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LAMPIRAN

Lampiran 1 Contoh Dataset Cabai Katokkon



Grade A



Grade B



Defect



Fly Bites

Lampiran 2 *Source Code* Training Model YOLOv8 dan YOLOv9

✓ Install Ultralytics

```
from google.colab import drive
drive.mount('/content/drive')
%cd /content/drive/MyDrive/Colab Notebooks/YOLOv8

# Ultralytics version for MODEL CONVERSION
# !git clone https://github.com/ultralytics/ultralytics
%pip install -q ultralytics
%cd ultralytics

# MODEL TRAINING
%pip install ultralytics
```

✓ Import Dataset from Roboflow

```
#YOLOv8 & YOLOv9 dataset used the same format
!pip install roboflow

from roboflow import Roboflow
rf = Roboflow(api_key="YKs50cm1gTqydIgP567E")
project = rf.workspace("unhas-9yowg").project("katokkon_ab_defect_flybites")
version = project.version(4)
dataset = version.download("yolov8")
```

✓ Train YOLOv8

```
# hyperparameter tuning

from ultralytics import YOLO
model = YOLO('yolov8n.pt')

model.tune(data='/content/katokkon_ab_defect_flybites-4/data.yaml', epochs=50, iterations=25, optimizer='AdamW', plots=False, save=False)

# yolov8
!yolo task=detect \
  mode=train \
  model=yolov8n.pt \
  data='/content/katokkon_ab_defect_flybites-4/data.yaml' \
  epochs= 100 \
  imgsz=320
```

✓ Train YOLOv9

```
# yolov9
!yolo task=detect \
  mode=train \
  model=yolov9c.pt \
  data='/content/katokkon_ab_defect_flybites-4/data.yaml' \
  epochs= 300 \
  patience=20 \
  imgsz=320
```

✓ Tensorboard

```
%load_ext tensorboard
%tensorboard --logdir runs
```

✓ Export model to ONNX format

```
!yolo task=detect mode=export model="/content/drive/MyDrive/Colab Notebooks/YOLOv8/runs/detect/train7-first100epochnewdatasetlabelw-ba
```

✓ export to ncnn & tf lite

```
!yolo task=detect mode=export model="/content/drive/MyDrive/Colab Notebooks/YOLOv8/train4.pt" format=tf lite simplify=True opset=13 imgsz:

!yolo task=detect mode=export model="/content/drive/MyDrive/Colab Notebooks/YOLOv8/v9train2.pt" format=tf lite simplify=True opset=13 img
```

Lampiran 3 *Source Code* Training Model YOLOv7

✓ Install Dependencies

```
!python --version

# Download YOLOv7 repository and install requirements
!git clone https://github.com/WongKinYiu/yolov7
%cd yolov7
!pip install -r requirements.txt
```

✓ Download dataset

```
# chili dataset
!pip install roboflow

from roboflow import Roboflow
rf = Roboflow(api_key="YKs50cm1gTqydIgP567E")
project = rf.workspace("unhas-9yowg").project("katokkon_ab_defect_flybites")
version = project.version(4)
dataset = version.download("yolov7")
```

✓ Training YOLOv7

```
# download COCO starting checkpoint
%cd /content/yolov7
!wget https://github.com/WongKinYiu/yolov7/releases/download/v0.1/yolov7_training.pt
```

✓ generate confusion matrix yolov7: edit utils/metrics.py – line 168

```
# run this cell to begin training
%cd /content/yolov7
!python train.py --batch 16 --epochs 50 --data {dataset.location}/data.yaml --weights 'yolov7_training.pt' --device 0
```

	from	n	params	module	arguments
0	-1	1	928	models.common.Conv	[3, 32, 3, 1]
1	-1	1	18560	models.common.Conv	[32, 64, 3, 2]
2	-1	1	36992	models.common.Conv	[64, 64, 3, 1]
3	-1	1	73984	models.common.Conv	[64, 128, 3, 2]
4	-1	1	8320	models.common.Conv	[128, 64, 1, 1]
5	-2	1	8320	models.common.Conv	[128, 64, 1, 1]
6	-1	1	36992	models.common.Conv	[64, 64, 3, 1]
7	-1	1	36992	models.common.Conv	[64, 64, 3, 1]
8	-1	1	36992	models.common.Conv	[64, 64, 3, 1]
9	-1	1	36992	models.common.Conv	[64, 64, 3, 1]
10	[-1, -3, -5, -6]	1	0	models.common.Concat	[1]
11	-1	1	66048	models.common.Conv	[256, 256, 1, 1]
12	-1	1	0	models.common.MP	[1]
13	-1	1	33024	models.common.Conv	[256, 128, 1, 1]
14	-3	1	33024	models.common.Conv	[256, 128, 1, 1]
15	-1	1	147712	models.common.Conv	[128, 128, 3, 2]
16	[-1, -3]	1	0	models.common.Concat	[1]
17	-1	1	33024	models.common.Conv	[256, 128, 1, 1]
18	-2	1	33024	models.common.Conv	[256, 128, 1, 1]
19	-1	1	147712	models.common.Conv	[128, 128, 3, 1]
20	-1	1	147712	models.common.Conv	[128, 128, 3, 1]

```

35          -1 1 598336 models.common.Conv [256, 256, 3, 1]
36 [-1, -3, -5, -6] 1 0 models.common.Concat [1]
37          -1 1 1058624 models.common.Conv [1024, 1024, 1, 1]
38          -1 1 0 models.common.MP [1]
39          -1 1 525312 models.common.Conv [1024, 512, 1, 1]
40          -3 1 525312 models.common.Conv [1024, 512, 1, 1]
41          -1 1 2368320 models.common.Conv [512, 512, 3, 2]
42          [-1, -3] 1 0 models.common.Concat [1]
43          -1 1 262656 models.common.Conv [1024, 256, 1, 1]
44          -2 1 262656 models.common.Conv [1024, 256, 1, 1]
45          -1 1 598336 models.common.Conv [256, 256, 3, 1]
46          -1 1 598336 models.common.Conv [256, 256, 3, 1]
47          -1 1 598336 models.common.Conv [256, 256, 3, 1]
48          -1 1 598336 models.common.Conv [256, 256, 3, 1]
49 [-1, -3, -5, -6] 1 0 models.common.Concat [1]
50          -1 1 1058624 models.common.Conv [1024, 1024, 1, 1]
51          -1 1 7609344 models.common.SPPCSPC [1024, 512, 1]
52          -1 1 131584 models.common.Conv [512, 256, 1, 1]
53          -1 1 0 torch.nn.modules.upsampling.Upsample [None, 2, 'nearest']
54          37 1 262656 models.common.Conv [1024, 256, 1, 1]

```

▼ Evaluation

```

# Run evaluation
!python detect.py --weights runs/train/exp/weights/best.pt --conf 0.1 --source {dataset.location}/test/images

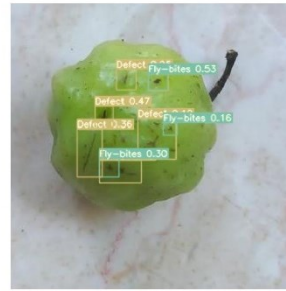
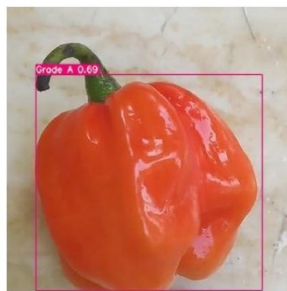
#display inference on ALL test images

import glob
from IPython.display import Image, display

i = 0
limit = 10000 # max images to print
for imageName in glob.glob('/content/yolov7/runs/detect/exp/*.jpg'): #assuming JPG
    if i < limit:
        display(Image(filename=imageName))
        print("\n")
    i = i + 1

```

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Lampiran 4 *Source Code* Inferensi *Raspberry Pi*

```

import cv2
import argparse
import time
import serial

from ultralytics import YOLO
import supervision as sv

def parse_arguments() -> argparse.Namespace:
    parser = argparse.ArgumentParser(description="YOLOv8 Live")
    parser.add_argument(
        "--webcam-resolution",
        default=[1280, 720],
        nargs=2,
        type=int
    )
    parser.add_argument(
        "--modeldir",
        default="yolov8n.pt",
        help='Folder the .pt file is located in'
    )
    parser.add_argument(
        "--source",
        default=0,
        type=int
    )
    args = parser.parse_args()
    return args

def write_list_to_file(my_list, filename):
    """
    Writes a list to a text file with each element on a new line.

    Args:
    my_list: The list to be written to the file.
    filename: The name of the text file.
    """
    with open(filename, 'w') as f:
        for item in my_list:
            f.seek(0, 2)
            f.write(str(item) + '\n') # Convert each item to string and add newline

results_list = []

def main():
    args = parse_arguments()
    frame_width, frame_height = args.webcam_resolution
    MODEL_NAME = args.modeldir

```

```

CAMERA_SOURCE = args.source

cap = cv2.VideoCapture(CAMERA_SOURCE)
cap.set(cv2.CAP_PROP_FRAME_WIDTH, frame_width)
cap.set(cv2.CAP_PROP_FRAME_HEIGHT, frame_height)

model = YOLO(MODEL_NAME)

box_annotator = sv.BoxAnnotator(
    thickness=2,
    text_thickness=2,
    text_scale=1
)

#calculating fps
prev_frame_time = 0
new_frame_time = 0

#loop through image from frame
while True:

    ret, frame = cap.read()

    result = model(frame, agnostic_nms=True)
    object_classes = result[0].boxes.cls.to('cpu').tolist()
    speed_dict = result[0].speed['inference']
    results_list.append(speed_dict)

    #print out detection results
    for i in object_classes:
        print(i)

    detections = sv.Detections.from_ultralytics(result[0])

    labels = [
        f"{model.model.names[class_id]} {confidence:.2f}"
        for class_id, confidence
        in zip(detections.class_id, detections.confidence)
    ]

    frame = box_annotator.annotate(
        scene=frame,
        detections=detections,
        labels=labels
    )

    font = cv2.FONT_HERSHEY_SIMPLEX
    new_frame_time = time.time()

    #calculating fps
    fps = 1/(new_frame_time-prev_frame_time)
    prev_frame_time = new_frame_time
    fps = float("{:.2f}".format(fps))
    fps = str(fps)

```

```
cv2.putText(frame,fps,(7,35), font, 1, (100, 255, 0), 1, cv2.LINE_AA)

cv2.imshow("yolo", frame)

if (cv2.waitKey(30) == 27):
    break

#write_list_to_file(results_list, "yolov8-inference.txt")

if __name__ == "__main__":
    main()
```

Lampiran 5 *Source Code* Inferensi *Jetson Nano*

Source Code inferensi YOLOv7 pada *Jetson Nano*:

<https://github.com/williamadam/Katokkon-Classification/tree/master/programs/Jetson-nano>

Lampiran 6 Program Aplikasi Android

Source Code Aplikasi Android: <https://github.com/williamadamm/Katokkon-Classification/tree/master/programs/Android/ncnn-android-yolov8>

Lampiran 7 Dokumentasi Pengujian

