

# CS562 MF Query Evaluator Final Project

Aaron Camacho and Walker Bove

# Introduction

This presentation will describe the code that makes up our MF Query Evaluator, as well as provide a few sample queries to demonstrate how it works.

The Evaluator was written in Python, and generates Python code files that access the database and compute the results. Due to the simplicity of Python, the entire evaluator exists within a single Python file.

# Purpose

While the SQL language can express... creating subgroups within the GROUP BY clause can become very wordy and inefficient, requiring the creation of multiple tables/views and joins (a notoriously expensive operation).

The MF Query Generator was designed to compute these queries without the view creation and expensive join costs normally associated with these types of queries.

# Code Overview

# Query 1

“Find for each customer the average sale in ‘NY’, the average sale in ‘CT’ and the average sale in ‘NJ’, if New York’s average is greater than the other two”

MF Input:

S: cust, avg\_1\_quant, avg\_2\_quant, avg\_3\_quant

N: 3

V: cust

F: avg\_1\_quant, avg\_2\_quant, avg\_3\_quant

Θ: record["state"]=='NY', record["state"]=='NJ',  
record["state"]=='CT'

G: aggs.avg\_1\_quant[0]/aggs.avg\_1\_quant[1] >  
aggs.avg\_2\_quant[0]/aggs.avg\_2\_quant[1] and  
aggs.avg\_1\_quant[0]/aggs.avg\_1\_quant[1] >  
aggs.avg\_3\_quant[0]/aggs.avg\_3\_quant[1]

1 → NY

2 → NJ

3 → CT

# Output and Equivalent SQL Query

# Query 2

“Find for each customer the total number of sales for apples and the total number of sales for cherries if the customer has purchased more apples than cherries”

MF Input:

S: cust, sum\_1\_quant, sum\_2\_quant

N: 2

V: cust

F: sum\_1\_quant, sum\_2\_quant

$\Theta$ : record["prod"]=="Apple",  
record["prod"]=="Cherry"

G: aggs.sum\_1\_quant > aggs.sum\_2\_quant

1  $\rightarrow$  Apples

2  $\rightarrow$  Cherries

# Output and Equivalent SQL Query



# Query 3

“For each customer and product combination, find the minimum amount of product purchased in 2017, 2018, and 2019”

MF Input:

S: cust, prod, min\_1\_quant, min\_2\_quant,  
min\_3\_quant

N: 3

V: cust, prod

F: min\_1\_quant, min\_2\_quant, min\_3\_quant

$\Theta$ : record["year"]==2017, record["year"]==2018,  
record["year"]==2019

G: True

1  $\rightarrow$  2017

2  $\rightarrow$  2018

3  $\rightarrow$  2019

# Output and Equivalent SQL Query

# Query 4

“For each customer and product,  
find the average sales quantities  
for each sales quarter”

MF Input:

S: cust, prod, avg\_1\_quant, avg\_2\_quant, avg\_3\_quant,  
avg\_4\_quant

N: 4

V: cust, prod

F: avg\_1\_quant, avg\_2\_quant, avg\_3\_quant, avg\_4\_quant

$\Theta$ : record["month"]>=1 and record["month"]<=3,  
record["month"]>=4 and record["month"]<=6,  
record["month"]>=7 and record["month"]<=9,  
record["month"]>=10 and record["month"]<=12

G: True

1 → Quarter 1

2 → Quarter 2

3 → Quarter 3

4 → Quarter 4

# Output and Equivalent SQL Query

# Things to Improve

- Support for EMF queries
- Error checking (presence of tables, valid aggregation, etc.)
- Minimal scanning detection (instead of all grouping variables being done one at a time)
- More advanced input processing (make the input rely less on existing Python syntax)