SF BAY AREA BIKE SHARE

MSAN697

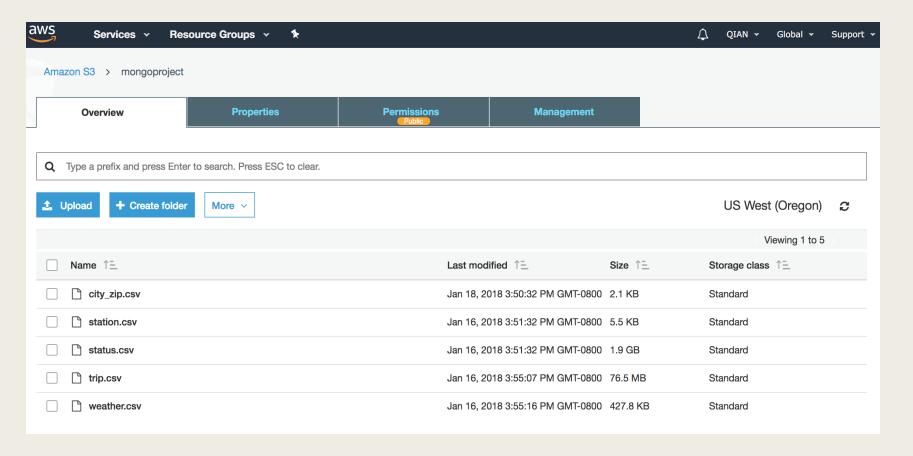
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Overview

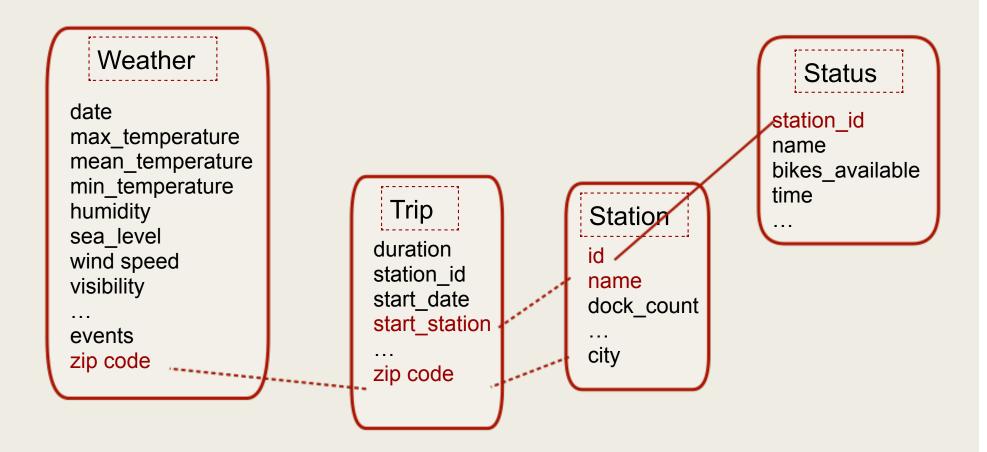
- Kaggle competition- "SF Bay Area Bike Share"
- The emergence of bike-share system transform the way we travel.
- We want to predict number of daily bike-share trips and analyze how much effect weather has on this number



Launch S3 to store data



Understanding Data Structure



Imported data from S3 into MongoDB on EC2

```
AWS — ec2-user@ip-172-31-40-253:~ — ssh -i

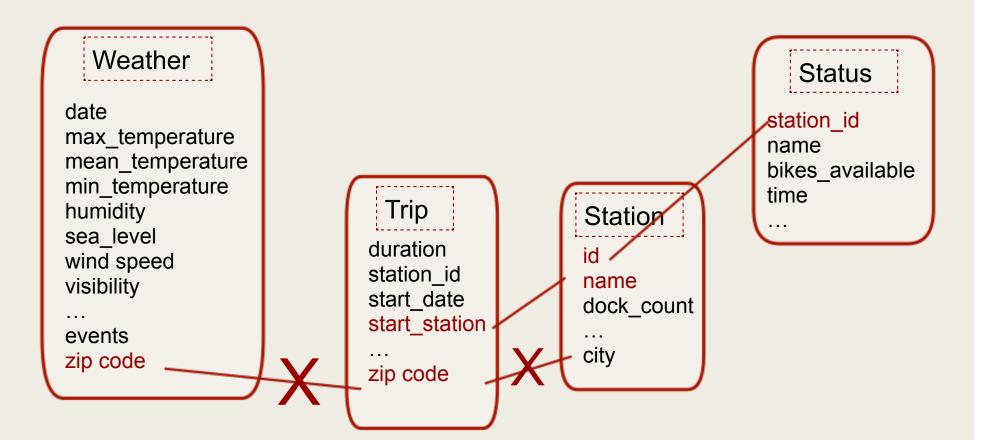
> db.trip.findOne()

{
    "_id" : ObjectId("5a5fa3e005daa8eea48c197b"),
    "id" : 4576,
    "duration" : 63,
    "start_date" : "8/29/2013 14:13",
    "start_station_name" : "South Van Ness at Market",
    "start_station_id" : 66,
    "end_date" : "8/29/2013 14:14",
    "end_station_name" : "South Van Ness at Market",
    "end_station_id" : 66,
    "bike_id" : 520,
    "subscription_type" : "Subscriber",
    "zip_code" : 94127
}
```

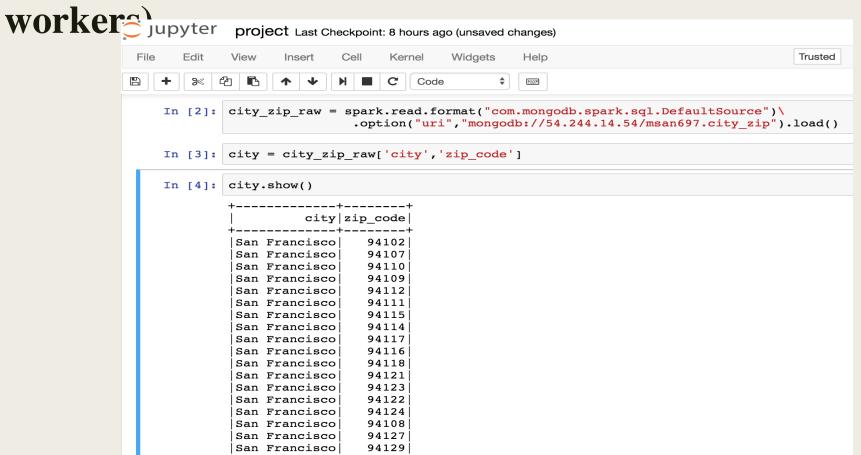


94127 not accurate

Understanding Data Structure



Import data from MongoDB on EMR (1 master 2



Create New DataFrame

Column types should be changed

```
311: trip = spark.read.format("com.mongodb.spark.sql.DefaultSource").
                                                                                          Assigned
32]: trip.printSchema()
                                                                                          defaultly
    root
                                                                                               should be
      -- id: struct (nullable = true)
                                                                                               deleted
          -- oid: string (nullable = true)
       -- bike id: integer (nullable = true)
       -- duration: integer (nullable = true)
       -- end date: string (nullable = true)
       -- end station id: integer (nullable = true)
       -- end station name: string (nullable = true)
       -- id: integer (nullable = true)
                                                                                          Type
       -- start date: string (nullable = true)-
                                                                                               string --
       -- start station id: integer (nullable = true)
                                                                                               timestamp
       -- start station name: string (nullable = true)
       -- subscription type: string (nullable = true)
      -- zip code: string (nullable = true)
```

Create New DataFrame

step 1 : create schema

new DF

Create New DataFrame

```
root
|-- id: long (nullable = true)
|-- duration: double (nullable = true)
|-- date: timestamp (nullable = true)
|-- start_station_name: string (nullable = true)
|-- start_station_id: long (nullable = true)
|-- end_date: timestamp (nullable = true)
|-- end_station_name: string (nullable = true)
|-- end_station_id: long (nullable = true)
|-- bike_id: long (nullable = true)
|-- subscription_type: string (nullable = true)
|-- zip_code: long (nullable = true)
```

```
weatherdf.printSchema()
root
 -- date: date (nullable = true)
  -- max temperature f: float (nullable = true)
  -- mean temperature f: float (nullable = true)
  -- min temperature f: float (nullable = true)
  -- max dew point f: float (nullable = true)
  -- mean_dew_point_f: float (nullable = true)
  -- min dew point f: float (nullable = true)
  -- max humidity: float (nullable = true)
  -- mean humidity: float (nullable = true)
  -- min humidity: float (nullable = true)
  -- max sea level pressure inches: float (nullable = true)
  -- mean sea level pressure inches: float (nullable = true)
  -- min sea level pressure inches: float (nullable = true)
  -- max visibility miles: float (nullable = true)
  -- mean visibility miles: float (nullable = true)
  -- min_visibility_miles: float (nullable = true)
  -- max wind Speed mph: float (nullable = true)
  -- mean wind Speed mph: float (nullable = true)
  -- max gust speed mph: float (nullable = true)
  -- precipitation inches: float (nullable = true)
  -- cloud cover: float (nullable = true)
  -- events: string (nullable = true)
  -- wind_dir_degrees: float (nullable = true)
  -- zip code: long (nullable = true)
```

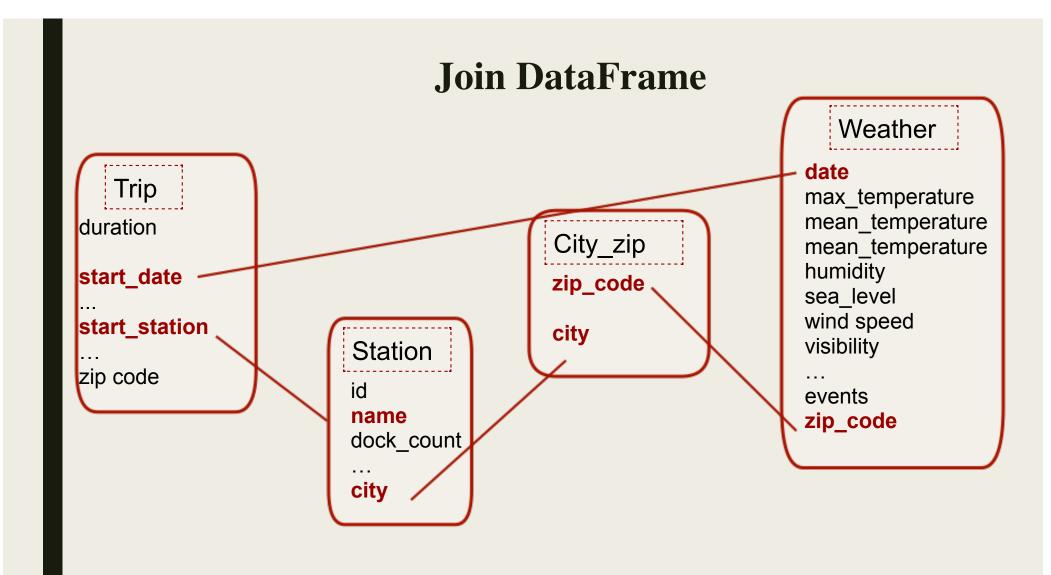
Weather date max_temperature mean_temperature ... events zip code Trip duration start_date ... end_station ... zip code

Extract day of week

```
def dow(x):
    v = x.strftime('%w')
    return v

dow = udf(dow, StringType())
```

tripdf = tripdf.withColumn("day_of_week",dow(tripdf['date']))



SparkSQL

```
weather_city = weatherdf.join(city_zip, 'zip_code').drop('zip_code')
                             trip_station = tripdf.join(stationdf, (tripdf.start_station_name == stationdf.name))
                             complete_table = trip_station.join(weather_city, ['date', 'city'])
                             sqlContext.sql('select count(1) as num_trips, \
                                             date, city from raw_data\
                                             group by 2, 3').show(5)
                              num_trips
                                              date
                                                            city
Response
                                      7 2013-09-04
                                                       Palo Alto
                                      7 2013-09-16 Mountain View
                                    448 2013-10-26 San Francisco
                                    865 2013-10-28 San Francisco
                                     14 2014-01-17 Mountain View
                             only showing top 5 rows
```

Imputing & Encoding



Machine Learning



Conclusion

The GBT model has a R2 score of 9.4

Most Important features

- San Francisco,
- Mean Wind Speed

```
feat_impo.sortBy(lambda x: x[1], ascending=False).take(10)
```

```
[('san_francisco', 0.2224607674272947),
  ('mean_wind_Speed_mph', 0.075666132179973408),
  ('min_dew_point_f', 0.062387996140445011),
  ('max_wind_Speed_mph', 0.054416037663545698),
  ('max_sea_level_pressure_inches', 0.046911619313796782),
  ('min_humidity', 0.043752369316787332),
  ('cloud_cover', 0.040819665641833432),
  ('max_dew_point_f', 0.039471005872858705),
  ('wind_dir_degrees', 0.037685852085101973),
  ('mean_humidity', 0.036768024362484095)]
```

Lesson learned

- Different operating systems have different characters representing the new line. We cannot discern difference except displaying in vim. The difference will interfere mongoimport command.
- AWS EC2 has its architecture and space distribution that we usually have to go to a memory space much larger than the data we need to store
 - t2.small (2GB), t2.medium(4GB), t2.xlarge(16GB) EBS
 - m3.2xlarge, 30GB 2*80 SSD
- s3 and ec2 put in the same region has higher connection success rate

Lesson Learned

Work with pyspark dataframe efficiently

- ☐ Create dataframe with RDD and schema
- pros: safe, easy to change column type
- cons: heavy boring finger workout when the dataframe is wide
- load data using spark.read.format()
- pros: generate dataframe automatically without specifying schema
- cons: sometimes hard to coerce to desired column type