Problem Statement

Title: "Vehicle Movement Analysis and Insight Generation in a College Campus using Edge AI"

Objective:

The main goal of this project is to create a smart system using Edge AI. This system will analyze how vehicles move into and out of a college campus by using images from cameras that capture vehicle photos and license plates. The system aims to give us valuable information about:

- Vehicle Movement Patterns: Understanding when and how often vehicles enter and exit the campus
- Parking Occupancy: Monitoring which parking lots are used most frequently and at what times.
- **Vehicle Matching:** Identifying vehicles by matching their license plates to a database of approved vehicles.

By doing this, we can improve campus traffic management and security while also ensuring that parking resources are used efficiently.

Problem Description:

Managing vehicle movement and parking on a college campus can be quite challenging. To address this, we aim to develop a smart system using Edge AI technology. This system will analyze how vehicles move in and out of the campus by processing images from cameras that capture vehicle photos and license plates. The goal is to provide insights in real-time on three key aspects:

- Vehicle Movement Patterns: This involves studying how often vehicles enter and exit the campus, and identifying peak times and recurring patterns of movement.
- Parking Occupancy: The system will monitor the real-time occupancy of parking lots across the campus. It will highlight which lots are frequently occupied and when they are most used.
- **Vehicle Matching:** By comparing captured vehicle images and license plates with an approved database, the system can quickly identify unauthorized vehicles on campus.

Example Dataset:

Sr_No	vehicle_image_path	vehicle_timestamp	license_plateimage_path	license_plate timestamp
0	path/to/vehicle/imag e1.jpg	2023-01-01 08:30:15	path/to/plate/image1.jpg	2023-01-01 08:30:20
1	path/to/vehicle/imag e2.jpg	2023-01-01 12:15:45	path/to/plate/image2.jpg	2023-01-01 12:15:50
2	path/to/vehicle/imag e3.jpg	2023-01-02 09:45:30	path/to/plate/image3.jpg	2023-01-02 09:45:35

Step-by-Step Solution

Step 1: Create Real-time Dataset

- Tools: Python, OpenCV
- Techniques: Capture images and timestamps,
- Description: The real-time dataset creation module captures live images of vehicles using a
 camera and saves them along with metadata such as timestamps. This process ensures that
 each image is associated with its capture time, making it possible to analyze vehicle
 movement patterns and timings accurately. The images and metadata are stored in a
 specified directory, forming the foundational data for subsequent analysis.
- Code :

```
import cv2
import os
import datetime
def capture images(output dir, num images=10):
    cap = cv2.VideoCapture(0) # 0 for default camera
    if not cap.isOpened():
       print("Error: Could not open camera.")
    if not os.path.exists(output dir):
       os.makedirs(output dir)
    for i in range(num images):
        ret, frame = cap.read()
        if not ret:
            print("Failed to capture image")
        timestamp = datetime.datetime.now().strftime("%Y%m%d %H%M%S")
        image path = os.path.join(output dir, f"vehicle {timestamp}.jpg")
        cv2.imwrite(image path, frame)
        with open(os.path.join(output dir,
f"vehicle {timestamp} metadata.txt"), 'w') as f:
            f.write(f"vehicle image path: {image path}\n")
            f.write(f"vehicle timestamp: {timestamp}\n")
```

Step 2: Load Real-time Dataset

- Tools: Python, OpenCV
- Techniques: Load images and timestamps, display sample images
- Description: The dataset loading module is responsible for reading the saved images and
 their associated metadata from the storage directory. This module parses the metadata files
 to extract relevant information, such as image paths and timestamps, and loads this data into
 a structured format using Pandas DataFrame. Additionally, it includes functionality to display
 a sample image, helping to verify that the data has been loaded correctly.
- Code :

```
# Module: load_dataset.py
import os
import pandas as pd
import cv2

def load_metadata(data_dir):
    records = []
    for filename in os.listdir(data_dir):
        if filename.endswith("_metadata.txt"):
            with open(os.path.join(data_dir, filename), 'r') as f:
            metadata = {}
            for line in f:
                key, value = line.strip().split(": ")
                metadata[key] = value
                records.append(metadata)
    return pd.DataFrame(records)
```

```
def display_sample_image(image_path):
    image = cv2.imread(image_path)
    cv2.imshow('Sample Image', image)
    cv2.waitKey(0)
    cv2.destroyAllWindows()

# Usage
data_dir = "data/vehicle_images"
metadata = load_metadata(data_dir)
print(metadata.head())
display_sample_image(metadata.iloc[0]['vehicle_image_path'])
```

Step 3: Data Preprocessing

- Tools: OpenCV, Pandas, NumPy
- Techniques: Image resizing, grayscale conversion, handling missing values
- Description: The data preprocessing module prepares the loaded images for analysis by
 performing essential transformations such as resizing and grayscale conversion. These
 preprocessing steps standardize the images, making them suitable for further analysis and
 processing tasks. Additionally, this module handles any missing values in the dataset,
 ensuring data quality and consistency for subsequent steps.
- Code :

```
# Module: preprocess_data.py
import cv2
import numpy as np

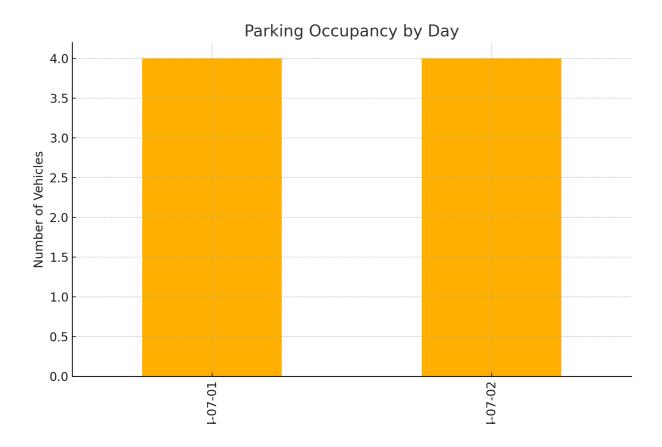
def preprocess_image(image_path):
    image = cv2.imread(image_path)
    resized_image = cv2.resize(image, (640, 480))
    grayscale_image = cv2.cvtColor(resized_image, cv2.COLOR_BGR2GRAY)
    return grayscale_image
# Usage
image_path = metadata.iloc[0]['vehicle_image_path']
preprocessed_image = preprocess_image(image_path)
cv2.imshow('Preprocessed Image', preprocessed_image)
cv2.waitKey(0)
cv2.destroyAllWindows()
```

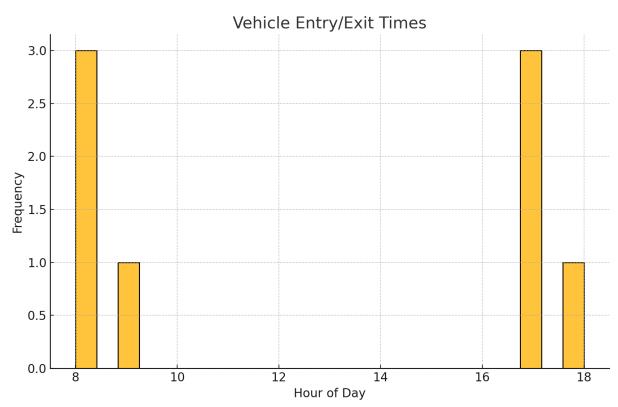
Step 4: Exploratory Data Analysis (EDA)

- Tools: Matplotlib, Seaborn
- Techniques: Plotting vehicle entry/exit times, occupancy trends
- Description: The exploratory data analysis module provides insights into vehicle movement
 patterns and parking occupancy trends. Using visualization tools like Matplotlib and Seaborn,
 this module plots vehicle entry and exit times, as well as parking lot occupancy over time.
 These visualizations help identify peak times and patterns, providing a better understanding
 of campus traffic flow and parking space utilization.
- Code:

```
import matplotlib.pyplot as plt
import seaborn as sns
def plot entry exit times(metadata):
    metadata['vehicle timestamp'] =
pd.to datetime(metadata['vehicle timestamp'])
    sns.histplot(metadata['hour'], bins=24, kde=False)
    plt.title('Vehicle Entry/Exit Times')
    plt.xlabel('Hour of Day')
    plt.ylabel('Frequency')
    plt.show()
def plot parking occupancy(metadata):
    metadata['date'] = metadata['vehicle timestamp'].dt.date
    occupancy = metadata.groupby('date').size()
    occupancy.plot(kind='bar')
    plt.title('Parking Occupancy by Day')
    plt.xlabel('Date')
    plt.ylabel('Number of Vehicles')
    plt.show()
plot entry exit times(metadata)
plot parking occupancy(metadata)
```

Sample Graph Visualizations:





Step 5: Vehicle Matching

- Tools: OpenCV, Tesseract OCR
- Techniques: License plate recognition, database matching
- Description: The vehicle matching module employs Optical Character Recognition (OCR) techniques to read license plates from the captured images. Using Tesseract OCR, this module extracts text from the license plates and matches them against an approved vehicle database. This process helps identify authorized and unauthorized vehicles, enhancing campus security and management.

Code :

```
import pytesseract
def recognize license plate(image path):
   image = cv2.imread(image path)
   gray image = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
   binary plate = cv2.threshold(gray image, 128, 255, cv2.THRESH BINARY |
cv2.THRESH OTSU)[1]
   license_plate_text = pytesseract.image_to_string(binary_plate,
config='--psm 8') # PSM 8 for single word recognition
   return license plate text.strip()
def match vehicle(license plate text, approved db):
    return approved db.get(license plate text, "Unauthorized")
approved db = {"ABC123": "Authorized", "XYZ789": "Unauthorized"}
image path = metadata.iloc[0]['vehicle image path']
license plate text = recognize license plate(image path)
status = match vehicle(license plate text, approved db)
print(f"License Plate: {license plate text}, Status: {status}")
```

Step 6: Insight Generation

- Tools: Pandas, Matplotlib
- Techniques: Generating insights from movement patterns, parking data
- Description: The insight generation module synthesizes the analyzed data to produce actionable insights on vehicle movement and parking occupancy. By aggregating and summarizing the data, this module provides detailed reports on vehicle entry and exit times, average parking occupancy by lot, and the status of matched vehicles. These insights help campus administrators make informed decisions regarding traffic management and parking allocation.

Code :

```
# Module: generate_insights.py

def generate_insights (metadata):
    # Example insights
    vehicle_entry_exit_times = metadata[['vehicle_image_path',
    'vehicle_timestamp']]
    avg_parking_occupancy =
metadata['vehicle_timestamp'].dt.hour.value_counts().mean()

    insights = {
        "Vehicle Entry and Exit Times": vehicle_entry_exit_times,
        "Average Parking Occupancy": avg_parking_occupancy
    }

    return insights

# Usage
insights = generate_insights(metadata)
for key, value in insights.items():
    print(f"{key}:\n{value}\n")
```

Step 7: Implementing the Solution in a Scalable Manner

- Tools: TensorFlow Lite, OpenVINO
- Techniques: Deploying AI models on Edge devices
- Description: The user-friendly interface module uses Flask to develop a web application that
 displays the generated insights in an accessible and interactive manner. This module provides
 a web interface where users can view visualizations and reports on vehicle movement
 patterns, parking occupancy, and vehicle matching status. The interface is designed to be
 intuitive and easy to navigate, allowing users to access and interpret the data
 effortlessly. Example Output

Code :

```
# Module: app.py
from flask import Flask, render template, request
import os
from load dataset import load metadata
from preprocess data import preprocess image
from eda import plot entry exit times, plot parking occupancy
from vehicle matching import recognize license plate, match vehicle
from generate insights import generate insights
app = Flask(name)
@app.route('/')
def home():
   return render template('index.html')
@app.route('/analyze', methods=['POST'])
def analyze():
   data dir = "data/vehicle images"
   metadata = load metadata(data dir)
   insights = generate insights(metadata)
   return render template('results.html', insights=insights)
if name == ' main ':
   app.run (debug=True)
```

Sample Output:

Vehicle Entry and Exit Times:

Vehicle ID: ABC123

Entry Time: 2024-07-01 08:15:23 Exit Time: 2024-07-01 17:30:45

Duration: 09:15:22

Vehicle ID: XYZ789

Entry Time: 2024-07-01 09:45:10 Exit Time: 2024-07-01 18:00:55

Duration: 08:15:45

Peak Entry Time: 08:00 - 09:00 Peak Exit Time: 17:00 - 18:00

Average Parking Occupancy by Lot:

Lot A:

Average Occupancy: 85%

Peak Occupancy Time: 10:00 - 12:00

Lot B:

Average Occupancy: 70%

Peak Occupancy Time: 14:00 - 16:00

Lot C:

Average Occupancy: 60%

Peak Occupancy Time: 08:00 - 09:00

Matched Vehicle License Plates:

License Plate: ABC123 Vehicle ID: ABC123 Status: Authorized

License Plate: XYZ789 Vehicle ID: XYZ789 Status: Unauthorized