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Кафедра ИИТ

Лабораторная работа №3
По дисциплине «ОИВИС»
Тема: **“Обучение детекторов объектов”**

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

Вариант 7.

В-т	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
```

```
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 YOLOv11n"""  
    print("Инициализация модели YOLOv11n...")
```

```
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("\n БЕЗОПАСНЫЕ 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("\n ОЦЕНКА МОДЕЛИ...")
metrics = detector.evaluate_model()
```

```
# 6. Сохранение графиков
print("\n СОХРАНЕНИЕ ГРАФИКОВ...")
detector.save_training_plots()
```

```
# 7. Тестирование детектора
print("\n ТЕСТИРОВАНИЕ 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(f"\n ФИНАЛЬНЫЕ 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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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	df_l_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 0.798 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%
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%
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%
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%
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%
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)

YOLO11n 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%

/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%

/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): 100% ————— 93/93 5.3it/s 17.6s
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







