

xmqrqhaqt

January 6, 2025

1 8.1. Script Principal

```
[3]: #Análisis conjunto de datos
carpeta_imagenes = r"C:
    ↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\images"
carpeta_etiquetas = r"C:
    ↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\labels"
carpeta_mascaras = r"C:
    ↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\masks"

carpeta_img_totas = r"C:
    ↳\Users\crome\Desktop\TFM\Dades\osaconservation\Limpieza\Dataset_Completo"

import os
import numpy as np
import matplotlib.pyplot as plt

total_imagenes = os.listdir(carpeta_img_totas)

print("Total de imágenes: ", len(total_imagenes))

print("*****")
print("Imágenes procesadas:")
#Cargamos las imágenes
imagenes = os.listdir(carpeta_imagenes)
etiquetas = os.listdir(carpeta_etiquetas)
mascaras = os.listdir(carpeta_mascaras)

#Totales
print("Total de imágenes: ", len(imagenes))
#% de imágenes del total
print("Porcentaje de imágenes del total: ", len(imagenes)/
    ↳len(total_imagenes)*100)
print("Total de etiquetas: ", len(etiquetas))
print("Total de mascarar: ", len(mascaras))

#Total imágenes sin etiqueta
```

```

imagenes_sin_etiqueta = [i for i in imagenes if i not in etiquetas]
print("Total de imagenes sin etiqueta: ", len(imagenes_sin_etiqueta))

#Total imagenes sin mascara
imagenes_sin_mascara = [i for i in imagenes if i not in mascarar]
print("Total de imagenes sin mascara: ", len(imagenes_sin_mascara))

#Total etiquetas sin imagen
etiquetas_sin_imagen = [i for i in etiquetas if i not in imagenes]
print("Total de etiquetas sin imagen: ", len(etiquetas_sin_imagen))

#Total mascarar sin imagen
mascarar_sin_imagen = [i for i in mascarar if i not in imagenes]
print("Total de mascarar sin imagen: ", len(mascarar_sin_imagen))

#Total de imagenes con etiqueta y mascara
imagenes_con_etiqueta = [i for i in imagenes if i in etiquetas]
imagenes_con_mascara = [i for i in imagenes if i in mascarar]
imagenes_con_etiqueta_y_mascara = [i for i in imagenes_con_etiqueta if i in
    ↪imagenes_con_mascara]
print("Total de imagenes con etiqueta y mascara: ",
    ↪len(imagenes_con_etiqueta_y_mascara))

print("*****")
# En train
carpeta_imagenes_train = r"C:
    ↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train\images"
carpeta_etiquetas_train = r"C:
    ↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train\labels"
carpeta_mascarar_train = r"C:
    ↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train\masks"

imagenes_train = os.listdir(carpeta_imagenes_train)
etiquetas_train = os.listdir(carpeta_etiquetas_train)
mascarar_train = os.listdir(carpeta_mascarar_train)

#Totales
print("Total de imagenes train: ", len(imagenes_train))
print("Total de etiquetas train: ", len(etiquetas_train))
print("Total de mascarar train: ", len(mascarar_train))

print("*****")
# En val
carpeta_imagenes_val = r"C:
    ↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\images"

```

```

carpeta_etiquetas_val = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\labels"
carpeta_mascaras_val = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\masks"

imagenes_val = os.listdir(carpeta_imagenes_val)
etiquetas_val = os.listdir(carpeta_etiquetas_val)
mascaras_val = os.listdir(carpeta_mascaras_val)

#Totales
print("Total de imagenes val: ", len(imagenes_val))
print("Total de etiquetas val: ", len(etiquetas_val))
print("Total de mascarar val: ", len(mascaras_val))

```

Total de imagenes: 4303

Imagenes procesadas:

Total de imagenes: 561

Porcentaje de imagenes del total: 13.03741575644899

Total de etiquetas: 561

Total de mascarar: 566

Total de imagenes sin etiqueta: 561

Total de imagenes sin mascara: 561

Total de etiquetas sin imagen: 561

Total de mascarar sin imagen: 566

Total de imagenes con etiqueta y mascara: 0

Total de imagenes train: 449

Total de etiquetas train: 449

Total de mascarar train: 452

Total de imagenes val: 112

Total de etiquetas val: 112

Total de mascarar val: 114

1. Introducción

Este notebook está diseñado para preparar datos, dividirlos en conjuntos de entrenamiento/validación, y entrenar un modelo YOLOv8 para detectar huellas de tortugas. Los pasos incluyen:

Configuración de rutas y organización de archivos. Visualización de etiquetas en las imágenes. Conversión de etiquetas en formato YOLO. División del dataset. Entrenamiento del modelo YOLOv8.

2. Configuración inicial

```
[ ]: # Importar librerías necesarias
import os
import json
import cv2
import shutil
import random
from ultralytics import YOLO

# Configuración de rutas principales
CARPETA_MEZCLADA = r"C:\Users\crome\Desktop\TFM\Dades\osaconservation\Para_Etiquetar_Huellas"
CARPETA_IMAGENES = os.path.join(CARPETA_MEZCLADA, "images")
CARPETA_JSON = os.path.join(CARPETA_MEZCLADA, "json")

# Crear carpetas separadas para imágenes y JSON
os.makedirs(CARPETA_IMAGENES, exist_ok=True)
os.makedirs(CARPETA_JSON, exist_ok=True)

# Mover archivos a sus respectivas carpetas
for archivo in os.listdir(CARPETA_MEZCLADA):
    ruta_archivo = os.path.join(CARPETA_MEZCLADA, archivo)
    if archivo.lower().endswith(('.jpg', '.jpeg', '.png')):
        shutil.move(ruta_archivo, os.path.join(CARPETA_IMAGENES, archivo))
    elif archivo.lower().endswith('.json'):
        shutil.move(ruta_archivo, os.path.join(CARPETA_JSON, archivo))

print("Archivos organizados:")
print(f"- Imágenes movidas a: {CARPETA_IMAGENES}")
print(f"- Archivos JSON movidos a: {CARPETA_JSON}")
```

3. Visualización de etiquetas

```
[ ]: import json
import cv2
import os

def visualizar_etiquetas(carpeta_imagenes, carpeta_json, carpeta_salida):
    """
    Dibuja las etiquetas de huellas en las imágenes basándose en los archivos
    JSON.
    """
    os.makedirs(carpeta_salida, exist_ok=True)

    for archivo_json in os.listdir(carpeta_json):
        if archivo_json.endswith(".json"):
            ruta_json = os.path.join(carpeta_json, archivo_json)
```

```

        ruta_imagen = os.path.join(carpeta_imagenes, archivo_json.replace("↪json", ".JPG"))

        if os.path.exists(ruta_imagen):
            # Leer la imagen
            imagen = cv2.imread(ruta_imagen)

            # Leer el archivo JSON
            with open(ruta_json, 'r') as f:
                datos = json.load(f)

            # Dibujar las etiquetas en la imagen
            for shape in datos.get("shapes", []):
                if shape["shape_type"] == "linestrip":
                    puntos = shape["points"]
                    for i in range(len(puntos) - 1):
                        x1, y1 = int(puntos[i][0]), int(puntos[i][1])
                        x2, y2 = int(puntos[i + 1][0]), int(puntos[i + 1][1])
                        ↪1) cv2.line(imagen, (x1, y1), (x2, y2), (0, 255, 0), ↪2) # Línea verde

            # Guardar la imagen con etiquetas
            ruta_salida = os.path.join(carpeta_salida, archivo_json.replace("↪json", ".JPG"))
            ↪cv2.imwrite(ruta_salida, imagen)

CARPETA_SALIDA_VISUALIZACION = r"C:
↪\Users\crome\Desktop\TFM\Dades\osaconservation\Imagenes_Con_Etiquetas"
visualizar_etiquetas(CARPETA_IMAGENES, CARPETA_JSON, ↪
↪CARPETA_SALIDA_VISUALIZACION)

```

4. Generar etiquetas YOLO y vacías

```

[21]: import os
import json
import shutil

def convertir_json_a_yolo(ruta_json, archivo_salida_txt, ↪
↪carpeta_salida_etiquetas, ancho_imagen, alto_imagen):
    """
    Convierte un archivo JSON a formato YOLO con normalización y lo guarda.
    """
    with open(ruta_json, 'r') as f:
        datos = json.load(f)

    yolo_labels = []

```

```

for shape in datos["shapes"]:
    if shape["shape_type"] == "linestrip":
        puntos = shape["points"]
        if len(puntos) >= 2: # Asegurar que haya al menos dos puntos
            x1, y1 = puntos[0]
            x2, y2 = puntos[-1]

            # Normalizar coordenadas
            xc = ((x1 + x2) / 2) / ancho_imagen
            yc = ((y1 + y2) / 2) / alto_imagen
            w = abs(x2 - x1) / ancho_imagen
            h = abs(y2 - y1) / alto_imagen

            # Verificar que las coordenadas están en el rango [0, 1]
            if 0 <= xc <= 1 and 0 <= yc <= 1 and 0 <= w <= 1 and 0 <= h <= 1:
↪1:
                yolo_labels.append(f"0 {xc:.6f} {yc:.6f} {w:.6f} {h:.6f}")
            else:
                print(f"Coordenadas fuera de rango en {archivo_salida_txt}:↪
↪{xc}, {yc}, {w}, {h}")

        ruta_salida = os.path.join(carpeta_salida_etiquetas, archivo_salida_txt)
        ruta_salida = ruta_salida.replace('.png', '.txt').replace('.jpg', '.txt').
↪replace('.jpeg', '.txt').replace('.JPG', '.txt')

        # Guardar etiquetas YOLO solo si hay datos válidos
        if yolo_labels:
            with open(ruta_salida, 'w') as f:
                f.write("\n".join(yolo_labels))

def procesar_imagenes_y_etiquetas(carpeta_imagenes, carpeta_json,↪
↪carpeta_salida_imagenes, carpeta_salida_etiquetas):
    """
    Copia imágenes con etiquetas JSON asociadas y convierte los JSON en formato↪
    ↪YOLO.
    Omite imágenes sin JSON asociado.
    """
    os.makedirs(carpeta_salida_imagenes, exist_ok=True)
    os.makedirs(carpeta_salida_etiquetas, exist_ok=True)

    for archivo in os.listdir(carpeta_imagenes):
        if archivo.lower().endswith(('.jpg', '.jpeg', '.png', '.JPG')):
            archivo_json = archivo.replace('.jpg', '.json').replace('.jpeg', '.
↪json').replace('.png', '.json').replace('.JPG', '.json')
            ruta_json = os.path.join(carpeta_json, archivo_json)
            ruta_imagen = os.path.join(carpeta_imagenes, archivo)

```

```

        if os.path.exists(ruta_json): # Solo procesar imágenes con JSON
↪ asociado

            # Obtener dimensiones de la imagen
            import cv2
            imagen = cv2.imread(ruta_imagen)
            alto_imagen, ancho_imagen = imagen.shape[:2]

            # Copiar imagen
            ruta_destino_imagen = os.path.join(carpeta_salida_imagenes,
↪ archivo)

            shutil.copy(ruta_imagen, ruta_destino_imagen)

            # Convertir JSON a YOLO
            archivo_txt = archivo.replace('.jpg', '.txt').replace('.jpeg',
↪ '.txt').replace('.png', '.txt').replace('.JPG', '.txt')
            convertir_json_a_yolo(ruta_json, archivo_txt,
↪ carpeta_salida_etiquetas, ancho_imagen, alto_imagen)

def revisar_imagenes_sin_etiquetas(carpeta_imagenes, carpeta_salida_etiquetas):
    """
    Revisa si hay imágenes sin etiquetas YOLO asociadas.
    """
    etiquetas_faltantes = []
    for archivo in os.listdir(carpeta_imagenes):
        if archivo.lower().endswith(('.jpg', '.jpeg', '.png', '.JPG')):
            archivo_txt = archivo.replace('.jpg', '.txt').replace('.jpeg', '.
↪ txt').replace('.png', '.txt').replace('.JPG', '.txt')
            ruta_txt = os.path.join(carpeta_salida_etiquetas, archivo_txt)
            if not os.path.exists(ruta_txt):
                etiquetas_faltantes.append(archivo)
    return etiquetas_faltantes

# Configuración de carpetas
CARPETA_IMAGENES = r"C:
↪ \Users\crome\Desktop\TFM\Dades\osaconservation\Para_Etiquetar_Huellas\images"
CARPETA_JSON = r"C:
↪ \Users\crome\Desktop\TFM\Dades\osaconservation\Para_Etiquetar_Huellas\json"
CARPETA_SALIDA_IMAGENES = r"C:
↪ \Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\images"
CARPETA_SALIDA_ETIQUETAS = r"C:
↪ \Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\labels"

# Procesar imágenes y etiquetas
procesar_imagenes_y_etiquetas(CARPETA_IMAGENES, CARPETA_JSON,
↪ CARPETA_SALIDA_IMAGENES, CARPETA_SALIDA_ETIQUETAS)

```

```

# Revisar imágenes sin etiquetas YOLO asociadas
imagenes_sin_etiquetas = []
    ↪revisar_imagenes_sin_etiquetas(CARPETA_SALIDA_IMAGENES,
    ↪CARPETA_SALIDA_ETIQUETAS)
print(f"Imágenes sin etiquetas YOLO txt asociadas: {imagenes_sin_etiquetas}")

# Eliminar imágenes sin etiquetas YOLO asociadas
for archivo in imagenes_sin_etiquetas:
    ruta_imagen = os.path.join(CARPETA_SALIDA_IMAGENES, archivo)
    os.remove(ruta_imagen)

print("Proceso completado: imágenes y etiquetas procesadas correctamente.")

```

Imágenes sin etiquetas YOLO txt asociadas: ['108FTASK_MAX_2283.JPG', '108FTASK_MAX_2291.JPG', '108FTASK_MAX_2294.JPG', '108FTASK_MAX_2301.JPG', '108FTASK_MAX_2302.JPG']

Proceso completado: imágenes y etiquetas procesadas correctamente.

5. Dividir el dataset

```

[22]: def dividir_dataset(carpeta_dataset, carpeta_train, carpeta_val, val_ratio=0.2):
    """
    Divide un conjunto de datos en entrenamiento y validación.
    """
    os.makedirs(os.path.join(carpeta_train, "images"), exist_ok=True)
    os.makedirs(os.path.join(carpeta_train, "labels"), exist_ok=True)
    os.makedirs(os.path.join(carpeta_val, "images"), exist_ok=True)
    os.makedirs(os.path.join(carpeta_val, "labels"), exist_ok=True)

    imagenes = [f for f in os.listdir(os.path.join(carpeta_dataset, "images"))
    ↪if f.lower().endswith(('.jpg', '.jpeg', '.png'))]
    random.shuffle(imagenes)

    val_size = int(len(imagenes) * val_ratio)

    for i, imagen in enumerate(imagenes):
        etiqueta = imagen.replace(".jpg", ".txt").replace(".jpeg", ".txt").
        ↪replace(".png", ".txt").replace(".JPG", ".txt")
        set_destino = carpeta_val if i < val_size else carpeta_train

        shutil.copy(os.path.join(carpeta_dataset, "images", imagen), os.path.
        ↪join(set_destino, "images", imagen))
        shutil.copy(os.path.join(carpeta_dataset, "labels", etiqueta), os.path.
        ↪join(set_destino, "labels", etiqueta))

CARPETA_DATASET = r"C:
    ↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado"

```



```
CARPETA_TRAIN = os.path.join(CARPETA_DATASET, "train")
CARPETA_VAL = os.path.join(CARPETA_DATASET, "val")
dividir_dataset(CARPETA_DATASET, CARPETA_TRAIN, CARPETA_VAL)
```

6. Entrenar YOLO

```
[2]: import os
os.environ["KMP_DUPLICATE_LIB_OK"]="TRUE"
from ultralytics import YOLO
```

```
[2]: import os
os.environ["KMP_DUPLICATE_LIB_OK"]="TRUE"
from ultralytics import YOLO

def entrenar_yolo(data_path, modelo_salida, epochs=1000, img_size=640):
    """
    Entrena un modelo YOLOv8 con el dataset proporcionado.
    """
    modelo = YOLO("yolov8n.pt")
    modelo.train(
        data=data_path,
        epochs=epochs,
        imgsz=img_size,
        save_period=10,
        project=os.path.dirname(modelo_salida),
        name=os.path.basename(modelo_salida).split('.')[0]
    )

DATA_PATH = r"C:\Users\Carlos\Desktop\pepe\data_other_pc.yaml"
MODELO_SALIDA = r"C:\Users\Carlos\Desktop\pepe\yolov8_model.pt"
entrenar_yolo(DATA_PATH, MODELO_SALIDA)
```

Ultralytics 8.3.49 Python-3.10.13 torch-2.5.1+cpu CPU (13th Gen Intel Core(TM) i9-13900KF)

```
engine\trainer: task=detect, mode=train, model=yolov8n.pt,
data=C:\Users\Carlos\Desktop\pepe\data_other_pc.yaml, epochs=1000, time=None,
patience=100, batch=16, imgsz=640, save=True, save_period=10, cache=False,
device=None, workers=8, project=C:\Users\Carlos\Desktop\pepe,
name=yolov8_model6, 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=True, opset=None, workspace=None, 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, 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, copy_paste_mode=flip, auto_augment=randaugument,
erasing=0.4, crop_fraction=1.0, cfg=None, tracker=botsort.yaml,
save_dir=C:\Users\Carlos\Desktop\pepe\yolov8_model6
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, 3,011,043 parameters, 3,011,027 gradients, 8.2 GFLOPs

```

Transferred 319/355 items from pretrained weights

TensorBoard: Start with 'tensorboard --logdir
C:\Users\Carlos\Desktop\pepe\yolov8_model6', view at <http://localhost:6006/>
Freezing layer 'model.22.dfl.conv.weight'

train: Scanning

C:\Users\Carlos\Desktop\pepe\Dataset_Organizado\train\labels.cache... 449
images, 0 backgrounds, 0 corrupt: 100%| | 449/449 [00:00<?, ?it/s]

val: Scanning

C:\Users\Carlos\Desktop\pepe\Dataset_Organizado\val\labels.cache... 112 images,
0 backgrounds, 0 corrupt: 100%| | 112/112 [00:00<?, ?it/s]

Plotting labels to C:\Users\Carlos\Desktop\pepe\yolov8_model6\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 640 train, 640 val

Using 0 dataloader workers

Logging results to C:\Users\Carlos\Desktop\pepe\yolov8_model6

Starting training for 1000 epochs...

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
1/1000	0G	3.096	3.328	2.562	5	640:
100%	29/29	[00:37<00:00, 1.29s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.04it/s]				
	all	112	557	0.0075	0.452	0.0172
0.00467						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
2/1000	OG	2.805	2.891	2.243	11	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.00it/s]				
	all	112	557	0.0188	0.431	0.0279
0.00754						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
3/1000	OG	2.8	2.873	2.248	12	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.06it/s]				
	all	112	557	0.0286	0.0395	0.0071
0.00184						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
4/1000	OG	2.774	2.841	2.248	12	640:
100%	29/29	[00:37<00:00, 1.28s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.23s/it]				
	all	112	557	0.0317	0.118	0.017
0.00397						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
5/1000	OG	2.697	2.783	2.223	4	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.07s/it]				
	all	112	557	0.0941	0.111	0.0321
0.00871						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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        6/1000      OG      2.667      2.721      2.181      9      640:
100%|      | 29/29 [00:36<00:00, 1.26s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.04it/s]
          all      112      557      0.112      0.154      0.0511
0.0145

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      7/1000      OG      2.622      2.652      2.133      6      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.12it/s]
          all      112      557      0.112      0.133      0.0489
0.0141

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      8/1000      OG      2.572      2.647      2.111      5      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.12s/it]
          all      112      557      0.0688      0.0898      0.0191
0.00609

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      9/1000      OG      2.605      2.683      2.184      12      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.08s/it]
          all      112      557      0.107      0.106      0.0369
0.0108

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
     10/1000      OG      2.541      2.616      2.125      9      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.11it/s]

```

	all	112	557	0.11	0.133	0.0431
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0.0133

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
11/1000	OG	2.563	2.611	2.123	9	640:
100%	29/29	[00:36<00:00, 1.27s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.126	0.126	0.0411

0.0126

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
12/1000	OG	2.527	2.621	2.106	3	640:
100%	29/29	[00:37<00:00, 1.30s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.06s/it]				
	all	112	557	0.0981	0.115	0.0498

0.0157

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
13/1000	OG	2.464	2.625	2.129	15	640:
100%	29/29	[00:36<00:00, 1.27s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.12s/it]				
	all	112	557	0.0873	0.185	0.0413

0.0126

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
14/1000	OG	2.437	2.576	2.049	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.03it/s]				
	all	112	557	0.0992	0.156	0.0505

0.0175

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
15/1000	OG	2.497	2.528	2.051	14	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.17	0.135	0.0651
0.0193						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
16/1000	OG	2.423	2.531	2.051	7	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.11	0.115	0.042
0.0129						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
17/1000	OG	2.46	2.667	2.084	2	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.165	0.117	0.0642
0.0231						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
18/1000	OG	2.466	2.586	2.097	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.19it/s]				
	all	112	557	0.19	0.201	0.0975
0.0315						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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19/1000      OG      2.449      2.54      2.079      2      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.05it/s]
          all      112      557      0.136      0.178      0.0631
0.0181

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
20/1000      OG      2.435      2.534      2.033      3      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.10it/s]
          all      112      557      0.163      0.154      0.0713
0.0226

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
21/1000      OG      2.413      2.525      2.026      13      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.16it/s]
          all      112      557      0.158      0.194      0.0814
0.0267

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
22/1000      OG      2.416      2.52      2.043      13      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.05it/s]
          all      112      557      0.211      0.208      0.0935
0.0309

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
23/1000      OG      2.348      2.504      2.015      1      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.15it/s]

```


	all	112	557	0.158	0.194	0.089
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0.0293

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
24/1000	OG	2.361	2.429	2.015	9	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.05it/s]				
	all	112	557	0.14	0.176	0.0734

0.0241

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
25/1000	OG	2.29	2.748	1.922	0	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.16it/s]				
	all	112	557	0.186	0.221	0.101

0.0332

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
26/1000	OG	2.348	2.466	1.984	11	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.18it/s]				
	all	112	557	0.164	0.212	0.0704

0.0229

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
27/1000	OG	2.331	2.419	1.979	8	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.228	0.224	0.133

0.0415

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
28/1000	OG	2.354	2.486	1.978	4	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.18it/s]				
	all	112	557	0.239	0.174	0.102
0.0321						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
29/1000	OG	2.338	2.44	2.001	12	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.18it/s]				
	all	112	557	0.165	0.18	0.0729
0.024						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
30/1000	OG	2.299	2.431	1.98	1	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.07it/s]				
	all	112	557	0.18	0.176	0.0937
0.0291						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
31/1000	OG	2.271	2.458	1.958	4	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.165	0.156	0.0596
0.0192						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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32/1000      OG      2.284      2.373      1.96      17      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.04it/s]
          all      112      557      0.243      0.156      0.0826
0.0277

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
33/1000      OG      2.289      2.397      1.961      30      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.21it/s]
          all      112      557      0.214      0.176      0.116
0.0388

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
34/1000      OG      2.246      2.479      1.936      1      640:
100%|      | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.14it/s]
          all      112      557      0.198      0.178      0.0992
0.0309

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
35/1000      OG      2.295      2.412      1.987      3      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.12it/s]
          all      112      557      0.254      0.199      0.142
0.0471

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
36/1000      OG      2.234      2.366      1.937      18      640:
100%|      | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.25it/s]

```

	all	112	557	0.229	0.214	0.126
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0.0436

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
37/1000	OG	2.231	2.322	1.918	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.263	0.223	0.146

0.053

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
38/1000	OG	2.238	2.336	1.911	6	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.227	0.162	0.1

0.0367

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
39/1000	OG	2.194	2.305	1.909	6	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.17it/s]				
	all	112	557	0.22	0.219	0.113

0.0367

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
40/1000	OG	2.274	2.381	1.955	22	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.23it/s]				
	all	112	557	0.213	0.16	0.108

0.0366

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
41/1000	OG	2.218	2.328	1.894	11	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.20it/s]				
	all	112	557	0.204	0.144	0.0852
0.0274						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
42/1000	OG	2.133	2.604	1.839	0	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.20it/s]				
	all	112	557	0.276	0.215	0.131
0.043						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
43/1000	OG	2.249	2.325	1.905	14	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.20it/s]				
	all	112	557	0.282	0.233	0.168
0.0613						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
44/1000	OG	2.181	2.327	1.894	4	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.272	0.217	0.153
0.0584						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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45/1000      OG      2.235      2.354      1.906      13      640:
100%|      | 29/29 [00:36<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.25it/s]
      all      112      557      0.231      0.214      0.146
0.0501

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
46/1000      OG      2.275      2.346      1.937      9      640:
100%|      | 29/29 [00:36<00:00, 1.26s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.24it/s]
      all      112      557      0.255      0.189      0.117
0.0393

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
47/1000      OG      2.203      2.312      1.907      8      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.27it/s]
      all      112      557      0.228      0.197      0.132
0.044

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
48/1000      OG      2.213      2.333      1.877      2      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.25it/s]
      all      112      557      0.216      0.18      0.131
0.0429

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
49/1000      OG      2.175      2.296      1.869      7      640:
100%|      | 29/29 [00:35<00:00, 1.23s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.22it/s]

```

0.0539	all	112	557	0.276	0.237	0.156
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Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
50/1000	OG	2.213	2.274	1.886	12	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.286	0.241	0.172

0.0614

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
51/1000	OG	2.161	2.552	1.814	1	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.23it/s]				
	all	112	557	0.238	0.199	0.116

0.0405

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
52/1000	OG	2.208	2.295	1.912	9	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.19it/s]				
	all	112	557	0.249	0.229	0.152

0.0558

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
53/1000	OG	2.134	2.273	1.872	16	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.20it/s]				
	all	112	557	0.275	0.214	0.146

0.0512

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
54/1000	OG	2.162	2.29	1.864	3	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.23it/s]				
	all	112	557	0.26	0.219	0.169
0.0598						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
55/1000	OG	2.192	2.392	1.923	2	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.261	0.194	0.157
0.0538						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
56/1000	OG	2.189	2.333	1.876	4	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.254	0.214	0.139
0.0482						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
57/1000	OG	2.145	2.222	1.844	8	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.218	0.194	0.113
0.0383						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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58/1000      OG      2.156      2.244      1.846      11      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.22it/s]
          all        112        557        0.272        0.21        0.151
0.0522

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
59/1000      OG      2.173      2.256      1.857      3      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.24it/s]
          all        112        557        0.313        0.242        0.19
0.0663

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
60/1000      OG      2.115      2.237      1.845      6      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.21it/s]
          all        112        557        0.237        0.224        0.133
0.0493

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
61/1000      OG      2.143      2.24      1.838      8      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.28it/s]
          all        112        557        0.297        0.237        0.169
0.0589

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
62/1000      OG      2.131      2.226      1.843      5      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.28it/s]

```

	all	112	557	0.263	0.207	0.148
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0.0489

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
63/1000	OG	2.092	2.197	1.799	6	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.304	0.203	0.155

0.0515

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
64/1000	OG	2.097	2.228	1.835	2	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.19it/s]				
	all	112	557	0.26	0.228	0.165

0.0592

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
65/1000	OG	2.087	2.227	1.822	12	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.275	0.226	0.15

0.052

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
66/1000	OG	2.134	2.15	1.817	9	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.24it/s]				
	all	112	557	0.254	0.215	0.146

0.0543

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
67/1000	OG	2.116	2.216	1.813	16	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.283	0.262	0.183
0.0663						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
68/1000	OG	2.067	2.223	1.804	9	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.298	0.197	0.146
0.0538						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
69/1000	OG	2.206	2.399	1.844	1	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.276	0.248	0.179
0.0635						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
70/1000	OG	2.149	2.221	1.873	5	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.268	0.227	0.166
0.0613						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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71/1000      OG      2.101      2.163      1.821      4      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.21it/s]
      all      112      557      0.263      0.23      0.148
0.0484

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
72/1000      OG      2.099      2.195      1.81      9      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.27it/s]
      all      112      557      0.283      0.208      0.148
0.0501

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
73/1000      OG      2.04      2.198      1.777      3      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.21it/s]
      all      112      557      0.255      0.202      0.152
0.0537

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
74/1000      OG      2.085      2.211      1.811      5      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.29it/s]
      all      112      557      0.272      0.223      0.16
0.0574

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
75/1000      OG      2.047      2.149      1.78      10      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.19it/s]

```

	all	112	557	0.273	0.23	0.172
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0.0605

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
76/1000	OG	2.074	2.154	1.796	8	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.24it/s]				
	all	112	557	0.329	0.278	0.214

0.079

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
77/1000	OG	2.028	2.107	1.758	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.301	0.266	0.195

0.0723

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
78/1000	OG	2.119	2.174	1.821	6	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.274	0.239	0.171

0.0634

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
79/1000	OG	2.042	2.147	1.791	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.305	0.237	0.167

0.0591

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
80/1000	OG	2.086	2.176	1.814	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.278	0.247	0.179
0.0644						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
81/1000	OG	2.117	2.295	1.864	1	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.309	0.224	0.179
0.0665						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
82/1000	OG	2.051	2.139	1.791	14	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.279	0.235	0.189
0.0742						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
83/1000	OG	2.078	2.141	1.801	10	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.334	0.257	0.192
0.0703						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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      84/1000      OG      2.037      2.181      1.768      3      640:
100%|      | 29/29 [00:36<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.25it/s]
      all      112      557      0.294      0.226      0.181
0.062

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      85/1000      OG      2.01      2.08      1.737      9      640:
100%|      | 29/29 [00:36<00:00, 1.24s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.01s/it]
      all      112      557      0.231      0.206      0.127
0.0417

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      86/1000      OG      2.015      2.069      1.77      8      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.23it/s]
      all      112      557      0.266      0.226      0.154
0.0574

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      87/1000      OG      2.023      2.064      1.765      6      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.28it/s]
      all      112      557      0.308      0.255      0.212
0.0751

```

```

      Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
      88/1000      OG      2.071      2.176      1.79      4      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.20it/s]

```

	all	112	557	0.244	0.201	0.138
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0.0449

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
89/1000	OG	2.039	2.126	1.762	6	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.278	0.26	0.166

0.0593

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
90/1000	OG	1.994	2.017	1.747	55	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.366	0.248	0.206

0.073

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
91/1000	OG	2.021	2.079	1.771	9	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.321	0.257	0.203

0.0786

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
92/1000	OG	2.015	2.084	1.757	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.21it/s]				
	all	112	557	0.28	0.24	0.17

0.0606

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
93/1000	OG	1.934	2.019	1.698	4	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.308	0.273	0.188
0.0689						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
94/1000	OG	1.972	2.037	1.737	14	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.304	0.257	0.189
0.0686						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
95/1000	OG	1.989	2.087	1.745	3	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.363	0.233	0.215
0.0826						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
96/1000	OG	1.96	2.018	1.721	10	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.319	0.223	0.174
0.0609						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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97/1000      OG      2.015      2.041      1.733      21      640:
100%|      | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.28it/s]
          all      112      557      0.267      0.194      0.141
0.0537

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
98/1000      OG      1.992      2.1      1.786      8      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.27it/s]
          all      112      557      0.262      0.223      0.155
0.0579

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
99/1000      OG      2.002      2.044      1.74      12      640:
100%|      | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.31it/s]
          all      112      557      0.295      0.242      0.175
0.0642

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
100/1000      OG      1.975      2.181      1.772      1      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.29it/s]
          all      112      557      0.249      0.219      0.145
0.0496

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
101/1000      OG      1.958      1.989      1.698      14      640:
100%|      | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:03<00:00, 1.25it/s]

```

	all	112	557	0.291	0.21	0.136
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0.0451

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
102/1000	OG	1.991	2.061	1.74	10	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.334	0.239	0.191

0.0705

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
103/1000	OG	1.947	2.15	1.723	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.24it/s]				
	all	112	557	0.291	0.257	0.173

0.0599

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
104/1000	OG	1.992	2.059	1.722	12	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.374	0.228	0.198

0.0764

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
105/1000	OG	1.977	2.105	1.742	17	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.311	0.224	0.165

0.0582

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
106/1000	OG	1.942	2.005	1.716	8	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.329	0.215	0.17
0.0637						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
107/1000	OG	1.912	1.999	1.692	14	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.333	0.268	0.182
0.0669						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
108/1000	OG	1.904	1.99	1.685	2	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.27	0.214	0.144
0.0478						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
109/1000	OG	1.909	1.973	1.693	7	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.287	0.258	0.191
0.0678						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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110/1000      OG      1.966      2.056      1.723      2      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.26it/s]
          all      112      557      0.34      0.239      0.187
0.0637

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
111/1000      OG      1.965      2.067      1.696      7      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all      112      557      0.311      0.232      0.174
0.0619

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
112/1000      OG      1.926      1.998      1.695      10      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.29it/s]
          all      112      557      0.323      0.232      0.184
0.0609

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
113/1000      OG      1.912      1.928      1.669      13      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.27it/s]
          all      112      557      0.339      0.25      0.2
0.0761

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
114/1000      OG      1.907      2.052      1.697      3      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.25it/s]

```

	all	112	557	0.367	0.246	0.219
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0.0855

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
115/1000	OG	1.88	1.902	1.648	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.361	0.257	0.2

0.0748

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
116/1000	OG	1.945	1.945	1.68	11	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.23it/s]				
	all	112	557	0.283	0.241	0.18

0.0665

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
117/1000	OG	1.889	1.93	1.672	18	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.318	0.271	0.211

0.0787

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
118/1000	OG	1.926	1.989	1.678	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.366	0.289	0.209

0.0759

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
119/1000	OG	1.896	2.043	1.678	7	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.314	0.287	0.196
0.0726						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
120/1000	OG	1.888	1.93	1.663	14	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.378	0.263	0.191
0.0719						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
121/1000	OG	1.926	1.948	1.695	10	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.321	0.282	0.196
0.0749						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
122/1000	OG	1.938	1.965	1.702	2	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.273	0.197	0.146
0.0517						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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123/1000      OG      1.861      1.887      1.649      24      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.28it/s]
          all          112          557      0.301      0.264      0.185
0.0713

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
124/1000      OG      1.891      1.882      1.661      19      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557      0.329      0.237      0.191
0.0719

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
125/1000      OG      1.903      1.947      1.666      16      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all          112          557      0.315      0.251      0.197
0.0752

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
126/1000      OG      1.904      1.91      1.66      4      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all          112          557      0.262      0.232      0.159
0.0585

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
127/1000      OG      1.879      1.917      1.651      21      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.25it/s]

```


	all	112	557	0.283	0.192	0.153
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0.0507

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
128/1000	OG	1.899	1.931	1.654	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.314	0.269	0.223

0.0818

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
129/1000	OG	1.878	1.882	1.654	11	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.276	0.242	0.176

0.0588

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
130/1000	OG	1.888	1.918	1.666	31	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.317	0.251	0.173

0.0577

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
131/1000	OG	1.84	1.892	1.64	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.308	0.271	0.183

0.0652

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
132/1000	OG	1.873	1.838	1.614	11	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.311	0.215	0.181
0.0649						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
133/1000	OG	1.884	1.889	1.653	8	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.301	0.242	0.181
0.0668						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
134/1000	OG	1.859	1.897	1.655	4	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.306	0.226	0.165
0.0567						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
135/1000	OG	1.869	1.894	1.67	13	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.419	0.232	0.223
0.0774						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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136/1000      OG      1.816      2.053      1.575      1      640:
100%|         | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.30it/s]
          all      112      557      0.321      0.233      0.195
0.0675

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
137/1000      OG      1.892      1.879      1.646      44      640:
100%|         | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.30it/s]
          all      112      557      0.375      0.233      0.212
0.0732

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
138/1000      OG      1.838      1.947      1.622      1      640:
100%|         | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.25it/s]
          all      112      557      0.365      0.228      0.201
0.0756

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
139/1000      OG      1.794      1.829      1.597      4      640:
100%|         | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.32it/s]
          all      112      557      0.338      0.212      0.199
0.0717

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
140/1000      OG      1.83      1.808      1.621      9      640:
100%|         | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.25it/s]

```

	all	112	557	0.293	0.223	0.166
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0.0551

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
141/1000	OG	1.85	1.828	1.633	8	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.316	0.21	0.177

0.0603

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
142/1000	OG	1.834	1.844	1.619	4	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.24it/s]				
	all	112	557	0.275	0.239	0.165

0.0558

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
143/1000	OG	1.803	1.788	1.592	11	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.333	0.246	0.205

0.0803

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
144/1000	OG	1.89	1.868	1.609	6	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.3	0.28	0.195

0.0677

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
145/1000	OG	1.741	2.087	1.538	0	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.345	0.253	0.212
0.0825						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
146/1000	OG	1.816	1.826	1.583	1	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.33	0.243	0.195
0.0685						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
147/1000	OG	1.846	1.88	1.647	12	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.282	0.246	0.188
0.0678						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
148/1000	OG	1.8	1.826	1.584	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.355	0.248	0.21
0.0758						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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149/1000      OG      1.824      1.825      1.61      4      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.27it/s]
          all      112      557      0.247      0.231      0.157
0.0534

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
150/1000      OG      1.78      1.783      1.588      6      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all      112      557      0.346      0.224      0.184
0.0635

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
151/1000      OG      1.881      1.849      1.644      21      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.24it/s]
          all      112      557      0.372      0.251      0.204
0.0696

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
152/1000      OG      1.821      1.793      1.593      20      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all      112      557      0.31      0.233      0.182
0.0625

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
153/1000      OG      1.806      1.793      1.6      11      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.26it/s]

```

	all	112	557	0.312	0.232	0.179
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0.059

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
154/1000	OG	1.811	1.823	1.594	10	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.357	0.255	0.215

0.0715

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
155/1000	OG	1.795	1.818	1.595	16	640:
100%	29/29	[00:38<00:00, 1.34s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.23it/s]				
	all	112	557	0.266	0.217	0.154

0.05

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
156/1000	OG	1.803	1.767	1.58	10	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.329	0.237	0.184

0.0623

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
157/1000	OG	1.782	1.754	1.567	10	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.22it/s]				
	all	112	557	0.306	0.219	0.17

0.059

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
158/1000	OG	1.796	1.77	1.566	10	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.287	0.215	0.157
0.0561						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
159/1000	OG	1.824	1.773	1.593	14	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.274	0.239	0.165
0.0623						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
160/1000	OG	1.834	1.809	1.62	7	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.267	0.266	0.171
0.0624						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
161/1000	OG	1.813	1.792	1.619	6	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.345	0.275	0.222
0.0773						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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162/1000      OG      1.857      1.851      1.602      9      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all      112      557      0.342      0.28      0.213
0.0691

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
163/1000      OG      1.752      1.757      1.562      1      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all      112      557      0.278      0.226      0.172
0.0628

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
164/1000      OG      1.777      1.767      1.568      5      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.27it/s]
          all      112      557      0.294      0.211      0.157
0.0538

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
165/1000      OG      1.828      1.802      1.601      12      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.25it/s]
          all      112      557      0.319      0.257      0.204
0.0741

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
166/1000      OG      1.765      1.797      1.581      2      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.23it/s]

```

	all	112	557	0.285	0.285	0.181
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0.0613

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
167/1000	OG	1.793	1.788	1.599	13	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.342	0.238	0.189

0.0632

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
168/1000	OG	1.795	1.795	1.608	5	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.274	0.262	0.189

0.064

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
169/1000	OG	1.751	1.738	1.556	3	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.335	0.244	0.191

0.0651

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
170/1000	OG	1.765	1.743	1.56	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.338	0.273	0.23

0.0854

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
171/1000	OG	1.715	1.696	1.544	4	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.336	0.309	0.223
0.0807						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
172/1000	OG	1.716	1.705	1.533	7	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.356	0.246	0.208
0.0698						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
173/1000	OG	1.737	1.722	1.579	4	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.275	0.285	0.201
0.0708						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
174/1000	OG	1.758	1.749	1.568	19	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.409	0.259	0.232
0.0816						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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175/1000      OG      1.762      1.758      1.543      22      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.27it/s]
          all        112        557        0.307        0.293        0.225
0.0801

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
176/1000      OG      1.789      1.774      1.579      10      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all        112        557        0.337        0.282        0.204
0.0751

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
177/1000      OG      1.771      1.738      1.566      26      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.24it/s]
          all        112        557        0.349        0.278        0.223
0.0797

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
178/1000      OG      1.753      2.035      1.531      1      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.29it/s]
          all        112        557        0.361        0.289        0.226
0.0826

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
179/1000      OG      1.753      1.788      1.559      3      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.26it/s]

```

	all	112	557	0.35	0.246	0.204
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0.0767

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
180/1000	OG	1.764	1.737	1.616	1	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.352	0.242	0.183

0.0627

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
181/1000	OG	1.76	1.783	1.582	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.423	0.244	0.223

0.0807

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
182/1000	OG	1.732	1.648	1.518	6	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.381	0.266	0.228

0.078

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
183/1000	OG	1.78	1.811	1.586	6	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.344	0.28	0.227

0.0766

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
184/1000	OG	1.71	1.791	1.535	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.356	0.242	0.197
0.0752						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
185/1000	OG	1.716	1.659	1.541	29	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.336	0.241	0.19
0.0617						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
186/1000	OG	1.698	1.67	1.525	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.341	0.264	0.206
0.0691						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
187/1000	OG	1.712	1.673	1.553	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.318	0.235	0.183
0.065						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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188/1000      OG      1.709      1.659      1.534      20      640:
100%|         | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.25it/s]
          all          112          557      0.298      0.264      0.185
0.0703

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
189/1000      OG      1.726      1.704      1.549          11      640:
100%|         | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557      0.357      0.237      0.196
0.068

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
190/1000      OG      1.755      1.694      1.569           6      640:
100%|         | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.26it/s]
          all          112          557      0.396      0.226      0.186
0.0658

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
191/1000      OG      1.641      1.845      1.474           0      640:
100%|         | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.30it/s]
          all          112          557      0.449      0.264      0.235
0.086

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
192/1000      OG      1.719      1.685      1.545          10      640:
100%|         | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|         | 4/4 [00:03<00:00, 1.30it/s]

```

	all	112	557	0.274	0.242	0.176
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0.0614

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
193/1000	OG	1.652	1.63	1.497	10	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.328	0.291	0.204

0.0693

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
194/1000	OG	1.729	1.692	1.546	7	640:
100%	29/29	[00:35<00:00, 1.22s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.388	0.235	0.211

0.0732

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
195/1000	OG	1.717	1.704	1.53	8	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.367	0.248	0.214

0.084

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
196/1000	OG	1.647	1.673	1.483	1	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.372	0.264	0.213

0.0781

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
197/1000	OG	1.67	1.646	1.512	14	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.344	0.262	0.222
0.0759						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
198/1000	OG	1.706	1.697	1.539	13	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.314	0.244	0.196
0.0635						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
199/1000	OG	1.701	1.668	1.512	6	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.354	0.242	0.203
0.0669						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
200/1000	OG	1.689	1.598	1.501	6	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.426	0.217	0.209
0.0719						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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201/1000      OG      1.678      1.632      1.512      13      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557      0.369      0.219      0.196
0.0686

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
202/1000      OG      1.699      1.761      1.534          2      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all          112          557      0.356      0.269      0.209
0.0741

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
203/1000      OG      1.631      1.56      1.47          15      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all          112          557      0.382      0.239      0.214
0.0773

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
204/1000      OG      1.655      1.615      1.483          5      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all          112          557      0.311      0.23      0.186
0.0649

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
205/1000      OG      1.652      1.603      1.485          2      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]

```

	all	112	557	0.356	0.287	0.208
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0.0741

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
206/1000	OG	1.648	1.578	1.478	10	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.354	0.266	0.208

0.069

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
207/1000	OG	1.67	1.619	1.472	14	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.385	0.215	0.182

0.0582

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
208/1000	OG	1.722	1.701	1.532	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.325	0.271	0.189

0.0637

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
209/1000	OG	1.679	1.61	1.498	6	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.365	0.248	0.209

0.0725

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
210/1000	OG	1.594	1.861	1.439	0	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.322	0.259	0.186
0.0638						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
211/1000	OG	1.661	1.602	1.486	12	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.371	0.232	0.207
0.0699						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
212/1000	OG	1.586	1.839	1.43	0	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.361	0.268	0.223
0.0758						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
213/1000	OG	1.669	1.63	1.494	4	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.398	0.289	0.228
0.0783						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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214/1000      OG      1.626      1.527      1.462      11      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.41        0.241        0.209
0.0763

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
215/1000      OG      1.564      1.639      1.406          0      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.367        0.246        0.212
0.0736

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
216/1000      OG      1.644      1.552      1.494          4      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.331        0.285        0.232
0.0827

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
217/1000      OG      1.623      1.538      1.464          9      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.27it/s]
          all        112        557        0.358        0.289        0.219
0.0783

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
218/1000      OG      1.655      1.604      1.496          6      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]

```

	all	112	557	0.276	0.287	0.19
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0.0687

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
219/1000	OG	1.623	1.62	1.461	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.385	0.233	0.195

0.0652

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
220/1000	OG	1.695	1.618	1.525	5	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.322	0.242	0.177

0.0594

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
221/1000	OG	1.665	1.605	1.497	10	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.317	0.219	0.169

0.0589

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
222/1000	OG	1.622	1.566	1.469	17	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.318	0.273	0.191

0.0668

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
223/1000	OG	1.654	1.606	1.485	2	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.315	0.26	0.181
0.0617						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
224/1000	OG	1.749	1.606	1.544	6	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.358	0.26	0.204
0.071						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
225/1000	OG	1.696	1.596	1.5	4	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.342	0.215	0.169
0.0587						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
226/1000	OG	1.643	1.562	1.476	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.344	0.257	0.2
0.0753						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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227/1000      OG      1.554      1.455      1.406      1      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all      112      557      0.393      0.221      0.185
0.0623

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
228/1000      OG      1.639      1.551      1.474      9      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.25it/s]
          all      112      557      0.403      0.248      0.192
0.069

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
229/1000      OG      1.643      1.566      1.464      6      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all      112      557      0.367      0.259      0.214
0.075

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
230/1000      OG      1.612      1.544      1.451      6      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.26it/s]
          all      112      557      0.37      0.273      0.21
0.0721

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
231/1000      OG      1.644      1.59      1.483      9      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]

```


0.0698	all	112	557	0.356	0.284	0.206
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Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
232/1000	OG	1.597	1.539	1.48	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.279	0.244	0.17

0.0609

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
233/1000	OG	1.599	1.52	1.47	2	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.422	0.215	0.203

0.0715

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
234/1000	OG	1.552	1.772	1.397	0	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.363	0.253	0.219

0.0768

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
235/1000	OG	1.579	1.552	1.432	7	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.339	0.244	0.216

0.0742

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
236/1000	OG	1.615	1.571	1.478	20	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.33it/s]				
	all	112	557	0.317	0.273	0.208
0.0737						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
237/1000	OG	1.601	1.491	1.447	30	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.318	0.259	0.193
0.0677						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
238/1000	OG	1.608	1.508	1.447	5	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.33it/s]				
	all	112	557	0.355	0.241	0.193
0.0655						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
239/1000	OG	1.543	1.732	1.405	0	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.307	0.262	0.178
0.0577						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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240/1000      OG      1.667      1.593      1.497      31      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.29it/s]
          all        112        557        0.365        0.276        0.213
0.0679

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
241/1000      OG      1.634      1.534      1.469          5      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.25it/s]
          all        112        557        0.359        0.248        0.206
0.0687

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
242/1000      OG      1.578      1.473      1.432         12      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all        112        557        0.327        0.269        0.188
0.0658

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
243/1000      OG      1.604      1.512      1.449         11      640:
100%|        | 29/29 [00:35<00:00, 1.23s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.421        0.224        0.207
0.0712

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
244/1000      OG      1.579      1.506      1.435         10      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]

```

0.0629	all	112	557	0.348	0.215	0.178
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Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
245/1000	OG	1.574	1.472	1.425	14	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.355	0.242	0.193

0.0635

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
246/1000	OG	1.596	1.501	1.44	9	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.368	0.25	0.194

0.0666

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
247/1000	OG	1.565	1.44	1.402	11	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.36	0.26	0.212

0.0701

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
248/1000	OG	1.59	1.508	1.441	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.329	0.217	0.172

0.0603

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
249/1000	OG	1.631	1.577	1.468	2	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.33it/s]				
	all	112	557	0.387	0.23	0.201
0.0747						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
250/1000	OG	1.575	1.492	1.438	8	640:
100%	29/29	[00:35<00:00, 1.22s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.371	0.257	0.193
0.0668						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
251/1000	OG	1.56	1.472	1.415	7	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.292	0.237	0.176
0.0626						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
252/1000	OG	1.596	1.466	1.43	20	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.278	0.276	0.185
0.0629						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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253/1000      OG      1.543      1.449      1.431      14      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all        112      557      0.362      0.25      0.193
0.0688

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
254/1000      OG      1.532      1.705      1.419          1      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.29it/s]
          all        112      557      0.4      0.259      0.208
0.0728

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
255/1000      OG      1.609      1.506      1.468          7      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all        112      557      0.437      0.244      0.224
0.0794

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
256/1000      OG      1.564      1.483      1.429          9      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all        112      557      0.432      0.235      0.221
0.0748

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
257/1000      OG      1.592      1.469      1.458          5      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]

```

0.07	all	112	557	0.38	0.248	0.203
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Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
258/1000	OG	1.539	1.494	1.419	2	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.406	0.233	0.206

0.0728

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
259/1000	OG	1.565	1.448	1.433	11	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.369	0.262	0.215

0.0757

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
260/1000	OG	1.551	1.434	1.398	15	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.386	0.206	0.188

0.0643

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
261/1000	OG	1.562	1.516	1.408	2	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.28it/s]				
	all	112	557	0.326	0.253	0.198

0.0693

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
262/1000	OG	1.533	1.45	1.396	5	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.331	0.276	0.2
0.0674						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
263/1000	OG	1.566	1.472	1.415	4	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.25it/s]				
	all	112	557	0.319	0.233	0.163
0.056						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
264/1000	OG	1.561	1.481	1.418	3	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.33it/s]				
	all	112	557	0.306	0.287	0.21
0.0691						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
265/1000	OG	1.559	1.478	1.418	7	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.333	0.26	0.194
0.0646						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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```

266/1000      OG      1.589      1.486      1.435      25      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all        112        557        0.312        0.212        0.152
0.0511

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
267/1000      OG      1.601      1.527      1.466          6      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.39         0.233        0.205
0.0683

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
268/1000      OG      1.582      1.494      1.424         10      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.30it/s]
          all        112        557        0.344        0.282        0.209
0.0683

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
269/1000      OG      1.528      1.482          1.4          2      640:
100%|        | 29/29 [00:36<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all        112        557        0.307        0.255        0.183
0.06

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
270/1000      OG      1.531      1.441      1.406          5      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]

```

	all	112	557	0.343	0.264	0.2
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0.0671

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
271/1000	OG	1.53	1.432	1.381	3	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.302	0.248	0.17

0.057

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
272/1000	OG	1.545	1.441	1.41	3	640:
100%	29/29	[00:36<00:00, 1.26s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.337	0.245	0.199

0.069

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
273/1000	OG	1.518	1.394	1.393	13	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.356	0.264	0.202

0.0699

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
274/1000	OG	1.524	1.404	1.419	2	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.345	0.244	0.192

0.0671

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
275/1000	OG	1.549	1.509	1.449	14	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.319	0.285	0.204
0.0707						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
276/1000	OG	1.546	1.448	1.405	5	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.24it/s]				
	all	112	557	0.341	0.233	0.194
0.067						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
277/1000	OG	1.568	1.413	1.401	11	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.345	0.26	0.198
0.0649						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
278/1000	OG	1.586	1.529	1.433	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.372	0.257	0.212
0.0699						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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```

279/1000      OG      1.53      1.417      1.413          9      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557          0.369          0.259          0.205
0.0699

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
280/1000      OG      1.505      1.391      1.387          4      640:
100%|        | 29/29 [00:36<00:00, 1.25s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557          0.316          0.237          0.171
0.0599

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
281/1000      OG      1.523      1.374      1.396          1      640:
100%|        | 29/29 [00:36<00:00, 1.26s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]
          all          112          557          0.372          0.235          0.196
0.0681

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
282/1000      OG      1.522      1.384      1.399          6      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.32it/s]
          all          112          557          0.381          0.242          0.216
0.0757

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss  Instances      Size
283/1000      OG      1.521      1.353      1.38          13      640:
100%|        | 29/29 [00:35<00:00, 1.24s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|        | 4/4 [00:03<00:00, 1.31it/s]

```

	all	112	557	0.49	0.214	0.2
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0.0725

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
284/1000	OG	1.521	1.41	1.379	3	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.378	0.25	0.199

0.0707

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
285/1000	OG	1.541	1.442	1.402	5	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.343	0.276	0.212

0.0747

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
286/1000	OG	1.487	1.375	1.364	6	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.32it/s]				
	all	112	557	0.295	0.293	0.206

0.0679

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
287/1000	OG	1.533	1.42	1.405	25	640:
100%	29/29	[00:36<00:00, 1.25s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.29it/s]				
	all	112	557	0.352	0.278	0.213

0.0747

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
288/1000	OG	1.562	1.438	1.428	8	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.31it/s]				
	all	112	557	0.329	0.237	0.177
0.059						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
289/1000	OG	1.531	1.432	1.434	25	640:
100%	29/29	[00:36<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.26it/s]				
	all	112	557	0.358	0.269	0.206
0.0723						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
290/1000	OG	1.536	1.39	1.407	4	640:
100%	29/29	[00:35<00:00, 1.23s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.30it/s]				
	all	112	557	0.316	0.266	0.189
0.0701						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
291/1000	OG	1.448	1.431	1.329	0	640:
100%	29/29	[00:35<00:00, 1.24s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.27it/s]				
	all	112	557	0.415	0.244	0.207
0.0741						

EarlyStopping: Training stopped early as no improvement observed in last 100 epochs. Best results observed at epoch 191, best model saved as best.pt.

To update EarlyStopping(patience=100) pass a new patience value, i.e. `patience=300` or use `patience=0` to disable EarlyStopping.

291 epochs completed in 3.191 hours.

Optimizer stripped from

C:\Users\Carlos\Desktop\pepe\yolov8_model6\weights\last.pt, 6.3MB

Optimizer stripped from

C:\Users\Carlos\Desktop\pepe\yolov8_model6\weights\best.pt, 6.3MB

Validating C:\Users\Carlos\Desktop\pepe\yolov8_model6\weights\best.pt...

Ultralytics 8.3.49 Python-3.10.13 torch-2.5.1+cpu CPU (13th Gen Intel Core(TM) i9-13900KF)

Model summary (fused): 168 layers, 3,005,843 parameters, 0 gradients, 8.1 GFLOPs

	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%		4/4	[00:02<00:00,	1.51it/s]		
	all	112	557	0.45	0.265	0.234

0.0859

Speed: 0.4ms preprocess, 15.8ms inference, 0.0ms loss, 0.6ms postprocess per image

Results saved to C:\Users\Carlos\Desktop\pepe\yolov8_model6

```
[11]: # Cargar el modelo entrenado
modelo = YOLO(r"C:
↪\Users\crome\Desktop\TFM\Modelos_Finales\yolov8_model6\weights\best.pt")
```

```
[12]: from ultralytics import YOLO
import cv2
import matplotlib.pyplot as plt

# Ruta a una imagen de prueba
imagen_prueba = r"C:
↪\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\images\108FTASK_MAX_2379.
↪JPG"

# Realizar la predicción
resultados = modelo(imagen_prueba)

# Dibujar las predicciones sobre la imagen
imagen_con_predicciones = resultados[0].plot()
%matplotlib inline
# Mostrar la imagen con predicciones
plt.figure(figsize=(10, 10))
plt.imshow(cv2.cvtColor(imagen_con_predicciones, cv2.COLOR_BGR2RGB))
plt.axis("off")
plt.show()
```

image 1/1 C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\images\108FTASK_MAX_2379.JPG: 480x640 9 turtle_tracks, 61.9ms
Speed: 4.0ms preprocess, 61.9ms inference, 2.0ms postprocess per image at shape (1, 3, 480, 640)



```
[15]: # Validar el modelo en el conjunto de validación
modelo.val(data=r"C:\Users\crome\Desktop\TFM\Codigo\data.yaml")
```

Ultralytics 8.3.39 Python-3.10.4 torch-2.1.0+cpu CPU (Intel Core(TM) i7-10750H 2.60GHz)

val: Scanning C:\Users\crome\Desktop\TFM\Dades\osaconservation\Data
set_Organizado\val\labels... 112 images, 0 backgrounds, 0 corrupt:
100%| | 112/112 [00:00<00:00, 499.22it/s]

val: New cache created: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\labels.cache

	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%		7/7	[00:08<00:00,	1.22s/it]		
	all	112	557	0.45	0.265	0.234

0.0859

Speed: 1.1ms preprocess, 57.5ms inference, 0.0ms loss, 1.5ms postprocess per image

Results saved to runs\detect\val9

[15]: 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 0x000002B441B312A0>
curves: ['Precision-Recall(B)', 'F1-Confidence(B)', 'Precision-Confidence(B)', 'Recall-Confidence(B)']
curves_results: [[array([
0, 0.001001, 0.002002, 0.003003, 0.004004, 0.005005, 0.006006, 0.007007, 0.008008, 0.009009,
0.01001, 0.011011, 0.012012, 0.013013, 0.014014, 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, 0.064064, 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, 0.088088, 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,	0.5956,	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,
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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,
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0.81582,					
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0.82783,	0.82883,	0.82983,	0.83083,	0.83183,	0.83283,
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0.83984,					

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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,
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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,
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0.98398,					
	0.98498,	0.98599,	0.98699,	0.98799,	0.98899,
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0.43143,	0.42938,	0.42938,	0.42737,	0.42737,	0.41114,
	0.41114,	0.41114,	0.395,	0.395,	0.395,
0.395,	0.395,	0.395,	0.38517,	0.38517,	0.38517,
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	0.36069,	0.34917,	0.34917,	0.34917,	0.34917,
0.34908,	0.34818,	0.34818,	0.34818,	0.34818,	0.33592,
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	0.084908,	0.084908,	0.084123,	0.084123,	0.083602,
0.083602,	0.082278,	0.079967,	0.079967,	0.078753,	0.078753,
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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]), array([[
0.03013,	0.030143,	0.039055,	0.046731,	0.053453,	0.059525,
0.065657,	0.071127,	0.076038,	0.080541,	0.085267,	0.08993,
0.092917,	0.09632,	0.09895,	0.10202,	0.10396,	0.1066,
0.11018,	0.11318,	0.11655,	0.1198,	0.12358,	
	0.12696,	0.13039,	0.13457,	0.13807,	0.14135,
0.14398,	0.1462,	0.14868,	0.15162,	0.15545,	0.15782,
0.16094,	0.16349,	0.16566,	0.16647,	0.16946,	0.17303,
0.17411,	0.17663,	0.17795,	0.18079,	0.18223,	0.18427,

	0.18583,	0.18914,	0.18931,	0.19289,	0.194,
0.19748,	0.20009,	0.20309,	0.20345,	0.20497,	0.20745,
0.20672,	0.20854,	0.21005,	0.21176,	0.21405,	0.2163,
0.21854,	0.21893,	0.22129,	0.22252,	0.22279,	0.2243,
	0.22547,	0.22771,	0.23018,	0.23157,	0.23343,
0.2362,	0.2375,	0.23862,	0.24164,	0.24262,	0.24441,
0.24493,	0.24823,	0.24815,	0.24829,	0.24974,	0.25201,
0.25398,	0.25638,	0.25789,	0.2597,	0.2622,	0.26385,
	0.26238,	0.2637,	0.26567,	0.26779,	0.27021,
0.27118,	0.27046,	0.27182,	0.27129,	0.26947,	0.27166,
0.2727,	0.27403,	0.27509,	0.27589,	0.27692,	0.27744,
0.27938,	0.27982,	0.27993,	0.28104,	0.27938,	0.28116,
	0.28251,	0.28321,	0.28642,	0.28841,	0.28921,
0.29063,	0.28977,	0.29092,	0.29293,	0.29359,	0.29605,
0.29836,	0.29797,	0.29788,	0.29913,	0.30016,	0.30076,
0.30058,	0.30184,	0.30449,	0.30566,	0.30706,	0.30729,
	0.30842,	0.30935,	0.31024,	0.30974,	0.31119,
0.31259,	0.31322,	0.31492,	0.31628,	0.31549,	0.31263,
0.31341,	0.31456,	0.31526,	0.31557,	0.31716,	0.31775,
0.31888,	0.32051,	0.32132,	0.32216,	0.32303,	0.32341,
	0.32278,	0.32179,	0.32216,	0.32204,	0.32309,
0.32358,	0.32465,	0.32518,	0.32564,	0.32226,	0.32294,
0.32464,	0.3262,	0.32542,	0.32302,	0.32238,	0.32287,
0.32347,	0.32242,	0.32305,	0.3217,	0.32214,	0.32371,
	0.32134,	0.32186,	0.32336,	0.32405,	0.3249,
0.32518,	0.32541,	0.32724,	0.32449,	0.32486,	0.32553,
0.32436,	0.32374,	0.32204,	0.32264,	0.32313,	0.32367,
0.32412,	0.3245,	0.32509,	0.32604,	0.32676,	0.32616,
	0.32778,	0.3266,	0.32631,	0.32652,	0.32541,
0.32622,	0.32666,	0.3272,	0.32578,	0.32553,	0.32561,
0.32626,	0.32711,	0.32827,	0.32872,	0.32949,	0.32927,
0.32707,	0.32574,	0.32603,	0.32685,	0.32702,	0.32797,
	0.33007,	0.32708,	0.32725,	0.32762,	0.32652,
0.32699,	0.32811,	0.32822,	0.32832,	0.32849,	0.32939,
0.33125,	0.32854,	0.32924,	0.33072,	0.33247,	0.33297,
0.33397,	0.33276,	0.33346,	0.33196,	0.33263,	0.33118,
	0.3319,	0.33334,	0.33361,	0.33392,	0.33312,
0.33169,	0.33262,	0.33323,	0.33403,	0.3342,	0.33433,
0.33446,	0.33386,	0.33322,	0.33259,	0.33414,	0.33265,
0.33423,	0.33446,	0.33344,	0.33363,	0.33419,	0.33481,
	0.33346,	0.33363,	0.33414,	0.33422,	0.3343,
0.33438,	0.33446,	0.33542,	0.33565,	0.33367,	0.33228,
0.33229,	0.33254,	0.33071,	0.33097,	0.33141,	0.33157,
0.33172,	0.32972,	0.32835,	0.32863,	0.32886,	0.32739,
	0.32762,	0.32787,	0.32809,	0.3285,	0.32752,
0.32659,	0.32703,	0.32745,	0.32784,	0.32819,	0.32895,
0.32912,	0.32928,	0.32916,	0.32831,	0.32803,	0.32712,

0.32622,	0.32535,	0.32455,	0.3247,	0.32488,	0.32571,
	0.32642,	0.32292,	0.32352,	0.3219,	0.3223,
0.32271,	0.32266,	0.32129,	0.32132,	0.3187,	0.31368,
0.31141,	0.31168,	0.31211,	0.31262,	0.31289,	0.31308,
0.31326,	0.31414,	0.31427,	0.3144,	0.31504,	0.31555,
	0.31575,	0.31586,	0.31597,	0.31533,	0.31441,
0.31558,	0.31518,	0.31382,	0.314,	0.31399,	0.31131,
0.30815,	0.30476,	0.30349,	0.30094,	0.29963,	0.29776,
0.2926,	0.29143,	0.29105,	0.2913,	0.29144,	0.29158,
	0.2918,	0.2921,	0.29223,	0.29237,	0.29134,
0.28936,	0.28829,	0.28849,	0.28864,	0.28869,	0.28751,
0.28712,	0.28762,	0.28771,	0.2878,	0.2879,	0.28774,
0.28571,	0.28551,	0.28531,	0.28511,	0.2849,	0.2847,
	0.2845,	0.2843,	0.28409,	0.28389,	0.28369,
0.28349,	0.28194,	0.28146,	0.28213,	0.2797,	0.27805,
0.27819,	0.27832,	0.27844,	0.27869,	0.27912,	0.27696,
0.27657,	0.27619,	0.2758,	0.27541,	0.27502,	0.27511,
	0.27317,	0.27177,	0.27185,	0.27194,	0.27203,
0.27213,	0.27017,	0.27057,	0.27027,	0.26652,	0.2658,
0.26545,	0.26509,	0.26474,	0.26439,	0.26404,	0.26377,
0.26401,	0.26429,	0.2628,	0.25971,	0.25987,	0.26003,
	0.25778,	0.25794,	0.25569,	0.25596,	0.25645,
0.25575,	0.25434,	0.25433,	0.25448,	0.25352,	0.25277,
0.25073,	0.25082,	0.25091,	0.251,	0.24994,	0.24855,
0.24757,	0.2466,	0.24663,	0.24684,	0.24506,	0.24257,
	0.24268,	0.2428,	0.24009,	0.23921,	0.23834,
0.237,	0.23552,	0.2364,	0.23664,	0.23413,	0.23427,
0.23442,	0.23444,	0.23398,	0.23351,	0.23305,	0.23258,
0.23195,	0.23077,	0.22967,	0.23,	0.23015,	0.23031,
	0.22976,	0.22917,	0.22857,	0.22798,	0.22779,
0.22788,	0.22796,	0.22805,	0.2282,	0.22837,	0.22439,
0.21807,	0.21567,	0.21577,	0.21586,	0.21595,	0.21539,
0.21461,	0.21384,	0.21338,	0.21352,	0.21371,	0.2128,
	0.21123,	0.20871,	0.20892,	0.20924,	0.20937,
0.20949,	0.20959,	0.20965,	0.20971,	0.20977,	0.20983,
0.20988,	0.21014,	0.21032,	0.21046,	0.20939,	0.20684,
0.20354,	0.20192,	0.20116,	0.2004,	0.19965,	0.1991,
	0.19855,	0.198,	0.19745,	0.19699,	0.1972,
0.19479,	0.19179,	0.19199,	0.19218,	0.19235,	0.19169,
0.19096,	0.19023,	0.18925,	0.1868,	0.18619,	0.18557,
0.18496,	0.18434,	0.18414,	0.18432,	0.18476,	0.1828,
	0.18254,	0.17986,	0.17722,	0.17497,	0.17448,
0.17452,	0.17456,	0.1746,	0.17464,	0.17468,	0.17472,
0.1748,	0.17441,	0.17402,	0.17363,	0.17323,	0.17284,
0.17245,	0.17223,	0.172,	0.17123,	0.17046,	0.16969,
	0.1696,	0.16966,	0.16973,	0.16979,	0.16815,
0.167,	0.16717,	0.16723,	0.16728,	0.16733,	0.16737,

0.16742,	0.16748,	0.1676,	0.16771,	0.16659,	0.16538,
0.16435,	0.16346,	0.16258,	0.16193,	0.16201,	0.1621,
	0.16081,	0.15888,	0.15767,	0.15625,	0.15318,
0.15209,	0.151,	0.15087,	0.14935,	0.14792,	0.14805,
0.14808,	0.14765,	0.14721,	0.14678,	0.14635,	0.14591,
0.14548,	0.14518,	0.14526,	0.14534,	0.14511,	0.1443,
	0.14348,	0.14267,	0.14063,	0.13755,	0.1363,
0.13589,	0.13548,	0.13507,	0.13465,	0.13424,	0.13383,
0.13296,	0.1311,	0.13055,	0.13062,	0.13068,	0.13074,
0.13081,	0.13088,	0.13102,	0.13126,	0.13103,	0.12941,
	0.12834,	0.12848,	0.12721,	0.12595,	0.12508,
0.12449,	0.12391,	0.12333,	0.12274,	0.12209,	0.12123,
0.12036,	0.1195,	0.1159,	0.11539,	0.11488,	0.11437,
0.11386,	0.11334,	0.11276,	0.11194,	0.11113,	0.11031,
	0.10933,	0.10807,	0.10681,	0.10617,	0.1056,
0.10504,	0.10447,	0.10391,	0.10341,	0.10308,	0.10275,
0.10241,	0.10208,	0.10175,	0.10141,	0.10108,	0.10075,
0.10042,	0.098062,	0.096569,	0.095731,	0.094893,	0.094055,
	0.094078,	0.093644,	0.09321,	0.092776,	0.092342,
0.091908,	0.091474,	0.091039,	0.086469,	0.08442,	0.08391,
0.0834,	0.08289,	0.082379,	0.081869,	0.081358,	0.078322,
0.075253,	0.074734,	0.074772,	0.074809,	0.074861,	0.074931,
	0.074966,	0.074981,	0.074996,	0.075011,	0.075025,
0.07504,	0.075054,	0.075069,	0.075083,	0.074382,	0.073519,
0.072655,	0.07179,	0.070683,	0.069576,	0.068481,	0.067942,
0.067404,	0.066865,	0.066326,	0.065786,	0.065247,	0.063129,
	0.06193,	0.060255,	0.058623,	0.058633,	0.058644,
0.058654,	0.058664,	0.058674,	0.058684,	0.058694,	0.058703,
0.058713,	0.058689,	0.058262,	0.057836,	0.057409,	0.056982,
0.056555,	0.056128,	0.055701,	0.053714,	0.048414,	0.047639,
	0.046864,	0.046089,	0.045313,	0.045223,	0.04523,
0.045237,	0.045243,	0.04525,	0.045256,	0.045263,	0.045269,
0.045275,	0.045282,	0.045288,	0.045294,	0.045328,	0.045194,
0.039314,	0.034718,	0.034043,	0.033367,	0.032691,	0.032015,
	0.030541,	0.028169,	0.028176,	0.028182,	0.028189,
0.028195,	0.028201,	0.028206,	0.028212,	0.028218,	0.027436,
0.026487,	0.025538,	0.024694,	0.024431,	0.024167,	0.023904,
0.023641,	0.023377,	0.023114,	0.02285,	0.022586,	0.022323,
	0.022059,	0.021795,	0.021531,	0.021267,	0.021247,
0.021256,	0.021265,	0.021272,	0.021086,	0.020656,	0.020225,
0.019794,	0.019362,	0.018931,	0.018499,	0.018068,	0.017604,
0.017062,	0.01652,	0.015978,	0.015436,	0.014893,	0.014351,
	0.014236,	0.014236,	0.014237,	0.014238,	0.014239,
0.01424,	0.01424,	0.014241,	0.014242,	0.014243,	0.014243,
0.014244,	0.014245,	0.014246,	0.014246,	0.014247,	0.014248,
0.014248,	0.014249,	0.01425,	0.014251,	0.014251,	0.014252,
	0.014252,	0.014253,	0.014254,	0.014254,	0.014255,

[illegible]

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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,	0.5956,	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]], array([[
0.015408,	0.015415,	0.020122,	0.024245,	0.02791,	0.031283,
0.034716,	0.03782,	0.040643,	0.043276,	0.04604,	0.04877,

0.050602,	0.052659,	0.054302,	0.056194,	0.057493,	0.059177,
0.061391,	0.063332,	0.065507,	0.067599,	0.07008,	
	0.072293,	0.074565,	0.077353,	0.079712,	0.081956,
0.083872,	0.085541,	0.087351,	0.089505,	0.092186,	0.09399,
0.096208,	0.098255,	0.10005,	0.10097,	0.10326,	0.10593,
0.1071,	0.10922,	0.11044,	0.11264,	0.11398,	0.1157,
	0.11705,	0.11968,	0.12035,	0.12325,	0.12445,
0.12732,	0.1295,	0.13203,	0.13282,	0.1343,	0.13643,
0.13652,	0.13831,	0.13983,	0.14156,	0.14382,	0.14586,
0.14791,	0.14872,	0.15091,	0.15254,	0.15355,	0.15499,
	0.15638,	0.15854,	0.16094,	0.1626,	0.16444,
0.16721,	0.16882,	0.17029,	0.17337,	0.17473,	0.1766,
0.1775,	0.18102,	0.18166,	0.1826,	0.18458,	0.18707,
0.18925,	0.19193,	0.19408,	0.19614,	0.19901,	0.20091,
	0.20069,	0.20224,	0.20457,	0.2071,	0.21001,
0.21145,	0.21143,	0.21335,	0.21361,	0.21314,	0.2159,
0.21722,	0.21955,	0.22091,	0.22195,	0.22329,	0.22434,
0.2272,	0.2285,	0.22938,	0.23088,	0.23013,	0.23254,
	0.23439,	0.23536,	0.23982,	0.24263,	0.24462,
0.24676,	0.24629,	0.24796,	0.2509,	0.25279,	0.25646,
0.25995,	0.2602,	0.26122,	0.26316,	0.26476,	0.26569,
0.26647,	0.26845,	0.27268,	0.27456,	0.27683,	0.27837,
	0.28024,	0.28177,	0.28325,	0.28366,	0.28609,
0.28848,	0.28954,	0.29247,	0.29482,	0.29478,	0.29334,
0.29527,	0.29732,	0.29856,	0.29912,	0.30199,	0.30307,
0.30513,	0.30812,	0.30962,	0.31118,	0.31282,	0.31352,
	0.31403,	0.31366,	0.31478,	0.3158,	0.31782,
0.31875,	0.32085,	0.32188,	0.32278,	0.32072,	0.32271,
0.32614,	0.3293,	0.32926,	0.32804,	0.32904,	0.33008,
0.33132,	0.33109,	0.33243,	0.33149,	0.33253,	0.33589,
	0.33494,	0.33608,	0.33937,	0.34089,	0.34276,
0.34339,	0.34391,	0.34812,	0.34638,	0.34721,	0.34874,
0.34842,	0.34838,	0.34781,	0.34923,	0.35037,	0.35163,
0.35271,	0.35361,	0.35501,	0.35728,	0.35902,	0.36016,
	0.36414,	0.36391,	0.36593,	0.36729,	0.36635,
0.3685,	0.36964,	0.37101,	0.37025,	0.37092,	0.37275,
0.37447,	0.37671,	0.3798,	0.38101,	0.38309,	0.38432,
0.38306,	0.3819,	0.38327,	0.38554,	0.38602,	0.38867,
	0.39462,	0.39298,	0.39353,	0.39461,	0.39499,
0.39638,	0.3997,	0.4,	0.4003,	0.4008,	0.40351,
0.40915,	0.40819,	0.41084,	0.41546,	0.42105,	0.42265,
0.42588,	0.42591,	0.42859,	0.42805,	0.43113,	0.43,
	0.43242,	0.43733,	0.43827,	0.43933,	0.44,
0.44127,	0.44456,	0.44678,	0.44963,	0.45025,	0.45073,
0.4512,	0.45068,	0.45012,	0.44955,	0.45531,	0.4551,
0.46106,	0.46196,	0.46222,	0.4644,	0.46658,	0.46902,
	0.46799,	0.47022,	0.47227,	0.47259,	0.47291,

0.47324,	0.47356,	0.47739,	0.47839,	0.47662,	0.47537,
0.47701,	0.47805,	0.4768,	0.47789,	0.47976,	0.48039,
0.48103,	0.48068,	0.47989,	0.4811,	0.48207,	0.48118,
	0.48348,	0.48455,	0.48552,	0.48756,	0.48668,
0.48588,	0.48782,	0.48971,	0.49143,	0.49302,	0.49647,
0.49722,	0.49797,	0.49949,	0.49872,	0.49993,	0.4991,
0.49828,	0.49896,	0.49858,	0.49931,	0.50017,	0.50409,
	0.5075,	0.50664,	0.50961,	0.50991,	0.51192,
0.51398,	0.51538,	0.51411,	0.51554,	0.51506,	0.51036,
0.50854,	0.50998,	0.51231,	0.51506,	0.51653,	0.51754,
0.51856,	0.52339,	0.52412,	0.52484,	0.5284,	0.5313,
	0.53244,	0.53308,	0.53371,	0.53351,	0.53499,
0.54184,	0.54316,	0.5421,	0.54321,	0.54408,	0.54155,
0.53855,	0.5353,	0.53408,	0.53365,	0.53238,	0.53054,
0.52751,	0.52634,	0.52677,	0.52839,	0.5293,	0.5302,
	0.5317,	0.53368,	0.53458,	0.53548,	0.5348,
0.53281,	0.53369,	0.53502,	0.5361,	0.53881,	0.53761,
0.53893,	0.54242,	0.54308,	0.54375,	0.54441,	0.54476,
0.54271,	0.54251,	0.5423,	0.54209,	0.54189,	0.54168,
	0.54148,	0.54127,	0.54106,	0.54086,	0.54065,
0.54045,	0.53886,	0.54026,	0.54521,	0.54373,	0.54434,
0.54541,	0.54636,	0.54731,	0.54922,	0.55258,	0.55346,
0.55306,	0.55265,	0.55225,	0.55185,	0.55145,	0.55277,
	0.55212,	0.55889,	0.55963,	0.56036,	0.56109,
0.56198,	0.56288,	0.56641,	0.56835,	0.56438,	0.56361,
0.56323,	0.56286,	0.56248,	0.5621,	0.56173,	0.56172,
0.56391,	0.56642,	0.56626,	0.56327,	0.56476,	0.56625,
	0.56518,	0.56678,	0.56576,	0.56842,	0.57324,
0.57404,	0.57249,	0.57372,	0.57528,	0.57481,	0.58029,
0.58201,	0.58298,	0.58394,	0.58491,	0.58428,	0.58623,
0.58514,	0.58405,	0.58578,	0.58821,	0.58949,	0.59094,
	0.59231,	0.59369,	0.59125,	0.59025,	0.58925,
0.5877,	0.58851,	0.59963,	0.60277,	0.60033,	0.60224,
0.60415,	0.60863,	0.60809,	0.60755,	0.60701,	0.60648,
0.60574,	0.60436,	0.60331,	0.60783,	0.61005,	0.61228,
	0.61175,	0.61105,	0.61035,	0.60966,	0.61012,
0.61138,	0.61263,	0.61388,	0.6161,	0.6186,	0.61432,
0.6113,	0.60858,	0.61008,	0.61159,	0.61309,	0.61271,
0.61176,	0.6108,	0.61095,	0.61323,	0.61643,	0.61924,
	0.61729,	0.61513,	0.61885,	0.62452,	0.62691,
0.62898,	0.63084,	0.63191,	0.63297,	0.63404,	0.6351,
0.63616,	0.64088,	0.64423,	0.64689,	0.64676,	0.64364,
0.6395,	0.63746,	0.63649,	0.63551,	0.63455,	0.63384,
	0.63313,	0.63241,	0.6317,	0.63156,	0.63574,
0.63401,	0.63023,	0.63455,	0.63864,	0.64256,	0.64196,
0.64098,	0.64001,	0.63869,	0.63538,	0.63453,	0.63368,
0.63283,	0.63198,	0.63383,	0.63813,	0.65198,	0.6493,

	0.65983,	0.65949,	0.66268,	0.65954,	0.65968,
0.6608,	0.66192,	0.66304,	0.66415,	0.66527,	0.66639,
0.67442,	0.67388,	0.67333,	0.67279,	0.67224,	0.6717,
0.67115,	0.67345,	0.67843,	0.67735,	0.67628,	0.67521,
	0.6767,	0.6788,	0.68091,	0.68301,	0.68121,
0.68196,	0.68744,	0.68965,	0.69125,	0.69285,	0.69445,
0.69604,	0.69808,	0.70224,	0.7064,	0.70512,	0.70347,
0.70204,	0.70081,	0.69957,	0.69941,	0.70269,	0.70598,
	0.70643,	0.71343,	0.71173,	0.70971,	0.70526,
0.70364,	0.70202,	0.7117,	0.70984,	0.70819,	0.71394,
0.71864,	0.718,	0.71735,	0.7167,	0.71606,	0.71541,
0.71476,	0.71533,	0.71928,	0.72324,	0.72538,	0.72416,
	0.72294,	0.72172,	0.73035,	0.72563,	0.72368,
0.72303,	0.72237,	0.72171,	0.72106,	0.7204,	0.71974,
0.71832,	0.71527,	0.71696,	0.72092,	0.72487,	0.72888,
0.733,	0.73711,	0.7461,	0.76204,	0.76874,	0.76635,
	0.76839,	0.77997,	0.77813,	0.77628,	0.775,
0.77412,	0.77324,	0.77236,	0.77148,	0.77048,	0.76913,
0.76777,	0.76642,	0.76056,	0.75969,	0.75882,	0.75795,
0.75708,	0.75621,	0.7552,	0.75376,	0.75232,	0.75088,
	0.74909,	0.74677,	0.74445,	0.74322,	0.74214,
0.74105,	0.73996,	0.73887,	0.7379,	0.73723,	0.73656,
0.73589,	0.73522,	0.73455,	0.73388,	0.73321,	0.73254,
0.73187,	0.72692,	0.72371,	0.72186,	0.72001,	0.71816,
	0.73675,	0.73579,	0.73483,	0.73387,	0.7329,
0.73194,	0.73098,	0.73002,	0.71922,	0.71418,	0.71286,
0.71154,	0.71021,	0.70889,	0.70757,	0.70624,	0.69794,
0.68909,	0.69285,	0.69941,	0.70597,	0.7153,	0.72822,
	0.73507,	0.73795,	0.74083,	0.7437,	0.74658,
0.74946,	0.75234,	0.75521,	0.75809,	0.75678,	0.75451,
0.75225,	0.74999,	0.74688,	0.74378,	0.7407,	0.73908,
0.73746,	0.73584,	0.73421,	0.73259,	0.73097,	0.72412,
	0.74083,	0.74444,	0.73991,	0.74322,	0.74653,
0.74983,	0.75314,	0.75645,	0.75976,	0.76306,	0.76637,
0.76968,	0.77262,	0.77124,	0.76987,	0.76849,	0.76712,
0.76574,	0.76437,	0.76299,	0.75607,	0.73599,	0.73265,
	0.72931,	0.72597,	0.72263,	0.72526,	0.72873,
0.7322,	0.73567,	0.73914,	0.74261,	0.74608,	0.74955,
0.75302,	0.75649,	0.75995,	0.76342,	0.78328,	0.81183,
0.78898,	0.76717,	0.76341,	0.75964,	0.75588,	0.75212,
	0.74282,	0.72773,	0.73657,	0.74541,	0.75425,
0.76309,	0.77193,	0.78077,	0.78961,	0.79845,	0.795,
0.78895,	0.78289,	0.77745,	0.77536,	0.77326,	0.77117,
0.76907,	0.76698,	0.76489,	0.76279,	0.7607,	0.7586,
	0.75651,	0.75441,	0.75232,	0.75022,	0.7715,
0.79557,	0.81964,	0.84371,	0.85585,	0.85293,	0.85001,
0.84709,	0.84417,	0.84124,	0.83832,	0.8354,	0.83183,

0.077077,	0.078078,	0.079079,	0.08008,	0.081081,	0.082082,
0.083083,	0.084084,	0.085085,	0.086086,	0.087087,	0.088088,
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,	0.5956,	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,					
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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,					
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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,
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0.996,	0.997,	0.998,	0.999,	1]), array([[
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0.5368,	0.53142,	0.52783,	0.52603,	0.52244,	
	0.52065,	0.51885,	0.51706,	0.51526,	0.51346,
0.50808,	0.50269,	0.4991,	0.49551,	0.49551,	0.49192,
0.49192,	0.48654,	0.48115,	0.47397,	0.47217,	0.47217,
0.46499,	0.4614,	0.45781,	0.45781,	0.45422,	0.45242,
	0.45063,	0.45063,	0.44345,	0.44345,	0.43986,
0.43986,	0.43986,	0.43986,	0.43447,	0.43268,	0.43268,
0.42549,	0.4237,	0.4219,	0.42011,	0.41831,	0.41831,
0.41831,	0.41472,	0.41472,	0.41113,	0.40575,	0.40575,
	0.40395,	0.40395,	0.40395,	0.40215,	0.40215,
0.40215,	0.40036,	0.39856,	0.39856,	0.39677,	0.39677,
0.39497,	0.39483,	0.39138,	0.38779,	0.386,	0.386,
0.386,	0.386,	0.3842,	0.3842,	0.3842,	0.3842,
	0.37882,	0.37882,	0.37882,	0.37882,	0.37882,
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0.36625,	0.36445,	0.36445,	0.36445,	0.36445,	0.36348,
0.36266,	0.36086,	0.35907,	0.35907,	0.35548,	0.35548,
	0.35548,	0.35548,	0.35548,	0.35548,	0.35368,
0.35348,	0.35189,	0.35189,	0.35189,	0.35009,	0.35009,
0.35009,	0.34857,	0.3465,	0.3465,	0.3465,	0.3465,
0.3447,	0.3447,	0.3447,	0.3447,	0.3447,	0.34291,
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0.33393,	0.33393,	0.33393,	0.33393,	0.33393,	0.33393,
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0.32316,	0.32316,	0.32167,	0.31816,	0.31598,	0.31598,
0.31598,	0.31418,	0.31418,	0.31248,	0.31239,	0.31239,
	0.3088,	0.3088,	0.3088,	0.3088,	0.3088,
0.3088,	0.3088,	0.30872,	0.30521,	0.30521,	0.30521,
0.30341,	0.30236,	0.29982,	0.29982,	0.29982,	0.29982,
0.29982,	0.29982,	0.29982,	0.29982,	0.29982,	0.29803,
	0.29803,	0.29623,	0.29443,	0.2939,	0.29271,
0.29264,	0.29264,	0.29264,	0.29084,	0.29005,	0.28905,
0.28905,	0.28905,	0.28905,	0.28905,	0.28905,	0.28801,
0.28536,	0.28397,	0.28366,	0.28366,	0.28366,	0.28366,
	0.28366,	0.28011,	0.28007,	0.28007,	0.27828,
0.27828,	0.27828,	0.27828,	0.27828,	0.27828,	0.27828,
0.27828,	0.2749,	0.27469,	0.27469,	0.27469,	0.27469,

0.27469,	0.27304,	0.27289,	0.2711,	0.27077,	0.2693,
	0.2693,	0.2693,	0.2693,	0.2693,	0.26802,
0.26571,	0.26571,	0.26571,	0.26571,	0.26571,	0.26571,
0.26571,	0.26513,	0.26453,	0.26393,	0.26391,	0.26212,
0.26212,	0.26212,	0.26078,	0.26032,	0.26032,	0.26032,
	0.25901,	0.25853,	0.25853,	0.25853,	0.25853,
0.25853,	0.25853,	0.25853,	0.25851,	0.25669,	0.2554,
0.25494,	0.25494,	0.25314,	0.25314,	0.25314,	0.25314,
0.25314,	0.25092,	0.24955,	0.24955,	0.24955,	0.2481,
	0.24776,	0.24776,	0.24776,	0.24769,	0.24681,
0.24596,	0.24596,	0.24596,	0.24596,	0.24596,	0.24596,
0.24596,	0.24596,	0.24546,	0.2447,	0.24409,	0.24329,
0.24249,	0.24137,	0.24057,	0.24057,	0.24057,	0.24057,
	0.24057,	0.23698,	0.23698,	0.23519,	0.23519,
0.23519,	0.23485,	0.23366,	0.23339,	0.23073,	0.22643,
0.22442,	0.22442,	0.22442,	0.22442,	0.22442,	0.22442,
0.22442,	0.22442,	0.22442,	0.22442,	0.22442,	0.22442,
	0.22442,	0.22442,	0.22442,	0.22381,	0.22262,
0.22262,	0.222,	0.22083,	0.22083,	0.22067,	0.21844,
0.21582,	0.21301,	0.21197,	0.20955,	0.20849,	0.20695,
0.20244,	0.20149,	0.20108,	0.20108,	0.20108,	0.20108,
	0.20108,	0.20108,	0.20108,	0.20108,	0.2002,
0.19861,	0.19749,	0.19749,	0.19749,	0.19717,	0.19622,
0.19569,	0.19569,	0.19569,	0.19569,	0.19569,	0.19551,
0.19389,	0.19373,	0.19357,	0.19341,	0.19325,	0.19309,
	0.19293,	0.19277,	0.19261,	0.19245,	0.19229,
0.19213,	0.19091,	0.19031,	0.19031,	0.18827,	0.18671,
0.18671,	0.18671,	0.18671,	0.18671,	0.18671,	0.18469,
0.18439,	0.18409,	0.18379,	0.18349,	0.18319,	0.18312,
	0.18148,	0.17953,	0.17953,	0.17953,	0.17953,
0.17953,	0.17774,	0.17774,	0.17729,	0.17445,	0.17391,
0.17364,	0.17338,	0.17311,	0.17284,	0.17258,	0.17235,
0.17235,	0.17235,	0.1711,	0.16876,	0.16876,	0.16876,
	0.16697,	0.16697,	0.16517,	0.16517,	0.16517,
0.16453,	0.16349,	0.16338,	0.16338,	0.16262,	0.16158,
0.15978,	0.15978,	0.15978,	0.15978,	0.15897,	0.15771,
0.157,	0.15629,	0.15619,	0.15619,	0.15468,	0.1526,
	0.1526,	0.1526,	0.15062,	0.15,	0.14938,
0.14843,	0.14722,	0.14722,	0.14722,	0.14542,	0.14542,
0.14542,	0.14518,	0.14486,	0.14453,	0.1442,	0.14388,
0.14344,	0.14261,	0.14183,	0.14183,	0.14183,	0.14183,
	0.14144,	0.14103,	0.14062,	0.1402,	0.14004,
0.14004,	0.14004,	0.14004,	0.14004,	0.14004,	0.13726,
0.1327,	0.13106,	0.13106,	0.13106,	0.13106,	0.13066,
0.13013,	0.12961,	0.12926,	0.12926,	0.12926,	0.12848,
	0.12742,	0.12567,	0.12567,	0.12567,	0.12567,
0.12567,	0.12567,	0.12567,	0.12567,	0.12567,	0.12567,

0.12567,	0.12567,	0.12567,	0.12567,	0.12491,	0.12322,
0.12103,	0.11996,	0.11946,	0.11895,	0.11846,	0.1181,
	0.11774,	0.11738,	0.11702,	0.1167,	0.1167,
0.11507,	0.11311,	0.11311,	0.11311,	0.11311,	0.11267,
0.11219,	0.11172,	0.11108,	0.1095,	0.1091,	0.1087,
0.10831,	0.10791,	0.10772,	0.10772,	0.10763,	0.10637,
	0.10592,	0.10413,	0.10228,	0.10086,	0.10054,
0.10054,	0.10054,	0.10054,	0.10054,	0.10054,	0.10054,
0.10041,	0.10017,	0.09992,	0.099674,	0.099428,	0.099181,
0.098935,	0.098743,	0.098481,	0.098001,	0.097521,	0.097041,
	0.096948,	0.096948,	0.096948,	0.096948,	0.095913,
0.095153,	0.095153,	0.095153,	0.095153,	0.095153,	0.095153,
0.095153,	0.095153,	0.095153,	0.095153,	0.094452,	0.093705,
0.093066,	0.092521,	0.091977,	0.091562,	0.091562,	0.091562,
	0.090731,	0.089395,	0.088658,	0.08779,	0.085922,
0.085259,	0.084596,	0.084381,	0.083455,	0.082585,	0.082585,
0.082543,	0.082283,	0.082022,	0.081762,	0.081502,	0.081242,
0.080982,	0.08079,	0.08079,	0.08079,	0.080619,	0.080132,
	0.079646,	0.079159,	0.077809,	0.075976,	0.075236,
0.074992,	0.074748,	0.074505,	0.074261,	0.074017,	0.073774,
0.073257,	0.072165,	0.071813,	0.071813,	0.071813,	0.071813,
0.071813,	0.071813,	0.071813,	0.071813,	0.071619,	0.07067,
	0.070018,	0.070006,	0.069269,	0.068532,	0.068027,
0.067689,	0.067352,	0.067014,	0.066676,	0.066297,	0.065799,
0.0653,	0.064802,	0.062732,	0.062438,	0.062144,	0.061851,
0.061557,	0.061263,	0.060927,	0.060461,	0.059995,	0.05953,
	0.058966,	0.05825,	0.057534,	0.057167,	0.056846,
0.056526,	0.056205,	0.055884,	0.055601,	0.055413,	0.055225,
0.055036,	0.054848,	0.054659,	0.054471,	0.054282,	0.054094,
0.053906,	0.052577,	0.051736,	0.051265,	0.050794,	0.050323,
	0.050247,	0.050004,	0.049761,	0.049518,	0.049275,
0.049032,	0.04879,	0.048547,	0.046,	0.044861,	0.044579,
0.044296,	0.044013,	0.043731,	0.043448,	0.043165,	0.041489,
0.039799,	0.039497,	0.039497,	0.039497,	0.039497,	0.039497,
	0.039497,	0.039497,	0.039497,	0.039497,	0.039497,
0.039497,	0.039497,	0.039497,	0.039497,	0.039113,	0.038642,
0.038171,	0.037699,	0.037097,	0.036495,	0.0359,	0.035608,
0.035316,	0.035024,	0.034732,	0.034439,	0.034147,	0.033003,
	0.032316,	0.031398,	0.030521,	0.030521,	0.030521,
0.030521,	0.030521,	0.030521,	0.030521,	0.030521,	0.030521,
0.030521,	0.030503,	0.030275,	0.030046,	0.029818,	0.02959,
0.029362,	0.029134,	0.028906,	0.027846,	0.02503,	0.02462,
	0.02421,	0.0238,	0.02339,	0.023339,	0.023339,
0.023339,	0.023339,	0.023339,	0.023339,	0.023339,	0.023339,
0.023339,	0.023339,	0.023339,	0.023339,	0.023339,	0.023244,
0.020159,	0.017761,	0.017409,	0.017058,	0.016707,	0.016356,
	0.015591,	0.014363,	0.014363,	0.014363,	0.014363,


```

save_dir: WindowsPath('runs/detect/val9')
speed: {'preprocess': 1.1174529790878296, 'inference': 57.51746892929077,
'loss': 0.0, 'postprocess': 1.4801046677998133}
task: 'detect'

```

Implementación de U-Net

Porque la detección de huellas de tortugas es un problema de segmentación de instancias, se puede utilizar una arquitectura de red neuronal convolucional (CNN) llamada U-Net. U-Net es una red neuronal convolucional profunda que se utiliza para la segmentación de imágenes. La arquitectura de U-Net consta de un codificador y un decodificador. El codificador es una pila de capas convolucionales que se utilizan para extraer características de la imagen de entrada. El decodificador es una pila de capas convolucionales que se utilizan para generar una máscara de segmentación de la imagen de entrada.

1. Introducción

Este notebook implementa y entrena un modelo U-Net para segmentación de huellas. Sigue los pasos detallados para cargar, preprocesar los datos, definir la arquitectura, entrenar el modelo y evaluarlo.

2. Preparación Inicial

```

[1]: # Importar librerías necesarias
import os
import numpy as np
import tensorflow as tf
from tensorflow.keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D, concatenate
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import Adam
import cv2
import matplotlib.pyplot as plt

# Configuración inicial
IMAGE_SIZE = 256 # Tamaño al que redimensionaremos las imágenes y máscaras
BATCH_SIZE = 20 # Tamaño del lote
EPOCHS = 50 # Número de épocas
LEARNING_RATE = 1e-4 # Tasa de aprendizaje

```

```

[29]: import os
import cv2
import numpy as np
import json

# Configuración de directorios
IMAGENES_DIR = r"C:\Users\crome\Desktop\TFM\Dades\osaconservation\Para_Etiquetar_Huellas\images"

```

```

JSON_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Para_Etiquetar_Huellas\json"
MASCARAS_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\masks"

# Crear carpeta para las máscaras si no existe
os.makedirs(MASCARAS_DIR, exist_ok=True)

def crear_mascaras(imagenes_dir, json_dir, mascarar_dir, tamaño=(256, 256)):
    """
    Genera máscaras en escala de grises a partir de archivos JSON.
    """
    for json_file in os.listdir(json_dir):
        if json_file.endswith(".json"):
            ruta_json = os.path.join(json_dir, json_file)
            ruta_imagen = os.path.join(imagenes_dir, json_file.replace(".json",
↳".jpg"))

            # Leer el archivo JSON
            with open(ruta_json, 'r') as f:
                datos = json.load(f)

            # Leer dimensiones de la imagen
            imagen = cv2.imread(ruta_imagen)
            altura, ancho, _ = imagen.shape

            # Crear una máscara vacía
            mascara = np.zeros((altura, ancho), dtype=np.uint8)

            # Dibujar las líneas o formas en la máscara
            for shape in datos.get("shapes", []):
                if shape["shape_type"] == "linestrip":
                    puntos = np.array(shape["points"], dtype=np.int32)
                    cv2.polylines(mascara, [puntos], isClosed=False, color=255,
↳thickness=5)

            # Redimensionar la máscara al tamaño deseado
            mascara = cv2.resize(mascara, tamaño, interpolation=cv2.
↳INTER_NEAREST)

            # Guardar la máscara
            ruta_mascara = os.path.join(mascarar_dir, json_file.replace(".
↳json", ".png"))
            cv2.imwrite(ruta_mascara, mascara)

# Generar máscaras

```

```

crear_mascaras(IMAGENES_DIR, JSON_DIR, MASCARAS_DIR, tamaño=(256, 256))

print(f"Máscaras generadas en: {MASCARAS_DIR}")

```

Máscaras generadas en:

C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\masks

```

[25]: # poner maskara en train y val
import os
import shutil

# Configuración de directorios
MASCARAS_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\masks"
TRAIN_MASKS = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train\masks"
VAL_MASKS = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\masks"

# Crear directorios para las máscaras en train y val
os.makedirs(TRAIN_MASKS, exist_ok=True)
os.makedirs(VAL_MASKS, exist_ok=True)

# Mover las máscaras a train y val
mascaras = os.listdir(MASCARAS_DIR)
for i, mascara in enumerate(mascaras):
    if i % 5 == 0:
        shutil.move(os.path.join(MASCARAS_DIR, mascara), os.path.
↳join(VAL_MASKS, mascara))
    else:
        shutil.move(os.path.join(MASCARAS_DIR, mascara), os.path.
↳join(TRAIN_MASKS, mascara))

print(f"Máscaras movidas a: {TRAIN_MASKS} y {VAL_MASKS}")

```

Máscaras movidas a:

C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train\masks
y C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val\masks

```

[27]: import shutil
import random

# Configuración de directorios
DATASET_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado"
TRAIN_DIR = os.path.join(DATASET_DIR, "train")

```

```

VAL_DIR = os.path.join(DATASET_DIR, "val")

# Crear carpetas de entrenamiento y validación
os.makedirs(os.path.join(TRAIN_DIR, "images"), exist_ok=True)
os.makedirs(os.path.join(TRAIN_DIR, "masks"), exist_ok=True)
os.makedirs(os.path.join(VAL_DIR, "images"), exist_ok=True)
os.makedirs(os.path.join(VAL_DIR, "masks"), exist_ok=True)

# Listar imágenes y máscaras
imagenes = os.listdir(IMAGENES_DIR)
mascaras = os.listdir(MASCARAS_DIR)

# Ordenar y dividir aleatoriamente
random.seed(42)
datos = list(zip(imagenes, mascaras))
random.shuffle(datos)

# Dividir en 80% entrenamiento y 20% validación
split_index = int(len(datos) * 0.8)
train_data = datos[:split_index]
val_data = datos[split_index:]

# Copiar datos
def copiar_datos(datos, destino_imagenes, destino_mascaras):
    for img_file, mask_file in datos:
        shutil.copy(os.path.join(IMAGENES_DIR, img_file), os.path.
↪join(destino_imagenes, img_file))
        shutil.copy(os.path.join(MASCARAS_DIR, mask_file), os.path.
↪join(destino_mascaras, mask_file))

copiar_datos(train_data, os.path.join(TRAIN_DIR, "images"), os.path.
↪join(TRAIN_DIR, "masks"))
copiar_datos(val_data, os.path.join(VAL_DIR, "images"), os.path.join(VAL_DIR,
↪"masks"))

print("Datos organizados en:")
print(f"- Entrenamiento: {TRAIN_DIR}")
print(f"- Validación: {VAL_DIR}")

```

Datos organizados en:

- Entrenamiento:

C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train

- Validación:

C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val

3. Cargar y Preprocesar los Datos

```

[1]: import os
import cv2
import numpy as np

IMAGE_SIZE = 256 # Tamaño de las imágenes
TRAIN_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train"
VAL_DIR = r"C:
↳\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\val"

def load_data(image_path, mask_path):
    """
    Carga y preprocesa las imágenes y máscaras.
    """
    images = []
    masks = []

    # Iterar sobre las imágenes en la carpeta
    for img_file in os.listdir(image_path):
        # Construir la ruta de la imagen
        img_path = os.path.join(image_path, img_file)

        # Leer la imagen
        if not os.path.exists(img_path):
            print(f"Imagen no encontrada: {img_path}")
            continue

        img = cv2.imread(img_path, cv2.IMREAD_COLOR)
        if img is None:
            print(f"Error al cargar la imagen: {img_path}")
            continue

        # Construir el nombre y la ruta de la máscara correspondiente
        mask_file = img_file.replace(".jpg", ".png").replace(".jpeg", ".png").
↳replace(".JPG", ".png")
        mask_path_full = os.path.join(mask_path, mask_file)

        # Leer la máscara
        if not os.path.exists(mask_path_full):
            print(f"Máscara no encontrada: {mask_path_full}")
            continue

        mask = cv2.imread(mask_path_full, cv2.IMREAD_GRAYSCALE | cv2.
↳IMREAD_ANYDEPTH)
        if mask is None:
            print(f"Error al cargar la máscara: {mask_path_full}")
            continue

```

```

# Redimensionar y normalizar las imágenes y máscaras
img = cv2.resize(img, (IMAGE_SIZE, IMAGE_SIZE)) / 255.0
mask = cv2.resize(mask, (IMAGE_SIZE, IMAGE_SIZE)) / 255.0

# Expandir la dimensión de la máscara para que sea compatible
mask = np.expand_dims(mask, axis=-1)

# Agregar la imagen y la máscara a las listas
images.append(img)
masks.append(mask)

return np.array(images), np.array(masks)

# Cargar datos
train_images, train_masks = load_data(TRAIN_DIR + "/images", TRAIN_DIR + "/"
↪masks")
val_images, val_masks = load_data(VAL_DIR + "/images", VAL_DIR + "/masks")

# Mostrar forma de los datos cargados
print(f"Imágenes de entrenamiento: {train_images.shape}, Máscaras de_
↪entrenamiento: {train_masks.shape}")
print(f"Imágenes de validación: {val_images.shape}, Máscaras de validación:_
↪{val_masks.shape}")

```

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0726.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0742.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0762.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0790.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0801.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0829.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0872.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_0905.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\102FTASK_IRX_1001.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\103FTASK_IRX_1085.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\103FTASK_IRX_1138.png

Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\103FTASK_IRX_1196.png
Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\103FTASK_IRX_1268.png
Máscara no encontrada: C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\train/masks\103FTASK_IRX_1299.png
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Imágenes de entrenamiento: (360, 256, 256, 3), Máscaras de entrenamiento: (360, 256, 256, 1)
Imágenes de validación: (24, 256, 256, 3), Máscaras de validación: (24, 256, 256, 1)

4. Definir la Arquitectura U-Net

```
[3]: #Desfasaco
def build_unet(input_size=(256, 256, 3)):
    """
    Construye el modelo U-Net.
    """
    inputs = Input(input_size)

    # Encoder
    c1 = Conv2D(64, (3, 3), activation='relu', padding='same')(inputs)
    c1 = Conv2D(64, (3, 3), activation='relu', padding='same')(c1)
    p1 = MaxPooling2D((2, 2))(c1)

    c2 = Conv2D(128, (3, 3), activation='relu', padding='same')(p1)
    c2 = Conv2D(128, (3, 3), activation='relu', padding='same')(c2)
    p2 = MaxPooling2D((2, 2))(c2)

    c3 = Conv2D(256, (3, 3), activation='relu', padding='same')(p2)
    c3 = Conv2D(256, (3, 3), activation='relu', padding='same')(c3)
    p3 = MaxPooling2D((2, 2))(c3)

    c4 = Conv2D(512, (3, 3), activation='relu', padding='same')(p3)
    c4 = Conv2D(512, (3, 3), activation='relu', padding='same')(c4)
    p4 = MaxPooling2D((2, 2))(c4)

    # Bottleneck
    c5 = Conv2D(1024, (3, 3), activation='relu', padding='same')(p4)
    c5 = Conv2D(1024, (3, 3), activation='relu', padding='same')(c5)

    # Decoder
    u6 = UpSampling2D((2, 2))(c5)
    u6 = concatenate([u6, c4])
    c6 = Conv2D(512, (3, 3), activation='relu', padding='same')(u6)
    c6 = Conv2D(512, (3, 3), activation='relu', padding='same')(c6)

    u7 = UpSampling2D((2, 2))(c6)
    u7 = concatenate([u7, c3])
    c7 = Conv2D(256, (3, 3), activation='relu', padding='same')(u7)
    c7 = Conv2D(256, (3, 3), activation='relu', padding='same')(c7)

    u8 = UpSampling2D((2, 2))(c7)
    u8 = concatenate([u8, c2])
    c8 = Conv2D(128, (3, 3), activation='relu', padding='same')(u8)
    c8 = Conv2D(128, (3, 3), activation='relu', padding='same')(c8)

    u9 = UpSampling2D((2, 2))(c8)
```

```

u9 = concatenate([u9, c1])
c9 = Conv2D(64, (3, 3), activation='relu', padding='same')(u9)
c9 = Conv2D(64, (3, 3), activation='relu', padding='same')(c9)

outputs = Conv2D(1, (1, 1), activation='sigmoid')(c9)

model = Model(inputs, outputs)
return model

# Construir el modelo
model = build_unet()
model.compile(optimizer=Adam(learning_rate=LEARNING_RATE),
              loss='binary_crossentropy', metrics=['accuracy'])

# Resumen del modelo
model.summary()

```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 256, 256, 3	0	[]
)]		
conv2d (Conv2D)	(None, 256, 256, 64	1792	
['input_1[0][0]'])		
conv2d_1 (Conv2D)	(None, 256, 256, 64	36928	
['conv2d[0][0]'])		
max_pooling2d (MaxPooling2D)	(None, 128, 128, 64	0	
['conv2d_1[0][0]'])		
conv2d_2 (Conv2D)	(None, 128, 128, 12	73856	
['max_pooling2d[0][0]']	8)		
conv2d_3 (Conv2D)	(None, 128, 128, 12	147584	
['conv2d_2[0][0]']	8)		
max_pooling2d_1 (MaxPooling2D)	(None, 64, 64, 128)	0	

['conv2d_3[0][0]']

conv2d_4 (Conv2D) (None, 64, 64, 256) 295168
['max_pooling2d_1[0][0]']

conv2d_5 (Conv2D) (None, 64, 64, 256) 590080
['conv2d_4[0][0]']

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 256, 256, 3	0	[]
)]		
conv2d (Conv2D)	(None, 256, 256, 64	1792	
['input_1[0][0]'])		
conv2d_1 (Conv2D)	(None, 256, 256, 64	36928	
['conv2d[0][0]'])		
max_pooling2d (MaxPooling2D)	(None, 128, 128, 64	0	
['conv2d_1[0][0]'])		
conv2d_2 (Conv2D)	(None, 128, 128, 12	73856	
['max_pooling2d[0][0]']	8)		
conv2d_3 (Conv2D)	(None, 128, 128, 12	147584	
['conv2d_2[0][0]']	8)		
max_pooling2d_1 (MaxPooling2D)	(None, 64, 64, 128)	0	
['conv2d_3[0][0]']			
conv2d_4 (Conv2D)	(None, 64, 64, 256)	295168	
['max_pooling2d_1[0][0]']			
conv2d_5 (Conv2D)	(None, 64, 64, 256)	590080	
['conv2d_4[0][0]']			
max_pooling2d_2 (MaxPooling2D)	(None, 32, 32, 256)	0	
['conv2d_5[0][0]']			

```

conv2d_6 (Conv2D)          (None, 32, 32, 512) 1180160
['max_pooling2d_2[0][0]']

conv2d_7 (Conv2D)          (None, 32, 32, 512) 2359808
['conv2d_6[0][0]']

max_pooling2d_3 (MaxPooling2D) (None, 16, 16, 512) 0
['conv2d_7[0][0]']

conv2d_8 (Conv2D)          (None, 16, 16, 1024 4719616
['max_pooling2d_3[0][0]']
)

conv2d_9 (Conv2D)          (None, 16, 16, 1024 9438208
['conv2d_8[0][0]']
)

up_sampling2d (UpSampling2D) (None, 32, 32, 1024 0
['conv2d_9[0][0]']
)

concatenate (Concatenate)   (None, 32, 32, 1536 0
['up_sampling2d[0][0]',
)
'conv2d_7[0][0]']

conv2d_10 (Conv2D)          (None, 32, 32, 512) 7078400
['concatenate[0][0]']

conv2d_11 (Conv2D)          (None, 32, 32, 512) 2359808
['conv2d_10[0][0]']

up_sampling2d_1 (UpSampling2D) (None, 64, 64, 512) 0
['conv2d_11[0][0]']

concatenate_1 (Concatenate) (None, 64, 64, 768) 0
['up_sampling2d_1[0][0]',
'conv2d_5[0][0]']

conv2d_12 (Conv2D)          (None, 64, 64, 256) 1769728
['concatenate_1[0][0]']

conv2d_13 (Conv2D)          (None, 64, 64, 256) 590080
['conv2d_12[0][0]']

up_sampling2d_2 (UpSampling2D) (None, 128, 128, 25 0
['conv2d_13[0][0]']

```

6)

```

concatenate_2 (Concatenate)      (None, 128, 128, 38  0
['up_sampling2d_2[0][0]',
                                4)
'conv2d_3[0][0]']

conv2d_14 (Conv2D)                (None, 128, 128, 12  442496
['concatenate_2[0][0]']
                                8)

conv2d_15 (Conv2D)                (None, 128, 128, 12  147584
['conv2d_14[0][0]']
                                8)

up_sampling2d_3 (UpSampling2D)   (None, 256, 256, 12  0
['conv2d_15[0][0]']
                                8)

concatenate_3 (Concatenate)      (None, 256, 256, 19  0
['up_sampling2d_3[0][0]',
                                2)
'conv2d_1[0][0]']

conv2d_16 (Conv2D)                (None, 256, 256, 64  110656
['concatenate_3[0][0]']
                                )

conv2d_17 (Conv2D)                (None, 256, 256, 64  36928
['conv2d_16[0][0]']
                                )

conv2d_18 (Conv2D)                (None, 256, 256, 1)  65
['conv2d_17[0][0]']

```

```

=====
=====
Total params: 31,378,945
Trainable params: 31,378,945
Non-trainable params: 0
-----
-----

```

```

[4]: #Nuevo modelo con arquitectura U-Net:
from keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D,
    ↳concatenate, Dropout, BatchNormalization, Add, Multiply,
    ↳GlobalAveragePooling2D, Dense, Reshape
from keras.models import Model

```

```

from keras.optimizers import Adam
from keras import backend as K

# Función de pérdida personalizada (Dice Loss)
def perdida_dice(y_real, y_pred):
    numerador = 2 * K.sum(y_real * y_pred)
    denominador = K.sum(y_real + y_pred)
    return 1 - (numerador + 1) / (denominador + 1)

# Bloque Residual
def bloque_residual(x, filtros):
    salto = Conv2D(filtros, (1, 1), activation='relu', padding='same')(x)  #_
    ↪Asegurar que la forma coincida
    x = Conv2D(filtros, (3, 3), activation='relu', padding='same')(x)
    x = BatchNormalization()(x)
    x = Conv2D(filtros, (3, 3), activation='relu', padding='same')(x)
    x = BatchNormalization()(x)
    return Add()([x, salto])

# Bloque de Atención
def bloque_atencion(tensor_entrada, tensor_gating):
    forma = tensor_entrada.shape[-1]
    atencion = GlobalAveragePooling2D()(tensor_entrada)
    atencion = Dense(forma // 2, activation='relu')(atencion)
    atencion = Dense(forma, activation='sigmoid')(atencion)
    atencion = Reshape((1, 1, forma))(atencion)
    return Multiply()([tensor_entrada, atencion])

# Construcción del modelo U-Net
def construir_unet(tamano_entrada=(256, 256, 3)):
    entradas = Input(tamano_entrada)

    # Ajustar la entrada al número de filtros del primer bloque residual
    inicial = Conv2D(64, (1, 1), activation='relu', padding='same')(entradas)

    # Codificador (Encoder)
    c1 = bloque_residual(inicial, 64)
    p1 = MaxPooling2D((2, 2))(c1)

    c2 = bloque_residual(p1, 128)
    p2 = MaxPooling2D((2, 2))(c2)

    c3 = bloque_residual(p2, 256)
    p3 = MaxPooling2D((2, 2))(c3)

    c4 = bloque_residual(p3, 512)
    p4 = MaxPooling2D((2, 2))(c4)

```

```

# Bottleneck (cuello de botella)
c5 = Conv2D(1024, (3, 3), activation='relu', padding='same')(p4)
c5 = BatchNormalization()(c5)
c5 = Dropout(0.5)(c5) # Dropout para reducir sobreajuste
c5 = Conv2D(1024, (3, 3), activation='relu', padding='same')(c5)
c5 = BatchNormalization()(c5)

# Decodificador (Decoder)
u6 = UpSampling2D((2, 2))(c5) # Aumentar tamaño
u6 = bloque_atencion(u6, c4) # Aplicar atención
u6 = concatenate([u6, c4]) # Concatenar con la salida del codificador
c6 = bloque_residual(u6, 512)

u7 = UpSampling2D((2, 2))(c6)
u7 = bloque_atencion(u7, c3)
u7 = concatenate([u7, c3])
c7 = bloque_residual(u7, 256)

u8 = UpSampling2D((2, 2))(c7)
u8 = bloque_atencion(u8, c2)
u8 = concatenate([u8, c2])
c8 = bloque_residual(u8, 128)

u9 = UpSampling2D((2, 2))(c8)
u9 = bloque_atencion(u9, c1)
u9 = concatenate([u9, c1])
c9 = bloque_residual(u9, 64)

# Salida del modelo
salidas = Conv2D(1, (1, 1), activation='sigmoid')(c9)

modelo = Model(entradas, salidas)
return modelo

# Construcción y compilación del modelo
TASA_APRENDIZAJE = 0.001
modelo = construir_unet()
modelo.compile(optimizer=Adam(learning_rate=TASA_APRENDIZAJE),
               loss=perdida_dice, metrics=['accuracy'])

# Resumen del modelo
modelo.summary()

```

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_2 (InputLayer)	[(None, 256, 256, 3	0	[]
)]		
conv2d_19 (Conv2D)	(None, 256, 256, 64	256	
['input_2[0][0]'])		
conv2d_21 (Conv2D)	(None, 256, 256, 64	36928	
['conv2d_19[0][0]'])		
batch_normalization (BatchNorm	(None, 256, 256, 64	256	
['conv2d_21[0][0]'])		
alization)			
conv2d_22 (Conv2D)	(None, 256, 256, 64	36928	
['batch_normalization[0][0]']			

Layer (type)	Output Shape	Param #	Connected to
=====			
input_2 (InputLayer)	[(None, 256, 256, 3	0	[]
)]		
conv2d_19 (Conv2D)	(None, 256, 256, 64	256	
['input_2[0][0]'])		
conv2d_21 (Conv2D)	(None, 256, 256, 64	36928	
['conv2d_19[0][0]'])		
batch_normalization (BatchNorm	(None, 256, 256, 64	256	
['conv2d_21[0][0]'])		
alization)			
conv2d_22 (Conv2D)	(None, 256, 256, 64	36928	
['batch_normalization[0][0]'])		
batch_normalization_1 (BatchNo	(None, 256, 256, 64	256	
['conv2d_22[0][0]'])		
rmalization)			

```

conv2d_20 (Conv2D) (None, 256, 256, 64 4160
['conv2d_19[0][0]']
)

add (Add) (None, 256, 256, 64 0
['batch_normalization_1[0][0]',
)
'conv2d_20[0][0]']

max_pooling2d_4 (MaxPooling2D) (None, 128, 128, 64 0 ['add[0][0]']
)

conv2d_24 (Conv2D) (None, 128, 128, 12 73856
['max_pooling2d_4[0][0]']
8)

batch_normalization_2 (BatchNo (None, 128, 128, 12 512
['conv2d_24[0][0]']
rmalization) 8)

conv2d_25 (Conv2D) (None, 128, 128, 12 147584
['batch_normalization_2[0][0]']
8)

batch_normalization_3 (BatchNo (None, 128, 128, 12 512
['conv2d_25[0][0]']
rmalization) 8)

conv2d_23 (Conv2D) (None, 128, 128, 12 8320
['max_pooling2d_4[0][0]']
8)

add_1 (Add) (None, 128, 128, 12 0
['batch_normalization_3[0][0]',
8)
'conv2d_23[0][0]']

max_pooling2d_5 (MaxPooling2D) (None, 64, 64, 128) 0 ['add_1[0][0]']

conv2d_27 (Conv2D) (None, 64, 64, 256) 295168
['max_pooling2d_5[0][0]']

batch_normalization_4 (BatchNo (None, 64, 64, 256) 1024
['conv2d_27[0][0]']
rmalization)

conv2d_28 (Conv2D) (None, 64, 64, 256) 590080
['batch_normalization_4[0][0]']

```

```

batch_normalization_5 (BatchNormaliz (None, 64, 64, 256) 1024
['conv2d_28[0][0]']
rmalization)

conv2d_26 (Conv2D) (None, 64, 64, 256) 33024
['max_pooling2d_5[0][0]']

add_2 (Add) (None, 64, 64, 256) 0
['batch_normalization_5[0][0]',
'conv2d_26[0][0]']

max_pooling2d_6 (MaxPooling2D) (None, 32, 32, 256) 0 ['add_2[0][0]']

conv2d_30 (Conv2D) (None, 32, 32, 512) 1180160
['max_pooling2d_6[0][0]']

batch_normalization_6 (BatchNormaliz (None, 32, 32, 512) 2048
['conv2d_30[0][0]']
rmalization)

conv2d_31 (Conv2D) (None, 32, 32, 512) 2359808
['batch_normalization_6[0][0]']

batch_normalization_7 (BatchNormaliz (None, 32, 32, 512) 2048
['conv2d_31[0][0]']
rmalization)

conv2d_29 (Conv2D) (None, 32, 32, 512) 131584
['max_pooling2d_6[0][0]']

add_3 (Add) (None, 32, 32, 512) 0
['batch_normalization_7[0][0]',
'conv2d_29[0][0]']

max_pooling2d_7 (MaxPooling2D) (None, 16, 16, 512) 0 ['add_3[0][0]']

conv2d_32 (Conv2D) (None, 16, 16, 1024 4719616
['max_pooling2d_7[0][0]']
)

batch_normalization_8 (BatchNormaliz (None, 16, 16, 1024 4096
['conv2d_32[0][0]']
rmalization)
)

dropout (Dropout) (None, 16, 16, 1024 0
['batch_normalization_8[0][0]']
)

```



```

conv2d_33 (Conv2D)          (None, 16, 16, 1024) 9438208
['dropout[0][0]']
)

batch_normalization_9 (BatchNormaliz (None, 16, 16, 1024) 4096
['conv2d_33[0][0]']
rmalization)
)

up_sampling2d_4 (UpSampling2D) (None, 32, 32, 1024) 0
['batch_normalization_9[0][0]']
)

global_average_pooling2d (GlobalAverage (None, 1024) 0
['up_sampling2d_4[0][0]']
Pooling2D)

dense (Dense)              (None, 512) 524800
['global_average_pooling2d[0][0]']
]

dense_1 (Dense)            (None, 1024) 525312 ['dense[0][0]']

reshape (Reshape)         (None, 1, 1, 1024) 0
['dense_1[0][0]']

multiply (Multiply)       (None, 32, 32, 1024) 0
['up_sampling2d_4[0][0]',
)
'reshape[0][0]']

concatenate_4 (Concatenate) (None, 32, 32, 1536) 0
['multiply[0][0]',
)
'add_3[0][0]']

conv2d_35 (Conv2D)        (None, 32, 32, 512) 7078400
['concatenate_4[0][0]']

batch_normalization_10 (BatchNormaliz (None, 32, 32, 512) 2048
['conv2d_35[0][0]']
rmalization)

conv2d_36 (Conv2D)        (None, 32, 32, 512) 2359808
['batch_normalization_10[0][0]']

batch_normalization_11 (BatchNormaliz (None, 32, 32, 512) 2048
['conv2d_36[0][0]']
rmalization)

```

```

conv2d_34 (Conv2D)          (None, 32, 32, 512) 786944
['concatenate_4[0][0]']

add_4 (Add)                 (None, 32, 32, 512) 0
['batch_normalization_11[0][0]',
'conv2d_34[0][0]']

up_sampling2d_5 (UpSampling2D) (None, 64, 64, 512) 0      ['add_4[0][0]']

global_average_pooling2d_1 (GlobalAveragePooling2D) (None, 512) 0
['up_sampling2d_5[0][0]']

dense_2 (Dense)             (None, 256) 131328
['global_average_pooling2d_1[0][0]']

dense_3 (Dense)             (None, 512) 131584
['dense_2[0][0]']

reshape_1 (Reshape)         (None, 1, 1, 512) 0
['dense_3[0][0]']

multiply_1 (Multiply)       (None, 64, 64, 512) 0
['up_sampling2d_5[0][0]',
'reshape_1[0][0]']

concatenate_5 (Concatenate) (None, 64, 64, 768) 0      ['add_2[0][0]']
['multiply_1[0][0]',

conv2d_38 (Conv2D)          (None, 64, 64, 256) 1769728
['concatenate_5[0][0]']

batch_normalization_12 (Batch Normalization) (None, 64, 64, 256) 1024
['conv2d_38[0][0]']

conv2d_39 (Conv2D)          (None, 64, 64, 256) 590080
['batch_normalization_12[0][0]']

batch_normalization_13 (Batch Normalization) (None, 64, 64, 256) 1024
['conv2d_39[0][0]']

conv2d_37 (Conv2D)          (None, 64, 64, 256) 196864
['concatenate_5[0][0]']

```

```

add_5 (Add) (None, 64, 64, 256) 0
['batch_normalization_13[0][0]',
'conv2d_37[0][0]']

up_sampling2d_6 (UpSampling2D) (None, 128, 128, 25 0 ['add_5[0][0]']
6)

global_average_pooling2d_2 (GlobalAveragePooling2D) (None, 256) 0
['up_sampling2d_6[0][0]']

dense_4 (Dense) (None, 128) 32896
['global_average_pooling2d_2[0][0]']

dense_5 (Dense) (None, 256) 33024
['dense_4[0][0]']

reshape_2 (Reshape) (None, 1, 1, 256) 0
['dense_5[0][0]']

multiply_2 (Multiply) (None, 128, 128, 25 0
['up_sampling2d_6[0][0]',
6)
'reshape_2[0][0]']

concatenate_6 (Concatenate) (None, 128, 128, 38 0
['multiply_2[0][0]',
4)
'add_1[0][0]']

conv2d_41 (Conv2D) (None, 128, 128, 12 442496
['concatenate_6[0][0]']
8)

batch_normalization_14 (Batch Normalization) (None, 128, 128, 12 512
['conv2d_41[0][0]']
8)

conv2d_42 (Conv2D) (None, 128, 128, 12 147584
['batch_normalization_14[0][0]']
8)

batch_normalization_15 (Batch Normalization) (None, 128, 128, 12 512
['conv2d_42[0][0]']
8)

conv2d_40 (Conv2D) (None, 128, 128, 12 49280

```

```

['concatenate_6[0][0]']
8)

add_6 (Add) (None, 128, 128, 12 0
['batch_normalization_15[0][0]',
8)
'conv2d_40[0][0]']

up_sampling2d_7 (UpSampling2D) (None, 256, 256, 12 0 ['add_6[0][0]']
8)

global_average_pooling2d_3 (GlobalAveragePooling2D) (None, 128) 0
['up_sampling2d_7[0][0]']
globalAveragePooling2D)

dense_6 (Dense) (None, 64) 8256
['global_average_pooling2d_3[0][0]']

dense_7 (Dense) (None, 128) 8320
['dense_6[0][0]']

reshape_3 (Reshape) (None, 1, 1, 128) 0
['dense_7[0][0]']

multiply_3 (Multiply) (None, 256, 256, 12 0
['up_sampling2d_7[0][0]'],
8)
'reshape_3[0][0]']

concatenate_7 (Concatenate) (None, 256, 256, 19 0
['multiply_3[0][0]',
2) 'add[0][0]']

conv2d_44 (Conv2D) (None, 256, 256, 64 110656
['concatenate_7[0][0]']
)

batch_normalization_16 (Batch Normalization) (None, 256, 256, 64 256
['conv2d_44[0][0]']
ormalization)

conv2d_45 (Conv2D) (None, 256, 256, 64 36928
['batch_normalization_16[0][0]']
)

batch_normalization_17 (Batch Normalization) (None, 256, 256, 64 256
['conv2d_45[0][0]']
)

```

```

ormalization)
)

conv2d_43 (Conv2D) (None, 256, 256, 64 12352
['concatenate_7[0][0]']
)

add_7 (Add) (None, 256, 256, 64 0
['batch_normalization_17[0][0]',
)
'conv2d_43[0][0]']

conv2d_46 (Conv2D) (None, 256, 256, 1) 65 ['add_7[0][0]']

```

```

=====
=====

```

```

Total params: 34,055,937
Trainable params: 34,044,161
Non-trainable params: 11,776

```

```

-----
-----

```

1. Entrenar el Modelo

```

[5]: # Entrenar el modelo
      #importar ModelCheckpoint
      from tensorflow.keras.callbacks import ModelCheckpoint

      history = model.fit(
          train_images, train_masks,
          validation_data=(val_images, val_masks),
          batch_size=BATCH_SIZE,
          epochs=EPOCHS,
          #guardar el modelo
          callbacks=[ModelCheckpoint('model.h5', save_best_only=True)]
      )

```

```

Epoch 1/50
18/18 [=====] - 423s 23s/step - loss: 0.3989 -
accuracy: 0.9223 - val_loss: 0.1401 - val_accuracy: 0.9812
Epoch 2/50
18/18 [=====] - 417s 23s/step - loss: 0.1302 -
accuracy: 0.9787 - val_loss: 0.1065 - val_accuracy: 0.9812
Epoch 3/50
18/18 [=====] - 418s 23s/step - loss: 0.1109 -
accuracy: 0.9787 - val_loss: 0.1037 - val_accuracy: 0.9812
Epoch 4/50
18/18 [=====] - 417s 23s/step - loss: 0.1075 -
accuracy: 0.9787 - val_loss: 0.0977 - val_accuracy: 0.9812

```

Epoch 5/50
18/18 [=====] - 419s 23s/step - loss: 0.1052 - accuracy: 0.9787 - val_loss: 0.0990 - val_accuracy: 0.9812

Epoch 6/50
18/18 [=====] - 425s 24s/step - loss: 0.1055 - accuracy: 0.9787 - val_loss: 0.1068 - val_accuracy: 0.9812

Epoch 7/50
18/18 [=====] - 424s 24s/step - loss: 0.1066 - accuracy: 0.9787 - val_loss: 0.1032 - val_accuracy: 0.9812

Epoch 8/50
18/18 [=====] - 422s 23s/step - loss: 0.1037 - accuracy: 0.9787 - val_loss: 0.0943 - val_accuracy: 0.9812

Epoch 9/50
18/18 [=====] - 421s 23s/step - loss: 0.1026 - accuracy: 0.9787 - val_loss: 0.0966 - val_accuracy: 0.9812

Epoch 10/50
18/18 [=====] - 420s 23s/step - loss: 0.1028 - accuracy: 0.9787 - val_loss: 0.0932 - val_accuracy: 0.9812

Epoch 11/50
18/18 [=====] - 464s 26s/step - loss: 0.1016 - accuracy: 0.9787 - val_loss: 0.0976 - val_accuracy: 0.9812

Epoch 12/50
18/18 [=====] - 665s 37s/step - loss: 0.1028 - accuracy: 0.9787 - val_loss: 0.0927 - val_accuracy: 0.9812

Epoch 13/50
18/18 [=====] - 462s 25s/step - loss: 0.1016 - accuracy: 0.9787 - val_loss: 0.0986 - val_accuracy: 0.9812

Epoch 14/50
18/18 [=====] - 415s 23s/step - loss: 0.1049 - accuracy: 0.9787 - val_loss: 0.0957 - val_accuracy: 0.9812

Epoch 15/50
18/18 [=====] - 455s 25s/step - loss: 0.1049 - accuracy: 0.9787 - val_loss: 0.0975 - val_accuracy: 0.9812

Epoch 16/50
18/18 [=====] - 584s 32s/step - loss: 0.1023 - accuracy: 0.9787 - val_loss: 0.0968 - val_accuracy: 0.9812

Epoch 17/50
18/18 [=====] - 416s 23s/step - loss: 0.1025 - accuracy: 0.9787 - val_loss: 0.0989 - val_accuracy: 0.9812

Epoch 18/50
18/18 [=====] - 416s 23s/step - loss: 0.1034 - accuracy: 0.9787 - val_loss: 0.0977 - val_accuracy: 0.9812

Epoch 19/50
18/18 [=====] - 501s 28s/step - loss: 0.1017 - accuracy: 0.9787 - val_loss: 0.0942 - val_accuracy: 0.9812

Epoch 20/50
18/18 [=====] - 621s 34s/step - loss: 0.1022 - accuracy: 0.9787 - val_loss: 0.0931 - val_accuracy: 0.9812

Epoch 21/50
18/18 [=====] - 575s 32s/step - loss: 0.1019 -
accuracy: 0.9787 - val_loss: 0.0935 - val_accuracy: 0.9812
Epoch 22/50
18/18 [=====] - 567s 32s/step - loss: 0.1017 -
accuracy: 0.9787 - val_loss: 0.0924 - val_accuracy: 0.9812
Epoch 23/50
18/18 [=====] - 595s 33s/step - loss: 0.1018 -
accuracy: 0.9787 - val_loss: 0.0957 - val_accuracy: 0.9812
Epoch 24/50
18/18 [=====] - 421s 23s/step - loss: 0.1037 -
accuracy: 0.9787 - val_loss: 0.0935 - val_accuracy: 0.9812
Epoch 25/50
18/18 [=====] - 455s 25s/step - loss: 0.1010 -
accuracy: 0.9787 - val_loss: 0.0928 - val_accuracy: 0.9812
Epoch 26/50
18/18 [=====] - 461s 26s/step - loss: 0.1005 -
accuracy: 0.9787 - val_loss: 0.0952 - val_accuracy: 0.9812
Epoch 27/50
18/18 [=====] - 421s 23s/step - loss: 0.1015 -
accuracy: 0.9787 - val_loss: 0.0937 - val_accuracy: 0.9812
Epoch 28/50
18/18 [=====] - 425s 24s/step - loss: 0.1014 -
accuracy: 0.9787 - val_loss: 0.0927 - val_accuracy: 0.9812
Epoch 29/50
18/18 [=====] - 429s 24s/step - loss: 0.1011 -
accuracy: 0.9787 - val_loss: 0.0927 - val_accuracy: 0.9812
Epoch 30/50
18/18 [=====] - 432s 24s/step - loss: 0.1008 -
accuracy: 0.9787 - val_loss: 0.0921 - val_accuracy: 0.9812
Epoch 31/50
18/18 [=====] - 439s 24s/step - loss: 0.1003 -
accuracy: 0.9787 - val_loss: 0.0922 - val_accuracy: 0.9812
Epoch 32/50
18/18 [=====] - 435s 24s/step - loss: 0.1003 -
accuracy: 0.9787 - val_loss: 0.0920 - val_accuracy: 0.9812
Epoch 33/50
18/18 [=====] - 427s 24s/step - loss: 0.1011 -
accuracy: 0.9787 - val_loss: 0.0935 - val_accuracy: 0.9812
Epoch 34/50
18/18 [=====] - 428s 24s/step - loss: 0.1016 -
accuracy: 0.9787 - val_loss: 0.0924 - val_accuracy: 0.9812
Epoch 35/50
18/18 [=====] - 426s 24s/step - loss: 0.1013 -
accuracy: 0.9787 - val_loss: 0.0946 - val_accuracy: 0.9812
Epoch 36/50
18/18 [=====] - 432s 24s/step - loss: 0.1008 -
accuracy: 0.9787 - val_loss: 0.0928 - val_accuracy: 0.9812

```

Epoch 37/50
18/18 [=====] - 436s 24s/step - loss: 0.1006 -
accuracy: 0.9787 - val_loss: 0.0932 - val_accuracy: 0.9812
Epoch 38/50
18/18 [=====] - 427s 24s/step - loss: 0.1007 -
accuracy: 0.9787 - val_loss: 0.0922 - val_accuracy: 0.9812
Epoch 39/50
18/18 [=====] - 428s 24s/step - loss: 0.1001 -
accuracy: 0.9787 - val_loss: 0.0922 - val_accuracy: 0.9812
Epoch 40/50
18/18 [=====] - 426s 24s/step - loss: 0.1003 -
accuracy: 0.9787 - val_loss: 0.0922 - val_accuracy: 0.9812
Epoch 41/50
18/18 [=====] - 425s 24s/step - loss: 0.1007 -
accuracy: 0.9787 - val_loss: 0.0937 - val_accuracy: 0.9812
Epoch 42/50
18/18 [=====] - 430s 24s/step - loss: 0.1007 -
accuracy: 0.9787 - val_loss: 0.0923 - val_accuracy: 0.9812
Epoch 43/50
18/18 [=====] - 432s 24s/step - loss: 0.1016 -
accuracy: 0.9787 - val_loss: 0.0929 - val_accuracy: 0.9812
Epoch 44/50
18/18 [=====] - 424s 24s/step - loss: 0.1014 -
accuracy: 0.9787 - val_loss: 0.0928 - val_accuracy: 0.9812
Epoch 45/50
18/18 [=====] - 423s 23s/step - loss: 0.1005 -
accuracy: 0.9787 - val_loss: 0.0918 - val_accuracy: 0.9812
Epoch 46/50
18/18 [=====] - 452s 25s/step - loss: 0.0999 -
accuracy: 0.9787 - val_loss: 0.0918 - val_accuracy: 0.9812
Epoch 47/50
18/18 [=====] - 429s 24s/step - loss: 0.1001 -
accuracy: 0.9787 - val_loss: 0.0928 - val_accuracy: 0.9812
Epoch 48/50
18/18 [=====] - 413s 23s/step - loss: 0.1002 -
accuracy: 0.9787 - val_loss: 0.0918 - val_accuracy: 0.9812
Epoch 49/50
18/18 [=====] - 410s 23s/step - loss: 0.0999 -
accuracy: 0.9787 - val_loss: 0.0921 - val_accuracy: 0.9812
Epoch 50/50
18/18 [=====] - 457s 26s/step - loss: 0.0993 -
accuracy: 0.9787 - val_loss: 0.0912 - val_accuracy: 0.9812

```

```

[6]: # guardar modelo
model.save("modelo_unet.h5")

# guardar el modelo entrenado

```



```

model.save("modelo_unet.h5")

# guardar el historial de entrenamiento
import pickle
with open("history_unet.pkl", "wb") as f:
    pickle.dump(history.history, f)

```

```

[7]: #Validar el modelo

# #Primero matriz de confusión
# from sklearn.metrics import confusion_matrix
# #cargar modelo
# import matplotlib.pyplot as plt
# from keras.models import load_model
# import time
# import numpy as np
# model = load_model("modelo_unet.h5")

# model.summary()

# # cargar el historial de entrenamiento
# import pickle
# with open("history_unet.pkl", "rb") as f:
#     history = pickle.load(f)

# # Predecir máscaras en el conjunto de validación
# predicciones = model.predict(val_images)

# # Convertir las máscaras a valores binarios
# predicciones_binarias = (predicciones > 0.5).astype(np.uint8)

# # Calcular la matriz de confusión
# confusion_matrices = []

# for i in range(len(val_masks)):
#     y_true = val_masks[i].flatten()
#     y_pred = predicciones_binarias[i].flatten()
#     confusion_matrices.append(confusion_matrix(y_true, y_pred))

# # Calcular la matriz de confusión promedio
# confusion_matrix_avg = np.mean(confusion_matrices, axis=0)

# # Mostrar la matriz de confusión promedio
# print("Matriz de confusión promedio:")
# print(confusion_matrix_avg)

# # Segundo métricas de evaluación

```

```

# from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score

# # Calcular métricas de evaluación
# accuracy = accuracy_score(y_true, y_pred)
# precision = precision_score(y_true, y_pred)
# recall = recall_score(y_true, y_pred)
# f1 = f1_score(y_true, y_pred)

# # Mostrar métricas de evaluación
# print(f"Precisión: {precision:.4f}")
# print(f"Recall: {recall:.4f}")
# print(f"F1 Score: {f1:.4f}")

# # Tercero visualización de resultados
# import matplotlib.pyplot as plt

# # Seleccionar una muestra aleatoria de índices
# indices = np.random.choice(range(len(val_images)), size=5, replace=False)

# # Mostrar las imágenes y las máscaras correspondientes
# plt.figure(figsize=(15, 10))
# for i, idx in enumerate(indices, 1):
#     # Imagen original
#     plt.subplot(5, 3, 3 * i - 2)
#     plt.imshow(val_images[idx])
#     plt.axis("off")
#     plt.title("Imagen")

#     # Máscara real
#     plt.subplot(5, 3, 3 * i - 1)
#     plt.imshow(val_masks[idx].squeeze(), cmap="gray")
#     plt.axis("off")
#     plt.title("Máscara Real")

#     # Máscara predicha
#     plt.subplot(5, 3, 3 * i)
#     plt.imshow(predicciones_binarias[idx].squeeze(), cmap="gray")
#     plt.axis("off")
#     plt.title("Máscara Predicha")

# plt.tight_layout()
# plt.show()

# Cuarto visualización de métricas
# import matplotlib.pyplot as plt

```

```

# Mostrar la pérdida y la precisión en el entrenamiento y la validación
# plt.figure(figsize=(12, 4))

# plt.subplot(1, 2, 1)
# plt.plot(history.history['loss'], label='Pérdida de entrenamiento')
# plt.plot(history.history['val_loss'], label='Pérdida de validación')
# plt.xlabel('Épocas')

# plt.subplot(1, 2, 2)
# plt.plot(history.history['accuracy'], label='Precisión de entrenamiento')
# plt.plot(history.history['val_accuracy'], label='Precisión de validación')
# plt.xlabel('Épocas')

# plt.legend()
# plt.tight_layout()
# plt.show()

# # Quinto visualización de máscaras
# import matplotlib.pyplot as plt

# # Seleccionar una muestra aleatoria de índices
# indices = np.random.choice(range(len(val_images)), size=5, replace=False)

# # Mostrar las imágenes y las máscaras correspondientes
# plt.figure(figsize=(15, 10))

# # for i, idx in enumerate(indices, 1):
# #     # Imagen original
# #     plt.subplot(5, 3, 3 * i - 2)
# #     plt.imshow(val_images[idx])
# #     plt.axis("off")
# #     plt.title("Imagen")

# #     # Máscara real
# #     plt.subplot(5, 3, 3 * i - 1)
# #     plt.imshow(val_masks[idx].squeeze(), cmap="gray")
# #     plt.axis("off")
# #     plt.title("Máscara Real")

# #     # Máscara predicha
# #     plt.subplot(5, 3, 3 * i)
# #     plt.imshow(predicciones_binarias[idx].squeeze(), cmap="gray")
# #     plt.axis("off")
# #     plt.title("Máscara Predicha")

# plt.tight_layout()
# plt.show()

```

```
[ ]: # Entrenar el modelo
# history_avanzado = modelo.fit(
#     train_images, train_masks,
#     validation_data=(val_images, val_masks),
#     batch_size=BATCH_SIZE,
#     epochs=EPOCHS
# )
```

6. Evaluar el Modelo

```
[4]: #cargar modelo
import matplotlib.pyplot as plt
from keras.models import load_model
import time
import numpy as np
model = load_model("modelo_unet.h5")

model.summary()

# cargar el historial de entrenamiento
import pickle
with open("history_unet.pkl", "rb") as f:
    history = pickle.load(f)

# Evaluar en el conjunto de validación
inicio = time.time()
predictions = model.predict(val_images)
fin = time.time()

# Calcular el tiempo de inferencia
tiempo_total_por_frame = fin - inicio

print(f"Tiempo de inferencia: {fin - inicio:.4f} segundos")
fps = len(val_images) / (fin - inicio)
print(f"FPS: {fps:.2f} frames por segundo")

# Mostrar predicciones
def plot_predictions(images, masks, preds, num=5):
    """
    Muestra imágenes, máscaras reales y predicciones.
    """
    for i in range(num):
        plt.figure(figsize=(12, 4))
        plt.subplot(1, 3, 1)
        plt.title("Imagen")
```

```

plt.imshow(images[i])
plt.axis('off')

plt.subplot(1, 3, 2)
plt.title("Máscara Real")
plt.imshow(masks[i].squeeze(), cmap='gray')
plt.axis('off')

plt.subplot(1, 3, 3)
plt.title("Predicción")
plt.imshow(preds[i].squeeze(), cmap='gray')
plt.axis('off')

plt.show()

plot_predictions(val_images, val_masks, predictions)

#calculamos las métricas
from sklearn.metrics import accuracy_score, precision_score

# Convertir las máscaras a valores binarios
binary_preds = (predictions > 0.5).astype(np.uint8)

# Calcular métricas
accuracy = accuracy_score(val_masks.flatten(), binary_preds.flatten())

# Mostrar métricas
print(f"Precisión: {accuracy:.4f}")

# Visualizar las métricas
import matplotlib.pyplot as plt

# Mostrar la pérdida y la precisión en el entrenamiento y la validación
plt.figure(figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(history['loss'], label='Pérdida de entrenamiento')
plt.plot(history['val_loss'], label='Pérdida de validación')
plt.xlabel('Épocas')
plt.ylabel('Pérdida')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history['accuracy'], label='Precisión de entrenamiento')
plt.plot(history['val_accuracy'], label='Precisión de validación')
plt.xlabel('Épocas')
plt.ylabel('Precisión')

```

```

plt.legend()

plt.tight_layout()
plt.show()

#matriz de confusión
from sklearn.metrics import confusion_matrix
# Calcular la matriz de confusión
predicciones_binarias = (predictions > 0.5).astype(np.uint8)

confusion_matrices = []

for i in range(len(val_masks)):
    y_true = val_masks[i].flatten()
    y_pred = predicciones_binarias[i].flatten()
    confusion_matrices.append(confusion_matrix(y_true, y_pred))

# Calcular la matriz de confusión promedio
confusion_matrix_avg = np.mean(confusion_matrices, axis=0)

# Mostrar la matriz de confusión promedio
print("Matriz de confusión promedio:")
print(confusion_matrix_avg)

```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 256, 256, 3 0)]		[]
conv2d (Conv2D) ['input_1[0][0]']	(None, 256, 256, 64 1792)		
conv2d_1 (Conv2D) ['conv2d[0][0]']	(None, 256, 256, 64 36928)		
max_pooling2d (MaxPooling2D) ['conv2d_1[0][0]']	(None, 128, 128, 64 0)		
conv2d_2 (Conv2D) ['max_pooling2d[0][0]']	(None, 128, 128, 12 73856)		

```

8)

conv2d_3 (Conv2D)          (None, 128, 128, 12) 147584
['conv2d_2[0][0]']

8)

max_pooling2d_1 (MaxPooling2D) (None, 64, 64, 128) 0
['conv2d_3[0][0]']

conv2d_4 (Conv2D)          (None, 64, 64, 256) 295168
['max_pooling2d_1[0][0]']

conv2d_5 (Conv2D)          (None, 64, 64, 256) 590080
['conv2d_4[0][0]']

max_pooling2d_2 (MaxPooling2D) (None, 32, 32, 256) 0
['conv2d_5[0][0]']

conv2d_6 (Conv2D)          (None, 32, 32, 512) 1180160
['max_pooling2d_2[0][0]']

conv2d_7 (Conv2D)          (None, 32, 32, 512) 2359808
['conv2d_6[0][0]']

max_pooling2d_3 (MaxPooling2D) (None, 16, 16, 512) 0
['conv2d_7[0][0]']

conv2d_8 (Conv2D)          (None, 16, 16, 1024) 4719616
['max_pooling2d_3[0][0]']
)

conv2d_9 (Conv2D)          (None, 16, 16, 1024) 9438208
['conv2d_8[0][0]']
)

up_sampling2d (UpSampling2D) (None, 32, 32, 1024) 0
['conv2d_9[0][0]']
)

concatenate (Concatenate)   (None, 32, 32, 1536) 0
['up_sampling2d[0][0]',
)
'conv2d_7[0][0]']

conv2d_10 (Conv2D)          (None, 32, 32, 512) 7078400
['concatenate[0][0]']

conv2d_11 (Conv2D)          (None, 32, 32, 512) 2359808

```

```

['conv2d_10[0][0]']

up_sampling2d_1 (UpSampling2D) (None, 64, 64, 512) 0
['conv2d_11[0][0]']

concatenate_1 (Concatenate) (None, 64, 64, 768) 0
['up_sampling2d_1[0][0]',
'conv2d_5[0][0]']

conv2d_12 (Conv2D) (None, 64, 64, 256) 1769728
['concatenate_1[0][0]']

conv2d_13 (Conv2D) (None, 64, 64, 256) 590080
['conv2d_12[0][0]']

up_sampling2d_2 (UpSampling2D) (None, 128, 128, 25 0
['conv2d_13[0][0]']

6)

concatenate_2 (Concatenate) (None, 128, 128, 38 0
['up_sampling2d_2[0][0]',

4)
'conv2d_3[0][0]']

conv2d_14 (Conv2D) (None, 128, 128, 12 442496
['concatenate_2[0][0]']

8)

conv2d_15 (Conv2D) (None, 128, 128, 12 147584
['conv2d_14[0][0]']

8)

up_sampling2d_3 (UpSampling2D) (None, 256, 256, 12 0
['conv2d_15[0][0]']

8)

concatenate_3 (Concatenate) (None, 256, 256, 19 0
['up_sampling2d_3[0][0]',

2)
'conv2d_1[0][0]']

conv2d_16 (Conv2D) (None, 256, 256, 64 110656
['concatenate_3[0][0]']

)

conv2d_17 (Conv2D) (None, 256, 256, 64 36928
['conv2d_16[0][0]']

)

```



```
conv2d_18 (Conv2D)          (None, 256, 256, 1) 65  
['conv2d_17[0][0]']
```

```
=====
```

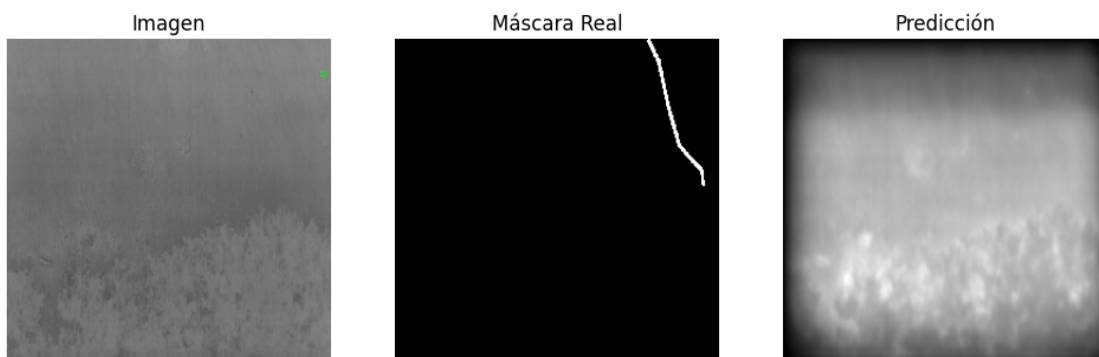
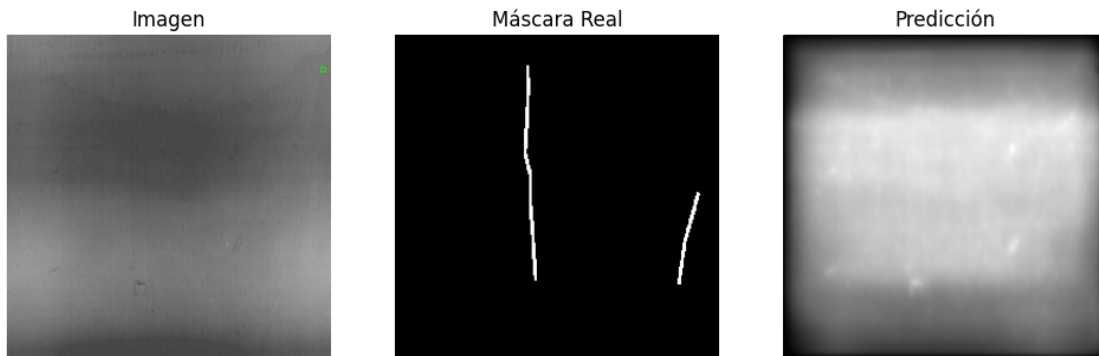
```
=====
```

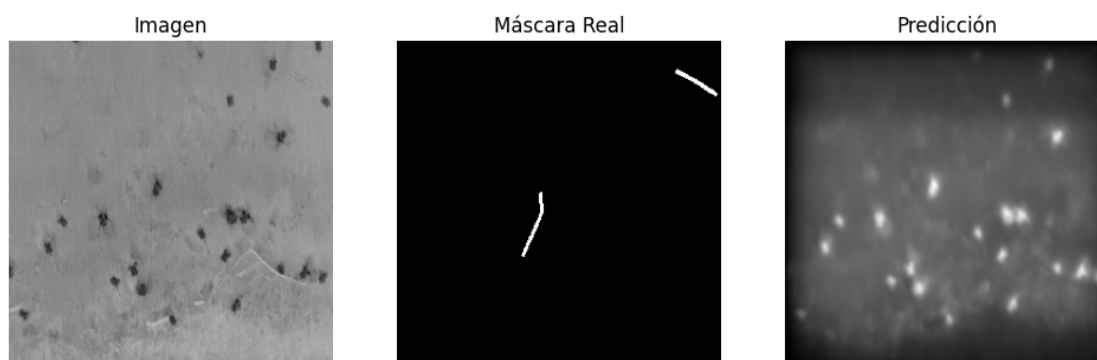
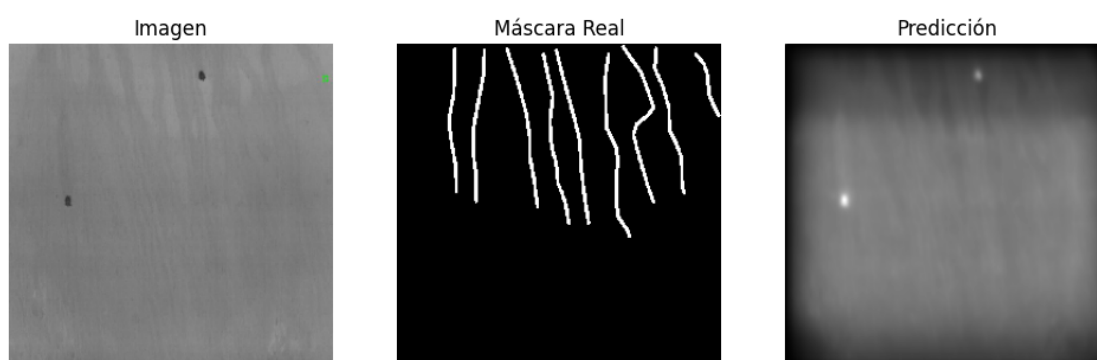
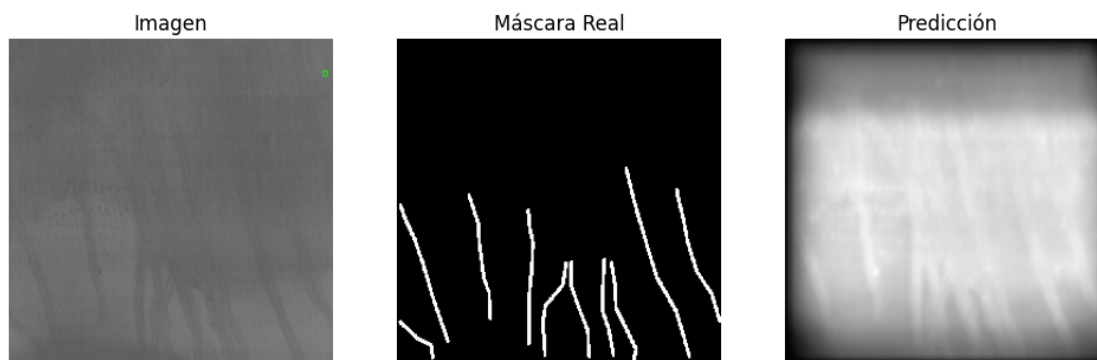
Total params: 31,378,945
Trainable params: 31,378,945
Non-trainable params: 0

```
-----
```

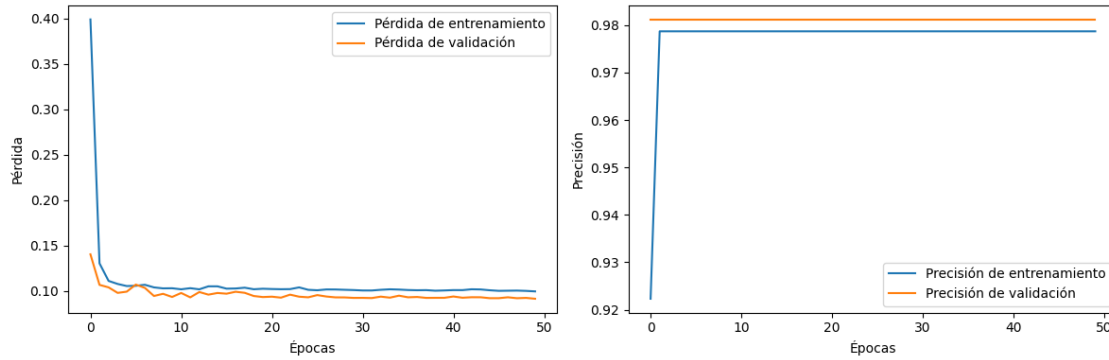
```
-----
```

1/1 [=====] - 13s 13s/step
Tiempo de inferencia: 12.6713 segundos
FPS: 1.89 frames por segundo





Precisión: 0.9812



Matriz de confusión promedio:

```
[[64304.08333333  0.      ]
 [ 1231.91666667  0.      ]]
```

```
[12]: from tensorflow.keras.preprocessing.image import load_img, img_to_array
import numpy as np
import matplotlib.pyplot as plt

# Cargar la imagen
image_path = r"C:\Users\crome\Desktop\TFM\Dades\osaconservation\Dataset_Organizado\images\108FTASK_MAX_2379.
↪JPG"
image = load_img(image_path, target_size=(256, 256)) # Asumiendo que el modelo
↪usa imágenes de 256x256

# Convertir a array y normalizar si es necesario
image_array = img_to_array(image) / 255.0 # Normalización a [0,1] si el modelo
↪fue entrenado así

# Añadir una dimensión extra para el batch
image_batch = np.expand_dims(image_array, axis=0)

# Realizar la predicción
prediction = model.predict(image_batch)

# Visualizar la imagen y su predicción
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
plt.title("Imagen de Entrada")
plt.imshow(image_array)
plt.axis('off')

plt.subplot(1, 2, 2)
plt.title("Predicción del Modelo")
```

```
plt.imshow(prediction.squeeze(), cmap='gray') # Mostrar la predicción
plt.axis('off')

plt.show()
```

1/1 [=====] - 0s 424ms/step

Imagen de Entrada



Predicción del Modelo

