Members:

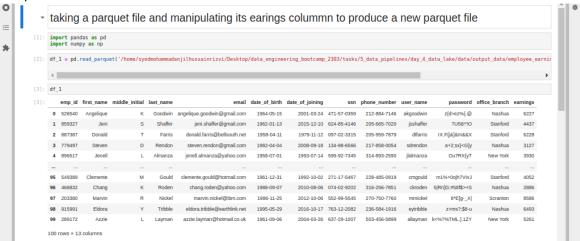
- 1. Syed Mohammad Anjil Hussain Rizvi (2303.KHI.DEG.031)
- 2. Huzefa Anver (2303.KHI.DEG.002)

Simulate earning predictions for two more days:

Manipulations to the data are made in Untitled.ipynb file. It can be found inside the same folder as this pdf file.

Data has been manipulated using pandas. The manipulation we are talking about Is simply multiplying the earnings column with 4.

Parquet file loaded:

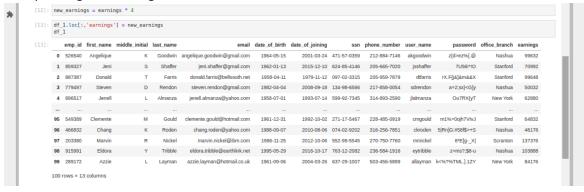


Accessing the earnings column:

```
O [4]: earnings= df_l.loc[:,'earnings'] earnings

[4]: 0 6227
1 4437
2 6228
3 3127
4 3939
...
95 4852
96 2886
97 8586
98 6493
99 5261
Name: earnings, Length: 100, dtype: int64
```

Manipulating the earnings column



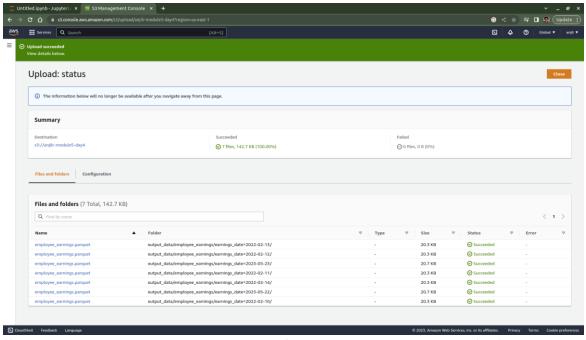
Creating a new parquet file:



All of these steps were repeated on another existing parquet file to create one more parquet file for another day.

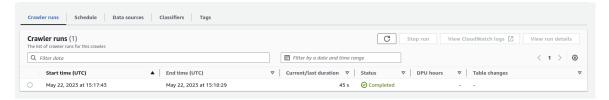
Data lake preparation:

1. Contents loaded into an s3 bucket:

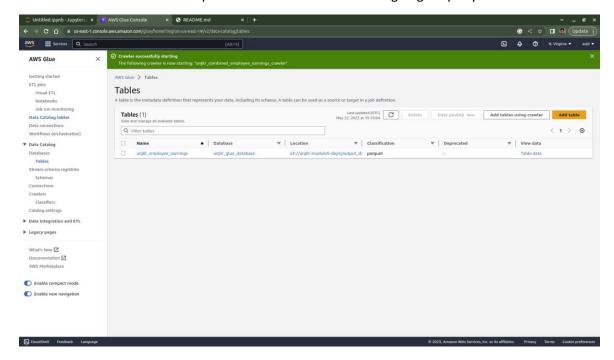


(folders with year 2023 in the title are the new folders containing new parquet files)

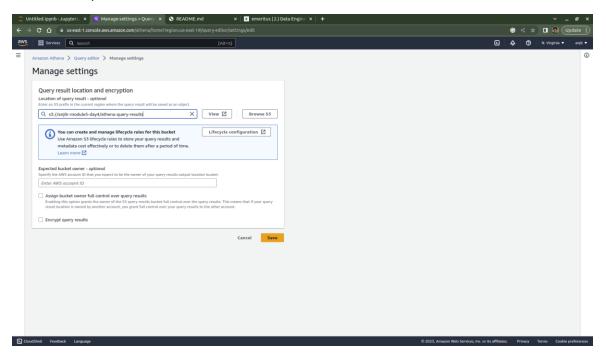
2. Crawler created and spun up:



3. We can see that it has produced our table that we are going to query from:

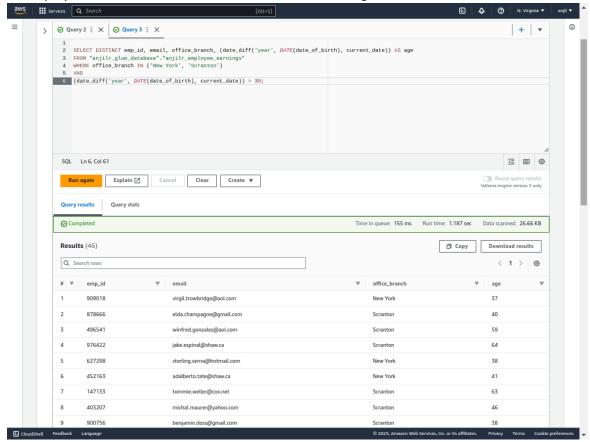


4. Query result location set inside s3 bucket:



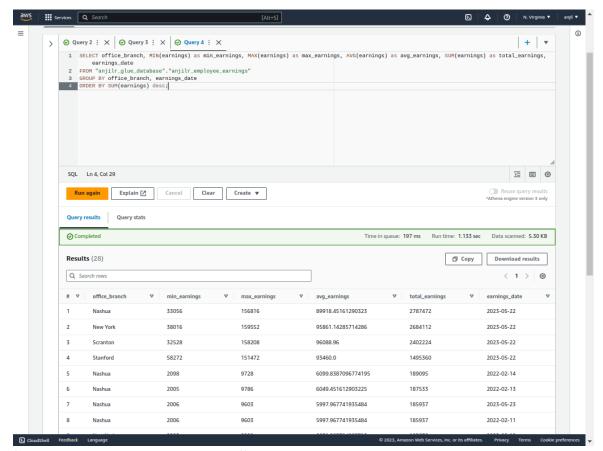
Querying using Athena:

All employees from offices 'New York' and 'Scranton' with age > 30:



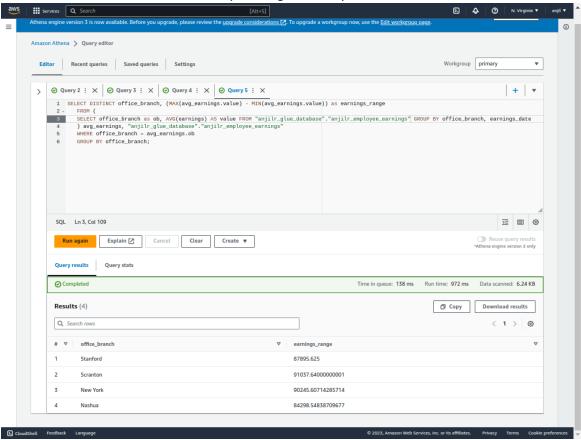
(no changes anticipated between this run and queries run before manipulation of data)

Min, max, average and total earnings for each office and each day - sorted by total earnings, highest to lowest:



(in this one however we can expect different query results between this run and the same query run before data manipulation; this query is related to earnings which is why we can expect a difference. In fact, we can already see the data points from new parquet files being reflected in the first few rows of the query)

Difference between worst and best day earnings for every office branch:



Assignment Query:

Create a new query in Athena that calculates the % change in earnings for every employee from a given day compared to the previous day.



⊘ Com	pleted							Time in queue: 156 m	Run time: 852 ms Data scanned: 2.62 KB
Results (100) Q. Search rows									Download results < 1 > ③
# ▽	emp_id	▽	earnings_date	▽	earnings	∇	previous_earnings	▽	percentage_change
1	138911		2022-02-14		6709		6112		9.767670157068062
2	143711		2022-02-14		8447		9462		-10.72711900232509
3	147133		2022-02-14		6348		6502		-2.368501999384805
4	149972		2022-02-14		4881		7841		-37.750286953194745
5	155097		2022-02-14		3945		8825		-55.297450424929174
6	160938		2022-02-14		3469		9033		-61.596368869699994
7	163409		2022-02-14		5323		7281		-26.891910451861005
8	170637		2022-02-14		8950		8601		4.057667713056621
9	174955		2022-02-14		8857		2409		267.6629306766293
10	184257		2022-02-14		5190		8394		-38.17012151536812
11	203380		2022-02-14		7741		9635		-19.6574987026466
12	215719		2022-02-14		6625		9023		-26.576526654106175
13	220965		2022-02-14		9378		6721		39.532807617914
14	233136		2022-02-14		6499		8704		-25.333180147058826
15	235295		2022-02-14		5760		7327		-21.38665210863928
16	242388		2022-02-14		3467		8227		-57.85827154491309
17	289172		2022-02-14		5868		4817		21.818559269254724
18	299088		2022-02-14		4627		7951		-41.80606213054961
19	312726		2022-02-14		6055		3109		94.75715664200708
20	314661		2022-02-14		8483		4480		89.35267857142857