Week4 Lab01 Simonsen

June 2, 2024

```
[]: from pyspark.sql.types import *
     import pyspark.sql.functions as F
[]: fire_schema = StructType([StructField('CallNumber', IntegerType(), True),
                          StructField('UnitID', StringType(), True),
                          StructField('IncidentNumber', IntegerType(), True),
                          StructField('CallType', StringType(), True),
                          StructField('CallDate', StringType(), True),
                          StructField('WatchDate', StringType(), True),
                          StructField('CallFinalDisposition', StringType(), True),
                          StructField('AvailableDtTm', StringType(), True),
                          StructField('Address', StringType(), True),
                          StructField('City', StringType(), True),
                          StructField('Zipcode', IntegerType(), True),
                          StructField('Battalion', StringType(), True),
                          StructField('StationArea', StringType(), True),
                          StructField('Box', StringType(), True),
                          StructField('OriginalPriority', StringType(), True),
                          StructField('Priority', StringType(), True),
                          StructField('FinalPriority', IntegerType(), True),
                          StructField('ALSUnit', BooleanType(), True),
                          StructField('CallTypeGroup', StringType(), True),
                          StructField('NumAlarms', IntegerType(), True),
                          StructField('UnitType', StringType(), True),
                          StructField('UnitSequenceInCallDispatch', IntegerType(),
      →True).
                          StructField('FirePreventionDistrict', StringType(), True),
                          StructField('SupervisorDistrict', StringType(), True),
                          StructField('Neighborhood', StringType(), True),
                          StructField('Location', StringType(), True),
                          StructField('RowID', StringType(), True),
                          StructField('Delay', FloatType(), True)])
[]: sf_fire_file = "dbfs:/FileStore/Merrimack/Week_4/sf_fire_calls.csv"
     fire_df = spark.read.csv(sf_fire_file,header=True, schema=fire_schema)
[]: fire_df.show(5)
```

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+							
+							
+							
CallNumber UnitI							
WatchDate CallFin	-		vailableDt				
Address City Zipc				-	-	-	
riority ALSUnit C				_	_		
PreventionDistric	t Superviso	rDistrict	Nei	ghborhood	L	ocati	lon
RowID Delay							
+							
+							
+							
+							. – – –
20110016 T1							
Other 01/11/2002	•	•			B04	1	
38 3362	3	3			NULLI		1
TRUCK	01	21	01	41	NOLL	5 l	- 1
Pacific Heights (37.789584067	•	20110016-T1	= •		01	
20110022 M1				t 01/11/2002	01/10/2002	ı	
Other 01/11/2002					B10 l	•	
42 6495	3	3	3	true	NULL		1
MEDIC	·	1		10	·		·
10 Bayview Hunter	s P (37.7	73376236738	397 0201	10022-M17	4.7		
20110023 M4				t 01/11/2002	01/10/2002	I	
Other 01/11/2002	02:39: M	ARKET ST/MO	CALLIS	SF 94102	B03		
01 1455	3	3	3	true	NULL		1
MEDIC		2		3		6	
Tenderloin (37.78	11772186856	0201100	23-M41 2.4	333334			
20110032 E1	1 200	03250 \	ehicle Fir	e 01/11/2002	01/10/2002	1	
Other 01/11/2002	04:16: AI	PPLETON AV/	MISSI	SF 94110	B06		
32 5626	3	3	3	false	NULL		1
ENGINE		1		6		9	
Bernal Heights (3							
20110043 BO		03259		ıs 01/11/2002	01/10/2002	1	
Other 01/11/2002					B04		
03 3223	3	3	3	•	NULL		1
CHIEF	,	2		4		2	
Western Addition							
+		•					
+_		·		•	•		
+		•		·	·		•
+			·		·		
only showing top	·	· -	,-	-	,		

```
[]: fire_ts_df = (fire_df
                 .withColumn('IncidentDate',F.to_timestamp(F.col('CallDate'),'MM/dd/
     →yyyy'))
                 .drop('CallDate')
                 .withColumn('OnWatchDate',F.to_timestamp(F.col('WatchDate'),'MM/dd/
     .drop('WatchDate')
                 .withColumn('AvailableDtTS',F.to_timestamp(F.
     .drop('AvailableDtTm')
                 )
[]: fire_ts_df.cache()
    fire_ts_df.columns
[]: ['CallNumber',
      'UnitID',
      'IncidentNumber',
      'CallType',
      'CallFinalDisposition',
      'Address',
     'City',
      'Zipcode',
      'Battalion',
      'StationArea',
      'Box',
      'OriginalPriority',
      'Priority',
      'FinalPriority',
      'ALSUnit',
      'CallTypeGroup',
      'NumAlarms',
      'UnitType',
      'UnitSequenceInCallDispatch',
      'FirePreventionDistrict',
      'SupervisorDistrict',
      'Neighborhood',
      'Location',
      'RowID',
     'Delay',
      'IncidentDate',
      'OnWatchDate',
      'AvailableDtTS']
[]: (fire_ts_df
         .select('IncidentDate','OnWatchDate','AvailableDtTS')
         .show(5, False)
```

```
|OnWatchDate
    +----+
    |2002-01-11 00:00:00|2002-01-10 00:00:00|2002-01-11 01:51:44|
    |2002-01-11 00:00:00|2002-01-10 00:00:00|2002-01-11 03:01:18|
    |2002-01-11 00:00:00|2002-01-10 00:00:00|2002-01-11 02:39:50|
    |2002-01-11 00:00:00|2002-01-10 00:00:00|2002-01-11 04:16:46|
    |2002-01-11 00:00:00|2002-01-10 00:00:00|2002-01-11 06:01:58|
    +----+
    only showing top 5 rows
[]: # 1)
               What were all the different types of fire calls in 2018?
    (fire_ts_df
        .select("CallType")
        .where(F.col('CallType').isNotNull())
        .where(F.year('IncidentDate') == 2018 )
        .distinct()
        .show(truncate=False)
    |CallType
    |Elevator / Escalator Rescue
    |Alarms
    |Odor (Strange / Unknown)
    |Citizen Assist / Service Call
    | HazMat
    |Vehicle Fire
    Other
    |Outside Fire
    |Traffic Collision
    |Assist Police
    |Gas Leak (Natural and LP Gases)|
    | Water Rescue
    |Electrical Hazard
    |Structure Fire
    |Medical Incident
    |Fuel Spill
    |Smoke Investigation (Outside) |
    |Train / Rail Incident
    |Explosion
    |Suspicious Package
```

```
[]: (fire_ts_df
     .select("CallType").where(F.col("CallType").isNotNull())
     .groupBy("CallType")
     .count()
     .orderBy("count", ascending=False)
     .show(n=10, truncate=False))
    +----+
    |CallType
    +----+
                               |113794|
    |Medical Incident
    |Structure Fire
                               |23319 |
    | Alarms
                               |19406 |
    |Traffic Collision
                               |7013 |
    |Citizen Assist / Service Call |2524 |
    Other
                                |2166 |
    |Outside Fire
                                12094 I
    |Vehicle Fire
                                |854 |
    |Gas Leak (Natural and LP Gases)|764
    |Water Rescue
   +----+
   only showing top 10 rows
[]: # 2) What months within the year 2018 saw the highest number of fire calls?
     → ANSWER: October, May, and March had the highest number of fire calls.
    (fire_ts_df
     .select(F.month('IncidentDate').alias('Month'))
     .where(F.year('IncidentDate')==2018)
     .groupBy('Month')
     .count()
     .orderBy('count',ascending=False)
     .show()
    )
    +----+
    |Month|count|
   +----+
     10 | 1068 |
        5 | 1047 |
      3 | 1029 |
      8 | 1021 |
       1 | 1007 |
       6 | 974 |
       7 | 974 |
        9 | 951 |
```

```
11 | 199 |
   +----+
[]: #3) Which neighborhood in San Francisco generated the most fire calls in
    →2018? ANSWER: Tenderloin generated the most fire calls in 2018.
    (fire_ts_df
     .select('Neighborhood')
     .where(F.year('IncidentDate')==2018)
     .groupBy('Neighborhood')
     .count()
     .orderBy('count',ascending=False)
     .show(n=5, truncate=False)
   +----+
   Neighborhood
   +----+
   Tenderloin
                             |1393 |
   |South of Market
                             |1053 |
   Mission
                             |913 |
   |Financial District/South Beach|772 |
   |Bayview Hunters Point
   +----+
   only showing top 5 rows
[]: | # 4) Which neighborhoods had the worst response times to fire calls in
     →2018? ANSWER: In 2018, Chinatown, Financial District/South Beach, and
    → Tenderloin had the worst response time to fire calls.
    fire_df_response = fire_ts_df.withColumnRenamed("Delay","ResponseDelayedMins")
    (fire_df_response
     .select('Neighborhood','ResponseDelayedMins')
     .where(F.year('IncidentDate')==2018)
     .orderBy('ResponseDelayedMins',ascending=False)
     .show(n=10,truncate=False)
   +----+
   Neighborhood
                             |ResponseDelayedMins|
   +-----+
   Chinatown
                             1491.26666
   |Financial District/South Beach|406.63333
```

4 | 947 | 2 | 919 |

```
Tenderloin
                          1340.48334
|Haight Ashbury
                          175.86667
                          |155.8
|Bayview Hunters Point
|Financial District/South Beach|135.51666
|Pacific Heights
                          1129.01666
|Potrero Hill
                          1109.8
|Inner Sunset
                          106.13333
|South of Market
                          194.71667
+----+
only showing top 10 rows
```

```
+----+
|Week|count|
+----+
|22 |259 |
|40 |255 |
|43 |250 |
|25 |249 |
|1 |246 |
+----+
only showing top 5 rows
```

```
(fire_ts_df
              .select('Neighborhood','ZipCode')
              .where((F.col('Zipcode') == 94102) | (F.col('Zipcode') == 94103) | (F.col('Zipcode') == 94103)
              \rightarrowcol('Zipcode') == 94110))
              .groupBy('Neighborhood','Zipcode')
              .count()
              .orderBy('count',ascending=False)
              .show(n=5, truncate=False)
          +----+
          |ZipCode|count|
          +----+
          194102 | 21840 |
          194103 | 20897 |
          |94110 |14801| |
          |94109 |14686|
          |94124 | |9236 |
          +----+
          only showing top 5 rows
          |Neighborhood |Zipcode|count|
          +----+
          lTenderloin
                                               |94102 |17084|
          |South of Market|94103 |13762|
          Mission
                                                194110 | 10444 |
          Mission
                                               |94103 |5445 |
          |Bernal Heights | 94110 | 3109 |
          +----+
          only showing top 5 rows
[]: | # 7)
                                        How can we use Parquet files or SQL tables to store this data and
             →read it back? ANSWER: The below code stores a dbfs file path as a path
              \rightarrowvariable, and then executes the code. The code itself first repartitions the
              →data to avoid creating small, sharded data files for each unit type. Then, the
              →code underneath creates an individual parquet file for each UnitType, saving
              \rightarroweach parquet file in the path variable specified. The same execution could
              →also be represented in SQL to write a table to the hive_metastore. To do this, ⊔
              \rightarrowall that would really change is the last line of code, instead running a_{\sqcup}
             →commmand that looks like .saveAsTable('hive_metastore_schema').
            path = "dbfs:/FileStore/Merrimack/Week_4/data_parquet_week4"
            (fire_df
```

```
.repartition("UnitType")
.write.format('parquet')
.partitionBy("UnitType")
.mode('overwrite')
.option("header", "true")
.save(path)
)
```