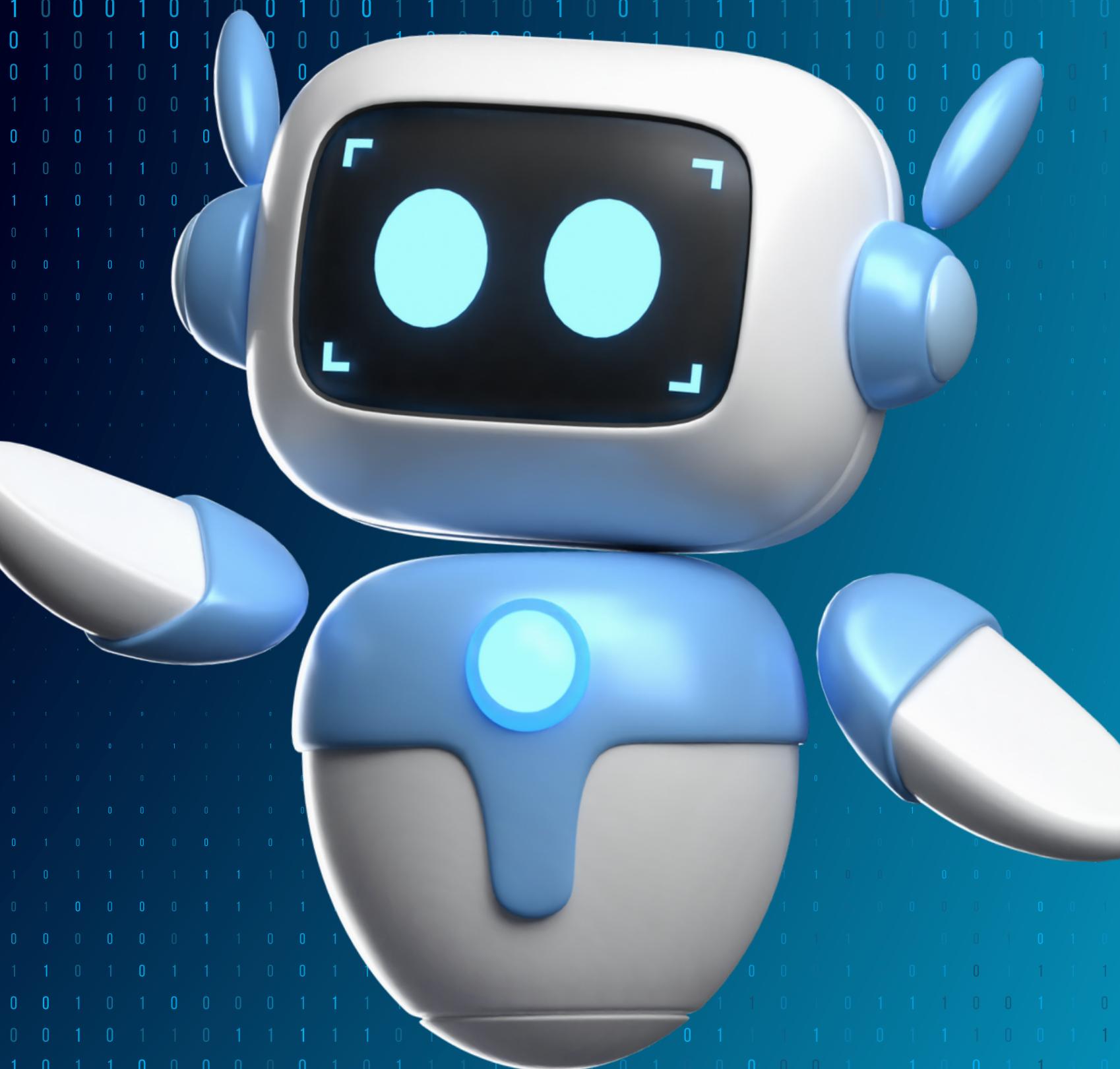


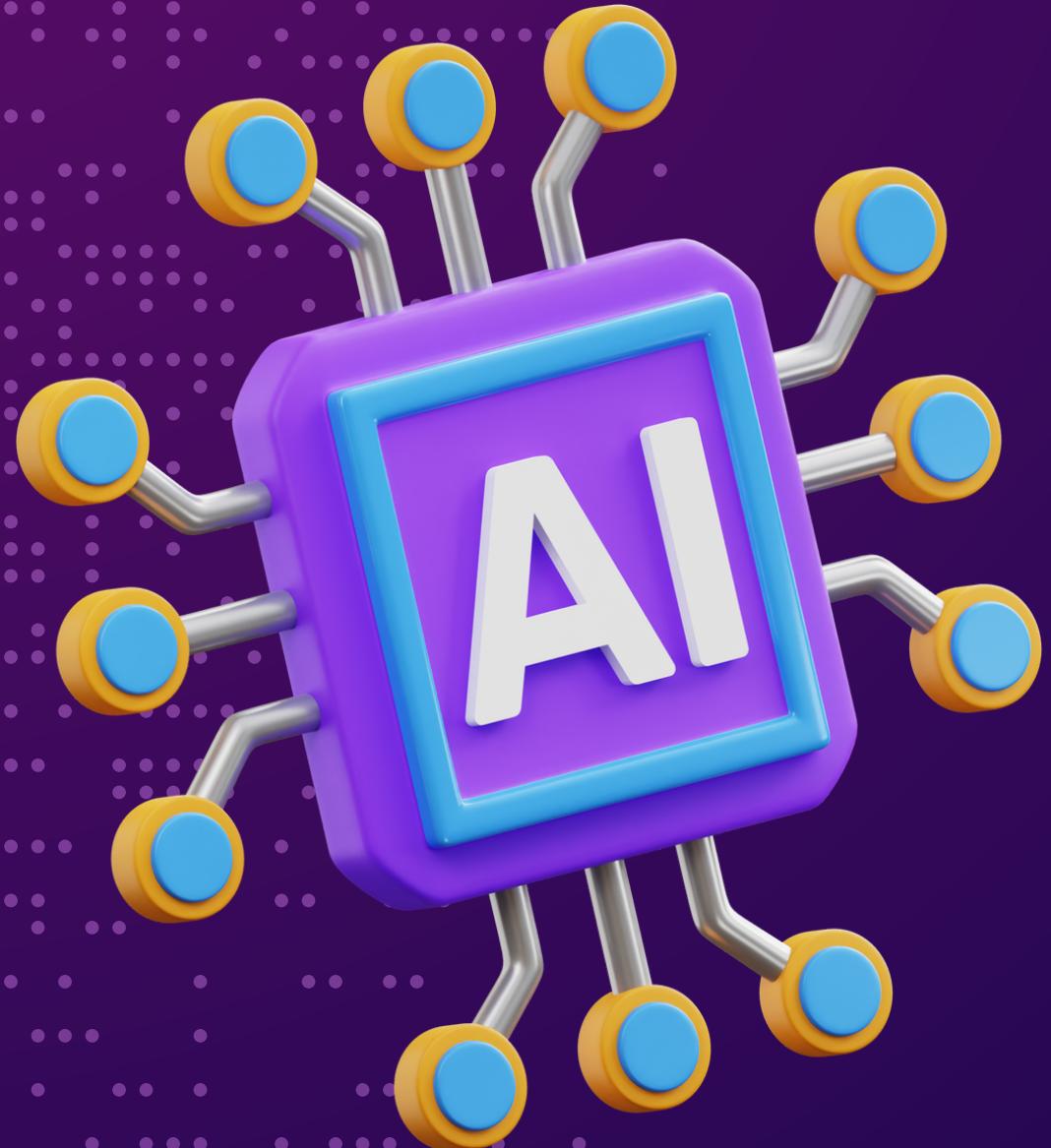
# YVR HACKATHON PRESENTATION

Fearless Trailblazers

Team: Shikha, Abdul, Bruce,  
Pahul



# CONTENTS

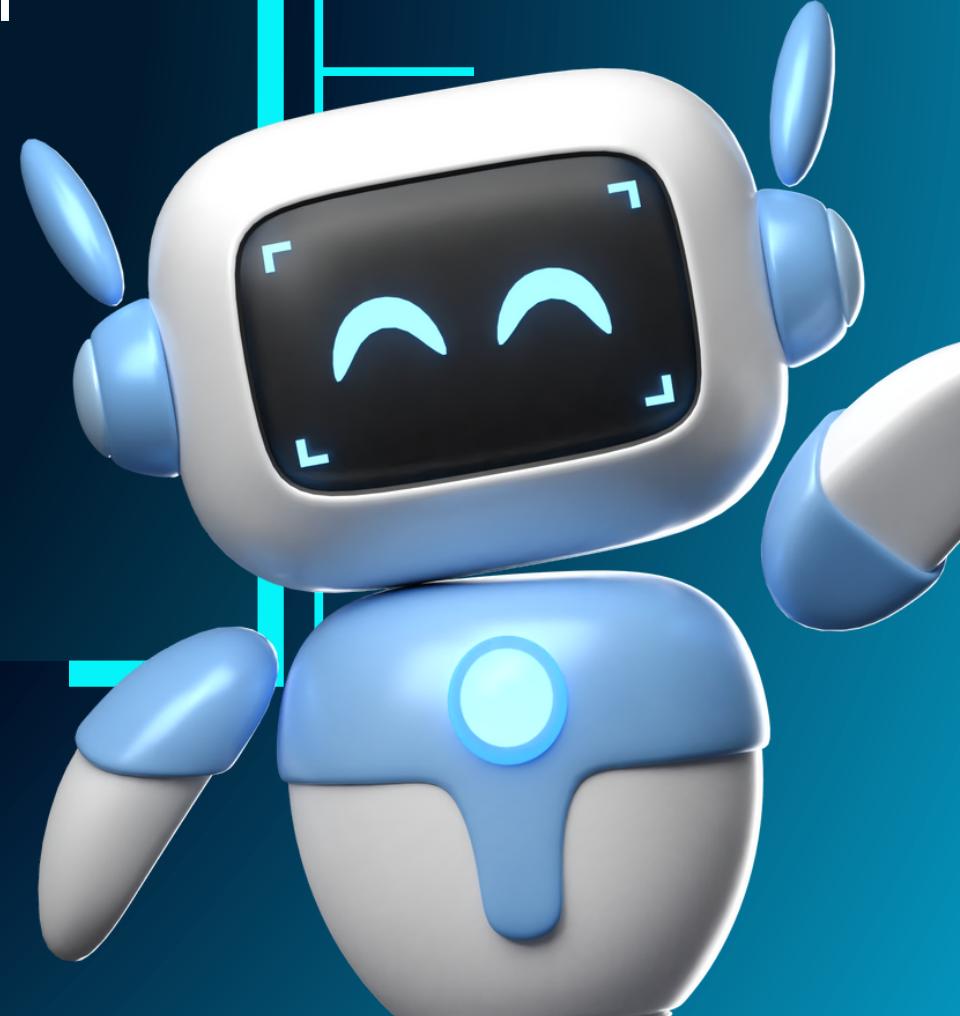


- ① Challenge
- ② Objective
- ③ Proposed Solution & Code Overview
- ④ Benefits
- ⑤ Next Steps

1

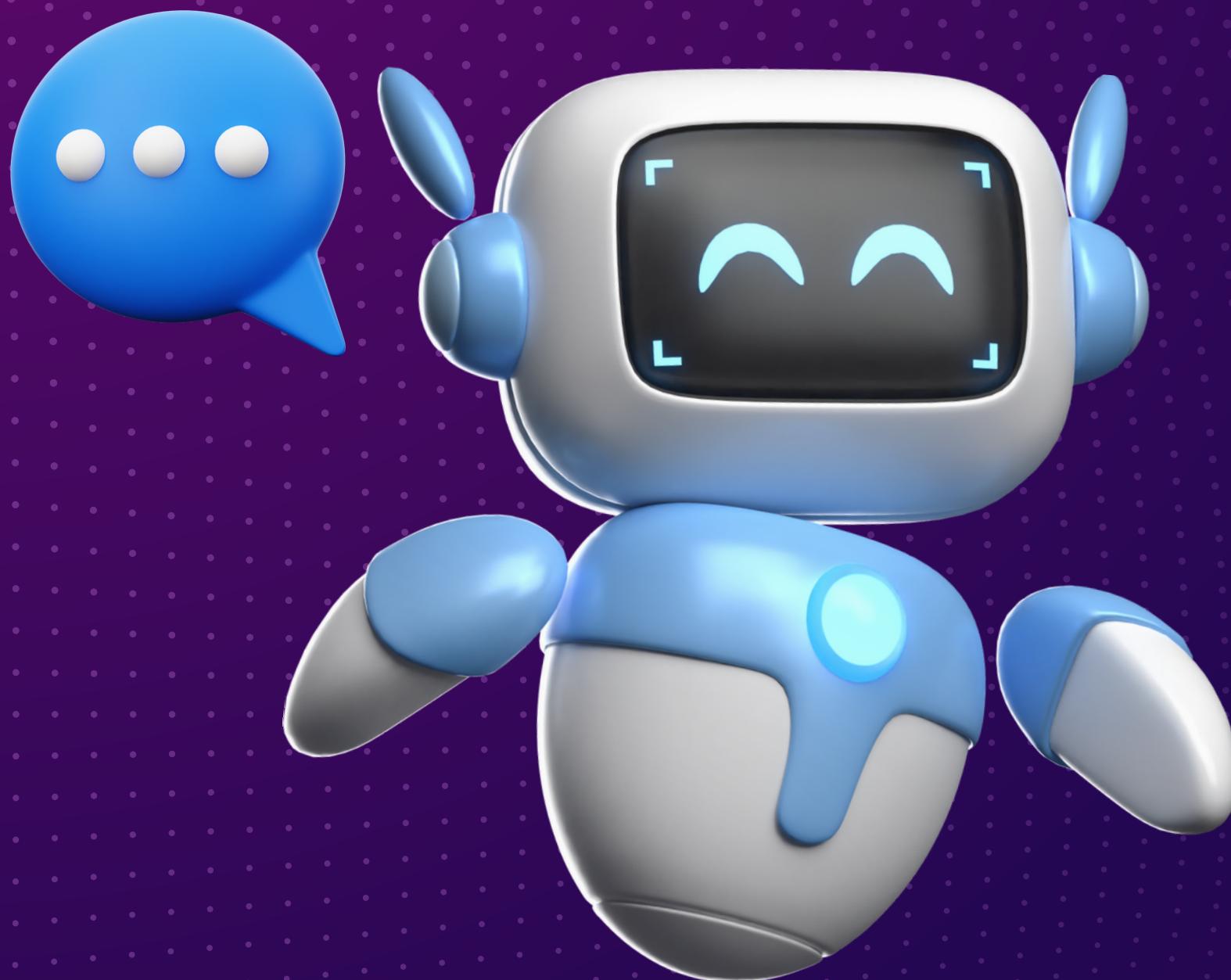
# OBJECTIVE

Utilize intelligent sensors, computer vision, and machine learning for real-time insights into cleanliness, unattended baggage, and passenger volume.



2

## PROPOSED SOLUTION



LEVERAGE YOLOV7

TRAIN SPECIALIZED MODELS

TARGETED DATASETS



## THE FEARLESS TRAILBLAZERS PROPOSE A SOLUTION LEVERAGING ADVANCEMENTS IN OBJECT DETECTION TO ENHANCE THE CROW'S NEST SYSTEM

### MACHINE LEARNING MODEL

We will train a YOLOv7 object detection model for trash detection using the TACO dataset and a similar approach for other datasets.

### TARGETED DATASETS

TACO: Trash Images (Fine-grained)  
COCO: Objects (Bags/Luggage)  
AVIDBAGS: Attended/Unattended Bags



# CODE OVERVIEW



# Set Up Environment

This section installs necessary Python packages for the project, including PyTorch and the Ultralytics YOLOv5 library, which are essential for machine learning and object detection tasks

```
!pip install torch torchvision
```

```
!pip install ultralytics
```

# Download the TACO Dataset

This part clones the TACO dataset repository from GitHub, changes the current directory to the TACO directory, and then runs a script to download the dataset images and annotations.

```
!git clone https://github.com/pedropro/TACO.git  
%cd TACO  
!python download.py
```

# 5 PREPARE DATASET FOR TRAINING

## SETTING THE DATASET PATH

```
dataset_path = './data/annotations.json'  
dataset_dir = os.path.dirname(dataset_path)
```

## LOADING ANNOTATIONS

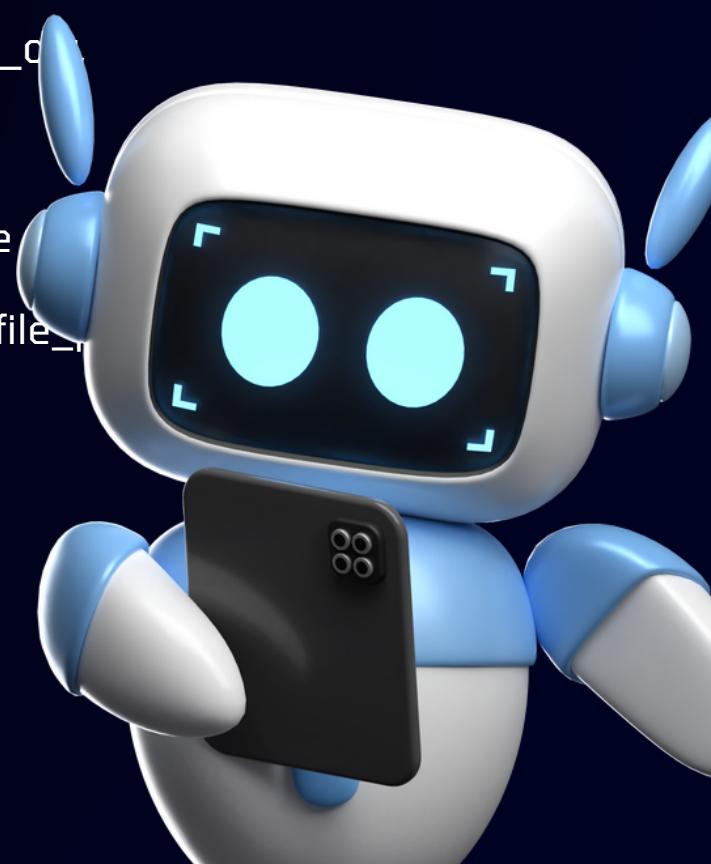
```
# Load annotations with open(dataset_path, 'r') as f:  
annotations = json.load(f) nr_images =  
len(annotations['images'])
```

## ITERATING OVER IMAGES

```
for i, image in enumerate(annotations['images']):  
    file_name = image['file_name'] url_original =  
    image['flickr_url'] file_path = os.path.join(dataset_dir,  
    file_name)
```

## DOWNLOADING AND SAVING IMAGES

```
if not os.path.isfile(file_path):# Load and Save Image  
    = requests.get(url_original) img =  
    Image.open(BytesIO(response.content)) img.save(file_
```



# GENERATE YOLO-FORMATTED LABELS

5

## LOADING ANNOTATIONS

```
with open('/content/TACO/data/annotations.json') as  
f: data = json.load(f)
```

## CREATING CATEGORY NAMES LIST

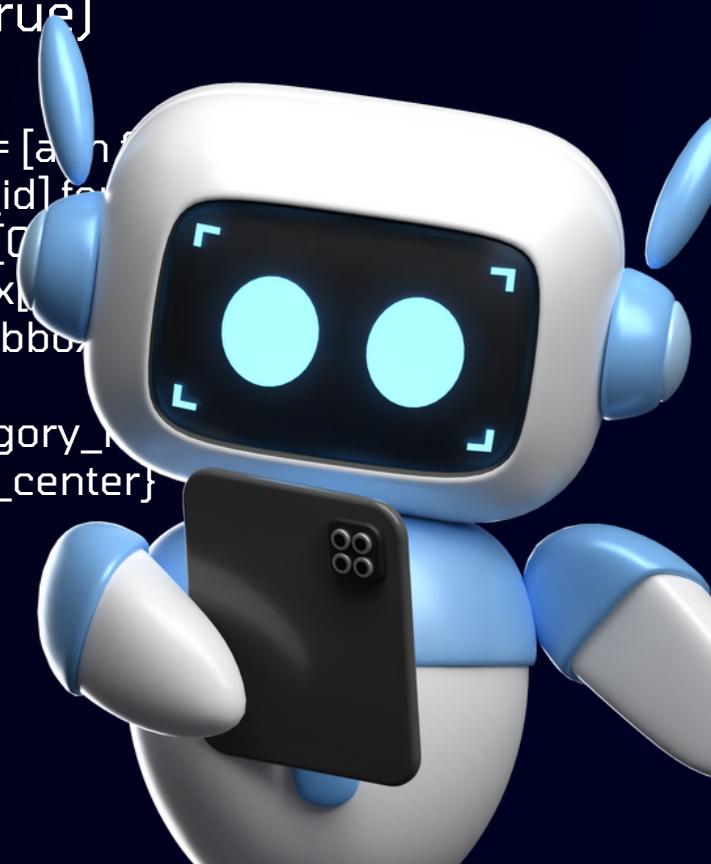
```
category_names = [cat['name'] for cat in  
sorted(data['categories'], key=lambda x: x['id'])]
```

## CREATING LABELS DIRECTORY

```
labels_base_dir = '/content/TACO/labels'  
os.makedirs(labels_base_dir, exist_ok=True)
```

## WRITING YOLO-FORMAT LABELS

```
with open(label_path, 'w') as label_file: annotations = [a for  
ann in data['annotations'] if ann['image_id'] == img_id for  
a in ann['bbox']] x_center = (bbox[0] + bbox[2] / 2) / img['width'] y_center = (bbox[1] + bbox[3] / 2) / img['height'] width = bbox[2] / img['width'] height = bbox[3] / img['height'] class_id = category_names.index(data['categories'][ann['category_id'] - 1]['name']) label_file.write(f"{class_id} {x_center} {y_center} {width} {height}\n")
```



# CONFIGURE YOLOV5 FOR TRAINING

5

## GENERATING CLASS NAMES

```
class_names = [f'class{i+1}' for i in range(60)]
```

## CONVERTING CLASS NAMES TO YAML FORMAT

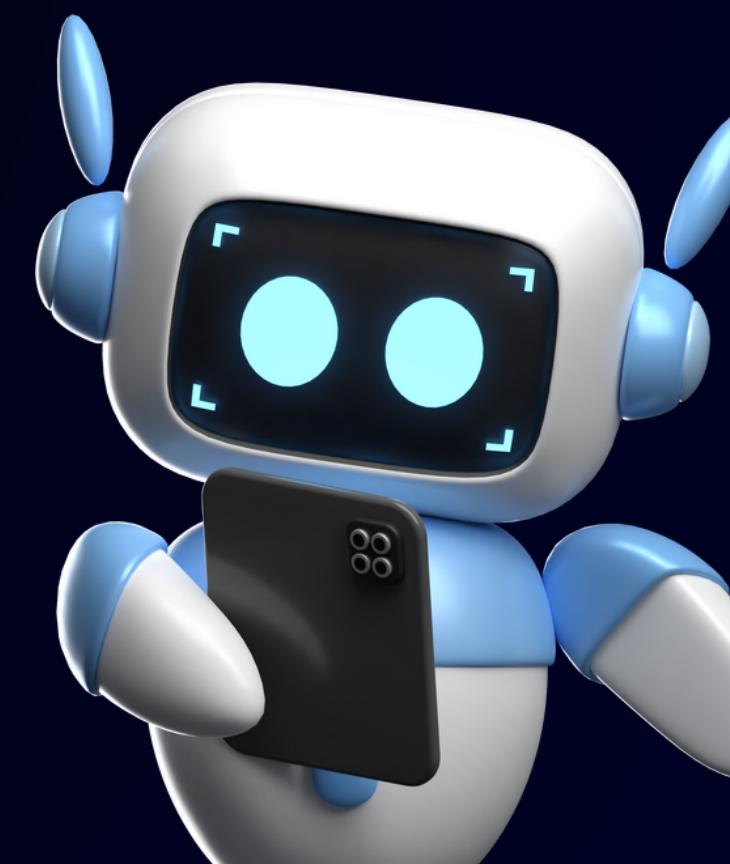
```
class_names_str = str(class_names)
```

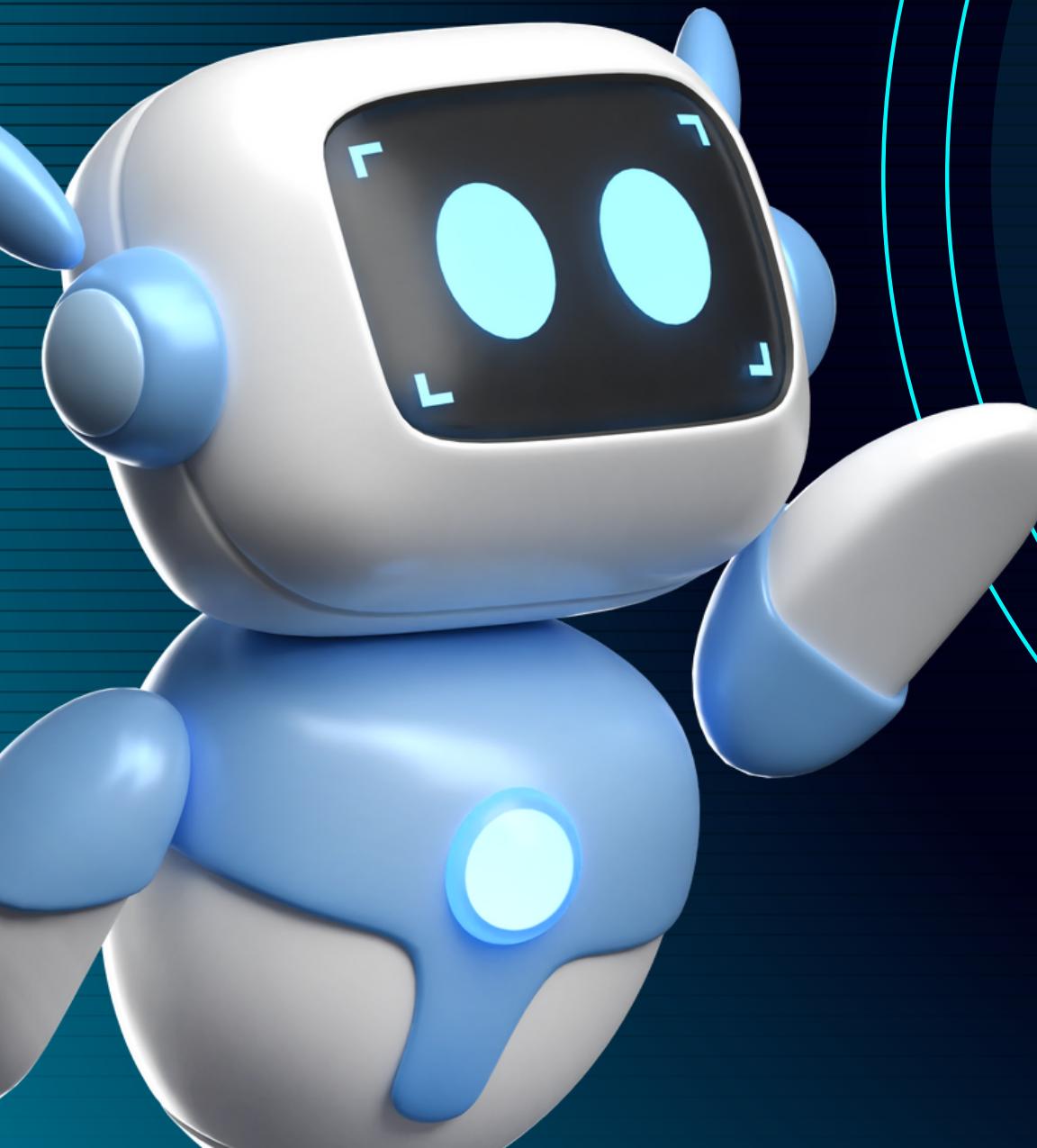
## CREATING YAML CONTENT

```
yaml_content = f""" train: /content/TACO/train val:  
/content/TACO/val nc: 60 names:  
{class_names_str} """
```

## WRITING YAML CONTENT TO FILE

```
with open('taco.yaml', 'w') as f:  
    f.write(yaml_content.strip())
```

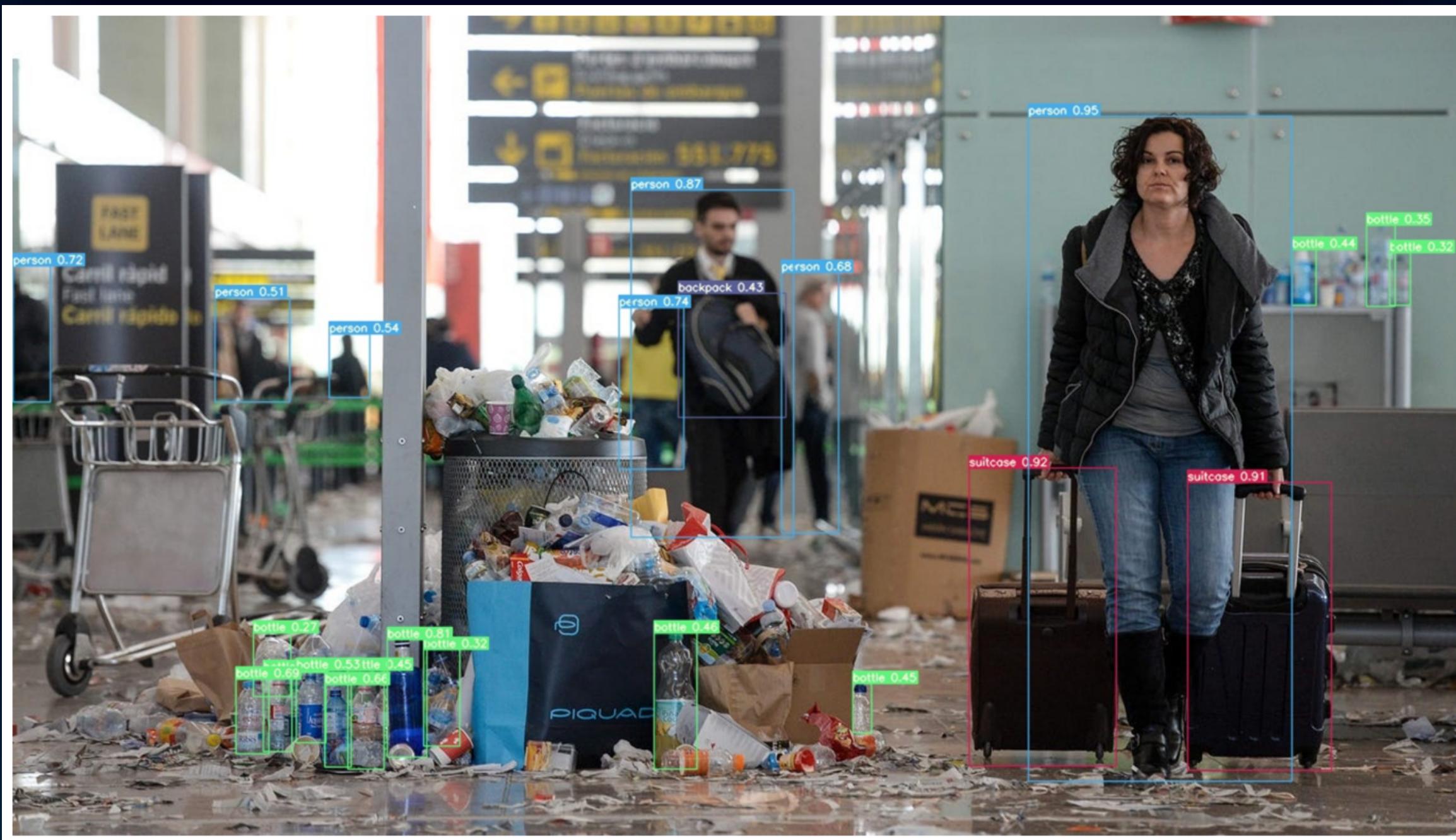




# INITIATE MODEL TRAINING

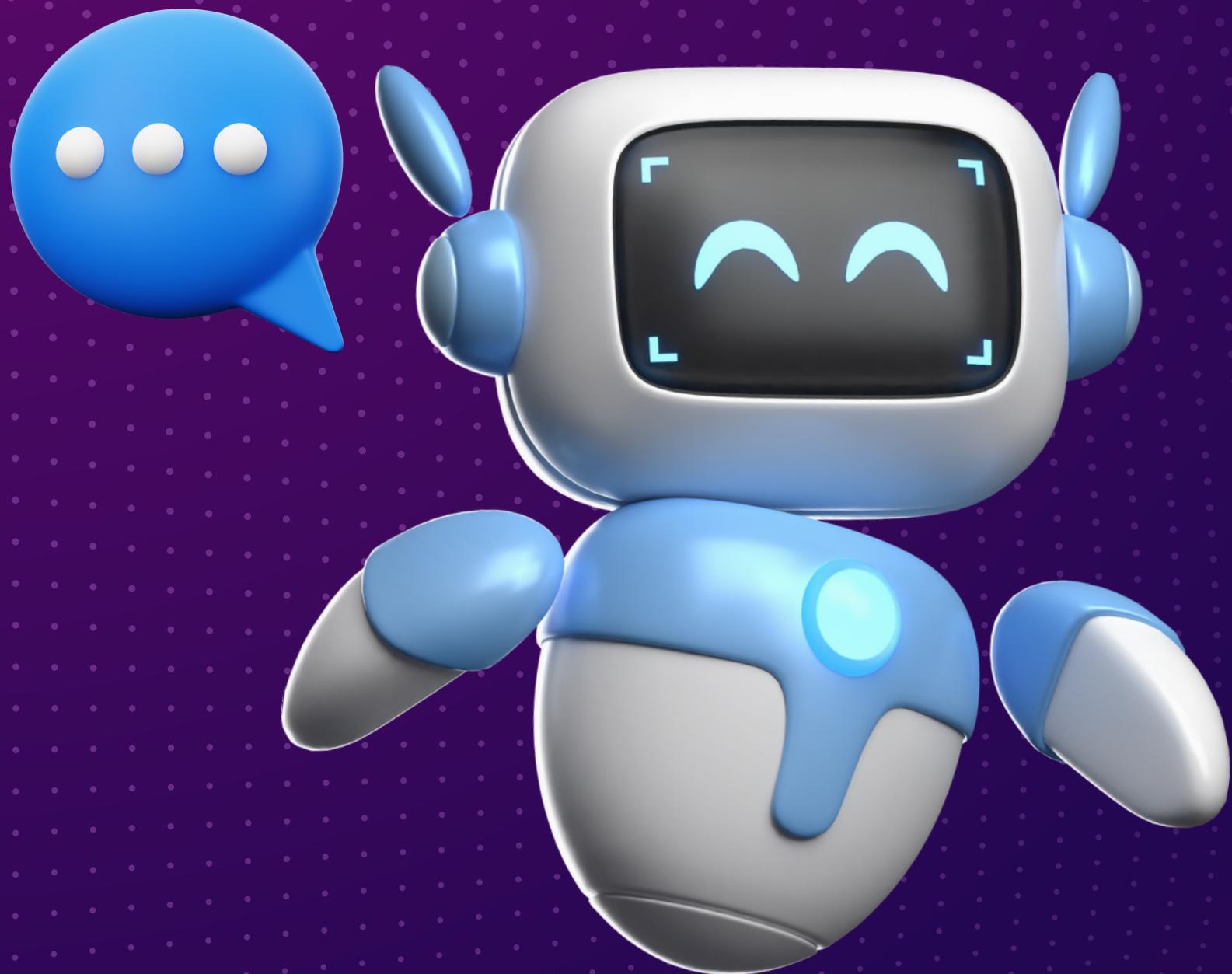
```
!PYTHON TRAIN.PY --IMG 640 --BATCH 16 --EPOCHS 10 --DATA  
TACO.YAML --CFG MODELS/YOLOV5S.YAML --WEIGHTS YOLOV5S.PT  
--NAME TACO_MODEL
```

# Example Output- by using YOLOv7



2

## BENEFITS



REAL-TIME AWARENESS: DETECT ISSUES & VOLUME INSTANTLY.

AUTOMATED MONITORING: FREE CROW'S NEST FOR CRITICAL TASKS.

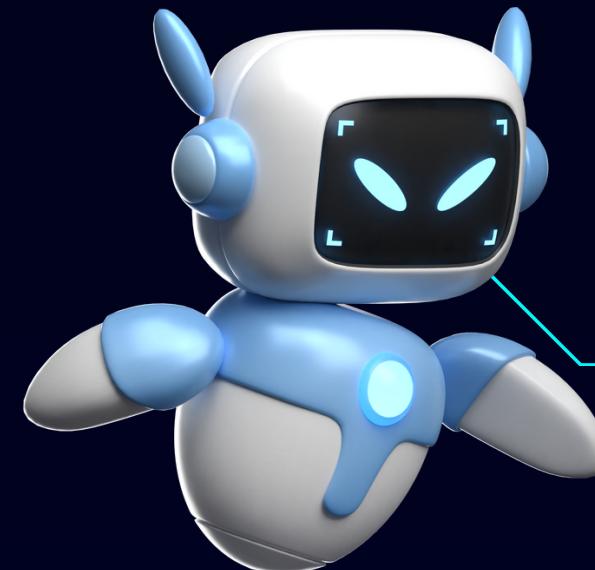
SUPERIOR ACCURACY: MACHINE LEARNING OUTPERFORMS MANUAL METHODS.

4

## NEXT STEPS



REFINE THE YOLOV7 MODEL FOR OPTIMAL PERFORMANCE ON YVR-SPECIFIC DATA.



INTEGRATE THE MODEL AND CONDUCT PILOT TESTING WITHIN A DESIGNATED AREA



THANK YOU