

License Plate Detection Model Training

Importing necessary libraries and dependencies

```
In [4]: !pip install opencv-python
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: opencv-python in c:\programdata\anaconda3\lib\site-packages (4.10.0.84)
Requirement already satisfied: numpy>=1.21.2 in c:\programdata\anaconda3\lib\site-packages (from opencv-python) (1.26.4)
```

```
In [6]: import os
import cv2
import shutil
import numpy as np
import pandas as pd
from glob import glob
import matplotlib.pyplot as plt
import xml.etree.ElementTree as xet
from sklearn.model_selection import train_test_split
```

```
In [8]: !pip install torch
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: torch in c:\programdata\anaconda3\lib\site-packages (2.3.1)
Requirement already satisfied: filelock in c:\programdata\anaconda3\lib\site-packages (from torch) (3.13.1)
Requirement already satisfied: typing-extensions>=4.8.0 in c:\programdata\anaconda3\lib\site-packages (from torch) (4.11.0)
Requirement already satisfied: sympy in c:\programdata\anaconda3\lib\site-packages (from torch) (1.12)
Requirement already satisfied: networkx in c:\programdata\anaconda3\lib\site-packages (from torch) (3.2.1)
Requirement already satisfied: Jinja2 in c:\programdata\anaconda3\lib\site-packages (from torch) (3.1.4)
Requirement already satisfied: fsspec in c:\programdata\anaconda3\lib\site-packages (from torch) (2024.3.1)
Requirement already satisfied: mkl<=2021.4.0,>=2021.1.1 in c:\programdata\anaconda3\lib\site-packages (from torch) (2021.4.0)
Requirement already satisfied: intel-openmp==2021.* in c:\programdata\anaconda3\lib\site-packages (from mkl<=2021.4.0,>=2021.1.1->torch) (2021.4.0)
Requirement already satisfied: tbb==2021.* in c:\programdata\anaconda3\lib\site-packages (from mkl<=2021.4.0,>=2021.1.1->torch) (2021.13.0)
Requirement already satisfied: MarkupSafe>=2.0 in c:\programdata\anaconda3\lib\site-packages (from Jinja2->torch) (2.1.3)
Requirement already satisfied: mpmath>=0.19 in c:\programdata\anaconda3\lib\site-packages (from sympy->torch) (1.3.0)
```

Checking if GPU is available for training

```
In [13]: import torch

print(f'{torch.cuda.is_available() = }')
print(f'{torch.cuda.device_count() = }')
```

```
torch.cuda.is_available() = False
torch.cuda.device_count() = 0
```

```
In [15]: !pip install ultralytics
```

```

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: ultralytics in c:\programdata\anaconda3\lib\site-packages (8.2.50)
Requirement already satisfied: numpy<2.0.0,>=1.23.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (1.26.4)
Requirement already satisfied: matplotlib>=3.3.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (3.8.4)
Requirement already satisfied: opencv-python>=4.6.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (4.10.0.84)
Requirement already satisfied: pillow>=7.1.2 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (10.3.0)
Requirement already satisfied: pyyaml>=5.3.1 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (6.0.1)
Requirement already satisfied: requests>=2.23.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (2.32.2)
Requirement already satisfied: scipy>=1.4.1 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (1.13.1)
Requirement already satisfied: torch>=1.8.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (2.3.1)
Requirement already satisfied: torchvision>=0.9.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (0.18.1)
Requirement already satisfied: tqdm>=4.64.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (4.66.4)
Requirement already satisfied: psutil in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (5.9.0)
Requirement already satisfied: py-cpuinfo in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (9.0.0)
Requirement already satisfied: pandas>=1.1.4 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (2.2.2)
Requirement already satisfied: seaborn>=0.11.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (0.13.2)
Requirement already satisfied: ultralytics-thop>=2.0.0 in c:\programdata\anaconda3\lib\site-packages (from ultralytics) (2.0.0)
Requirement already satisfied: contourpy>=1.0.1 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (1.2.0)
Requirement already satisfied: cycler>=0.10 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (1.4.4)
Requirement already satisfied: packaging>=20.0 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (23.2)
Requirement already satisfied: pyparsing>=2.3.1 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (3.0.9)
Requirement already satisfied: python-dateutil>=2.7 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.3.0->ultralytics) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\programdata\anaconda3\lib\site-packages (from pandas>=1.1.4->ultralytics) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\programdata\anaconda3\lib\site-packages (from pandas>=1.1.4->ultralytics) (2023.3)
Requirement already satisfied: charset-normalizer<4,>=2 in c:\programdata\anaconda3\lib\site-packages (from requests>=2.23.0->ultralytics) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in c:\programdata\anaconda3\lib\site-packages (from requests>=2.23.0->ultralytics) (3.7)
Requirement already satisfied: urllib3<3,>=1.21.1 in c:\programdata\anaconda3\lib\site-packages (from requests>=2.23.0->ultralytics) (2.2.2)
Requirement already satisfied: certifi>=2017.4.17 in c:\programdata\anaconda3\lib\site-packages (from requests>=2.23.0->ultralytics) (2024.6.2)
Requirement already satisfied: filelock in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (3.13.1)
Requirement already satisfied: typing-extensions>=4.8.0 in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (4.11.0)
Requirement already satisfied: sympy in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (1.12)
Requirement already satisfied: networkx in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (3.2.1)
Requirement already satisfied: Jinja2 in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (3.1.4)
Requirement already satisfied: fsspec in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (2024.3.1)
Requirement already satisfied: mkl<=2021.4.0,>=2021.1.1 in c:\programdata\anaconda3\lib\site-packages (from torch>=1.8.0->ultralytics) (2021.4.0)
Requirement already satisfied: colorama in c:\programdata\anaconda3\lib\site-packages (from tqdm>=4.64.0->ultralytics) (0.4.6)
Requirement already satisfied: intel-openmp==2021.* in c:\programdata\anaconda3\lib\site-packages (from mkl<=2021.4.0,>=2021.1.1->torch>=1.8.0->ultralytics) (2021.4.0)
Requirement already satisfied: tbb==2021.* in c:\programdata\anaconda3\lib\site-packages (from mkl<=2021.4.0,>=2021.1.1->torch>=1.8.0->ultralytics) (2021.13.0)
Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib>=3.3.0->ultralytics) (1.16.0)
Requirement already satisfied: MarkupSafe>=2.0 in c:\programdata\anaconda3\lib\site-packages (from Jinja2->torch>=1.8.0->ultralytics) (2.1.3)
Requirement already satisfied: mpmath>=0.19 in c:\programdata\anaconda3\lib\site-packages (from sympy->torch>=1.8.0->ultralytics) (1.3.0)

```

In [17]: `!pip install -U ipywidgets`

```

Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: ipywidgets in c:\programdata\anaconda3\lib\site-packages (8.1.3)
Requirement already satisfied: comm>=0.1.3 in c:\programdata\anaconda3\lib\site-packages (from ipywidgets) (0.2.1)
Requirement already satisfied: ipython>=6.1.0 in c:\programdata\anaconda3\lib\site-packages (from ipywidgets) (8.25.0)
Requirement already satisfied: traitlets>=4.3.1 in c:\programdata\anaconda3\lib\site-packages (from ipywidgets) (5.14.3)
Requirement already satisfied: widgetsnbextension~4.0.11 in c:\programdata\anaconda3\lib\site-packages (from ipywidgets) (4.0.11)
Requirement already satisfied: jupyterlab-widgets~3.0.11 in c:\programdata\anaconda3\lib\site-packages (from ipywidgets) (3.0.11)
Requirement already satisfied: decorator in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (5.1.1)
Requirement already satisfied: jedi>=0.16 in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (0.18.1)
Requirement already satisfied: matplotlib-inline in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (0.1.6)
Requirement already satisfied: prompt-toolkit<3.1.0,>=3.0.41 in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (3.0.43)
Requirement already satisfied: pygments>=2.4.0 in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (2.15.1)
Requirement already satisfied: stack-data in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (0.2.0)
Requirement already satisfied: colorama in c:\programdata\anaconda3\lib\site-packages (from ipython>=6.1.0->ipywidgets) (0.4.6)
Requirement already satisfied: parso<0.9.0,>=0.8.0 in c:\programdata\anaconda3\lib\site-packages (from jedi>=0.16->ipython>=6.1.0->ipywidgets) (0.8.3)
Requirement already satisfied: wcwidth in c:\programdata\anaconda3\lib\site-packages (from prompt-toolkit<3.1.0,>=3.0.41->ipython>=6.1.0->ipywidgets) (0.2.5)
Requirement already satisfied: executing in c:\programdata\anaconda3\lib\site-packages (from stack-data->ipython>=6.1.0->ipywidgets) (0.8.3)
Requirement already satisfied: asttokens in c:\programdata\anaconda3\lib\site-packages (from stack-data->ipython>=6.1.0->ipywidgets) (2.0.5)
Requirement already satisfied: pure-eval in c:\programdata\anaconda3\lib\site-packages (from stack-data->ipython>=6.1.0->ipywidgets) (0.2.2)
Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-packages (from asttokens->stack-data->ipython>=6.1.0->ipywidgets) (1.16.0)

```

Importing data

```
In [20]: dataset_path = 'archive'
```

Function to extract the first sequence of digits from the given filename string

```
In [23]: import re

def the_number_in_the_string(filename):
    """
    Extracts the first sequence of digits from the given filename string and returns it as an integer.
    If no digits are found, returns 0.
    """
    # Search for the first occurrence of one or more digits in the filename
    match = re.search(r'(\d+)', filename)

    # If a match is found, return the matched number as an integer
    if match:
        return int(match.group(0))
    # If no match is found, return 0
    else:
        return 0
```

```
In [25]: labels_dict = dict(
    img_path=[],
    xmin=[],
    xmax=[],
    ymin=[],
    ymax=[],
    img_w=[],
    img_h=[]
)
```

```
In [27]: # Get the List of XML files from the annotations directory
xml_files = glob(f'{dataset_path}/annotations/*.xml')
```

Processing each XML file

```
In [30]: for filename in sorted(xml_files, key=the_number_in_the_string):
    # Parse the XML file
    info = xet.parse(filename)
    root = info.getroot()

    # Find the 'object' element in the XML and extract bounding box information
    member_object = root.find('object')
    labels_info = member_object.find('bndbox')
    xmin = int(labels_info.find('xmin').text)
    xmax = int(labels_info.find('xmax').text)
    ymin = int(labels_info.find('ymin').text)
    ymax = int(labels_info.find('ymax').text)

    # Get the image filename and construct the full path to the image
    img_name = root.find('filename').text
    img_path = os.path.join(dataset_path, 'images', img_name)

    # Append the extracted information to the respective Lists in the dictionary
    labels_dict['img_path'].append(img_path)
    labels_dict['xmin'].append(xmin)
    labels_dict['xmax'].append(xmax)
    labels_dict['ymin'].append(ymin)
    labels_dict['ymax'].append(ymax)

    # Read the image to get its dimensions
    height, width, _ = cv2.imread(img_path).shape
    labels_dict['img_w'].append(width)
    labels_dict['img_h'].append(height)
```

```
# Convert the dictionary to a pandas DataFrame
alldata = pd.DataFrame(labels_dict)

# Display the DataFrame
alldata
```

Out[30]:

	img_path	xmin	xmax	ymin	ymax	img_w	img_h
0	archive\images\Cars0.png	226	419	125	173	500	268
1	archive\images\Cars1.png	134	262	128	160	400	248
2	archive\images\Cars2.png	229	270	176	193	400	400
3	archive\images\Cars3.png	142	261	128	157	400	225
4	archive\images\Cars4.png	156	503	82	253	590	350
...
428	archive\images\Cars428.png	142	258	128	157	400	225
429	archive\images\Cars429.png	86	208	166	195	301	400
430	archive\images\Cars430.png	38	116	159	197	400	225
431	archive\images\Cars431.png	55	343	82	147	400	192
432	archive\images\Cars432.png	95	196	258	284	467	300

433 rows × 7 columns

Train-test split

```
In [33]: from sklearn.model_selection import train_test_split

# Split the data into training and test sets
# Use 10% of the data for the test set
train, test = train_test_split(alldata, test_size=1/10, random_state=42)

# Split the training data further into training and validation sets
# Use 8/9 of the remaining data for the training set, resulting in an 80/10/10 split overall
train, val = train_test_split(train, train_size=8/9, random_state=42)

# Print the number of samples in each set
print(f'''
    len(train) = {len(train)}
    len(val) = {len(val)}
    len(test) = {len(test)}
''')
```

len(train) = 345
len(val) = 44
len(test) = 44

Preparing data for model training

```
In [36]: # Remove the 'datasets' directory if it exists
if os.path.exists('datasets'):
    shutil.rmtree('datasets')
```

```
In [38]: def make_split_folder_in_yolo_format(split_name, split_df, output_folder):
    """
    Creates a folder structure for a dataset split (train/val/test) in YOLO format.
```

```

Parameters:
split_name (str): The name of the split (e.g., 'train', 'val', 'test').
split_df (pd.DataFrame): The DataFrame containing the data for the split.
output_folder (str): The root folder where the dataset will be created.

The function will create 'labels' and 'images' subdirectories under '{output_folder}/cars_license_plate/{split_name}',
and save the corresponding labels and images in YOLO format.
"""
labels_path = os.path.join(output_folder, 'cars_license_plate', split_name, 'labels')
images_path = os.path.join(output_folder, 'cars_license_plate', split_name, 'images')

# Create directories for labels and images
os.makedirs(labels_path, exist_ok=True)
os.makedirs(images_path, exist_ok=True)

# Iterate over each row in the DataFrame
for _, row in split_df.iterrows():
    img_name, img_extension = os.path.splitext(os.path.basename(row['img_path']))

    # Calculate YOLO format bounding box coordinates
    x_center = (row['xmin'] + row['xmax']) / 2 / row['img_w']
    y_center = (row['ymin'] + row['ymax']) / 2 / row['img_h']
    width = (row['xmax'] - row['xmin']) / row['img_w']
    height = (row['ymax'] - row['ymin']) / row['img_h']

    # Save the label in YOLO format
    label_path = os.path.join(labels_path, f'{img_name}.txt')
    with open(label_path, 'w') as file:
        file.write(f"0 {x_center:.4f} {y_center:.4f} {width:.4f} {height:.4f}\n")

    # Copy the image to the images directory
    shutil.copy(row['img_path'], os.path.join(images_path, img_name + img_extension))

print(f"Created '{images_path}' and '{labels_path}'")

```

```

In [40]: # Create YOLO format folders for train, validation, and test splits
make_split_folder_in_yolo_format("train", train, "datasets")
make_split_folder_in_yolo_format("val", val, "datasets")
make_split_folder_in_yolo_format("test", test, "datasets")

Created 'datasets\cars_license_plate\train\images' and 'datasets\cars_license_plate\train\labels'
Created 'datasets\cars_license_plate\val\images' and 'datasets\cars_license_plate\val\labels'
Created 'datasets\cars_license_plate\test\images' and 'datasets\cars_license_plate\test\labels'

```

```

In [42]: # Directory paths
image_dir = 'cars_license_plate_new/train/images'
label_dir = 'cars_license_plate_new/train/labels'

```

```

In [44]: # Get the first image file
image_files = sorted(os.listdir(image_dir))
first_image_file = image_files[0]

# Construct paths for the image and its corresponding label
image_path = os.path.join(image_dir, first_image_file)
label_path = os.path.join(label_dir, os.path.splitext(first_image_file)[0] + '.txt')

# Load the image using OpenCV
image = cv2.imread(image_path)
# Convert the image from BGR (OpenCV default) to RGB (matplotlib default)
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

# Read the label file to get bounding box information
with open(label_path, 'r') as f:
    lines = f.readlines()

```

```

In [46]: # Plot the bounding box on the image
for line in lines:

```

```

# Parse the label file line to extract bounding box information
class_id, x_center, y_center, width, height = map(float, line.strip().split())
img_height, img_width, _ = image.shape

# Convert YOLO format to bounding box format
x_center *= img_width
y_center *= img_height
width *= img_width
height *= img_height

# Calculate the top-left and bottom-right coordinates of the bounding box
x1 = int(x_center - width / 2)
y1 = int(y_center - height / 2)
x2 = int(x_center + width / 2)
y2 = int(y_center + height / 2)

# Draw the bounding box on the image using a green rectangle
cv2.rectangle(image, (x1, y1), (x2, y2), (0, 255, 0), 2)

# Display the image with bounding box using matplotlib
plt.imshow(image)
plt.axis('off') # Hide the axis
plt.show() # Display the image

```



```

In [48]: # Define the content of the datasets.yaml file
datasets_yaml = '''
path: cars_license_plate_new

train: train/images
val: val/images
test: test/images

# number of classes
nc: 1

# class names
names: ['license_plate']
'''

# Write the content to the datasets.yaml file
with open('datasets.yaml', 'w') as file:
    file.write(datasets_yaml)

```

Model Training and Evaluation

```

In [51]: from ultralytics import YOLO

```

```
model = YOLO('yolov8n.pt')

Downloading https://github.com/ultralytics/assets/releases/download/v8.2.0/yolov8n.pt to 'yolov8n.pt'...
100%[██████████] 6.23M/6.23M [00:03<00:00, 1.90MB/s]

In [53]: model.train(
    data="datasets.yaml", # Path to the dataset configuration file
    epochs=100,           # Number of training epochs
    batch=16,             # Batch size
    device='cpu',         # Use CPU for training
    imgsz=320,            # Image size (width and height) for training
    cache=True            # Cache images for faster training
)

New https://pypi.org/project/ultralytics/8.2.52 available Update with 'pip install -U ultralytics'
Ultralytics YOLOv8.2.50 Python-3.12.3 torch-2.3.1+cpu CPU (AMD Ryzen 7 5800H with Radeon Graphics)
engine\trainer: task=detect, mode=train, model=yolov8n.pt, data=datasets.yaml, epochs=100, time=None, patience=100, batch=16, imgsz=320, save=True, save_period=-1, cache=True, device=cpu, workers=8, project=None, name=train19, exist_ok=False, pretrained=True, optimizer=auto, verbose=True, seed=0, deterministic=True, single_cls=False, rect=False, cos_lr=False, close_mosaic=10, resume=False, amp=True, fraction=1.0, profile=False, freeze=None, multi_scale=False, overlap_mask=True, mask_ratio=4, dropout=0.0, val=True, split=val, save_json=False, save_hybrid=False, conf=None, iou=0.7, max_det=300, half=False, dnn=False, plots=True, source=None, vid_stride=1, stream_buffer=False, visualize=False, augment=False, agnostic_nms=False, classes=None, retina_masks=False, embed=None, show=False, save_frames=False, save_txt=False, save_conf=False, save_crop=False, show_labels=True, show_conf=True, show_boxes=True, line_width=None, format=torchscript, keras=False, optimize=False, int8=False, dynamic=False, simplify=False, opset=None, workspace=4, nms=False, lr0=0.01, lrf=0.01, momentum=0.937, weight_decay=0.0005, warmup_epochs=3.0, warmup_momentum=0.8, warmup_bias_lr=0.1, box=7.5, cls=0.5, dfl=1.5, pose=12.0, kobj=1.0, label_smoothing=0.0, nbs=64, hsv_h=0.015, hsv_s=0.7, hsv_v=0.4, degrees=0.0, translate=0.1, scale=0.5, shear=0.0, perspective=0.0, flipud=0.0, fliplr=0.5, bgr=0.0, mosaic=1.0, mixup=0.0, copy_paste=0.0, auto_augment=randaugument, erasing=0.4, crop_fraction=1.0, cfg=None, tracker=botsort.yaml, save_dir=runs\detect\train19
Overriding model.yaml nc=80 with nc=1

      from  n  params module                                arguments
0           -1  1    464  ultralytics.nn.modules.conv.Conv          [3, 16, 3, 2]
1           -1  1   4672 ultralytics.nn.modules.conv.Conv          [16, 32, 3, 2]
2           -1  1   7360 ultralytics.nn.modules.block.C2f          [32, 32, 1, True]
3           -1  1  18560 ultralytics.nn.modules.conv.Conv          [32, 64, 3, 2]
4           -1  2  49664 ultralytics.nn.modules.block.C2f          [64, 64, 2, True]
5           -1  1  73984 ultralytics.nn.modules.conv.Conv          [64, 128, 3, 2]
6           -1  2 197632 ultralytics.nn.modules.block.C2f          [128, 128, 2, True]
7           -1  1 295424 ultralytics.nn.modules.conv.Conv          [128, 256, 3, 2]
8           -1  1 460288 ultralytics.nn.modules.block.C2f          [256, 256, 1, True]
9           -1  1 164608 ultralytics.nn.modules.block.SPPF          [256, 256, 5]
10          -1  1      0  torch.nn.modules.upsampling.Upsample          [None, 2, 'nearest']
11         [-1, 6]  1      0  ultralytics.nn.modules.conv.Concat          [1]
12          -1  1 148224 ultralytics.nn.modules.block.C2f          [384, 128, 1]
13          -1  1      0  torch.nn.modules.upsampling.Upsample          [None, 2, 'nearest']
14         [-1, 4]  1      0  ultralytics.nn.modules.conv.Concat          [1]
15          -1  1  37248 ultralytics.nn.modules.block.C2f          [192, 64, 1]
16          -1  1  36992 ultralytics.nn.modules.conv.Conv          [64, 64, 3, 2]
17         [-1, 12]  1      0  ultralytics.nn.modules.conv.Concat          [1]
18          -1  1 123648 ultralytics.nn.modules.block.C2f          [192, 128, 1]
19          -1  1 147712 ultralytics.nn.modules.conv.Conv          [128, 128, 3, 2]
20         [-1, 9]  1      0  ultralytics.nn.modules.conv.Concat          [1]
21          -1  1 493056 ultralytics.nn.modules.block.C2f          [384, 256, 1]
22         [15, 18, 21] 1 751507 ultralytics.nn.modules.head.Detect          [1, [64, 128, 256]]

Model summary: 225 layers, 3011043 parameters, 3011027 gradients, 8.2 GFLOPs

Transferred 319/355 items from pretrained weights
TensorBoard: Start with 'tensorboard --logdir runs\detect\train19', view at http://localhost:6006/
Freezing layer 'model.22.dfl.conv.weight'

train: Scanning C:\Documents\Intel Unnati\7th Jul\cars_license_plate_new\train\labels.cache... 345 images, 0 backgrounds, 0 corrupt: 100%[██████████] 345/345 [00:00<?, ?it/s]
train: Caching images (0.1GB RAM): 100%[██████████] 345/345 [00:00<00:00, 853.42it/s]
val: Scanning C:\Documents\Intel Unnati\7th Jul\cars_license_plate_new\val\labels.cache... 44 images, 0 backgrounds, 0 corrupt: 100%[██████████] 44/44 [00:00<?, ?it/s]
val: Caching images (0.0GB RAM): 100%[██████████] 44/44 [00:00<00:00, 843.20it/s]

Plotting labels to runs\detect\train19\labels.jpg...
optimizer: 'optimizer=auto' found, ignoring 'lr0=0.01' and 'momentum=0.937' and determining best 'optimizer', 'lr0' and 'momentum' automatically...
optimizer: AdamW(lr=0.002, momentum=0.9) with parameter groups 57 weight(decay=0.0), 64 weight(decay=0.0005), 63 bias(decay=0.0)
TensorBoard: model graph visualization added
Image sizes 320 train, 320 val
Using 0 dataloader workers
Logging results to runs\detect\train19
Starting training for 100 epochs...

Epoch    GPU_mem  box_loss  cls_loss  dfl_loss  Instances    Size
```

1/100	0G	1.594	2.786	1.257	17	320: 100%	<div></div>	22/22 [00:22<00:00, 1.04s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.52it/s]
	all	44	44	0.00352	0.932	0.598	0.27	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
2/100	0G	1.579	1.518	1.137	16	320: 100%	<div></div>	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.07it/s]
	all	44	44	0.00463	0.886	0.272	0.15	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
3/100	0G	1.536	1.401	1.154	15	320: 100%	<div></div>	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.97it/s]
	all	44	44	0.677	0.295	0.317	0.149	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
4/100	0G	1.526	1.351	1.191	18	320: 100%	<div></div>	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.06it/s]
	all	44	44	0.733	0.561	0.706	0.372	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
5/100	0G	1.534	1.288	1.187	18	320: 100%	<div></div>	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.09it/s]
	all	44	44	0.679	0.568	0.629	0.298	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
6/100	0G	1.532	1.272	1.147	21	320: 100%	<div></div>	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.02it/s]
	all	44	44	0.882	0.679	0.753	0.359	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
7/100	0G	1.468	1.145	1.156	15	320: 100%	<div></div>	22/22 [00:21<00:00, 1.04it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.05it/s]
	all	44	44	0.703	0.295	0.357	0.164	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
8/100	0G	1.498	1.134	1.17	23	320: 100%	<div></div>	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.62it/s]
	all	44	44	0.488	0.591	0.512	0.273	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
9/100	0G	1.456	1.069	1.161	13	320: 100%	<div></div>	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.95it/s]
	all	44	44	0.834	0.727	0.741	0.402	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
10/100	0G	1.404	1.026	1.12	20	320: 100%	<div></div>	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.97it/s]
	all	44	44	0.832	0.705	0.79	0.424	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
11/100	0G	1.371	1.007	1.112	24	320: 100%	<div></div>	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.10it/s]
	all	44	44	0.812	0.882	0.881	0.506	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
12/100	0G	1.33	0.9416	1.072	21	320: 100%	<div></div>	22/22 [00:22<00:00, 1.02s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.03it/s]
	all	44	44	0.892	0.795	0.89	0.43	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
13/100	0G	1.311	0.8927	1.077	20	320: 100%	<div></div>	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.93it/s]
	all	44	44	0.859	0.864	0.877	0.423	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
14/100	0G	1.339	0.9124	1.065	19	320: 100%	<div></div>	22/22 [00:22<00:00, 1.00s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01<00:00, 1.78it/s]
	all	44	44	0.824	0.886	0.871	0.471	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size		
15/100	0G	1.322	0.9129	1.097	14	320: 100%	<div></div>	22/22 [00:22<00:00, 1.01s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00<00:00, 2.10it/s]
	all	44	44	0.919	0.841	0.877	0.462	

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
16/100	0G	1.32	0.8965	1.11	15	320: 100% ██████████ 22/22 [00:22<00:00, 1.04s/it]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.01it/s]
	all	44	44	0.822	0.818	0.79 0.408
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
17/100	0G	1.309	0.867	1.081	16	320: 100% ██████████ 22/22 [00:22<00:00, 1.04s/it]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.08it/s]
	all	44	44	0.914	0.795	0.844 0.438
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
18/100	0G	1.309	0.8775	1.105	15	320: 100% ██████████ 22/22 [00:21<00:00, 1.00it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.10it/s]
	all	44	44	0.846	0.871	0.866 0.499
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
19/100	0G	1.314	0.839	1.096	15	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.13it/s]
	all	44	44	0.916	0.886	0.883 0.491
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
20/100	0G	1.239	0.8222	1.053	15	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.09it/s]
	all	44	44	0.874	0.864	0.913 0.472
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
21/100	0G	1.193	0.7636	1.034	22	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.11it/s]
	all	44	44	0.948	0.834	0.892 0.484
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
22/100	0G	1.184	0.7828	1.035	13	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.03it/s]
	all	44	44	0.848	0.909	0.89 0.46
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
23/100	0G	1.159	0.7422	1.032	21	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:01<00:00, 1.96it/s]
	all	44	44	0.872	0.841	0.884 0.485
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
24/100	0G	1.118	0.7352	1.023	21	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.06it/s]
	all	44	44	0.901	0.864	0.89 0.459
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
25/100	0G	1.212	0.7647	1.035	15	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.10it/s]
	all	44	44	0.946	0.8	0.893 0.464
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
26/100	0G	1.165	0.7393	1.026	20	320: 100% ██████████ 22/22 [00:22<00:00, 1.01s/it]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:01<00:00, 1.96it/s]
	all	44	44	0.946	0.792	0.875 0.477
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
27/100	0G	1.137	0.7245	1.024	12	320: 100% ██████████ 22/22 [00:22<00:00, 1.02s/it]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:01<00:00, 1.95it/s]
	all	44	44	0.929	0.795	0.864 0.478
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
28/100	0G	1.126	0.7225	1.016	16	320: 100% ██████████ 22/22 [00:22<00:00, 1.02s/it]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.07it/s]
	all	44	44	0.901	0.828	0.887 0.466
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
29/100	0G	1.121	0.6994	1.002	24	320: 100% ██████████ 22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.08it/s]
	all	44	44	0.906	0.909	0.939 0.513
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
30/100	0G	1.093	0.6998	0.9954	16	320: 100% ██████████ 22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.09it/s]

	all	44	44	0.885	0.909	0.895	0.518
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
31/100	0G	1.108	0.6991	1.016	18	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.902	0.832	0.885	0.513
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
32/100	0G	1.094	0.6919	1.004	13	320: 100%	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.826	0.864	0.885	0.504
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
33/100	0G	1.079	0.686	0.9867	20	320: 100%	22/22 [00:22<00:00, 1.01s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.957	0.841	0.93	0.483
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
34/100	0G	1.05	0.6576	0.9923	17	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.939	0.864	0.91	0.504
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
35/100	0G	1.053	0.6701	0.975	16	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.894	0.818	0.891	0.502
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
36/100	0G	1.081	0.667	1.002	12	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.907	0.882	0.901	0.486
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
37/100	0G	1.092	0.6963	1.017	17	320: 100%	22/22 [00:22<00:00, 1.00s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.907	0.864	0.907	0.489
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
38/100	0G	1.092	0.6777	1.004	18	320: 100%	22/22 [00:23<00:00, 1.08s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.965	0.886	0.919	0.48
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
39/100	0G	1.038	0.631	0.9707	16	320: 100%	22/22 [00:22<00:00, 1.01s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.968	0.886	0.913	0.456
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
40/100	0G	1.043	0.6512	0.979	15	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.942	0.841	0.92	0.511
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
41/100	0G	1.052	0.6469	0.986	16	320: 100%	22/22 [00:22<00:00, 1.01s/it]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.906	0.878	0.918	0.465
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
42/100	0G	1.018	0.6178	0.9888	20	320: 100%	22/22 [00:21<00:00, 1.00it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.901	0.886	0.904	0.471
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
43/100	0G	1.027	0.6354	1.001	14	320: 100%	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.864	0.909	0.921	0.481
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
44/100	0G	0.9997	0.6243	0.9782	17	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.979	0.841	0.946	0.502
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	

45/100	0G	1.018	0.6106	0.9723	13	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.865	0.909	0.937	0.501
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
46/100	0G	1.021	0.6275	0.9809	21	320: 100%	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.951	0.883	0.938	0.504
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
47/100	0G	0.9959	0.6146	0.9574	12	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.951	0.882	0.926	0.518
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
48/100	0G	1.032	0.6227	0.9625	17	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.846	0.932	0.937	0.527
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
49/100	0G	0.9399	0.5842	0.9562	22	320: 100%	22/22 [00:21<00:00, 1.04it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.85	0.901	0.923	0.514
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
50/100	0G	0.9598	0.5801	0.9622	17	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.882	0.848	0.901	0.503
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
51/100	0G	0.9702	0.5745	0.9626	15	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.884	0.868	0.924	0.498
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
52/100	0G	0.9914	0.5738	0.9571	16	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.88	0.886	0.923	0.495
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
53/100	0G	0.936	0.5642	0.9431	15	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.895	0.886	0.919	0.521
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
54/100	0G	0.9161	0.5534	0.9469	16	320: 100%	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.842	0.886	0.921	0.531
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
55/100	0G	0.9298	0.5598	0.9424	16	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.925	0.843	0.929	0.525
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
56/100	0G	0.9411	0.5534	0.9417	19	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.92	0.909	0.93	0.516
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
57/100	0G	0.9141	0.5605	0.9548	23	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.925	0.909	0.919	0.496
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
58/100	0G	0.9359	0.5466	0.9403	20	320: 100%	22/22 [00:21<00:00, 1.01it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.88	0.841	0.918	0.533
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
59/100	0G	0.9207	0.5741	0.9429	17	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.917	0.818	0.918	0.549

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
60/100	0G	0.9311	0.5448	0.9368	11	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.08it/s]
	all	44	44	0.879	0.886	0.933 0.533
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
61/100	0G	0.903	0.5398	0.9373	14	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.12it/s]
	all	44	44	0.924	0.841	0.932 0.518
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
62/100	0G	0.838	0.5165	0.9353	16	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.11it/s]
	all	44	44	0.885	0.841	0.897 0.497
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
63/100	0G	0.8803	0.5269	0.9325	19	320: 100% ██████████ 22/22 [00:22<00:00, 1.03s/it]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.11it/s]
	all	44	44	0.918	0.909	0.936 0.513
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
64/100	0G	0.888	0.5474	0.9382	13	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.07it/s]
	all	44	44	0.9	0.886	0.927 0.485
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
65/100	0G	0.8801	0.5419	0.9461	17	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.04it/s]
	all	44	44	0.958	0.864	0.928 0.515
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
66/100	0G	0.8885	0.5272	0.9353	13	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.11it/s]
	all	44	44	0.88	0.909	0.921 0.515
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
67/100	0G	0.8464	0.5289	0.9348	12	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:01<00:00, 2.00it/s]
	all	44	44	0.88	0.909	0.936 0.526
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
68/100	0G	0.8672	0.5234	0.939	18	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.12it/s]
	all	44	44	0.907	0.888	0.922 0.509
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
69/100	0G	0.8536	0.5172	0.9371	16	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.12it/s]
	all	44	44	0.888	0.909	0.924 0.505
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
70/100	0G	0.8811	0.5273	0.9352	14	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.11it/s]
	all	44	44	0.886	0.909	0.935 0.526
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
71/100	0G	0.8502	0.5132	0.9182	18	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.13it/s]
	all	44	44	0.898	0.886	0.934 0.537
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
72/100	0G	0.8767	0.5215	0.9255	14	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.04it/s]
	all	44	44	0.909	0.864	0.934 0.52
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
73/100	0G	0.856	0.5127	0.9348	15	320: 100% ██████████ 22/22 [00:21<00:00, 1.04it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.12it/s]
	all	44	44	0.869	0.903	0.921 0.526
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
74/100	0G	0.8285	0.4919	0.9155	14	320: 100% ██████████ 22/22 [00:21<00:00, 1.03it/s]
	Class Images			Box(P	R	mAP50 mAP50-95): 100% ██████████ 2/2 [00:00<00:00, 2.01it/s]

	all	44	44	0.888	0.903	0.936	0.526
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
75/100	0G	0.8196	0.5036	0.9209	19	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.908	0.901	0.941	0.541
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
76/100	0G	0.8199	0.4919	0.9132	11	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.887	0.864	0.935	0.534
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
77/100	0G	0.7854	0.4747	0.9139	22	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.947	0.864	0.936	0.532
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
78/100	0G	0.8061	0.4751	0.918	16	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.95	0.859	0.938	0.534
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
79/100	0G	0.7829	0.4806	0.9164	17	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.899	0.909	0.925	0.538
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
80/100	0G	0.8195	0.4784	0.91	16	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.917	0.886	0.943	0.533
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
81/100	0G	0.7762	0.4836	0.9117	17	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.951	0.881	0.945	0.548
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
82/100	0G	0.7922	0.4702	0.9113	12	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.928	0.875	0.945	0.547
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
83/100	0G	0.7643	0.4657	0.8988	21	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.927	0.872	0.934	0.543
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
84/100	0G	0.7421	0.4407	0.9026	19	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.927	0.87	0.942	0.54
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
85/100	0G	0.7862	0.4687	0.9058	14	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.951	0.885	0.925	0.531
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
86/100	0G	0.7471	0.4645	0.9049	18	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.951	0.876	0.929	0.539
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
87/100	0G	0.7592	0.4708	0.9141	16	320: 100%	22/22 [00:21<00:00, 1.03it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.935	0.864	0.941	0.538
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
88/100	0G	0.7433	0.4475	0.9071	16	320: 100%	22/22 [00:21<00:00, 1.02it/s]
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%
	all	44	44	0.948	0.864	0.943	0.543
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	

89/100	0G	0.7288	0.4428	0.894	19	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.09it/s]
all	44	44	0.947	0.864	0.941	0.548	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
90/100	0G	0.7407	0.4416	0.8943	17	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.08it/s]
all	44	44	0.945	0.864	0.943	0.549	
Closing dataloader mosaic							
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
91/100	0G	0.9923	0.5605	0.9999	9	320: 100%	22/22 [00:22:00:00, 1.01s/it]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01:00:00, 1.78it/s]
all	44	44	0.952	0.841	0.933	0.535	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
92/100	0G	0.9761	0.5226	0.989	9	320: 100%	22/22 [00:23:00:00, 1.06s/it]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01:00:00, 1.97it/s]
all	44	44	0.95	0.855	0.933	0.55	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
93/100	0G	0.9603	0.5098	0.9773	9	320: 100%	22/22 [00:21:00:00, 1.04it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.10it/s]
all	44	44	0.967	0.864	0.944	0.556	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
94/100	0G	0.9583	0.5085	0.9641	9	320: 100%	22/22 [00:21:00:00, 1.04it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.11it/s]
all	44	44	0.956	0.841	0.943	0.544	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
95/100	0G	0.9543	0.4853	0.9829	9	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.09it/s]
all	44	44	0.973	0.822	0.939	0.549	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
96/100	0G	0.9593	0.4987	0.9799	9	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.06it/s]
all	44	44	0.949	0.838	0.937	0.548	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
97/100	0G	0.9177	0.4805	0.9647	9	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.02it/s]
all	44	44	0.921	0.864	0.934	0.547	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
98/100	0G	0.9341	0.4862	0.9712	9	320: 100%	22/22 [00:21:00:00, 1.05it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:01:00:00, 1.92it/s]
all	44	44	0.913	0.841	0.932	0.542	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
99/100	0G	0.9248	0.4716	0.9649	9	320: 100%	22/22 [00:20:00:00, 1.05it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.03it/s]
all	44	44	0.95	0.863	0.932	0.535	
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	
100/100	0G	0.9297	0.4793	0.9668	9	320: 100%	22/22 [00:21:00:00, 1.03it/s]
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.11it/s]
all	44	44	0.948	0.864	0.931	0.54	
100 epochs completed in 0.645 hours.							
Optimizer stripped from runs\detect\train19\weights\last.pt, 6.2MB							
Optimizer stripped from runs\detect\train19\weights\best.pt, 6.2MB							
Validating runs\detect\train19\weights\best.pt...							
Ultralytics YOLOv8.2.50 Python-3.12.3 torch-2.3.1+cpu CPU (AMD Ryzen 7 5800H with Radeon Graphics)							
Model summary (fused): 168 layers, 3005843 parameters, 0 gradients, 8.1 GFLOPs							
Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	2/2 [00:00:00:00, 2.31it/s]
all	44	44	0.968	0.864	0.944	0.556	
Speed: 0.6ms preprocess, 16.9ms inference, 0.0ms loss, 0.2ms postprocess per image							
Results saved to runs\detect\train19							

Out[53]: ultralytics.utils.metrics.DetMetrics object with attributes:

```

ap_class_index: array([0])
box: ultralytics.utils.metrics.Metric object
confusion_matrix: <ultralytics.utils.metrics.ConfusionMatrix object at 0x000002CE5D3F2CC0>
curves: ['Precision-Recall(B)', 'F1-Confidence(B)', 'Precision-Confidence(B)', 'Recall-Confidence(B)']
curves_results: [[array([
0.015015, 0.016016, 0.017017, 0.018018, 0.019019, 0.02002, 0.021021, 0.022022, 0.023023,
0.024024, 0.025025, 0.026026, 0.027027, 0.028028, 0.029029, 0.03003, 0.031031, 0.032032, 0.033033, 0.034034, 0.035035, 0.036036, 0.037037, 0.038038, 0.039039, 0.
04004, 0.041041, 0.042042, 0.043043, 0.044044, 0.045045, 0.046046, 0.047047,
0.048048, 0.049049, 0.05005, 0.051051, 0.052052, 0.053053, 0.054054, 0.055055, 0.056056, 0.057057, 0.058058, 0.059059, 0.06006, 0.061061, 0.062062, 0.063063, 0.0
64064, 0.065065, 0.066066, 0.067067, 0.068068, 0.069069, 0.07007, 0.071071,
0.072072, 0.073073, 0.074074, 0.075075, 0.076076, 0.077077, 0.078078, 0.079079, 0.08008, 0.081081, 0.082082, 0.083083, 0.084084, 0.085085, 0.086086, 0.087087, 0.0
88088, 0.089089, 0.09009, 0.091091, 0.092092, 0.093093, 0.094094, 0.095095,
0.096096, 0.097097, 0.098098, 0.099099, 0.1001, 0.1011, 0.1021, 0.1031, 0.1041, 0.10511, 0.10611, 0.10711, 0.10811, 0.10911, 0.11011, 0.11111, 0.
11211, 0.11311, 0.11411, 0.11512, 0.11612, 0.11712, 0.11812, 0.11912,
0.12012, 0.12112, 0.12212, 0.12312, 0.12412, 0.12513, 0.12613, 0.12713, 0.12813, 0.12913, 0.13013, 0.13113, 0.13213, 0.13313, 0.13413, 0.13514, 0.
13614, 0.13714, 0.13814, 0.13914, 0.14014, 0.14114, 0.14214, 0.14314,
0.14414, 0.14515, 0.14615, 0.14715, 0.14815, 0.14915, 0.15015, 0.15115, 0.15215, 0.15315, 0.15415, 0.15516, 0.15616, 0.15716, 0.15816, 0.15916, 0.
16016, 0.16116, 0.16216, 0.16316, 0.16416, 0.16517, 0.16617, 0.16717,
0.16817, 0.16917, 0.17017, 0.17117, 0.17217, 0.17317, 0.17417, 0.17518, 0.17618, 0.17718, 0.17818, 0.17918, 0.18018, 0.18118, 0.18218, 0.18318, 0.
18418, 0.18519, 0.18619, 0.18719, 0.18819, 0.18919, 0.19019, 0.19119,
0.19219, 0.19319, 0.19419, 0.1952, 0.1962, 0.1972, 0.1982, 0.1992, 0.2002, 0.2012, 0.2022, 0.2032, 0.2042, 0.20521, 0.20621, 0.20721, 0.
20821, 0.20921, 0.21021, 0.21121, 0.21221, 0.21321, 0.21421, 0.21522,
0.21622, 0.21722, 0.21822, 0.21922, 0.22022, 0.22122, 0.22222, 0.22322, 0.22422, 0.22523, 0.22623, 0.22723, 0.22823, 0.22923, 0.23023, 0.23123, 0.
23223, 0.23323, 0.23423, 0.23524, 0.23624, 0.23724, 0.23824, 0.23924,
0.24024, 0.24124, 0.24224, 0.24324, 0.24424, 0.24525, 0.24625, 0.24725, 0.24825, 0.24925, 0.25025, 0.25125, 0.25225, 0.25325, 0.25425, 0.25526, 0.
25626, 0.25726, 0.25826, 0.25926, 0.26026, 0.26126, 0.26226, 0.26326,
0.26426, 0.26527, 0.26627, 0.26727, 0.26827, 0.26927, 0.27027, 0.27127, 0.27227, 0.27327, 0.27427, 0.27528, 0.27628, 0.27728, 0.27828, 0.27928, 0.
28028, 0.28128, 0.28228, 0.28328, 0.28428, 0.28529, 0.28629, 0.28729,
0.28829, 0.28929, 0.29029, 0.29129, 0.29229, 0.29329, 0.29429, 0.2953, 0.2963, 0.2973, 0.2983, 0.2993, 0.3003, 0.3013, 0.3023, 0.3033,
0.3043, 0.30531, 0.30631, 0.30731, 0.30831, 0.30931, 0.31031, 0.31131,
0.31231, 0.31331, 0.31431, 0.31532, 0.31632, 0.31732, 0.31832, 0.31932, 0.32032, 0.32132, 0.32232, 0.32332, 0.32432, 0.32533, 0.32633, 0.32733, 0.
32833, 0.32933, 0.33033, 0.33133, 0.33233, 0.33333, 0.33433, 0.33534,
0.33634, 0.33734, 0.33834, 0.33934, 0.34034, 0.34134, 0.34234, 0.34334, 0.34434, 0.34535, 0.34635, 0.34735, 0.34835, 0.34935, 0.35035, 0.35135, 0.
35235, 0.35335, 0.35435, 0.35536, 0.35636, 0.35736, 0.35836, 0.35936,
0.36036, 0.36136, 0.36236, 0.36336, 0.36436, 0.36537, 0.36637, 0.36737, 0.36837, 0.36937, 0.37037, 0.37137, 0.37237, 0.37337, 0.37437, 0.37538, 0.
37638, 0.37738, 0.37838, 0.37938, 0.38038, 0.38138, 0.38238, 0.38338,
0.38438, 0.38539, 0.38639, 0.38739, 0.38839, 0.38939, 0.39039, 0.39139, 0.39239, 0.39339, 0.39439, 0.3954, 0.3964, 0.3974, 0.3984, 0.3994,
0.4004, 0.4014, 0.4024, 0.4034, 0.4044, 0.40541, 0.40641, 0.40741,
0.40841, 0.40941, 0.41041, 0.41141, 0.41241, 0.41341, 0.41441, 0.41542, 0.41642, 0.41742, 0.41842, 0.41942, 0.42042, 0.42142, 0.42242, 0.42342, 0.
42442, 0.42543, 0.42643, 0.42743, 0.42843, 0.42943, 0.43043, 0.43143,
0.43243, 0.43343, 0.43443, 0.43544, 0.43644, 0.43744, 0.43844, 0.43944, 0.44044, 0.44144, 0.44244, 0.44344, 0.44444, 0.44545, 0.44645, 0.44745, 0.
44845, 0.44945, 0.45045, 0.45145, 0.45245, 0.45345, 0.45445, 0.45546,
0.45646, 0.45746, 0.45846, 0.45946, 0.46046, 0.46146, 0.46246, 0.46346, 0.46446, 0.46547, 0.46647, 0.46747, 0.46847, 0.46947, 0.47047, 0.47147, 0.
47247, 0.47347, 0.47447, 0.47548, 0.47648, 0.47748, 0.47848, 0.47948,
0.48048, 0.48148, 0.48248, 0.48348, 0.48448, 0.48549, 0.48649, 0.48749, 0.48849, 0.48949, 0.49049, 0.49149, 0.49249, 0.49349, 0.49449, 0.4955,
0.4965, 0.4975, 0.4985, 0.4995, 0.5005, 0.5015, 0.5025, 0.5035,
0.5045, 0.50551, 0.50651, 0.50751, 0.50851, 0.50951, 0.51051, 0.51151, 0.51251, 0.51351, 0.51451, 0.51552, 0.51652, 0.51752, 0.51852, 0.51952, 0.
52052, 0.52152, 0.52252, 0.52352, 0.52452, 0.52553, 0.52653, 0.52753,
0.52853, 0.52953, 0.53053, 0.53153, 0.53253, 0.53353, 0.53453, 0.53554, 0.53654, 0.53754, 0.53854, 0.53954, 0.54054, 0.54154, 0.54254, 0.54354, 0.
54454, 0.54555, 0.54655, 0.54755, 0.54855, 0.54955, 0.55055, 0.55155,
0.55255, 0.55355, 0.55455, 0.55556, 0.55656, 0.55756, 0.55856, 0.55956, 0.56056, 0.56156, 0.56256, 0.56356, 0.56456, 0.56557, 0.56657, 0.56757, 0.
56857, 0.56957, 0.57057, 0.57157, 0.57257, 0.57357, 0.57457, 0.57558,
0.57658, 0.57758, 0.57858, 0.57958, 0.58058, 0.58158, 0.58258, 0.58358, 0.58458, 0.58559, 0.58659, 0.58759, 0.58859, 0.58959, 0.59059, 0.59159, 0.
59259, 0.59359, 0.59459, 0.59556, 0.59656, 0.59756, 0.59856, 0.5996,
0.6006, 0.6016, 0.6026, 0.6036, 0.6046, 0.60561, 0.60661, 0.60761, 0.60861, 0.60961, 0.61061, 0.61161, 0.61261, 0.61361, 0.61461, 0.61562, 0.
61662, 0.61762, 0.61862, 0.61962, 0.62062, 0.62162, 0.62262, 0.62362,
0.62462, 0.62563, 0.62663, 0.62763, 0.62863, 0.62963, 0.63063, 0.63163, 0.63263, 0.63363, 0.63463, 0.63564, 0.63664, 0.63764, 0.63864, 0.63964, 0.
64064, 0.64164, 0.64264, 0.64364, 0.64464, 0.64565, 0.64665, 0.64765,
0.64865, 0.64965, 0.65065, 0.65165, 0.65265, 0.65365, 0.65465, 0.65566, 0.65666, 0.65766, 0.65866, 0.65966, 0.66066, 0.66166, 0.66266, 0.66366, 0.
66466, 0.66567, 0.66667, 0.66767, 0.66867, 0.66967, 0.67067, 0.67167,
0.67267, 0.67367, 0.67467, 0.67568, 0.67668, 0.67768, 0.67868, 0.67968, 0.68068, 0.68168, 0.68268, 0.68368, 0.68468, 0.68569, 0.68669, 0.68769, 0.
68869, 0.68969, 0.69069, 0.69169, 0.69269, 0.69369, 0.69469, 0.6957,
0.6967, 0.6977, 0.6987, 0.6997, 0.7007, 0.7017, 0.7027, 0.7037, 0.7047, 0.70571, 0.70671, 0.70771, 0.70871, 0.70971, 0.71071, 0.71171, 0.
71271, 0.71371, 0.71471, 0.71572, 0.71672, 0.71772, 0.71872, 0.71972,
0.72072, 0.72172, 0.72272, 0.72372, 0.72472, 0.72573, 0.72673, 0.72773, 0.72873, 0.72973, 0.73073, 0.73173, 0.73273, 0.73373, 0.73473, 0.73574, 0.
73674, 0.73774, 0.73874, 0.73974, 0.74074, 0.74174, 0.74274,

```

16/29

localhost:8888/lab/tree/Documents/Intel Unnati/Final/ANPR.ipynb? 17/29

	0.33634,	0.33734,	0.33834,	0.33934,	0.34034,	0.34134,	0.34234,	0.34334,	0.34434,	0.34535,	0.34635,	0.34735,	0.34835,	0.34935,	0.35035,	0.35135,	0.
35235,	0.35335,	0.35435,	0.35536,	0.35636,	0.35736,	0.35836,	0.35936,										
	0.36036,	0.36136,	0.36236,	0.36336,	0.36436,	0.36537,	0.36637,	0.36737,	0.36837,	0.36937,	0.37037,	0.37137,	0.37237,	0.37337,	0.37437,	0.37538,	0.
37638,	0.37738,	0.37838,	0.37938,	0.38038,	0.38138,	0.38238,	0.38338,										
	0.38438,	0.38539,	0.38639,	0.38739,	0.38839,	0.38939,	0.39039,	0.39139,	0.39239,	0.39339,	0.39439,	0.3954,	0.3964,	0.3974,	0.3984,	0.3994,	
0.4004,	0.4014,	0.4024,	0.4034,	0.4044,	0.40541,	0.40641,	0.40741,										
	0.40841,	0.40941,	0.41041,	0.41141,	0.41241,	0.41341,	0.41441,	0.41542,	0.41642,	0.41742,	0.41842,	0.41942,	0.42042,	0.42142,	0.42242,	0.42342,	0.
42442,	0.42543,	0.42643,	0.42743,	0.42843,	0.42943,	0.43043,	0.43143,										
	0.43243,	0.43343,	0.43443,	0.43544,	0.43644,	0.43744,	0.43844,	0.43944,	0.44044,	0.44144,	0.44244,	0.44344,	0.44444,	0.44545,	0.44645,	0.44745,	0.
44845,	0.44945,	0.45045,	0.45145,	0.45245,	0.45345,	0.45445,	0.45546,										
	0.45646,	0.45746,	0.45846,	0.45946,	0.46046,	0.46146,	0.46246,	0.46346,	0.46446,	0.46547,	0.46647,	0.46747,	0.46847,	0.46947,	0.47047,	0.47147,	0.
47247,	0.47347,	0.47447,	0.47548,	0.47648,	0.47748,	0.47848,	0.47948,										
	0.48048,	0.48148,	0.48248,	0.48348,	0.48448,	0.48549,	0.48649,	0.48749,	0.48849,	0.48949,	0.49049,	0.49149,	0.49249,	0.49349,	0.49449,	0.4955,	
0.4965,	0.4975,	0.4985,	0.4995,	0.5005,	0.5015,	0.5025,	0.5035,										
	0.5045,	0.50551,	0.50651,	0.50751,	0.50851,	0.50951,	0.51051,	0.51151,	0.51251,	0.51351,	0.51451,	0.51552,	0.51652,	0.51752,	0.51852,	0.51952,	0.
52052,	0.52152,	0.52252,	0.52352,	0.52452,	0.52553,	0.52653,	0.52753,										
	0.52853,	0.52953,	0.53053,	0.53153,	0.53253,	0.53353,	0.53453,	0.53554,	0.53654,	0.53754,	0.53854,	0.53954,	0.54054,	0.54154,	0.54254,	0.54354,	0.
54454,	0.54555,	0.54655,	0.54755,	0.54855,	0.54955,	0.55055,	0.55155,										
	0.55255,	0.55355,	0.55455,	0.55556,	0.55656,	0.55756,	0.55856,	0.55956,	0.56056,	0.56156,	0.56256,	0.56356,	0.56456,	0.56557,	0.56657,	0.56757,	0.
56857,	0.56957,	0.57057,	0.57157,	0.57257,	0.57357,	0.57457,	0.57558,										
	0.57658,	0.57758,	0.57858,	0.57958,	0.58058,	0.58158,	0.58258,	0.58358,	0.58458,	0.58559,	0.58659,	0.58759,	0.58859,	0.58959,	0.59059,	0.59159,	0.
59259,	0.59359,	0.59459,	0.5956,	0.5966,	0.5976,	0.5986,	0.5996,										
	0.6006,	0.6016,	0.6026,	0.6036,	0.6046,	0.60561,	0.60661,	0.60761,	0.60861,	0.60961,	0.61061,	0.61161,	0.61261,	0.61361,	0.61461,	0.61562,	0.
61662,	0.61762,	0.61862,	0.61962,	0.62062,	0.62162,	0.62262,	0.62362,										
	0.62462,	0.62563,	0.62663,	0.62763,	0.62863,	0.62963,	0.63063,	0.63163,	0.63263,	0.63363,	0.63463,	0.63564,	0.63664,	0.63764,	0.63864,	0.63964,	0.
64064,	0.64164,	0.64264,	0.64364,	0.64464,	0.64565,	0.64665,	0.64765,										
	0.64865,	0.64965,	0.65065,	0.65165,	0.65265,	0.65365,	0.65465,	0.65566,	0.65666,	0.65766,	0.65866,	0.65966,	0.66066,	0.66166,	0.66266,	0.66366,	0.
66466,	0.66567,	0.66667,	0.66767,	0.66867,	0.66967,	0.67067,	0.67167,										
	0.67267,	0.67367,	0.67467,	0.67568,	0.67668,	0.67768,	0.67868,	0.67968,	0.68068,	0.68168,	0.68268,	0.68368,	0.68468,	0.68569,	0.68669,	0.68769,	0.
68869,	0.68969,	0.69069,	0.69169,	0.69269,	0.69369,	0.69469,	0.6957,										
	0.6967,	0.6977,	0.6987,	0.6997,	0.7007,	0.7017,	0.7027,	0.7037,	0.7047,	0.70571,	0.70671,	0.70771,	0.70871,	0.70971,	0.71071,	0.71171,	0.
71271,	0.71371,	0.71471,	0.71572,	0.71672,	0.71772,	0.71872,	0.71972,										
	0.72072,	0.72172,	0.72272,	0.72372,	0.72472,	0.72573,	0.72673,	0.72773,	0.72873,	0.72973,	0.73073,	0.73173,	0.73273,	0.73373,	0.73473,	0.73574,	0.
73674,	0.73774,	0.73874,	0.73974,	0.74074,	0.74174,	0.74274,	0.74374,										
	0.74474,	0.74575,	0.74675,	0.74775,	0.74875,	0.74975,	0.75075,	0.75175,	0.75275,	0.75375,	0.75475,	0.75576,	0.75676,	0.75776,	0.75876,	0.75976,	0.
76076,	0.76176,	0.76276,	0.76376,	0.76476,	0.76577,	0.76677,	0.76777,										
	0.76877,	0.76977,	0.77077,	0.77177,	0.77277,	0.77377,	0.77477,	0.77578,	0.77678,	0.77778,	0.77878,	0.77978,	0.78078,	0.78178,	0.78278,	0.78378,	0.
78478,	0.78579,	0.78679,	0.78779,	0.78879,	0.78979,	0.79079,	0.79179,										
	0.79279,	0.79379,	0.79479,	0.7958,	0.7968,	0.7978,	0.7988,	0.7998,	0.8008,	0.8018,	0.8028,	0.8038,	0.8048,	0.80581,	0.80681,	0.80781,	0.
80881,	0.80981,	0.81081,	0.81181,	0.81281,	0.81381,	0.81481,	0.81582,										
	0.81682,	0.81782,	0.81882,	0.81982,	0.82082,	0.82182,	0.82282,	0.82382,	0.82482,	0.82583,	0.82683,	0.82783,	0.82883,	0.82983,	0.83083,	0.83183,	0.
83283,	0.83383,	0.83483,	0.83584,	0.83684,	0.83784,	0.83884,	0.83984,										
	0.84084,	0.84184,	0.84284,	0.84384,	0.84484,	0.84585,	0.84685,	0.84785,	0.84885,	0.84985,	0.85085,	0.85185,	0.85285,	0.85385,	0.85485,	0.85586,	0.
85686,	0.85786,	0.85886,	0.85986,	0.86086,	0.86186,	0.86286,	0.86386,										
	0.86486,	0.86587,	0.86687,	0.86787,	0.86887,	0.86987,	0.87087,	0.87187,	0.87287,	0.87387,	0.87487,	0.87588,	0.87688,	0.87788,	0.87888,	0.87988,	0.
88088,	0.88188,	0.88288,	0.88388,	0.88488,	0.88589,	0.88689,	0.88789,										
	0.88889,	0.88989,	0.89089,	0.89189,	0.89289,	0.89389,	0.89489,	0.8959,	0.8969,	0.8979,	0.8989,	0.8999,	0.9009,	0.9019,	0.9029,	0.9039,	
0.9049,	0.90591,	0.90691,	0.90791,	0.90891,	0.90991,	0.91091,	0.91191,										
	0.91291,	0.91391,	0.91491,	0.91592,	0.91692,	0.91792,	0.91892,	0.91992,	0.92092,	0.92192,	0.92292,	0.92392,	0.92492,	0.92593,	0.92693,	0.92793,	0.
92893,	0.92993,	0.93093,	0.93193,	0.93293,	0.93393,	0.93493,	0.93594,										
	0.93694,	0.93794,	0.93894,	0.93994,	0.94094,	0.94194,	0.94294,	0.94394,	0.94494,	0.94595,	0.94695,	0.94795,	0.94895,	0.94995,	0.95095,	0.95195,	0.
95295,	0.95395,	0.95495,	0.95596,	0.95696,	0.95796,	0.95896,	0.95996,										
	0.96096,	0.96196,	0.96296,	0.96396,	0.96496,	0.96597,	0.96697,	0.96797,	0.96897,	0.96997,	0.97097,	0.97197,	0.97297,	0.97397,	0.97497,	0.97598,	0.
97698,	0.97798,	0.97898,	0.97998,	0.98098,	0.98198,	0.98298,	0.98398,										
	0.98498,	0.98599,	0.98699,	0.98799,	0.98899,	0.98999,	0.99099,	0.99199,	0.99299,	0.99399,	0.99499,	0.996,	0.997,	0.998,	0.999,	1]], arra	
y([[0.42424,	0.42424,	0.5455,	0.65221,	0.70683,	0.71717,	0.72036,	0.72294,	0.72757,	0.7361,	0.74648,	0.75777,	0.76149,	0.77058,	0.78585,	0.78776,	0.789
67,	0.79156,	0.79304,	0.79415,	0.79526,	0.79637,	0.79747,											
	0.79857,	0.79967,	0.81141,	0.81516,	0.81373,	0.8123,	0.81086,	0.80943,	0.80799,	0.80655,	0.80511,	0.80533,	0.81224,	0.81423,	0.81621,	0.81819,	
0.82005,	0.82061,	0.82118,	0.82175,	0.82231,	0.82288,	0.82344,											
	0.824,	0.82457,	0.82513,	0.82569,	0.82625,	0.82681,	0.82737,	0.82793,	0.82917,	0.83153,	0.83389,	0.83623,	0.83686,	0.83702,	0.83718,	0.83734,	
0.8375,	0.83765,	0.83781,	0.83797,	0.83813,	0.83829,	0.83845,											
	0.83861,	0.83876,	0.83892,	0.83908,	0.83924,	0.8394,	0.83956,	0.83971,	0.83987,	0.84003,	0.84019,	0.84035,	0.8405,	0.84066,	0.84082,	0.84098,	
0.84114,	0.84129,	0.84145,	0.84161,	0.84177,	0.84192,	0.84208,											
	0.84224,	0.8424,	0.84255,	0.84271,	0.84287,	0.84302,	0.84318,	0.84334,	0.8435,	0.84365,	0.84381,	0.84397,	0.84412,	0.84428,	0.84444,	0.84459,	
0.84475,	0.84491,	0.84506,	0.84522,	0.84544,	0.84622,	0.847,											
	0.84778,	0.84856,	0.84934,	0.85012,	0.8509,	0.85167,	0.85244,	0.85322,	0.85399,	0.85439,	0.85468,	0.85497,	0.85526,	0.85555,	0.85584,	0.85614,	
0.85643,	0.85672,	0.85701,	0.85729,	0.85758,	0.85787,	0.85816,											
	0.85845,	0.85874,	0.85903,	0.85932,	0.85961,	0.85989,	0.86018,	0.86047,	0.86076,	0.86105,	0.86133,	0.86162,	0.86191,	0.86219,	0.86248,	0.86277,	

0.86305,	0.86333,	0.8636,	0.86387,	0.86414,	0.86441,	0.86468,										
	0.86495,	0.86522,	0.86549,	0.86576,	0.86603,	0.8663,	0.86657,	0.86684,	0.86711,	0.86738,	0.86765,	0.86792,	0.86818,	0.86845,	0.86872,	0.86899,
0.86926,	0.86952,	0.86979,	0.87006,	0.87033,	0.87059,	0.87086,										
	0.87113,	0.87139,	0.87166,	0.87193,	0.87219,	0.86768,	0.86023,	0.8603,	0.86036,	0.86043,	0.86049,	0.86055,	0.86062,	0.86068,	0.86074,	0.86081,
0.86087,	0.86094,	0.861,	0.86106,	0.86113,	0.86119,	0.86126,										
	0.86132,	0.86138,	0.86145,	0.86151,	0.86157,	0.86164,	0.8617,	0.86177,	0.86183,	0.86189,	0.86196,	0.86202,	0.86208,	0.86215,	0.86221,	0.86227,
0.86234,	0.8624,	0.86247,	0.86253,	0.86259,	0.86266,	0.86272,										
	0.86278,	0.86285,	0.86291,	0.86297,	0.86304,	0.8631,	0.86316,	0.86323,	0.86329,	0.86336,	0.86342,	0.86348,	0.86355,	0.86361,	0.86367,	0.86374,
0.8638,	0.86386,	0.86393,	0.86399,	0.86405,	0.86412,	0.86418,										
	0.86424,	0.86431,	0.86437,	0.86443,	0.8645,	0.86456,	0.86462,	0.86469,	0.86475,	0.86481,	0.86488,	0.86494,	0.865,	0.86507,	0.86513,	0.86519,
0.86526,	0.86532,	0.86538,	0.86545,	0.86551,	0.86557,	0.86564,										
	0.8657,	0.86576,	0.86583,	0.86589,	0.86595,	0.86601,	0.86608,	0.86614,	0.8662,	0.86627,	0.86633,	0.86639,	0.86646,	0.86652,	0.86658,	0.86665,
0.86671,	0.86677,	0.86684,	0.8669,	0.86696,	0.86702,	0.86709,										
	0.86715,	0.86721,	0.86728,	0.86734,	0.8674,	0.86747,	0.86753,	0.86759,	0.86765,	0.86772,	0.86778,	0.86784,	0.86791,	0.86797,	0.86803,	0.86809,
0.86816,	0.86822,	0.86828,	0.86835,	0.86841,	0.86847,	0.86853,										
	0.8686,	0.86866,	0.86872,	0.86879,	0.86885,	0.86891,	0.86897,	0.86904,	0.8691,	0.86916,	0.86922,	0.86929,	0.86935,	0.86941,	0.86948,	0.86954,
0.86962,	0.86971,	0.8698,	0.8699,	0.86999,	0.87008,	0.87017,										
	0.87027,	0.87036,	0.87045,	0.87055,	0.87064,	0.87073,	0.87082,	0.87092,	0.87101,	0.8711,	0.87119,	0.87129,	0.87138,	0.87147,	0.87156,	0.87166,
0.87175,	0.87184,	0.87193,	0.87203,	0.87212,	0.87221,	0.8723,										
	0.8724,	0.87249,	0.87258,	0.87267,	0.87276,	0.87286,	0.87295,	0.87304,	0.87313,	0.87323,	0.87332,	0.87341,	0.8735,	0.87359,	0.87369,	0.87378,
0.87387,	0.87396,	0.87405,	0.87415,	0.87424,	0.87433,	0.87442,										
	0.87451,	0.87461,	0.8747,	0.87479,	0.87488,	0.87497,	0.87506,	0.87516,	0.87525,	0.87534,	0.87543,	0.87552,	0.87562,	0.87571,	0.8758,	0.87589,
0.87598,	0.87607,	0.87617,	0.87626,	0.87635,	0.87644,	0.87653,										
	0.87662,	0.87671,	0.87681,	0.8769,	0.87699,	0.87708,	0.87717,	0.87726,	0.87735,	0.87745,	0.87754,	0.87763,	0.87772,	0.87781,	0.8779,	0.87799,
0.87808,	0.87818,	0.87827,	0.87836,	0.87845,	0.87854,	0.87863,										
	0.87872,	0.87881,	0.8789,	0.879,	0.87909,	0.87935,	0.87972,	0.88008,	0.88045,	0.88081,	0.88118,	0.88154,	0.88191,	0.88227,	0.88264,	0.883,
0.88336,	0.88373,	0.88409,	0.88445,	0.88481,	0.88518,	0.88554,										
	0.8859,	0.88626,	0.88662,	0.88698,	0.88734,	0.8877,	0.88806,	0.88842,	0.88878,	0.88764,	0.88586,	0.88408,	0.88229,	0.8805,	0.8787,	0.8769,
0.87798,	0.88015,	0.88231,	0.88446,	0.88632,	0.8859,	0.88549,										
	0.88507,	0.88465,	0.88424,	0.88382,	0.8834,	0.88299,	0.88257,	0.88215,	0.88173,	0.88132,	0.8809,	0.88048,	0.88006,	0.87964,	0.87922,	0.8788,
0.87838,	0.87796,	0.87754,	0.87712,	0.8767,	0.87628,	0.87586,										
	0.87543,	0.87501,	0.87459,	0.87417,	0.87375,	0.8738,	0.87422,	0.87464,	0.87505,	0.87547,	0.87589,	0.8763,	0.87672,	0.87713,	0.87755,	0.87796,
0.87838,	0.87879,	0.8792,	0.87962,	0.88003,	0.88044,	0.88085,										
	0.88126,	0.88168,	0.88209,	0.8825,	0.88291,	0.88331,	0.88389,	0.8942,	0.89432,	0.89445,	0.89457,	0.8947,	0.89482,	0.89495,	0.89507,	0.8952,
0.89533,	0.89545,	0.89558,	0.8957,	0.89583,	0.89595,	0.89608,										
	0.8962,	0.89633,	0.89645,	0.89658,	0.8967,	0.89682,	0.89695,	0.89707,	0.8972,	0.89732,	0.89745,	0.89757,	0.8977,	0.89782,	0.89795,	0.89807,
0.8982,	0.89832,	0.89844,	0.89857,	0.89869,	0.89882,	0.89894,										
	0.89906,	0.89919,	0.89931,	0.89944,	0.89956,	0.89968,	0.89981,	0.89993,	0.90006,	0.90018,	0.9003,	0.90043,	0.90055,	0.90067,	0.9008,	0.90092,
0.90105,	0.90117,	0.90129,	0.90142,	0.90154,	0.90166,	0.90179,										
	0.90191,	0.90203,	0.90216,	0.90228,	0.9024,	0.90252,	0.90265,	0.90277,	0.90289,	0.90302,	0.90314,	0.90326,	0.90339,	0.90351,	0.90363,	0.90375,
0.90388,	0.904,	0.90412,	0.90424,	0.90437,	0.90449,	0.90461,										
	0.90473,	0.90497,	0.90523,	0.9055,	0.90577,	0.90603,	0.9063,	0.90656,	0.90683,	0.9071,	0.90736,	0.90763,	0.90789,	0.90816,	0.90842,	0.90868,
0.90895,	0.90921,	0.90948,	0.90974,	0.91,	0.91027,	0.91053,										
	0.91079,	0.91106,	0.91132,	0.91158,	0.91184,	0.91211,	0.91237,	0.91263,	0.91289,	0.91315,	0.91342,	0.91368,	0.91394,	0.9142,	0.91446,	0.91472,
0.91498,	0.91524,	0.9155,	0.91512,	0.91367,	0.91222,	0.91076,										
	0.9093,	0.90784,	0.90637,	0.9049,	0.90343,	0.90257,	0.90296,	0.90335,	0.90373,	0.90412,	0.90451,	0.9049,	0.90529,	0.90567,	0.90606,	0.90645,
0.90683,	0.90722,	0.9076,	0.90799,	0.90837,	0.90875,	0.90914,										
	0.90952,	0.9099,	0.91029,	0.91067,	0.91105,	0.91143,	0.91181,	0.91219,	0.91257,	0.91295,	0.91333,	0.91116,	0.90398,	0.89977,	0.89927,	0.89876,
0.89826,	0.89775,	0.89725,	0.89674,	0.89623,	0.89573,	0.89522,										
	0.89471,	0.8942,	0.89369,	0.89318,	0.89267,	0.89216,	0.89165,	0.89114,	0.89063,	0.89012,	0.8896,	0.88909,	0.88858,	0.88806,	0.88755,	0.88703,
0.88652,	0.88581,	0.88398,	0.88214,	0.88029,	0.87844,	0.87658,										
	0.87471,	0.87284,	0.87131,	0.87021,	0.8691,	0.868,	0.86689,	0.86578,	0.86467,	0.86355,	0.86243,	0.86132,	0.86019,	0.85907,	0.85794,	0.85583,
0.85127,	0.84667,	0.8421,	0.84149,	0.84088,	0.84027,	0.83965,										
	0.83904,	0.83843,	0.83782,	0.8372,	0.83659,	0.83597,	0.83536,	0.83474,	0.83412,	0.8335,	0.83289,	0.83227,	0.83165,	0.83103,	0.83041,	0.82978,
0.82916,	0.82854,	0.82791,	0.82729,	0.82663,	0.8262,	0.80965,										
	0.80831,	0.80698,	0.80564,	0.80429,	0.80295,	0.8016,	0.80024,	0.79889,	0.79753,	0.79617,	0.7948,	0.79378,	0.79284,	0.7919,	0.79095,	0.79001,
0.78906,	0.78811,	0.78717,	0.78622,	0.78526,	0.78431,	0.78336,										
	0.7824,	0.78144,	0.78048,	0.77952,	0.77856,	0.7772,	0.77414,	0.77107,	0.76798,	0.76487,	0.76175,	0.7577,	0.75308,	0.74842,	0.74373,	0.73918,
0.73464,	0.73007,	0.72546,	0.72212,	0.71905,	0.71596,	0.71286,										
	0.70974,	0.7066,	0.69957,	0.69129,	0.68434,	0.67918,	0.67398,	0.66874,	0.65995,	0.64871,	0.63675,	0.6244,	0.61144,	0.59772,	0.58283,	0.50333,
0.48674,	0.47749,	0.47052,	0.4635,	0.4564,	0.45143,	0.4465,										
	0.44155,	0.43656,	0.43154,	0.41001,	0.38157,	0.33327,	0.31826,	0.30273,	0.28636,	0.27328,	0.26898,	0.26467,	0.26033,	0.25597,	0.25159,	0.24719,
0.24276,	0.23559,	0.22384,	0.21194,	0.20256,	0.19817,	0.19376,										
	0.18932,	0.18487,	0.18039,	0.17589,	0.17136,	0.16682,	0.12324,	0.11865,	0.11403,	0.10939,	0.10473,	0.10004,	0.095335,	0.090603,	0.084051,	0.07146,
058703,	0.045776,	0,	0,	0,	0,	0,										0.
	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,
0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,	0,

20/29

	0.74474,	0.74575,	0.74675,	0.74775,	0.74875,	0.74975,	0.75075,	0.75175,	0.75275,	0.75375,	0.75475,	0.75576,	0.75676,	0.75776,	0.75876,	0.75976,	0.
76076,	0.76176,	0.76276,	0.76376,	0.76476,	0.76577,	0.76677,	0.76777,										
	0.76877,	0.76977,	0.77077,	0.77177,	0.77277,	0.77377,	0.77477,	0.77578,	0.77678,	0.77778,	0.77878,	0.77978,	0.78078,	0.78178,	0.78278,	0.78378,	0.
78478,	0.78579,	0.78679,	0.78779,	0.78879,	0.78979,	0.79079,	0.79179,										
	0.79279,	0.79379,	0.79479,	0.7958,	0.7968,	0.7978,	0.7988,	0.7998,	0.8008,	0.8018,	0.8028,	0.8038,	0.8048,	0.80581,	0.80681,	0.80781,	0.
80881,	0.80981,	0.81081,	0.81181,	0.81281,	0.81381,	0.81481,	0.81582,										
	0.81682,	0.81782,	0.81882,	0.81982,	0.82082,	0.82182,	0.82282,	0.82382,	0.82482,	0.82583,	0.82683,	0.82783,	0.82883,	0.82983,	0.83083,	0.83183,	0.
83283,	0.83383,	0.83483,	0.83584,	0.83684,	0.83784,	0.83884,	0.83984,										
	0.84084,	0.84184,	0.84284,	0.84384,	0.84484,	0.84585,	0.84685,	0.84785,	0.84885,	0.84985,	0.85085,	0.85185,	0.85285,	0.85385,	0.85485,	0.85586,	0.
85686,	0.85786,	0.85886,	0.85986,	0.86086,	0.86186,	0.86286,	0.86386,										
	0.86486,	0.86587,	0.86687,	0.86787,	0.86887,	0.86987,	0.87087,	0.87187,	0.87287,	0.87387,	0.87487,	0.87588,	0.87688,	0.87788,	0.87888,	0.87988,	0.
88088,	0.88188,	0.88288,	0.88388,	0.88488,	0.88589,	0.88689,	0.88789,										
	0.88889,	0.88989,	0.89089,	0.89189,	0.89289,	0.89389,	0.89489,	0.8959,	0.8969,	0.8979,	0.8989,	0.8999,	0.9009,	0.9019,	0.9029,	0.9039,	
0.9049,	0.90591,	0.90691,	0.90791,	0.90891,	0.90991,	0.91091,	0.91191,										
	0.91291,	0.91391,	0.91491,	0.91592,	0.91692,	0.91792,	0.91892,	0.91992,	0.92092,	0.92192,	0.92292,	0.92392,	0.92492,	0.92593,	0.92693,	0.92793,	0.
92893,	0.92993,	0.93093,	0.93193,	0.93293,	0.93393,	0.93493,	0.93594,										
	0.93694,	0.93794,	0.93894,	0.93994,	0.94094,	0.94194,	0.94294,	0.94394,	0.94494,	0.94595,	0.94695,	0.94795,	0.94895,	0.94995,	0.95095,	0.95195,	0.
95295,	0.95395,	0.95495,	0.95596,	0.95696,	0.95796,	0.95896,	0.95996,										
	0.96096,	0.96196,	0.96296,	0.96396,	0.96496,	0.96597,	0.96697,	0.96797,	0.96897,	0.96997,	0.97097,	0.97197,	0.97297,	0.97397,	0.97497,	0.97598,	0.
97698,	0.97798,	0.97898,	0.97998,	0.98098,	0.98198,	0.98298,	0.98398,										
	0.98498,	0.98599,	0.98699,	0.98799,	0.98899,	0.98999,	0.99099,	0.99199,	0.99299,	0.99399,	0.99499,	0.996,	0.997,	0.998,	0.999,	1]],	arra
y([[0.27273,	0.27273,	0.38186,	0.49533,	0.5612,	0.57435,	0.57845,	0.58179,	0.5878,	0.59902,	0.61289,	0.62825,	0.63339,	0.64607,	0.66783,	0.67059,	0.673
36,	0.67612,	0.67828,	0.67991,	0.68153,	0.68316,	0.68478,											
	0.68641,	0.68803,	0.70561,	0.7117,	0.71109,	0.71047,	0.70986,	0.70925,	0.70863,	0.70802,	0.7074,	0.70908,	0.71986,	0.72299,	0.72613,	0.72926,	
0.73221,	0.73312,	0.73402,	0.73493,	0.73584,	0.73674,	0.73765,											
	0.73855,	0.73946,	0.74036,	0.74127,	0.74217,	0.74308,	0.74398,	0.74489,	0.74689,	0.75074,	0.75459,	0.75844,	0.75947,	0.75973,	0.75999,	0.76025,	
0.76051,	0.76077,	0.76104,	0.7613,	0.76156,	0.76182,	0.76208,											
	0.76235,	0.76261,	0.76287,	0.76313,	0.76339,	0.76366,	0.76392,	0.76418,	0.76444,	0.7647,	0.76497,	0.76523,	0.76549,	0.76575,	0.76601,	0.76628,	
0.76654,	0.7668,	0.76706,	0.76732,	0.76759,	0.76785,	0.76811,											
	0.76837,	0.76863,	0.7689,	0.76916,	0.76942,	0.76968,	0.76994,	0.7702,	0.77047,	0.77073,	0.77099,	0.77125,	0.77151,	0.77178,	0.77204,	0.7723,	
0.77256,	0.77282,	0.77309,	0.77335,	0.77371,	0.77502,	0.77634,											
	0.77765,	0.77896,	0.78028,	0.78159,	0.78291,	0.78422,	0.78553,	0.78685,	0.78816,	0.78884,	0.78934,	0.78984,	0.79033,	0.79083,	0.79133,	0.79182,	
0.79232,	0.79282,	0.79331,	0.79381,	0.79431,	0.7948,	0.7953,											
	0.7958,	0.79629,	0.79679,	0.79728,	0.79778,	0.79828,	0.79877,	0.79927,	0.79977,	0.80026,	0.80076,	0.80126,	0.80175,	0.80225,	0.80275,	0.80324,	
0.80374,	0.80422,	0.80469,	0.80516,	0.80563,	0.8061,	0.80657,											
	0.80704,	0.80751,	0.80798,	0.80845,	0.80892,	0.80939,	0.80986,	0.81033,	0.8108,	0.81127,	0.81174,	0.81221,	0.81269,	0.81316,	0.81363,	0.8141,	
0.81457,	0.81504,	0.81551,	0.81598,	0.81645,	0.81692,	0.81739,											
	0.81786,	0.81833,	0.8188,	0.81927,	0.81974,	0.81858,	0.81636,	0.81648,	0.81659,	0.81671,	0.81682,	0.81694,	0.81705,	0.81717,	0.81728,	0.8174,	
0.81751,	0.81763,	0.81774,	0.81786,	0.81797,	0.81809,	0.8182,											
	0.81832,	0.81843,	0.81855,	0.81866,	0.81878,	0.81889,	0.81901,	0.81912,	0.81924,	0.81935,	0.81947,	0.81958,	0.8197,	0.81981,	0.81993,	0.82004,	
0.82016,	0.82027,	0.82039,	0.8205,	0.82062,	0.82073,	0.82085,											
	0.82096,	0.82108,	0.8212,	0.82131,	0.82143,	0.82154,	0.82166,	0.82177,	0.82189,	0.822,	0.82212,	0.82223,	0.82235,	0.82246,	0.82258,	0.82269,	
0.82281,	0.82292,	0.82304,	0.82315,	0.82327,	0.82338,	0.8235,											
	0.82361,	0.82373,	0.82384,	0.82396,	0.82407,	0.82419,	0.8243,	0.82442,	0.82453,	0.82465,	0.82476,	0.82488,	0.82499,	0.82511,	0.82522,	0.82534,	
0.82545,	0.82557,	0.82569,	0.8258,	0.82592,	0.82603,	0.82615,											
	0.82626,	0.82638,	0.82649,	0.82661,	0.82672,	0.82684,	0.82695,	0.82707,	0.82718,	0.8273,	0.82741,	0.82753,	0.82764,	0.82776,	0.82787,	0.82799,	
0.8281,	0.82822,	0.82833,	0.82845,	0.82856,	0.82868,	0.82879,											
	0.82891,	0.82902,	0.82914,	0.82925,	0.82937,	0.82948,	0.8296,	0.82971,	0.82983,	0.82994,	0.83006,	0.83017,	0.83029,	0.83041,	0.83052,	0.83064,	
0.83075,	0.83087,	0.83098,	0.8311,	0.83121,	0.83133,	0.83144,											
	0.83156,	0.83167,	0.83179,	0.8319,	0.83202,	0.83213,	0.83225,	0.83236,	0.83248,	0.83259,	0.83271,	0.83282,	0.83294,	0.83305,	0.83317,	0.83328,	
0.83343,	0.8336,	0.83377,	0.83394,	0.83411,	0.83428,	0.83445,											
	0.83462,	0.83479,	0.83496,	0.83514,	0.83531,	0.83548,	0.83565,	0.83582,	0.83599,	0.83616,	0.83633,	0.8365,	0.83667,	0.83684,	0.83701,	0.83718,	
0.83735,	0.83752,	0.83769,	0.83786,	0.83804,	0.83821,	0.83838,											
	0.83855,	0.83872,	0.83889,	0.83906,	0.83923,	0.8394,	0.83957,	0.83974,	0.83991,	0.84008,	0.84025,	0.84042,	0.84059,	0.84076,	0.84094,	0.84111,	
0.84128,	0.84145,	0.84162,	0.84179,	0.84196,	0.84213,	0.8423,											
	0.84247,	0.84264,	0.84281,	0.84298,	0.84315,	0.84332,	0.84349,	0.84366,	0.84384,	0.84401,	0.84418,	0.84435,	0.84452,	0.84469,	0.84486,	0.84503,	
0.8452,	0.84537,	0.84554,	0.84571,	0.84588,	0.84605,	0.84622,											
	0.84639,	0.84656,	0.84673,	0.84691,	0.84708,	0.84725,	0.84742,	0.84759,	0.84776,	0.84793,	0.8481,	0.84827,	0.84844,	0.84861,	0.84878,	0.84895,	
0.84912,	0.84929,	0.84946,	0.84963,	0.84981,	0.84998,	0.85015,											
	0.85032,	0.85049,	0.85066,	0.85083,	0.851,	0.85149,	0.85218,	0.85287,	0.85355,	0.85424,	0.85493,	0.85562,	0.8563,	0.85699,	0.85768,	0.85836,	
0.85905,	0.85974,	0.86043,	0.86111,	0.8618,	0.86249,	0.86318,											
	0.86386,	0.86455,	0.86524,	0.86593,	0.86661,	0.8673,	0.86799,	0.86868,	0.86936,	0.86927,	0.86886,	0.86844,	0.86803,	0.86761,	0.8672,	0.86678,	
0.86976,	0.87403,	0.8783,	0.88257,	0.88635,	0.88627,	0.88618,											
	0.88609,	0.88601,	0.88592,	0.88583,	0.88575,	0.88566,	0.88558,	0.88549,	0.8854,	0.88532,	0.88523,	0.88514,	0.88506,	0.88497,	0.88488,	0.8848,	
0.88471,	0.88462,	0.88454,	0.88445,	0.88436,	0.88428,	0.88419,											
	0.8841,	0.88402,	0.88393,	0.88384,	0.88376,	0.88421,	0.88506,	0.88592,	0.88678,	0.88763,	0.88849,	0.88935,	0.8902,	0.89106,	0.89192,	0.89277,	
0.89363,	0.89449,	0.89534,	0.8962,	0.89706,	0.89791,	0.89877,											
	0.89963,	0.90048,	0.90134,	0.9022,	0.90305,	0.90391,	0.90513,	0.927,	0.92727,	0.92754,	0.92781,	0.92808,	0.92835,	0.92862,	0.92889,	0.92916,	

localhost:8888/lab/tree/Documents/Intel Unnati/Final/ANPR.ipynb? 22/29

23/29

24/29

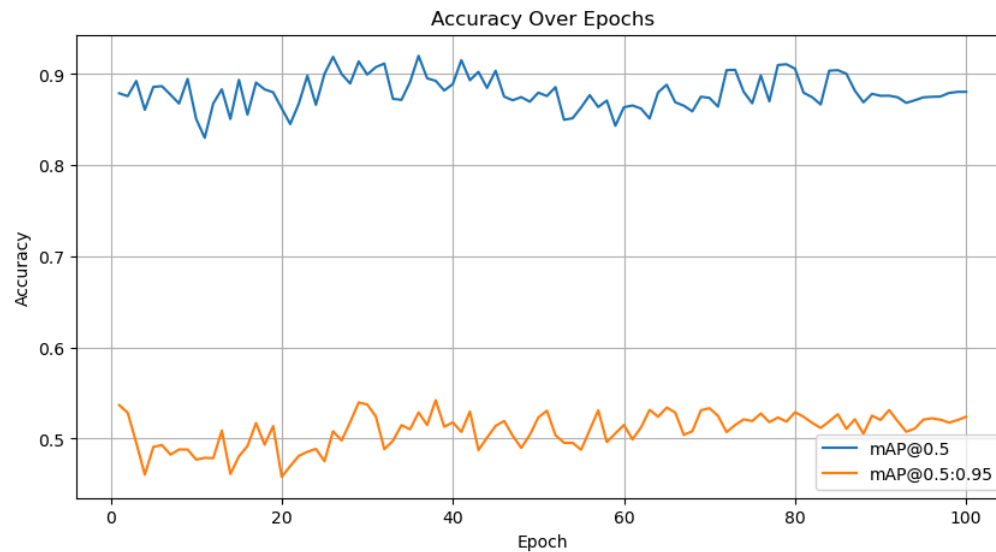

```
0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,
0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,
0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,      0,
                                0,      0,      0,      0,      0,      0,      0,      0,      0]], 'Confidence', 'Recall']]
fitness: 0.5945190545287014
keys: ['metrics/precision(B)', 'metrics/recall(B)', 'metrics/mAP50(B)', 'metrics/mAP50-95(B)']
maps: array([    0.55564])
names: {0: 'license_plate'}
plot: True
results_dict: {'metrics/precision(B)': 0.9681071891700327, 'metrics/recall(B)': 0.8636363636363636, 'metrics/mAP50(B)': 0.9444430446978715, 'metrics/mAP50-95(B)': 0.5556386111765713, 'fitness': 0.5945190545287014}
save_dir: WindowsPath('runs/detect/train19')
speed: {'preprocess': 0.6023320284756748, 'inference': 16.895288770849056, 'loss': 0.0, 'postprocess': 0.24994394995949484}
task: 'detect'
```

```
In [55]: # Find the most recent training log directory
log_dir = max(glob('runs/detect/train*'), key = the_number_in_the_string)

# Load the training results
results = pd.read_csv(os.path.join(log_dir, 'results.csv'))
results.columns = results.columns.str.strip() # Remove any leading/trailing whitespace from column names

# Extract epochs and accuracy metrics
epochs = results.index + 1 # Epochs are zero-indexed, so add 1
mAP_0_5 = results['metrics/mAP50(B)'] # Mean Average Precision at IoU=0.5
mAP_0_5_0_95 = results['metrics/mAP50-95(B)'] # Mean Average Precision at IoU=0.5:0.95

# Plot the accuracy over epochs
plt.figure(figsize=(10, 5))
plt.plot(epochs, mAP_0_5, label='mAP@0.5')
plt.plot(epochs, mAP_0_5_0_95, label='mAP@0.5:0.95')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.title('Accuracy Over Epochs')
plt.legend()
plt.grid(True)
plt.show()
```



```
In [57]: # Save the trained model
model.save('best_numberplatedetection.pt')
```

Testing the model

```
In [60]: def run(path_test_car):
        """
        Predicts and plots the bounding boxes on the given test image using the trained YOLO model.
        Also performs OCR on the detected bounding boxes to extract text.a
        """
        # Perform prediction on the test image using the model
        results = model.predict(path_test_car, device='cpu')

        # Load the image using OpenCV
        image = cv2.imread(path_test_car)
        # Convert the image from BGR (OpenCV default) to RGB (matplotlib default)
        image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

        # Extract the bounding boxes and labels from the results
        for result in results:
            for box in result.bboxes:
                # Get the coordinates of the bounding box
                x1, y1, x2, y2 = map(int, box.xyxy[0])
                # Get the confidence score of the prediction
                confidence = box.conf[0]

                # Draw the bounding box on the image
                cv2.rectangle(image, (x1, y1), (x2, y2), (0, 255, 0), 2)
                # Draw the confidence score near the bounding box
                cv2.putText(image, f'{confidence*100:.2f}%', (x1, y1 - 10),
                           cv2.FONT_HERSHEY_SIMPLEX, 0.9, (255, 0, 0), 2)

                # Crop the bounding box from the image for OCR
                roi = image[y1:y2, x1:x2]

            # Plot the image with bounding boxes
            plt.imshow(image)
            plt.axis('off') # Hide the axis
            plt.show() # Display the image
```

```
In [62]: print(test)
```

	img_path	xmin	xmax	ymin	ymax	img_w	img_h
425	archive\images\Cars425.png	211	301	103	132	500	232
75	archive\images\Cars75.png	209	218	196	201	400	248
181	archive\images\Cars181.png	69	128	195	209	225	400
30	archive\images\Cars30.png	155	200	169	182	400	240
364	archive\images\Cars364.png	61	115	140	167	400	242
408	archive\images\Cars408.png	160	187	217	228	400	267
253	archive\images\Cars253.png	25	64	124	137	400	210
155	archive\images\Cars155.png	150	203	177	190	400	267
168	archive\images\Cars168.png	82	137	95	125	400	225
415	archive\images\Cars415.png	133	261	157	192	400	300
152	archive\images\Cars152.png	92	199	258	284	467	300
70	archive\images\Cars70.png	179	213	115	133	400	226
203	archive\images\Cars203.png	116	271	112	165	400	267
335	archive\images\Cars335.png	156	221	163	198	399	400
39	archive\images\Cars39.png	225	328	180	213	400	300
395	archive\images\Cars395.png	241	288	182	205	400	301
274	archive\images\Cars274.png	196	253	149	169	400	300
72	archive\images\Cars72.png	156	255	182	208	400	332
9	archive\images\Cars9.png	164	272	224	247	442	333
281	archive\images\Cars281.png	151	243	169	183	400	247
287	archive\images\Cars287.png	148	262	180	213	400	332
362	archive\images\Cars362.png	139	162	172	179	400	300
353	archive\images\Cars353.png	93	201	189	221	400	246
55	archive\images\Cars55.png	143	210	171	208	400	299
289	archive\images\Cars289.png	163	271	140	173	400	307
198	archive\images\Cars198.png	166	273	225	246	442	333
148	archive\images\Cars148.png	135	218	134	158	400	225
392	archive\images\Cars392.png	144	260	150	209	400	262
341	archive\images\Cars341.png	47	115	161	192	400	267
268	archive\images\Cars268.png	213	271	133	168	300	225
126	archive\images\Cars126.png	103	293	99	201	400	300
400	archive\images\Cars400.png	303	340	191	217	400	267
305	archive\images\Cars305.png	361	434	167	183	450	270
93	archive\images\Cars93.png	347	367	122	132	400	251
428	archive\images\Cars428.png	142	258	128	157	400	225
77	archive\images\Cars77.png	229	309	151	178	400	270
298	archive\images\Cars298.png	166	233	148	184	400	268
140	archive\images\Cars140.png	283	424	239	286	435	290
323	archive\images\Cars323.png	123	279	28	65	400	225
429	archive\images\Cars429.png	86	208	166	195	301	400
225	archive\images\Cars225.png	189	384	161	215	600	375
154	archive\images\Cars154.png	70	264	197	265	375	500
78	archive\images\Cars78.png	149	249	210	239	400	300
73	archive\images\Cars73.png	119	283	22	69	400	225

```
In [66]: run(test.iloc[12].img_path)
```

image 1/1 C:\Documents\Intel Unnati\Final\archive\images\Cars203.png: 224x320 1 license_plate, 58.6ms
Speed: 0.0ms preprocess, 58.6ms inference, 0.0ms postprocess per image at shape (1, 3, 224, 320)



```
In [68]: run(test.iloc[20].img_path)
```

image 1/1 C:\Documents\Intel Unnati\Final\archive\images\Cars287.png: 288x320 1 license_plate, 41.9ms
Speed: 0.0ms preprocess, 41.9ms inference, 1.0ms postprocess per image at shape (1, 3, 288, 320)



```
In [70]: run(test.iloc[25].img_path)
```

image 1/1 C:\Documents\Intel Unnati\Final\archive\images\Cars198.png: 256x320 1 license_plate, 40.8ms
Speed: 0.0ms preprocess, 40.8ms inference, 0.0ms postprocess per image at shape (1, 3, 256, 320)



```
In [72]: run(test.iloc[30].img_path)
```

image 1/1 C:\Documents\Intel Unnati\Final\archive\images\Cars126.png: 256x320 1 license_plate, 29.5ms
Speed: 1.0ms preprocess, 29.5ms inference, 1.0ms postprocess per image at shape (1, 3, 256, 320)

