# Data Engineer Case Interview

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# Section 1: Pipeline

Language: Python3

Tools used: VisualStudioCode, Apache Airflow, Astro CLI, Docker

```
python_task = PythonOperator()
    task_id='Section1', python_callable=process_datasets,
    dag=dag,

# Define the DAG

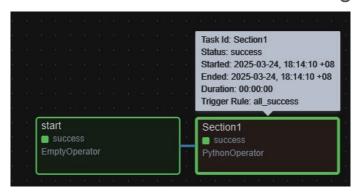
dag = DAG(
    'Section1', # DAG name
    default_args=default_args,
    description='process and output data files on daily interval',
    schedule_interval='0 1 * * * *', # Run every day at 1 AM
    catchup=False, # Don't backfill past DAG runs
)
```



### Dataset processing function

```
def process datasets():
   # Read Datasets-----
  df1 = pd.read csv('/opt/airflow/csv files/dataset1.csv')
  df2 = pd.read csv('/opt/airflow/csv files/dataset2.csv')
   print(df1.head())
  print(df2.head())
   # Process dataset1.csv-------
   # Clean NaN or empty names and price = 0
  df1 cleaned = df1[(df1['name'].notna()) & (df1['name'] != '') & (df1['price'] != 0)]
   # Split name into first name and last name
  df1 cleaned[['first name', 'last name']] = df1 cleaned['name'].str.split(' ', expand=True)
   # Remove name column
  df1 cleaned = df1 cleaned.drop(columns=['name'])
   # Reorder columns
  df1 cleaned = df1 cleaned[['first name', 'last name', 'price']]
  df1 cleaned['above 100'] = df1 cleaned['price'] > 100
   print(df1 cleaned.head(10))
```

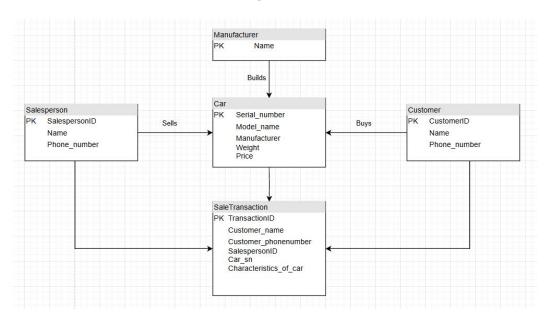
#### Successful execution of the dag



# Section 2: Database

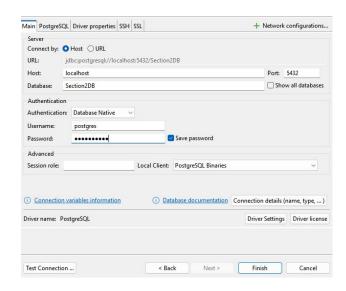
Language: PosgreSQL

Tools used: Docker Image, DBeaver, Draw.io

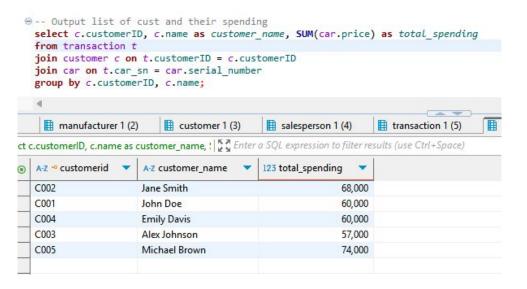


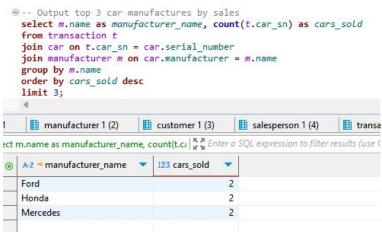
### Building the postgresql with docker image

```
C:\Users\daryl\OneDrive\Desktop\Job Applications 2025\SkillsFuture\Section2>docker build -t section2_postgres_image .
[+] Building 0.4s (7/7) FINISHED
                                                                                                   docker:desktop-linux
=> [internal] load build definition from Dockerfile
 => => transferring dockerfile: 340B
                                                                                                                   0.05
=> WARN: SecretsUsedInArgOrEnv: Do not use ARG or ENV instructions for sensitive data (ENV "POSTGRES_PASSWORD")
                                                                                                                   0.0s
 => [internal] load metadata for docker.io/library/postgres:latest
                                                                                                                   0.05
=> [internal] load .dockerignore
                                                                                                                   0.05
=> => transferring context: 2B
                                                                                                                   0.0s
=> [internal] load build context
                                                                                                                   0.05
=> => transferring context: 37B
                                                                                                                   0.05
=> [1/2] FROM docker.io/library/postgres:latest
                                                                                                                   0.05
=> CACHED [2/2] COPY tableScript.sql /docker-entrypoint-initdb.d/
                                                                                                                   0.05
 => exporting to image
                                                                                                                   0.1s
 => => exporting layers
                                                                                                                   0.05
=> => writing image sha256:17bf387356d2951fa1b00ca47e0ac048d62ef7af790b8d97ff3fcee847d69ada
                                                                                                                   0.05
=> => naming to docker.io/library/section2_postgres_image
                                                                                                                   0.05
View build details: docker-desktop://dashboard/build/desktop-linux/desktop-linux/uy6r4ceweo0gvljegd4eih7j2
1 warning found (use docker --debug to expand):
 - SecretsUsedInArgOrEnv: Do not use ARG or ENV instructions for sensitive data (ENV "POSTGRES_PASSWORD") (line 5)
C:\Users\dary\\OneDrive\Desktop\Job Applications 2025\SkillsFuture\Section2>docker run --name section2_postgres_containe
r -p 5432:5432 -d section2_postgres_image
73f7d7e1abad41b3c8d20c646294fd5a6643fe86aa0a1106cfb1f7c75f3ff89d
```



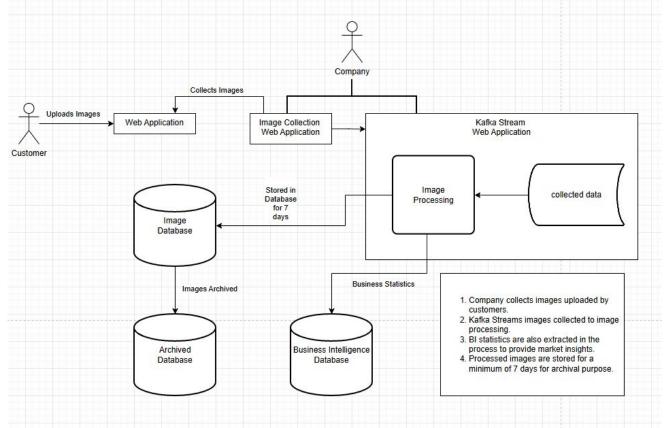
### Query 1 and 2 outputs





# Section 3: System Design

Tools used: Draw.io



# Section 4: API and Graph

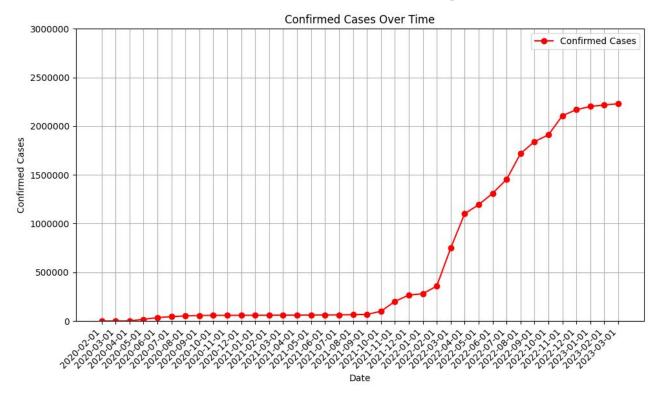
Language: Python3

Tools used: requests, datetime, dateutil, matplotlib

```
from datetime import datetime, timedelta
from dateutil.relativedelta import relativedelta
 of get covid data(start date, end date):
  covid data = []
  current date = start date
  while (current date <= end date):
       formatted date = current date.strftime('%Y-%m-%d')
       params = {
           "date": formatted_date,
      response = requests.get(url, params=params)
       if response.status code == 200:
          data = response.json().get("data", {})
          confirmed = data.get("confirmed", "0")
          covid data.append({"date": formatted date, "confirmed": confirmed})
          print(formatted date, confirmed)
          print(f"Error: {response.status code}")
          print(response.text)
      current date += relativedelta(months=1)
   return covid data
```

```
start_date = datetime(2020, 2, 1)
end date = datetime(2023, 3, 1)
data = get covid data(start date, end date)
# import plot lib
 import matplotlib.pyplot as plt
dates = [entry["date"] for entry in data]
values = [entry["confirmed"] for entry in data]
plt.figure(figsize=(10, 6))
plt.plot(dates, values, marker='o', color='r', linestyle='-', label="Confirmed Cases")
# Customizing the plot
plt.title('Confirmed Cases Over Time')
plt.xlabel('Date')
plt.ylabel('Confirmed Cases')
plt.xticks(rotation=45, ha='right') # Rotate the date labels for better readability
plt.ylim(0, 3000000)
plt.ticklabel format(style="plain", axis="y")
plt.grid(True)
plt.tight layout()
plt.legend()
plt.show()
```

Displayed Graph of Number of cases overtime in singapore in months



# Section 5: ML

Language: Python3

Tools used: SciKit-Learn, Jupyter, Visual Studio Code

0.05

0.04

0.03

1.00

1.00

- 0.8

- 0.6

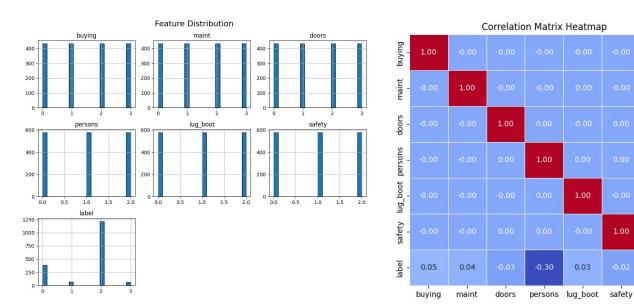
- 0.4

- 0.2

- 0.0

- -0.2

#### Visualisation to better understand dataset



### Label Encoding to numerical values

```
Click to add a breakpoint al data to numerical data
   le = LabelEncoder()
   # encoding to each categorical column
 vfor column in df.columns:
       df[column] = le.fit_transform(df[column])
   # printing df
   print(df.head())
   print(df.tail())
 ✓ 0.0s
  buying maint doors persons lug boot safety label
                    doors persons lug boot safety
                                                     label
1724
1726
1727
```

### Selecting features and data splitting

```
# import libs for model training

√ from sklearn.svm import SVC

 from sklearn.model_selection import train_test_split, GridSearchCV
 from sklearn.metrics import accuracy_score
 # Split data
 X = df.drop(columns=["buying", "persons"])
 y = df["buying"]
 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
  0.0s
           doors lug boot safety label
1724
1725
```

Training model and testing

Model used: Support Vector Machine (SVM) has best performing model compared to other models

```
# Model Training
   # Initialize the model
   svm = SVC(random state=42, kernel='linear')
   svm.fit(X train, y train)

√ 0.0s

                                    0 0
                 SVC
SVC(kernel='linear', random state=42)
   # Predict and evaluate
   y pred = svm.predict(X test)
   accuracy = accuracy score(y test, y pred)
   print(f"Accuracy: {accuracy:.2f}")

√ 0.0s

Accuracy: 0.26
```

### Predicting buying price

- Maintenance = High
- Number of doors = 4
- Lug Boot Size = Big
- Safety = High
- Class Value = Good

```
# Predict buying price
  # Load the model
  svm loaded = joblib.load("random forest model.pkl")
 ✓ 0.0s
   # Data for testing
   test = ['high', '4', 'big', 'high', 'good']
   enc = le.fit transform(test)
   enc_test = pd.DataFrame([enc], columns=['maint', 'doors', 'lug_boot', 'safety', 'label'])
  enc_test
 ✓ 0.0s
   maint doors lug_boot safety label
   pred = svm loaded.predict(enc test)
  pred_convert = unique_values[pred]
   print(pred)
  print(pred_convert)
 ✓ 0.0s
['vhigh']
```