# Vehicle Detection: Capstone Project

# Requirement

Identify all vehicles in an image — including buses, cars, and trucks — and draw bounding boxes around them.

# **Problem Overview**

Detecting vehicles on the road — and more critically, identifying their type — is a challenging task. It's akin to the complexity of distinguishing between a **cat** and a **dog** in an image. Once you can do that, you're halfway to solving the vehicle detection problem.

There are many pre-trained models like YOLO (You Only Look Once) and SSD (Single Shot Detector) that can detect and classify vehicles efficiently, often in as few as 10 lines of code.

However, our capstone project has a different requirement:

Build a vehicle detection model from scratch.

This is significantly more complex. Differentiating between a car and a traffic signal isn't trivial.

# The Real Challenge

Let's break it down:

#### Problem 1:

You can detect if an image contains a cat or dog. But...

Can you detect how many cats there are?

#### Problem 2:

Assuming you can identify a cat or dog, the next step is to draw a bounding box around each instance.

This involves:

- · Edge detection
- · Object localization
- · Drawing accurate bounding boxes

### Solution

While building your own model, you can still **train it using pre-trained architectures** like YOLO or SSD by preparing your own dataset.

#### Steps:

- 1. Prepare your dataset (images with vehicles).
- 2. Create annotations for each image, you need an annotation file. This is a 1:1 mapping.

#### **Annotation Details:**

Annotations are metadata that describe:

- · All vehicles in an image
- Their coordinates (x1, y1, x2, y2)
- Bounding box size (height, width)
- Positions (top-left, center, etc.)

#### **Annotation Formats:**

- .txt or .csv for YOLO
- .xml for SSD

## **How to Create Annotations**

- 1. Manual labeling (most accurate)
- 2. Labeling tools like LabelImg
  - o Upload one image at a time
  - o Draw bounding boxes manually
- 3. Automated scripts
- 4. Use YOLO pre-trained models to auto-label your dataset

**Note:** No tool can perfectly auto-annotate. Manual labeling still gives the best results.

# Final Steps

Once you have 1:1 annotations, you are 70% done.

Then, use **Keras** and **TensorFlow** to:

- Build your own object detection model
- Train it on your annotated dataset