

Министерство образования Республики Беларусь

Учреждение образования

«Брестский Государственный технический университет»

Кафедра ИИТ

Лабораторная работа №3

По дисциплине «ОИвИС»

Тема: “**Обучение детекторов объектов**”

Выполнил:

Студент 4 курса

Группы ИИ-23

Копач А. В.

Проверила:

Андренко К.В.

Брест 2025

Цель: осуществлять обучение нейросетевого детектора для решения задачи обнаружения заданных объектов

Вариант 7.

B-T	Детектор	Датасет
7	YOLOv11n	Люди: https://universe.roboflow.com/leo-ueno/people-detection-o4rdr/dataset/10

Код программы:

```
import os
import sys
import shutil
import numpy as np
import pandas as pd
import yaml
import cv2
import matplotlib.pyplot as plt
from ultralytics import YOLO
import torch
import glob
from datetime import datetime
import json
import gc

class CSVToYOLOConverter:
    def __init__(self, base_path):
        self.base_path = base_path
        self.class_name = "person"
        self.class_id = 0

    def convert_annotations(self, csv_path, images_dir, output_dir):
        """ Конвертация CSV аннотаций в YOLO формат """
        labels_dir = os.path.join(output_dir, 'labels')
        images_output_dir = os.path.join(output_dir, 'images')

        os.makedirs(labels_dir, exist_ok=True)
        os.makedirs(images_output_dir, exist_ok=True)

        print(f"Чтение CSV: {csv_path}")

        try:
            df = pd.read_csv(csv_path)
            print(f"Загружено {len(df)} аннотаций из {csv_path}")
        except Exception as e:
            print(f"Ошибка чтения CSV: {e}")
            return 0

        return df
```

```

processed_files = 0
missing_images = 0

for filename, group in df.groupby('filename'):
    src_image_path = os.path.join(images_dir, filename)

    if not os.path.exists(src_image_path):
        missing_images += 1
        continue

    dst_image_path = os.path.join(images_output_dir, filename)
    shutil.copy2(src_image_path, dst_image_path)

    label_filename = os.path.splitext(filename)[0] + '.txt'
    label_path = os.path.join(labels_dir, label_filename)

    try:
        with open(label_path, 'w') as f:
            for _, row in group.iterrows():
                x_center, y_center, width, height = self.convert_to_yolo_format(
                    row['xmin'], row['ymin'], row['xmax'], row['ymax'],
                    row['width'], row['height']
                )
                f.write(f'{self.class_id} {x_center:.6f} {y_center:.6f} {width:.6f} {height:.6f}\n')

        processed_files += 1
    except Exception as e:
        print(f'Ошибка создания аннотации для {filename}: {e}')

if missing_images > 0:
    print(f'Всего отсутствующих изображений: {missing_images}')

print(f'Обработано {processed_files} файлов для {output_dir}')
return processed_files

def convert_to_yolo_format(self, xmin, ymin, xmax, ymax, img_width, img_height):
    """Конвертация координат в YOLO формат"""
    x_center = (xmin + xmax) / 2 / img_width
    y_center = (ymin + ymax) / 2 / img_height
    width = (xmax - xmin) / img_width
    height = (ymax - ymin) / img_height

    x_center = max(0, min(1, x_center))
    y_center = max(0, min(1, y_center))
    width = max(0, min(1, width))
    height = max(0, min(1, height))

    return x_center, y_center, width, height

```

```
class PeopleDetectorTrainer:
    def __init__(self, dataset_path='lab3'):
        self.script_dir = os.path.dirname(os.path.abspath(__file__))
        self.dataset_path = os.path.join(self.script_dir, dataset_path)
        self.yolo_dataset_path = os.path.join(self.script_dir, 'yolo_dataset')
        self.model = None
        self.results = None
        self.timestamp = datetime.now().strftime("%Y%m%d_%H%M%S")
        self.output_dir = os.path.join(self.script_dir, 'training_results', self.timestamp)

    def setup_environment(self):
        """Проверка и настройка окружения"""
        print("Настройка окружения...")
        torch.backends.cudnn.benchmark = True
        if torch.cuda.is_available():
            gpu_name = torch.cuda.get_device_name(0)
            gpu_memory = torch.cuda.get_device_properties(0).total_memory / 1024 ** 3
            print(f"GPU доступен: {gpu_name} ({gpu_memory:.1f} GB)")
            self.device = 0
            # Очистка памяти
            torch.cuda.empty_cache()
            # Установка лимита памяти
            torch.cuda.set_per_process_memory_fraction(0.7) # Используем только 70% памяти
        else:
            print("GPU не доступен, используется CPU")
            self.device = None

        os.makedirs(self.output_dir, exist_ok=True)
        os.makedirs(os.path.join(self.output_dir, 'detections'), exist_ok=True)
        os.makedirs(os.path.join(self.output_dir, 'graphs'), exist_ok=True)
        os.makedirs(os.path.join(self.output_dir, 'models'), exist_ok=True)

    def prepare_dataset(self):
        """Подготовка датасета в YOLO формате"""
        print("Подготовка датасета в YOLO формате...")
        if not os.path.exists(self.dataset_path):
            print(f"Директория датасета не найдена: {self.dataset_path}")
            return False

        converter = CSVToYOLOConverter(self.dataset_path)

        yolo_dirs = ['train', 'val', 'test']
        for dir_name in yolo_dirs:
            os.makedirs(os.path.join(self.yolo_dataset_path, dir_name, 'images'), exist_ok=True)
```

```
os.makedirs(os.path.join(self.yolo_dataset_path, dir_name, 'labels'), exist_ok=True)

train_csv = os.path.join(self.dataset_path, 'train', '_annotations.csv')
train_images = os.path.join(self.dataset_path, 'train')
print("Конвертация тренировочных данных...")
converter.convert_annotations(train_csv, train_images,
                               os.path.join(self.yolo_dataset_path, 'train'))

val_csv = os.path.join(self.dataset_path, 'valid', '_annotations.csv')
val_images = os.path.join(self.dataset_path, 'valid')
print("Конвертация валидационных данных...")
converter.convert_annotations(val_csv, val_images,
                               os.path.join(self.yolo_dataset_path, 'val'))

test_csv = os.path.join(self.dataset_path, 'test', '_annotations.csv')
test_images = os.path.join(self.dataset_path, 'test')
print("Конвертация тестовых данных...")
converter.convert_annotations(test_csv, test_images,
                               os.path.join(self.yolo_dataset_path, 'test'))

self.create_data_yaml()
print("Датасет подготовлен в YOLO формате")
return True

def create_data_yaml(self):
    """Создание data.yaml файла"""
    data = {
        'path': os.path.abspath(self.yolo_dataset_path),
        'train': 'train/images',
        'val': 'val/images',
        'test': 'test/images',
        'nc': 1,
        'names': ['person']
    }

    yaml_path = os.path.join(self.yolo_dataset_path, 'data.yaml')
    with open(yaml_path, 'w') as f:
        yaml.dump(data, f, default_flow_style=False)

    print(f'data.yaml создан: {yaml_path}')

def analyze_dataset(self):
    """Анализ датасета"""
    print("Анализ датасета:")

    splits = ['train', 'val', 'test']
    for split in splits:
        images_dir = os.path.join(self.yolo_dataset_path, split, 'images')
```

```

labels_dir = os.path.join(self.yolo_dataset_path, split, 'labels')

if not os.path.exists(images_dir):
    print(f'Директория {images_dir} не найдена')
    continue

num_images = len([f for f in os.listdir(images_dir) if f.endswith('.jpg', '.png', '.jpeg')])]
num_labels = len([f for f in os.listdir(labels_dir) if f.endswith('.txt')]) if os.path.exists(
    labels_dir) else 0

total_objects = 0
if os.path.exists(labels_dir):
    for label_file in os.listdir(labels_dir):
        if label_file.endswith('.txt'):
            try:
                with open(os.path.join(labels_dir, label_file), 'r') as f:
                    total_objects += len(f.readlines())
            except:
                pass

print(f" {split.capitalize()}: {num_images} изображений, {num_labels} аннотаций, {total_objects} объектов")

def setup_model(self):
    """Инициализация модели YOLOv1In"""
    print("Инициализация модели YOLOv1In...")

    try:
        self.model = YOLO('yolo11n.pt')
        print("Модель успешно загружена")
        return True
    except Exception as e:
        print(f"Ошибка загрузки модели: {e}")
        return False

def train_model(self, epochs=50, imgsz=640, batch=12, patience=15):
    """
    Обучение модели с безопасными параметрами
    """
    print("Запуск обучения модели...")

    if self.model is None:
        print("Модель не инициализирована")
        return False

    try:
        # Безопасные параметры для избежания "Killed"
        self.results = self.model.train(

```

```
data=os.path.join(self.yolo_dataset_path, 'data.yaml'),
epochs=epochs,
imgsz=imgsz,
batch=batch, # Уменьшили batch size
device=self.device,
workers=4, # Уменьшили workers
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,
patience=patience,
save=True,
save_period=10,
cache=False, # Отключили кэш для экономии памяти
name=f'people_detection_{self.timestamp}',
exist_ok=True,
amp=True, # Mixed precision обязательно
# Минимальные аугментации
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,
mosaic=0.0, # Отключили mosaic для экономии памяти
mixup=0.0, # Отключили mixup
copy_paste=0.0
)
```

```
print("Обучение успешно завершено!")
return True
```

```
except Exception as e:
```

```
    print(f"Ошибка обучения: {e}")
    return False
```

```
def evaluate_model(self):
```

```
    """Оценка модели на тестовой выборке"""

```

```
print("Оценка модели на тестовой выборке...")

if self.model is None:
    print("Модель не инициализирована")
    return None

try:
    best_model_path = os.path.join(self.script_dir, 'runs', 'detect',
                                   f'people_detection_{self.timestamp}', 'weights', 'best.pt')
    if os.path.exists(best_model_path):
        self.model = YOLO(best_model_path)
        print("Загружена лучшая модель после обучения")

    # Копируем лучшую модель в папку результатов
    shutil.copy2(best_model_path, os.path.join(self.output_dir, 'models', 'best.pt'))

# Очистка памяти перед валидацией
if torch.cuda.is_available():
    torch.cuda.empty_cache()

metrics = self.model.val(
    data=os.path.join(self.yolo_dataset_path, 'data.yaml'),
    split='test',
    device=self.device,
    batch=8, # Уменьшили batch для валидации
    workers=2
)

print("Результаты оценки:")
print(f" mAP50-95: {metrics.box.map:.4f}")
print(f" mAP50: {metrics.box.map50:.4f}")
print(f" mAP75: {metrics.box.map75:.4f}")
print(f" Precision: {metrics.box.precision:.4f}")
print(f" Recall: {metrics.box.recall:.4f}")

self.save_metrics_to_file(metrics)

return metrics

except Exception as e:
    print(f"Ошибка оценки: {e}")
    return None

def save_metrics_to_file(self, metrics):
    """Сохранение метрик в файл"""
    metrics_data = {
        'mAP50-95': float(metrics.box.map),
        'mAP50': float(metrics.box.map50),
```

```

'mAP75': float(metrics.box.map75),
'precision': float(metrics.box.precision),
'recall': float(metrics.box.recall),
'training_time': self.timestamp,
'model': 'YOLOv11n',
'dataset': 'People Detection',
'epochs': 50,
'image_size': 640,
'batch_size': 12
}

metrics_path = os.path.join(self.output_dir, 'metrics.json')
with open(metrics_path, 'w', encoding='utf-8') as f:
    json.dump(metrics_data, f, indent=4, ensure_ascii=False)

print(f"Метрики сохранены: {metrics_path}")

def save_training_plots(self):
    """Сохранение графиков обучения"""
    if self.results is None:
        print("Нет результатов обучения для построения графиков")
        return

    try:
        results_dict = self.results.results_dict

        if results_dict:
            # Графики потерь
            plt.figure(figsize=(15, 10))

            metrics = ['train/box_loss', 'train/cls_loss', 'train/dfl_loss',
                       'val/box_loss', 'val/cls_loss', 'val/dfl_loss']

            for i, metric in enumerate(metrics, 1):
                plt.subplot(2, 3, i)
                if metric in results_dict:
                    plt.plot(results_dict[metric])
                    plt.title(metric.replace('/', ' ').title())
                    plt.xlabel('Эпоха')
                    plt.ylabel('Потери')
                    plt.grid(True, alpha=0.3)

            plt.tight_layout()
            plt.savefig(os.path.join(self.output_dir, 'graphs', 'training_losses.png'),
                        dpi=300, bbox_inches='tight', facecolor='white')
            plt.close()

            # Графики метрик

```

```

plt.figure(figsize=(15, 5))

metrics_to_plot = [
    ('metrics/precision(B)', 'Precision'),
    ('metrics/recall(B)', 'Recall'),
    ('metrics/mAP50(B)', 'mAP50'),
    ('metrics/mAP50-95(B)', 'mAP50-95')
]

for i, (metric, title) in enumerate(metrics_to_plot, 1):
    plt.subplot(1, 4, i)
    if metric in results_dict:
        plt.plot(results_dict[metric])
        plt.title(title)
        plt.xlabel('Эпоха')
        plt.ylabel('Значение')
        plt.grid(True, alpha=0.3)

plt.tight_layout()
plt.savefig(os.path.join(self.output_dir, 'graphs', 'training_metrics.png'),
           dpi=300, bbox_inches='tight', facecolor='white')
plt.close()

print(f"Графики обучения сохранены: {os.path.join(self.output_dir, 'graphs')}")
```

```

else:
    print("Нет данных истории обучения")
```

```
except Exception as e:
```

```
    print(f"Ошибка сохранения графиков: {e}")
```

```
class DetectionVisualizer:
```

```
    def __init__(self, model_path, output_dir):
        self.model = YOLO(model_path)
        self.output_dir = output_dir
        os.makedirs(self.output_dir, exist_ok=True)
```

```
    def test_on_dataset(self, dataset_path, num_images=5):
        """Тестирование на изображениях из датасета"""
        test_images_dir = os.path.join(dataset_path, 'test', 'images')
```

```
        if not os.path.exists(test_images_dir):
            print(f"Директория тестовых изображений не найдена: {test_images_dir}")
            return
```

```
        image_files = glob.glob(os.path.join(test_images_dir, "*.jpg"))
        image_files = image_files[:num_images]
```

```
print(f"Тестирование на {len(image_files)} изображениях из датасета...")  
  
for i, image_path in enumerate(image_files):  
    try:  
        # Очистка памяти перед каждой обработкой  
        if torch.cuda.is_available():  
            torch.cuda.empty_cache()  
  
        results = self.model.predict(image_path, conf=0.25, imgsz=640)  
        plotted_image = results[0].plot()  
        output_path = os.path.join(self.output_dir, f'dataset_test_{i + 1:03d}.jpg')  
        cv2.imwrite(output_path, plotted_image)  
  
        num_detections = len(results[0].boxes) if results[0].boxes else 0  
        print(f"Изображение {i + 1}/{len(image_files)}: обнаружено {num_detections} человек")  
  
    except Exception as e:  
        print(f"Ошибка обработки {image_path}: {e}")  
  
  
def main():  
    """Основная функция"""  
    print("ЗАПУСК ПРОГРАММЫ ОБНАРУЖЕНИЯ ЛЮДЕЙ")  
    print("=" * 50)  
  
    # Очистка памяти перед началом  
    if torch.cuda.is_available():  
        torch.cuda.empty_cache()  
  
    # 1. Инициализация  
    detector = PeopleDetectorTrainer(dataset_path='lab3')  
    detector.setup_environment()  
  
    # 2. Подготовка датасета  
    if not detector.prepare_dataset():  
        print("Не удалось подготовить датасет. Проверьте структуру папок.")  
        return  
  
    detector.analyze_dataset()  
  
    # 3. Инициализация модели  
    if not detector.setup_model():  
        return  
  
    # 4. Обучение модели  
    print("\nБЕЗОПАСНЫЕ ПАРАМЕТРЫ ОБУЧЕНИЯ:")  
    print(" - Размер изображения: 640x640")  
    print(" - Batch size: 12 (уменьшен для стабильности)")
```

```

print(" - Количество эпох: 50 (уменьшено)")
print(" - Workers: 4 (уменьшено)")
print(" - Mosaic: отключен (экономия памяти)")
print(" - MixUp: отключен (экономия памяти)")
print(" - Лимит памяти GPU: 70%")

if not detector.train_model(epochs=50, imgsz=640, batch=12):
    return

# 5. Оценка модели
print("\nОЦЕНКА МОДЕЛИ...")
metrics = detector.evaluate_model()

# 6. Сохранение графиков
print("\nСОХРАНЕНИЕ ГРАФИКОВ...")
detector.save_training_plots()

# 7. Тестирование детектора
print("\nТЕСТИРОВАНИЕ ДЕТЕКТОРА...")
best_model_path = os.path.join(detector.output_dir, 'models', 'best.pt')

if os.path.exists(best_model_path):
    visualizer = DetectionVisualizer(best_model_path,
                                      os.path.join(detector.output_dir, 'detections'))

    # Тестирование на датасете
    visualizer.test_on_dataset(detector.yolo_dataset_path, num_images=5)

    print(f"РЕЗУЛЬТАТЫ СОХРАНЕНЫ В: {detector.output_dir}")

if metrics:
    print("\nФИНАЛЬНЫЕ МЕТРИКИ МОДЕЛИ:")
    print(f" mAP50: {metrics.box.map50:.3f}")
    print(f" mAP50-95: {metrics.box.map:.3f}")
    print(f" Precision: {metrics.box.precision:.3f}")
    print(f" Recall: {metrics.box.recall:.3f}")
else:
    print("Лучшая модель не найдена")

if __name__ == "__main__":
    main()

```

Вывод программы:

Image sizes 640 train, 640 val

Using 4 dataloader workers

Logging results to /home/oppa/runs/detect/people_detection_20251111_214341
Starting training for 50 epochs...

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
1/50	0.25G	1.517	1.963	1.323	65	640: 100% ————— 1083/1083 5.8it/s 3:07
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 2.2it/s 26.2s
	all	1370	10660	0.675	0.465	0.504 0.243
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
2/50	2.27G	1.536	1.639	1.34	25	640: 100% ————— 1083/1083 7.4it/s 2:27
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 7.3it/s 8.0s
	all	1370	10660	0.624	0.449	0.489 0.232
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
3/50	3.13G	1.624	1.656	1.419	22	640: 100% ————— 1083/1083 8.0it/s 2:15
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 6.9it/s 8.4s
	all	1370	10660	0.568	0.395	0.411 0.191
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
4/50	3.14G	1.689	1.689	1.469	55	640: 100% ————— 1083/1083 7.6it/s 2:23
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 6.5it/s 9.0s
	all	1370	10660	0.642	0.414	0.463 0.223
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
5/50	3.14G	1.611	1.594	1.435	25	640: 100% ————— 1083/1083 7.8it/s 2:19
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 7.1it/s 8.2s
	all	1370	10660	0.655	0.429	0.493 0.243
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
6/50	3.14G	1.555	1.508	1.402	17	640: 100% ————— 1083/1083 7.6it/s 2:22
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 6.8it/s 8.6s
	all	1370	10660	0.612	0.449	0.496 0.252
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
7/50	3.14G	1.514	1.454	1.373	30	640: 100% ————— 1083/1083 7.6it/s 2:22
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 6.3it/s 9.3s
	all	1370	10660	0.683	0.484	0.544 0.28
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
8/50	3.14G	1.477	1.411	1.354	28	640: 100% ————— 1083/1083 7.6it/s 2:23
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 7.1it/s 8.2s
	all	1370	10660	0.691	0.485	0.557 0.295
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
9/50	3.14G	1.451	1.373	1.334	120	640: 100% ————— 1083/1083 7.6it/s 2:22
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 58/58 6.3it/s 9.2s

all 1370 10660 0.681 0.481 0.543 0.292

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

10/50 3.14G 1.423 1.338 1.318 19 640: 100% ————— 1083/1083 7.2it/s 2:31
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.0it/s 8.3s
all 1370 10660 0.732 0.476 0.542 0.299

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

11/50 3.14G 1.405 1.302 1.304 27 640: 100% ————— 1083/1083 7.8it/s 2:19
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.7it/s 8.6s
all 1370 10660 0.703 0.512 0.581 0.32

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

12/50 3.14G 1.384 1.286 1.293 20 640: 100% ————— 1083/1083 7.5it/s 2:24
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.3it/s 8.0s
all 1370 10660 0.712 0.515 0.591 0.327

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

13/50 3.14G 1.361 1.267 1.282 17 640: 100% ————— 1083/1083 7.4it/s 2:26
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.8it/s 8.6s
all 1370 10660 0.713 0.53 0.595 0.329

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

14/50 3.14G 1.35 1.243 1.272 26 640: 100% ————— 1083/1083 7.8it/s 2:18
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.3it/s 8.0s
all 1370 10660 0.75 0.516 0.602 0.338

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

15/50 3.14G 1.335 1.219 1.261 9 640: 100% ————— 1083/1083 7.9it/s 2:17
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.1it/s 8.1s
all 1370 10660 0.72 0.532 0.599 0.336

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

16/50 3.14G 1.326 1.199 1.251 36 640: 100% ————— 1083/1083 7.7it/s 2:21
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.7it/s 8.6s
all 1370 10660 0.752 0.536 0.619 0.348

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

17/50 3.14G 1.31 1.184 1.242 25 640: 100% ————— 1083/1083 7.4it/s 2:26
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.9it/s 8.4s
all 1370 10660 0.744 0.547 0.627 0.353

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

18/50 3.14G 1.298 1.168 1.238 23 640: 100% ————— 1083/1083 7.4it/s 2:25
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.6it/s 8.8s

all	1370	10660	0.748	0.54	0.623	0.353		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
19/50	3.14G	1.288	1.15	1.229	57	640: 100%	1083/1083	7.7it/s 2:22
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.9it/s 8.4s
all	1370	10660	0.763	0.554	0.632	0.359		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
20/50	3.14G	1.275	1.145	1.222	25	640: 100%	1083/1083	7.6it/s 2:22
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.8it/s 8.5s
all	1370	10660	0.742	0.552	0.635	0.368		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
21/50	3.14G	1.259	1.119	1.21	16	640: 100%	1083/1083	7.9it/s 2:18
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.9it/s 8.4s
all	1370	10660	0.76	0.563	0.647	0.372		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
22/50	3.14G	1.252	1.101	1.205	34	640: 100%	1083/1083	7.6it/s 2:23
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.5it/s 8.9s
all	1370	10660	0.768	0.562	0.642	0.373		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
23/50	3.14G	1.249	1.104	1.204	25	640: 100%	1083/1083	7.7it/s 2:22
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.5it/s 8.9s
all	1370	10660	0.759	0.562	0.648	0.375		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
24/50	3.14G	1.235	1.084	1.199	33	640: 100%	1083/1083	7.7it/s 2:21
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	7.0it/s 8.3s
all	1370	10660	0.775	0.56	0.647	0.375		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
25/50	3.14G	1.22	1.074	1.192	19	640: 100%	1083/1083	7.7it/s 2:20
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.4it/s 9.1s
all	1370	10660	0.773	0.557	0.65	0.383		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
26/50	3.14G	1.211	1.058	1.186	29	640: 100%	1083/1083	7.6it/s 2:23
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.8it/s 8.5s
all	1370	10660	0.761	0.58	0.665	0.392		
Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size								
27/50	3.14G	1.209	1.052	1.178	31	640: 100%	1083/1083	7.5it/s 2:24
Class	Images	Instances	Box(P)	R	mAP50	mAP50-95): 100%	58/58	6.7it/s 8.6s

all 1370 10660 0.773 0.573 0.665 0.393

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

28/50 3.14G 1.197 1.04 1.172 44 640: 100% ————— 1083/1083 7.7it/s 2:20

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.2it/s 8.1s
all 1370 10660 0.772 0.576 0.661 0.39

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

29/50 3.14G 1.188 1.032 1.171 35 640: 100% ————— 1083/1083 7.6it/s 2:22

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.1it/s 8.2s
all 1370 10660 0.771 0.582 0.667 0.393

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

30/50 3.14G 1.182 1.021 1.164 15 640: 100% ————— 1083/1083 7.6it/s 2:23

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.0it/s 8.3s
all 1370 10660 0.777 0.579 0.67 0.398

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

31/50 3.14G 1.169 1.01 1.159 21 640: 100% ————— 1083/1083 7.9it/s 2:17

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.0it/s 8.2s
all 1370 10660 0.777 0.581 0.671 0.401

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

32/50 3.14G 1.161 0.994 1.154 28 640: 100% ————— 1083/1083 7.7it/s 2:22

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.9it/s 8.4s
all 1370 10660 0.789 0.58 0.674 0.403

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

33/50 3.14G 1.154 0.9868 1.147 25 640: 100% ————— 1083/1083 7.6it/s 2:23

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.8it/s 8.5s
all 1370 10660 0.775 0.593 0.677 0.404

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

34/50 3.14G 1.145 0.9767 1.141 20 640: 100% ————— 1083/1083 7.9it/s 2:18

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.1it/s 8.2s
all 1370 10660 0.787 0.583 0.679 0.409

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

35/50 3.14G 1.136 0.967 1.138 17 640: 100% ————— 1083/1083 7.5it/s 2:24

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.0it/s 8.3s
all 1370 10660 0.787 0.587 0.68 0.41

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

36/50 3.14G 1.122 0.9551 1.133 20 640: 100% ————— 1083/1083 7.5it/s 2:25

Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.7it/s 8.6s

all 1370 10660 0.783 0.589 0.681 0.412

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

37/50 3.14G 1.124 0.9468 1.13 46 640: 100% ━━━━━━━━ 1083/1083 7.9it/s 2:18

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.9it/s 8.4s
all 1370 10660 0.795 0.585 0.684 0.416

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

38/50 3.14G 1.109 0.9319 1.124 17 640: 100% ━━━━━━━━ 1083/1083 8.1it/s 2:14

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.5it/s 8.9s
all 1370 10660 0.795 0.582 0.681 0.417

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

39/50 3.14G 1.099 0.926 1.12 23 640: 100% ━━━━━━━━ 1083/1083 7.7it/s 2:20

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.9it/s 8.4s
all 1370 10660 0.784 0.593 0.687 0.42

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

40/50 3.14G 1.093 0.9161 1.111 66 640: 100% ━━━━━━━━ 1083/1083 7.5it/s 2:25

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.8it/s 8.6s
all 1370 10660 0.784 0.598 0.692 0.424

Closing dataloader mosaic

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

41/50 3.14G 1.081 0.9017 1.107 46 640: 100% ━━━━━━━━ 1083/1083 7.5it/s 2:24

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.5it/s 8.9s
all 1370 10660 0.787 0.602 0.692 0.424

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

42/50 3.14G 1.075 0.8935 1.104 38 640: 100% ━━━━━━━━ 1083/1083 7.4it/s 2:26

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.7it/s 8.7s
all 1370 10660 0.795 0.592 0.69 0.424

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

43/50 3.14G 1.065 0.8872 1.102 54 640: 100% ━━━━━━━━ 1083/1083 7.2it/s 2:31

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.7it/s 8.7s
all 1370 10660 0.799 0.589 0.689 0.424

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

44/50 3.14G 1.057 0.8754 1.096 55 640: 100% ━━━━━━━━ 1083/1083 7.2it/s 2:31

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.2it/s 9.4s
all 1370 10660 0.803 0.587 0.69 0.425

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size

45/50 3.14G 1.047 0.8658 1.092 32 640: 100% ━━━━━━━━ 1083/1083 7.1it/s 2:33

Class Images Instances Box(P R mAP50 mAP50-95): 100% ━━━━━━ 58/58 6.5it/s 8.9s
all 1370 10660 0.803 0.587 0.69 0.425

all 1370 10660 0.798 0.593 0.691 0.426

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size
46/50 3.14G 1.043 0.8585 1.088 11 640: 100% ————— 1083/1083 7.7it/s 2:20
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.9it/s 8.4s
all 1370 10660 0.797 0.599 0.694 0.427

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size
47/50 3.14G 1.039 0.8528 1.085 29 640: 100% ————— 1083/1083 7.6it/s 2:22
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.0it/s 8.3s
all 1370 10660 0.796 0.603 0.695 0.428

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size
48/50 3.14G 1.022 0.8392 1.079 63 640: 100% ————— 1083/1083 7.4it/s 2:27
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.2it/s 9.3s
all 1370 10660 0.794 0.601 0.694 0.429

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size
49/50 3.14G 1.021 0.8362 1.078 24 640: 100% ————— 1083/1083 7.4it/s 2:27
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 6.4it/s 9.0s
all 1370 10660 0.79 0.605 0.695 0.429

Epoch GPU_mem box_loss cls_loss dfl_loss Instances Size
50/50 3.14G 1.014 0.8278 1.075 30 640: 100% ————— 1083/1083 7.3it/s 2:28
Class Images Instances Box(P R mAP50 mAP50-95): 100% ————— 58/58 7.4it/s 7.9s
all 1370 10660 0.794 0.606 0.696 0.43

50 epochs completed in 2.124 hours.

Optimizer stripped from /home/oppa/runs/detect/people_detection_20251111_214341/weights/last.pt, 5.4MB
Optimizer stripped from /home/oppa/runs/detect/people_detection_20251111_214341/weights/best.pt, 5.4MB

ОЦЕНКА МОДЕЛИ...

Оценка модели на тестовой выборке...

Ultralytics 8.3.227 🚀 Python-3.12.3 torch-2.8.0+cu129 CUDA:0 (NVIDIA GeForce RTX 5060, 8150MiB)

YOLOv1n summary (fused): 100 layers, 2,582,347 parameters, 0 gradients, 6.3 GFLOPs

val: Fast image access (ping: 0.4±0.3 ms, read: 47.3±23.6 MB/s, size: 35.6 KB)

val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 60 images, 0 backgrounds, 0 corrupt: 8%

————— 60/738 177.9it/s 0.1s<3.8val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 129 images,

0 backgrounds, 0 corrupt: 17% ————— 129/738 324.4it/s 0.2s<val: Scanning

/home/oppa/mamka/yolo_dataset/test/labels... 197 images, 0 backgrounds, 0 corrupt: 27% —————

197/738 426.6it/s 0.3s<val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 265 images, 0 backgrounds, 0

corrupt: 36% ————— 265/738 501.5it/s 0.4s<val: Scanning

/home/oppa/mamka/yolo_dataset/test/labels... 330 images, 0 backgrounds, 0 corrupt: 45% —————

330/738 543.2it/s 0.5s<val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 399 images, 0 backgrounds, 0

corrupt: 54% ————— 399/738 583.5it/s 0.6s<val: Scanning
 /home/oppa/mamka/yolo_dataset/test/labels... 466 images, 0 backgrounds, 0 corrupt: 63% —————
 466/738 606.8it/s 0.7s<val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 533 images, 0 backgrounds, 0
 corrupt: 72% ————— 533/738 622.6it/s 0.8s<val: Scanning
 /home/oppa/mamka/yolo_dataset/test/labels... 600 images, 0 backgrounds, 0 corrupt: 81% —————
 600/738 636.5it/s 0.9s<val: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 665 images, 0 backgrounds, 0
 corrupt: 90% ————— 665/738 638.9it/s 1.0s<val: Scanning
 /home/oppa/mamka/yolo_dataset/test/labels... 737 images, 0 backgrounds, 0 corrupt: 100% —————
 737/738 661.1it/s 1.1sval: Scanning /home/oppa/mamka/yolo_dataset/test/labels... 738 images, 0 backgrounds, 0 corrupt:
 100% ————— 738/738 661.3it/s 1.1s

val: New cache created: /home/oppa/mamka/yolo_dataset/test/labels.cache

Class	Images	Instances	Box(P)	R	mAP50	mAP50-95
all	738	2783	0.868	0.832	0.893	0.656

Speed: 0.8ms preprocess, 17.4ms inference, 0.0ms loss, 1.2ms postprocess per image

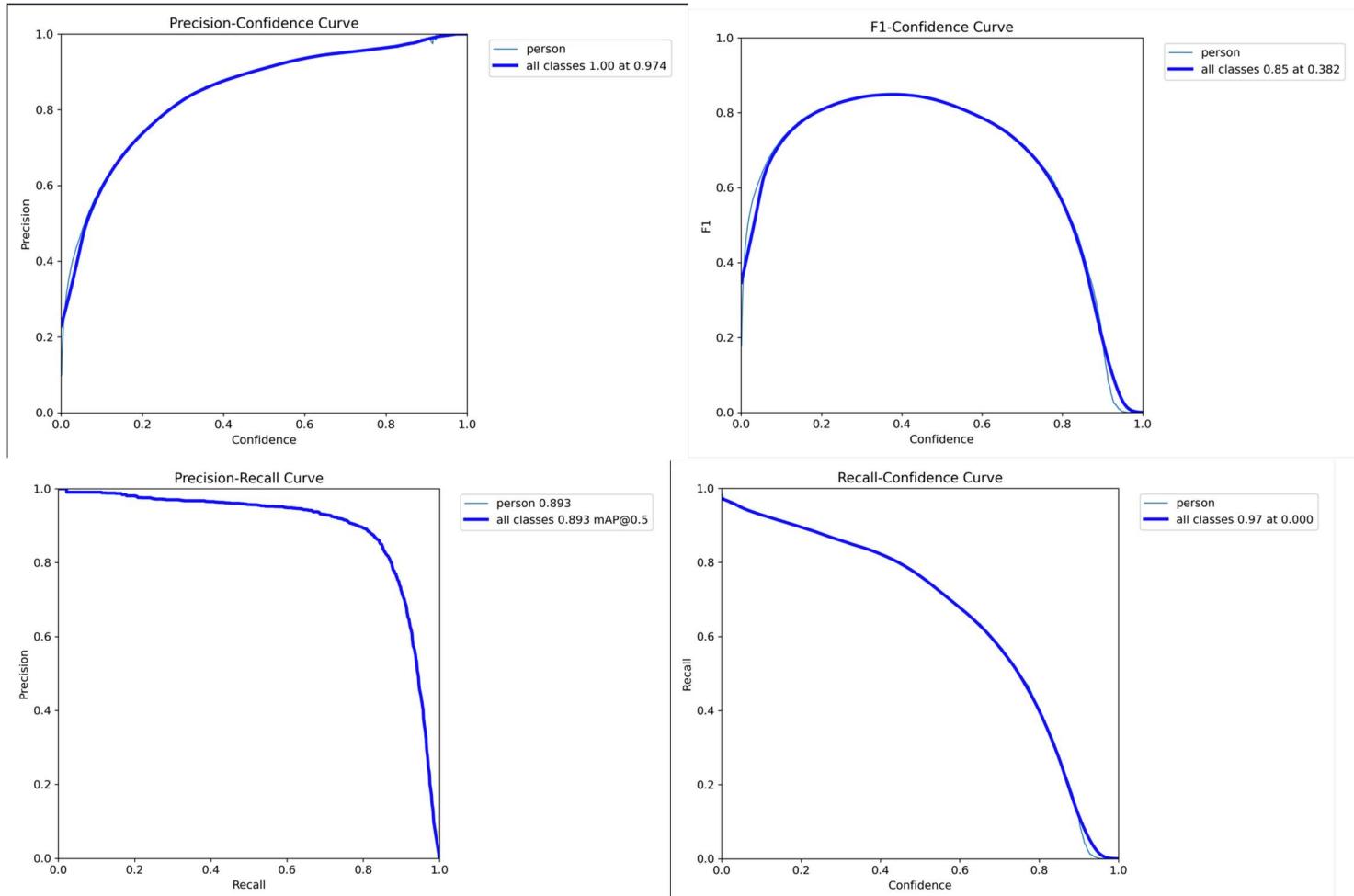
Results saved to /home/oppa/runs/detect/val

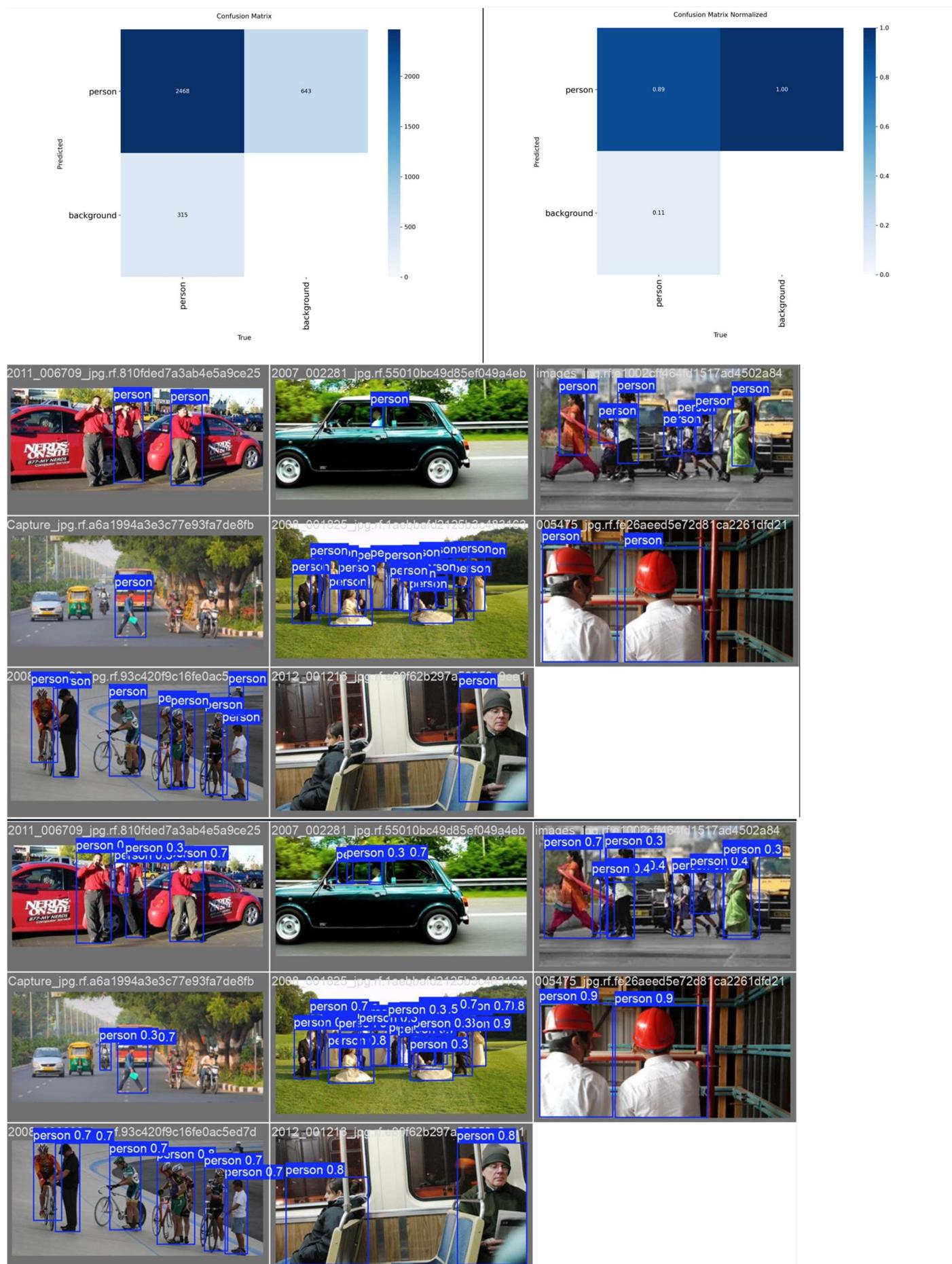
Результаты оценки:

mAP50-95: 0.6556

mAP50: 0.8933

mAP75: 0.7308









Вывод: осуществил обучение нейросетевого детектора для решения задачи обнаружения нных объектов.