

Day35 - March 11th 2024

1. Started my day as usual
2. Packed food and headed to library
3. Started marketing my profile for data engineer
4. Solved one medium leetcode problem

The screenshot shows a LeetCode submission for the 'Group Anagrams' problem. The submission is accepted with a runtime of 92ms and memory of 21.31 MB. The code is a Python3 solution using a defaultdict to group anagrams. The explanation section provides a detailed breakdown of the algorithm.

Runtime: 92 ms
Beats 45.98% of users with Python3

Memory: 21.31 MB
Beats 27.80% of users with Python3

Code:

```
class Solution:
    def groupAnagrams(self, strs: List[str]) -> List[List[str]]:
        res = {}
        for s in strs:
            count = [0]*26 #a...z
            for i in s:
                count[ord(i)-ord("a")] += 1
            key = tuple(count)

            if key in res:
                res[key].append(s)
            else:
                res[key] = [s]

        return list(res.values())
```

Explanation:

1. Here the idea is to use count of each char in string as key and the original char as value
2. So for example if we consider "eat", "tea" there count is same in count[0*26] so we will make this as key and append its string which serves as a value
3. The output of res would look like this:
defaultdict(- 4. In the end we just return values of our dict res

Testcase: **Test Result**

Accepted Runtime: 69 ms

Case 1 **Case 2** **Case 3**

Input:

```
strs =
```

5. Started learning spark practically

The collage illustrates the practical learning of Spark through video, hands-on notebooks, and documentation.

Top Left: YouTube Video
Title: spark practical (DataFrame API)
Channel: MANISH KUMAR
The video shows a Databricks notebook with the following code:

```
1 df.write.format("csv")\
2   .option("header", "true")\
3   .option("mode", "overwrite")\
4   .option("path", "/FileStore/tables/partition_by_address/")\
5   .partitionBy("address")\
6   .save()
```

It also displays Spark job output:

```
1 dbutils.fs.ls("/FileStore/tables/employee_write_data.csv")
Out[13]: True
```

Top Right: Databricks Notebook
Section: Partitioning and Bucketing
Code:

```
1 #partition by address(country)
2 df.write.format("csv")\
3   .option("header", "true")\
4   .option("mode", "overwrite")\
5   .option("path", "/FileStore/tables/partition_by_address/")\
6   .partitionBy("address")\
7   .save()
```

Spark job output:

```
1 dbutils.fs.ls("/FileStore/tables/partition_by_address/")
```

Bottom: Google Docs Document
Title: Spark_Practical_Day4
Section: 21. Expression
Sub-section: Expressions in PySpark DataFrames
Text: Expressions are powerful tools used to manipulate, transform, and analyze data within DataFrames. They can be constructed using various components:

- Column References:** These directly refer to existing columns within the DataFrame.
- Operators:** PySpark supports a rich set of operators for arithmetic calculations, comparisons, string manipulations, and more (e.g., `+`, `-`, `*`, `/`, `==`, `!=`, `LIKE`, `CONCAT`).
- Functions:** PySpark offers a wide range of built-in functions for data processing and analysis (e.g., `abs`, `sqrt`, `lower`, `upper`, `count`, `avg`, `sum`).
- SQL Expressions:** The `expr()` function allows you to write SQL-like expressions directly within PySpark (useful for functions or operations not readily available as built-in functions).

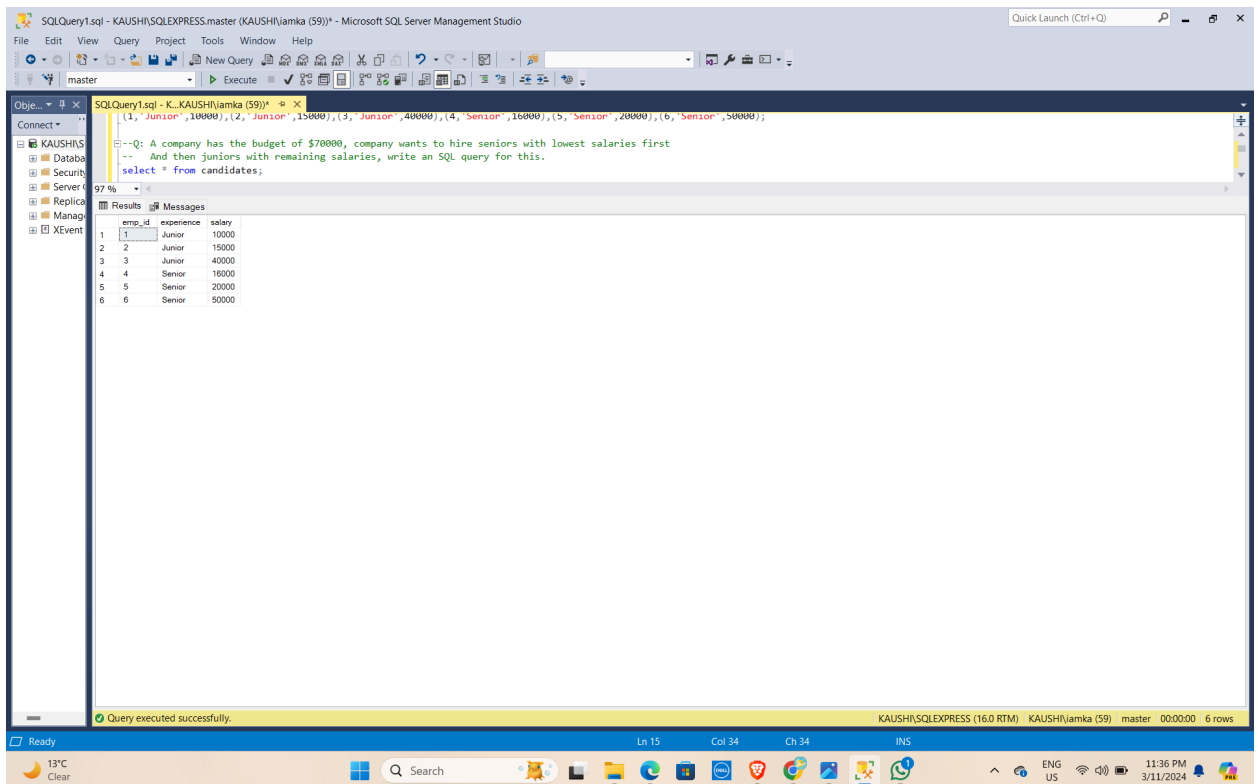
Code example for SQL expression:

```
1 employee_df.select(expr("id + 5")).show()
```

Output:

```
[1] Spark Jobs
-----
[(id + 5)]
-----
|
| 6|
| 7|
| 8|
| 9|
|10|
-----
```

6. Ended my day by solving complex SQL questions from Ankit'YT

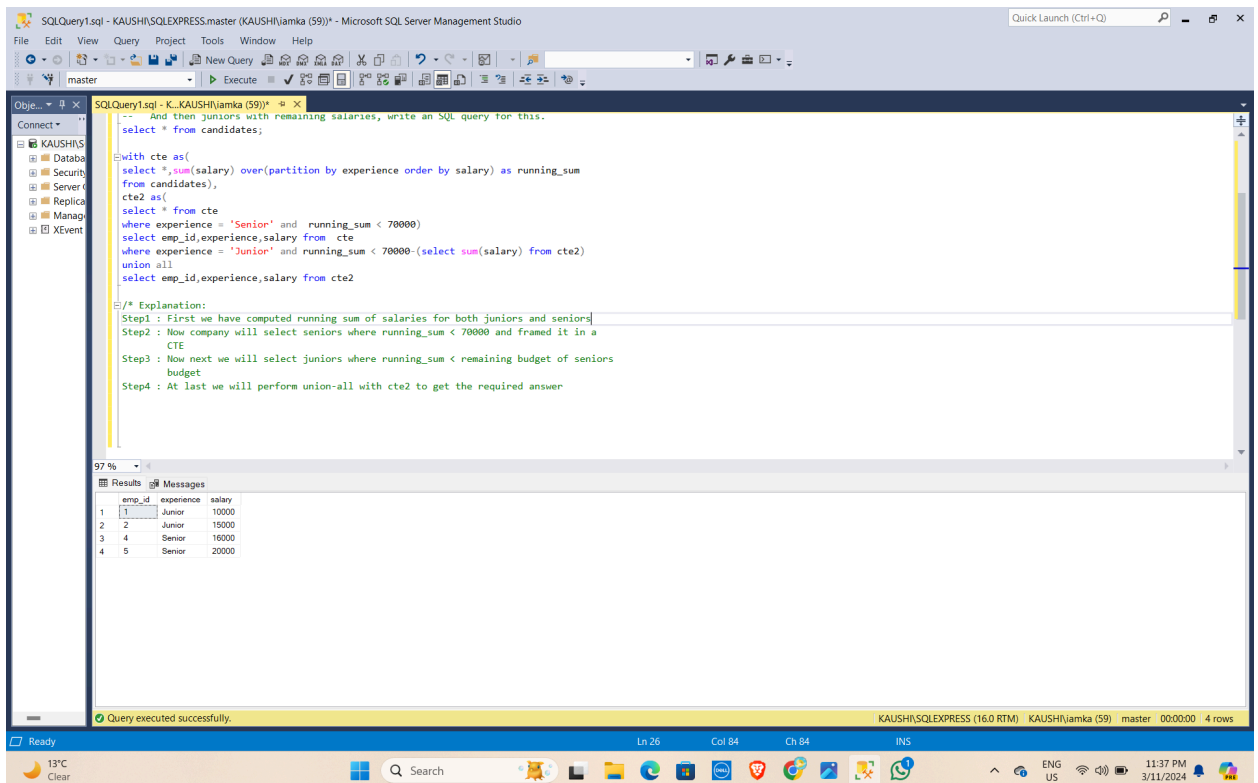


SQLQuery1.sql - KAUSHI\SQLEXPRESS:master (KAUSHI\jamka (59)) - Microsoft SQL Server Management Studio

```
--Q: A company has the budget of $70000, company wants to hire seniors with lowest salaries first
-- And then juniors with remaining salaries, write an SQL query for this.
select * from candidates;
```

emp_id	experience	salary
1	Junior	10000
2	Junior	15000
3	Junior	40000
4	Senior	16000
5	Senior	20000
6	Senior	50000

Query executed successfully.



SQLQuery1.sql - KAUSHI\SQLEXPRESS:master (KAUSHI\jamka (59)) - Microsoft SQL Server Management Studio

```
-- And then juniors with remaining salaries, write an SQL query for this.
select * from candidates;

with cte as(
select *,sum(salary) over(partition by experience order by salary) as running_sum
from candidates),
cte2 as(
select * from cte
where experience = 'Senior' and running_sum < 70000)
select emp_id,experience,salary from cte
where experience = 'Junior' and running_sum < 70000-(select sum(salary) from cte2)
union all
select emp_id,experience,salary from cte2

/* Explanation:
Step1 : First we have computed running sum of salaries for both juniors and seniors
Step2 : Now company will select seniors where running_sum < 70000 and framed it in a CTE
Step3 : Now next we will select juniors where running_sum < remaining budget of seniors budget
Step4 : At last we will perform union-all with cte2 to get the required answer
```

emp_id	experience	salary
1	Junior	10000
2	Junior	15000
4	Senior	16000
5	Senior	20000

Query executed successfully.