: 1

▶ Event Timeline



Task 2

Status: SUCCEEDED

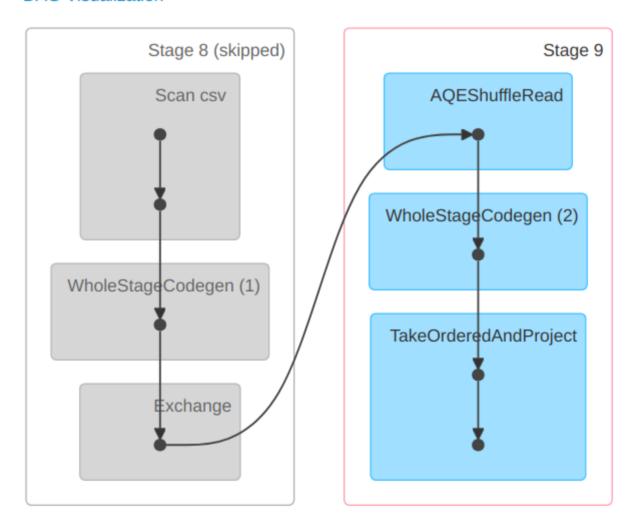
Submitted: 2024/12/12 19:39:53

Duration: 38 ms

Associated SQL Query: 9

Completed Stages: 1 Skipped Stages: 1

▶ Event Timeline



Task 3

Status: SUCCEEDED

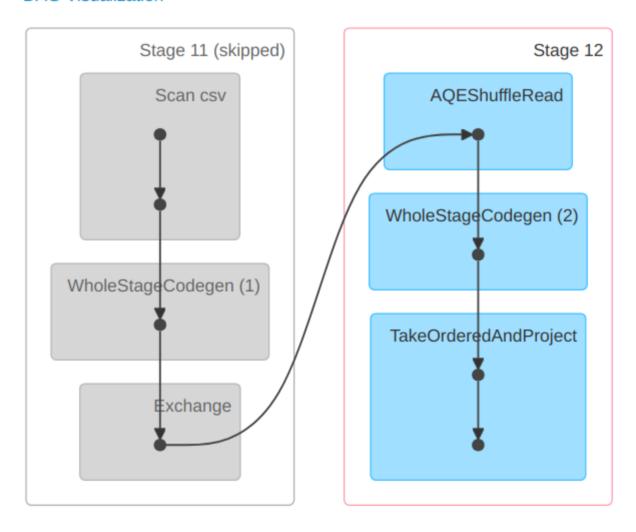
Submitted: 2024/12/12 19:39:54

Duration: 87 ms

Associated SQL Query: 10

Completed Stages: 1 Skipped Stages: 1

Event Timeline



Task 4

Status: SUCCEEDED

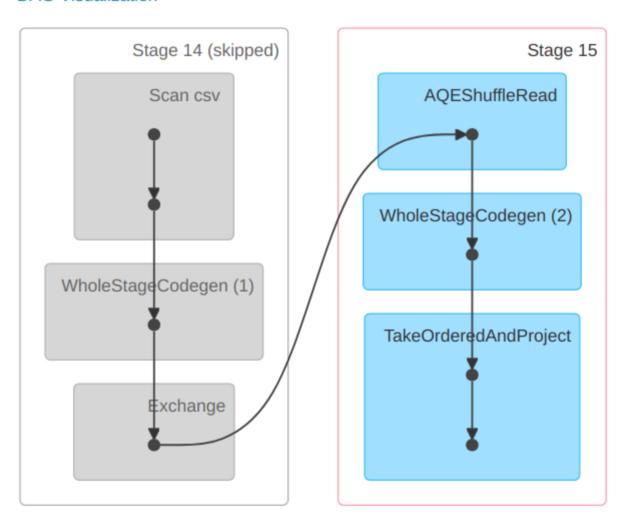
**Submitted:** 2024/12/12 19:39:55

Duration: 47 ms

Associated SQL Query: 11

Completed Stages: 1 Skipped Stages: 1

▶ Event Timeline



Task 5

Status: SUCCEEDED

Submitted: 2024/12/12 19:39:57

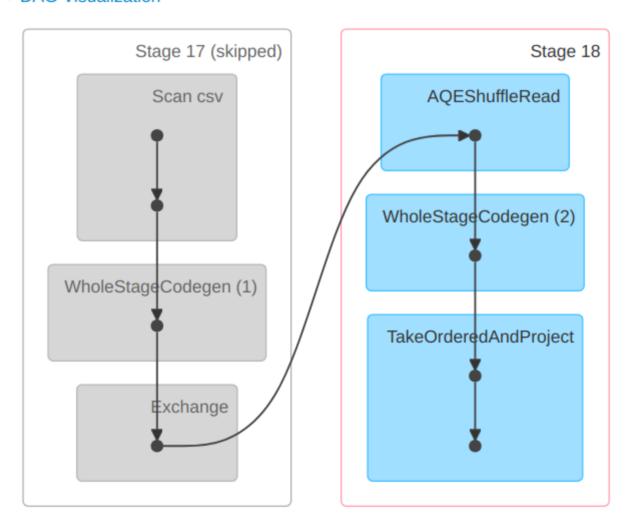
**Duration:** 58 ms

Associated SQL Query: 12

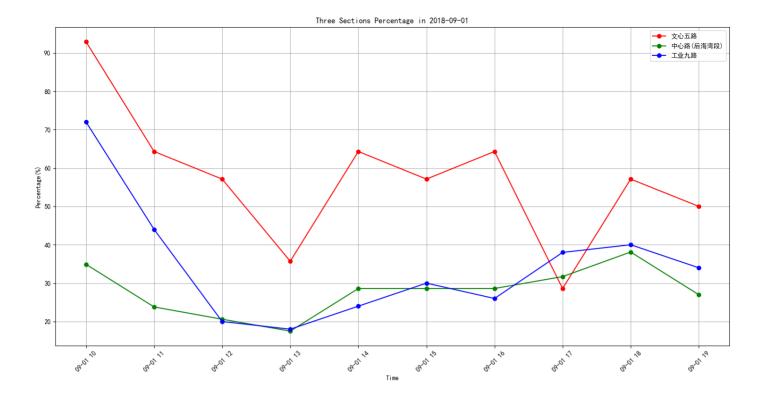
Completed Stages: 1 Skipped Stages: 1

▶ Event Timeline

▼ DAG Visualization



r5 Subtask Analysis



I selected three sections and analyzed and visualized them hourly on 2018-09-01.

I found that parking occupancy rates were higher in the morning and evening, probably because people had not started their day yet, so the cars were in the parking lot.

During the day, people would drive around, so parking occupancy rates were lower.