drone 0.7

드론 감지 및 분류 AI모델

AIM

전남대학교

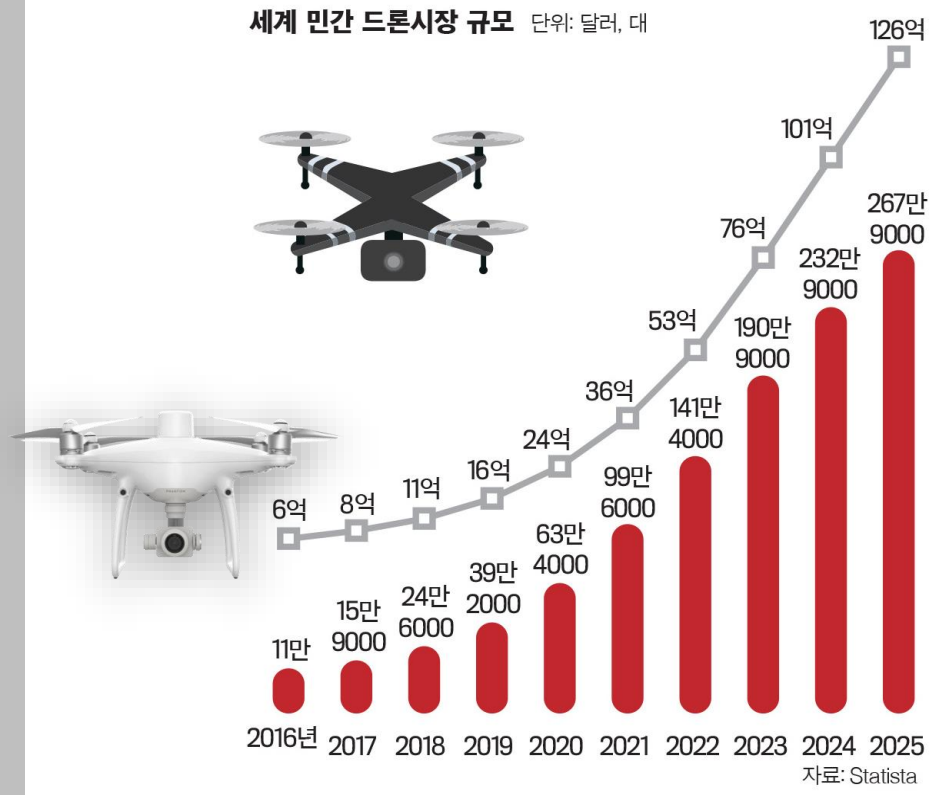
김주완(인공지능학부)
서가연(인공지능학부)
조하영(인공지능학부)

ENERGY+AI 2022 하계 마이크로 캡스톤 디자인

전남대학교 ENERGY+AI
핵심인재양성 교육연구단
CNU Energy+AI Education Research Center

목 차

연구 배경



드론에 대한 관심이 높아지면서
드론의 활용과 시장이 확대되고 있다.

드론이 상용화됨에 따라 불법촬영으로 인한
사생활 침해 문제가 발생하고 있다.

연구 목표

TASK1

Object Detection:

드론 객체 인식

GOAL

이미지(영상) 데이터에서 드론 객체 탐지

연구 목표

GOAL

드론의 종류를 구별하고, 카메라 객체 탐지

TASK2

Object Classification:

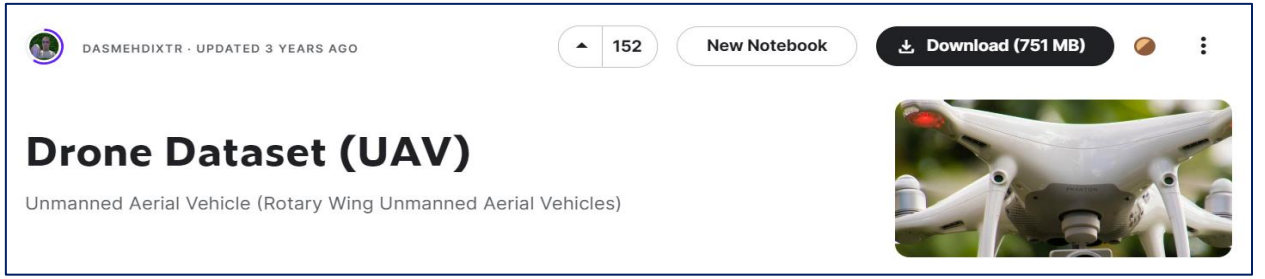
드론 모델 종류별 분류

데이터 수집

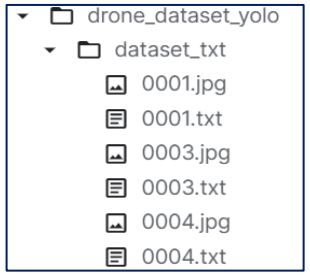
드론 데이터 수집 방법

플랫폼 활용

kaggle

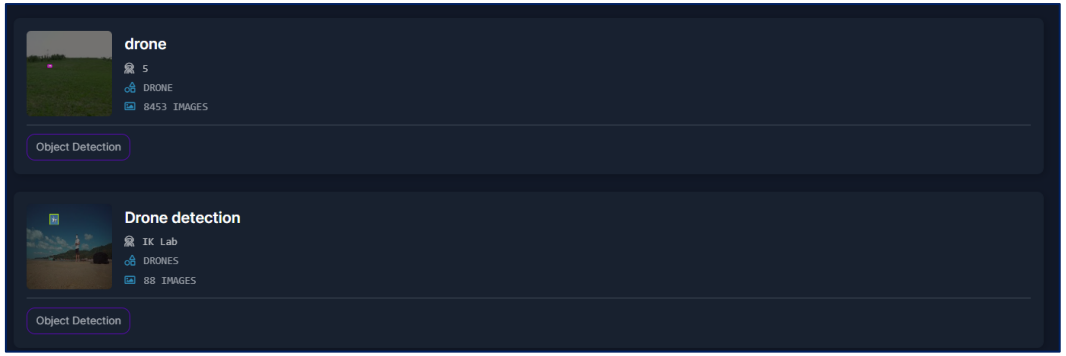


https://www.kaggle.com/datasets/dasmehdixtr/drone-dataset-uav



라벨링 된 데이터 다운로드

roboflow

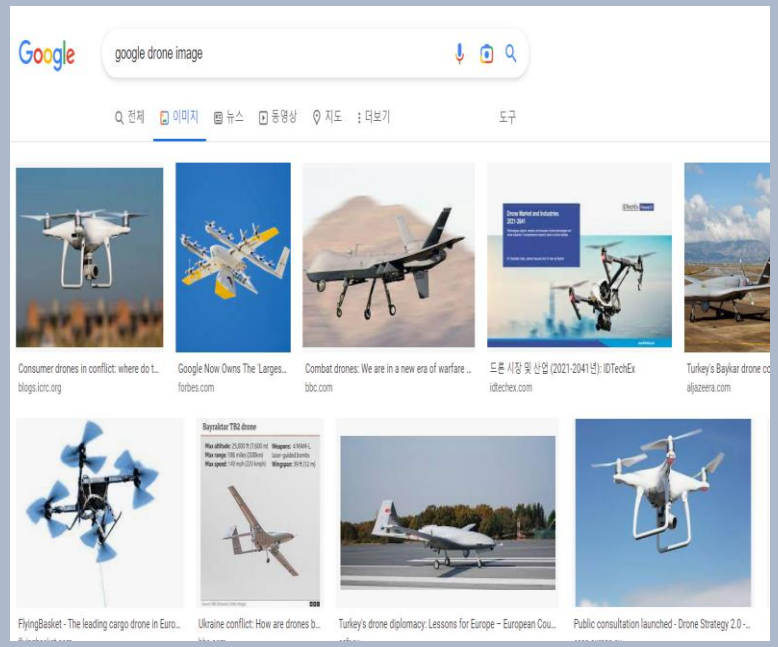


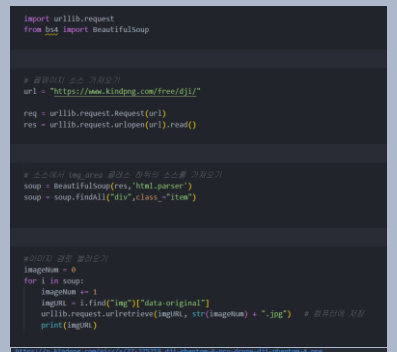
https://universe.roboflow.com/search?q=drone

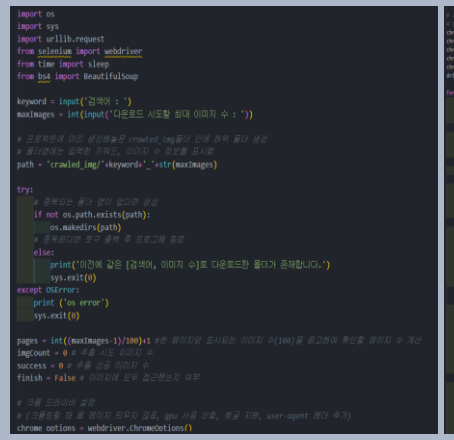
데이터 수집

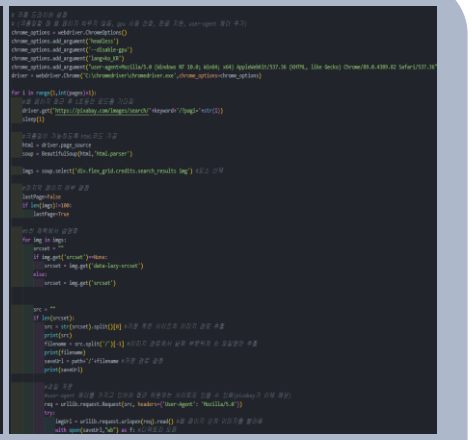
드론 데이터 수집 방법

웹크롤링





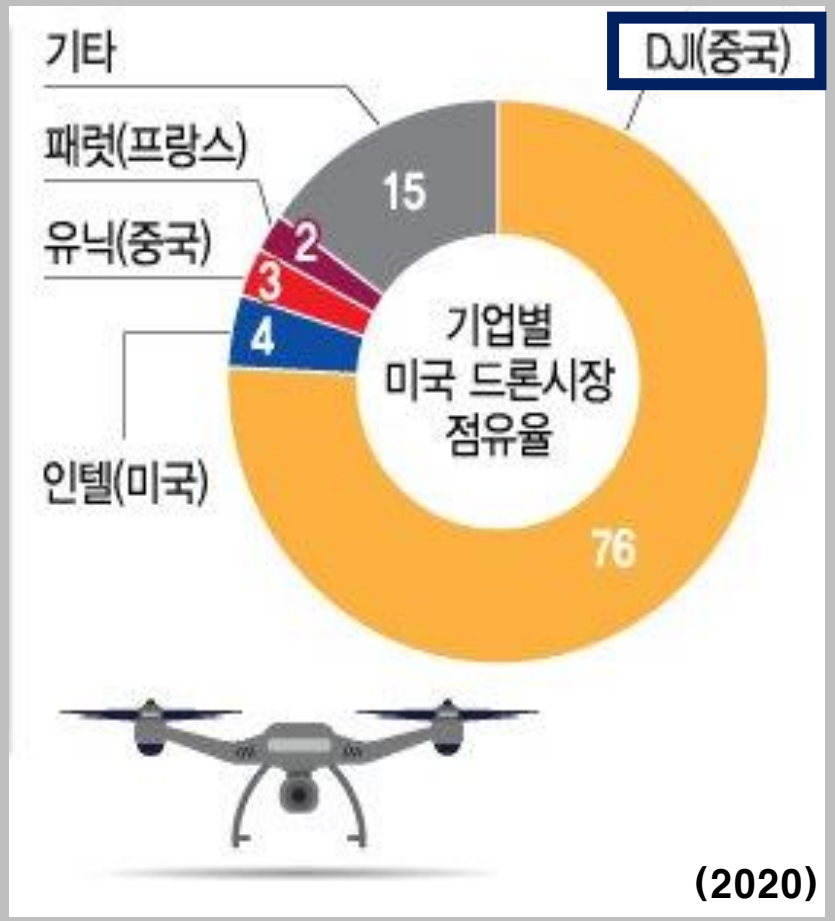




이미지 크롤링으로 드론 데이터 수집

드론 데이터 종류

미국 드론시장 내 점유율(%)



DJI mavic

DJI Phantom

DJI Inspire

데이터 라벨링

- **roboflow** 플랫폼 이용
- 수집한 데이터 셋: yolov5-pytorch로 export하여 모델에 적용

Detection

- class : 1 [“drone”]
- Kaggle + roboflow 6000장
- Annotation: bounding box

Classification

- class : 4
  [“DJI Phantom” , “DJI Mavic” , “DJI Inspire” , “camera”]
- 크롤링 데이터 144장
- Annotation: bounding box + polygon

데이터 전처리 및 증강

전처리 : resize(416*416)

original

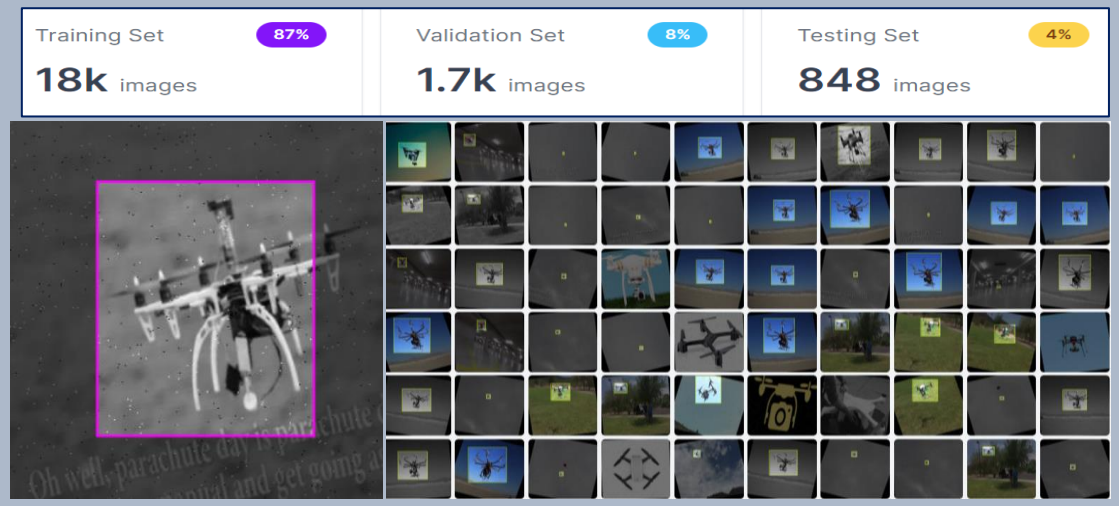
resized

416*416

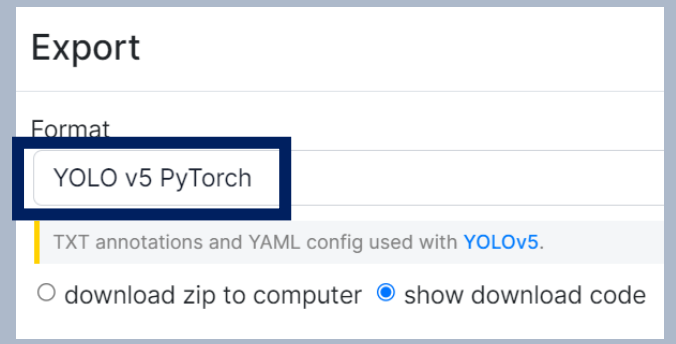
데이터 전처리 및 증강

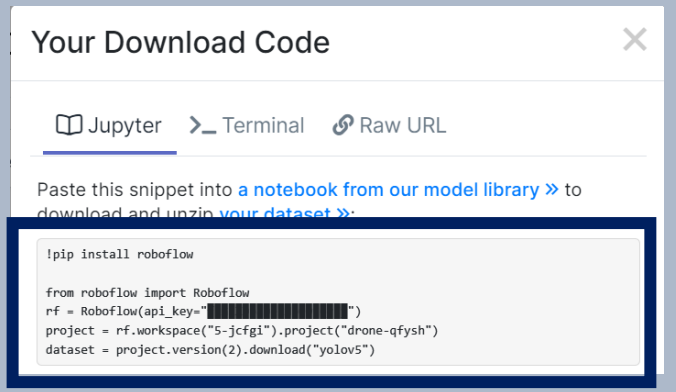
데이터 분리

train : valid : test = 70 : 20 : 10



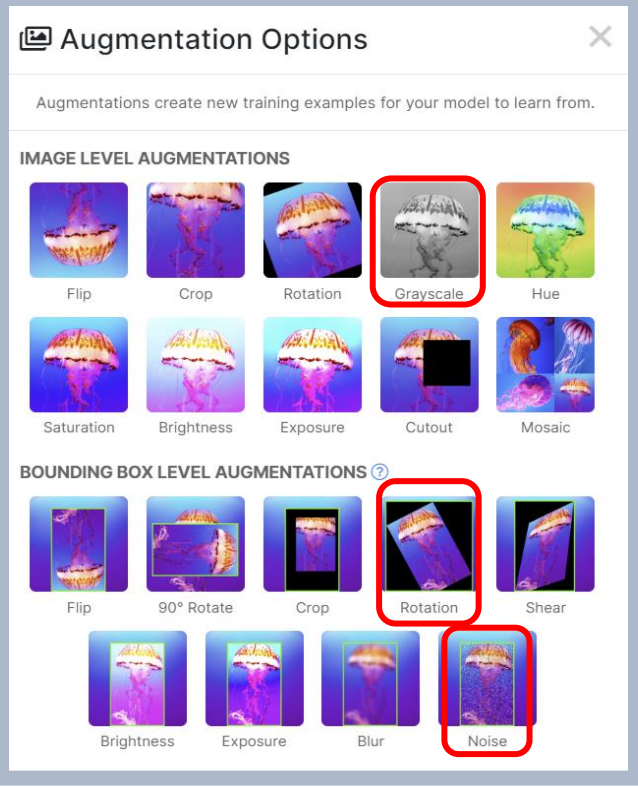
dataset



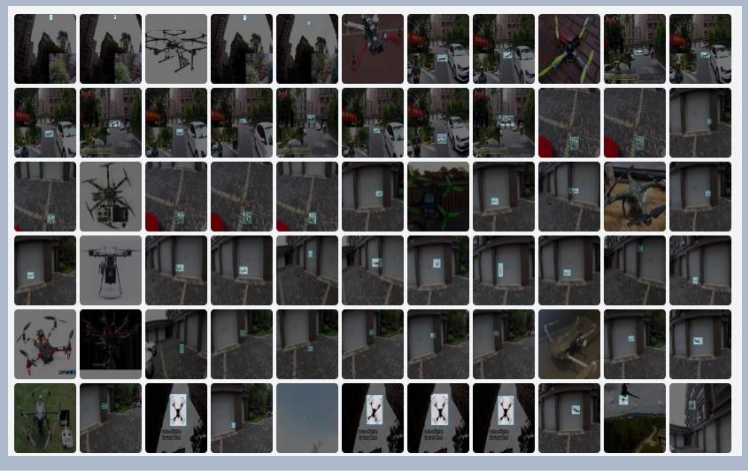


데이터 전처리 및 증강

데이터 증강(x3)



- **Rotation (−15° ~ +15°)**
- **Grayscale (Apply to 30% of images)**
- **Noise (5% of pixels)**



모델 선택

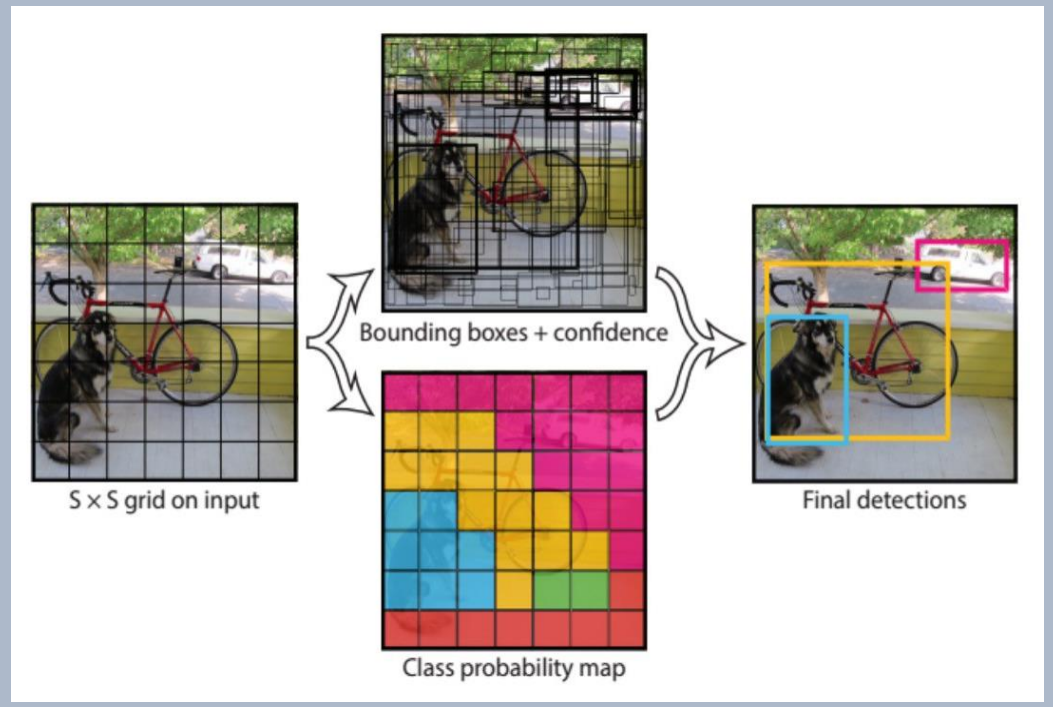
Object Detection : ~~CNN, RCNN, Faster-RCNN, SSD-net, ...~~

기존의 Object detection 알고리즘은 real-time으로 사용하기에는 속도가 느리다는 한계점

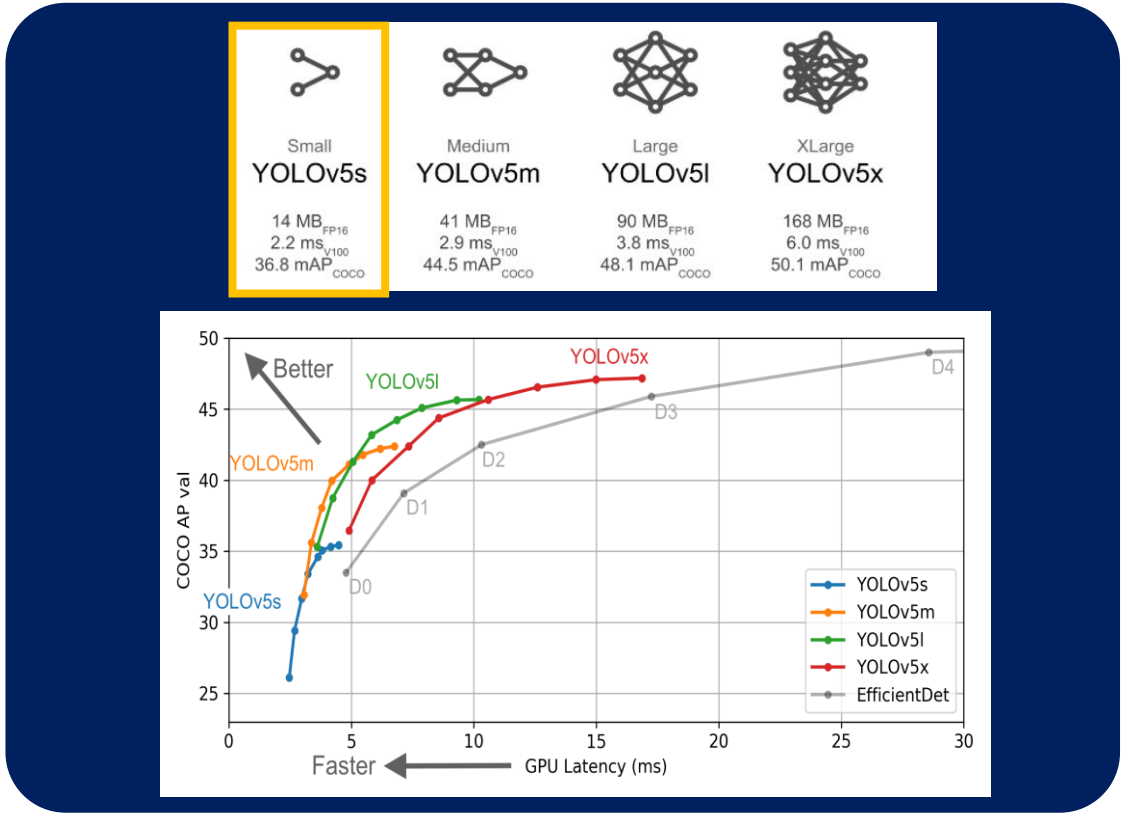
High performance & Fast detection speed

모델 선택

YOLO



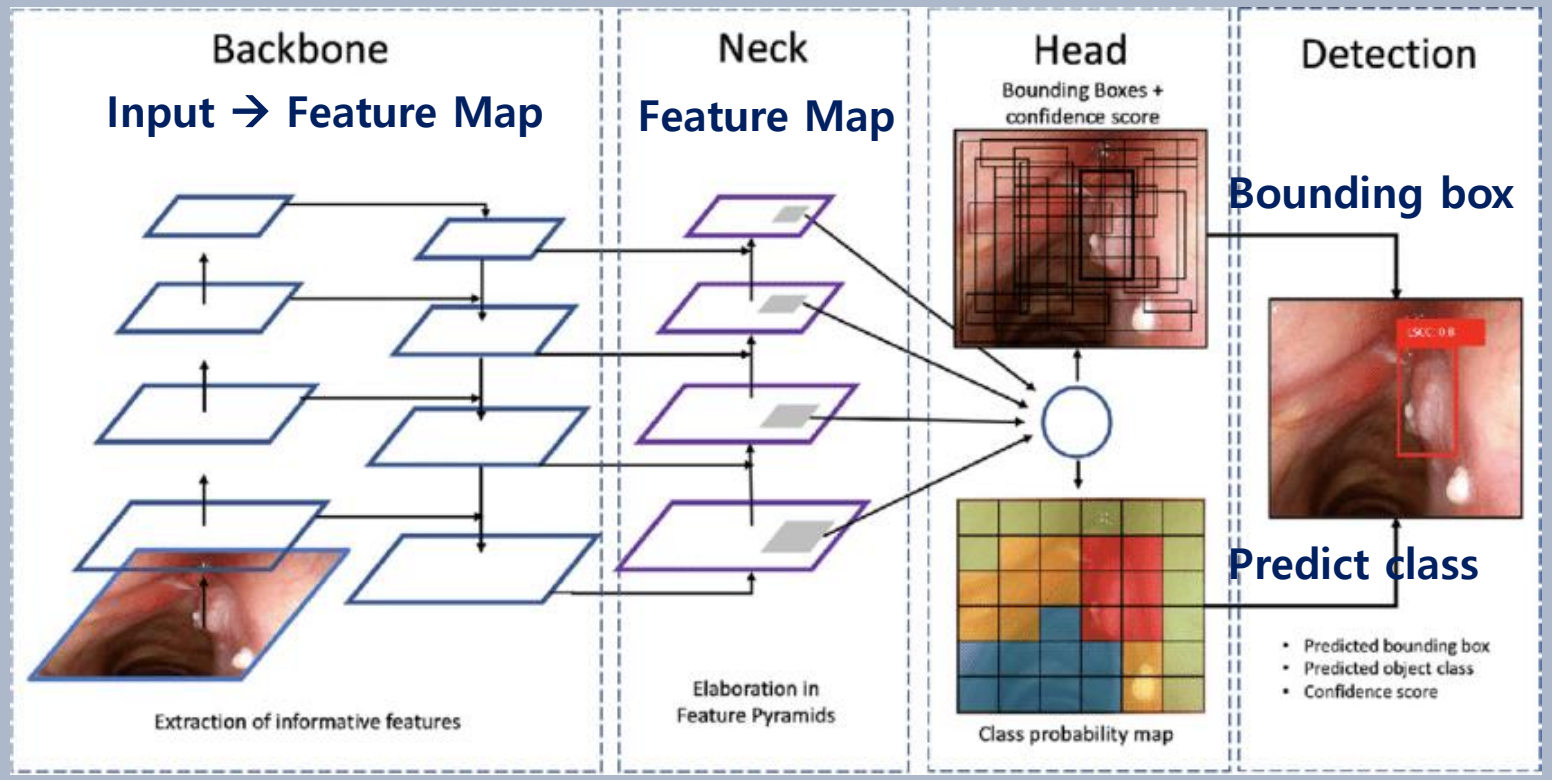
YOLOv5 PyTorch



성능이 좋진 않지만, detection speed(fps)에 강점이 있는 YOLOv5s 선택!

모델 구조

Architecture



모델링

Env Setup

Torch와 cuda 버전 확인

```
import torch
import os
from IPython.display import Image, clear_output

# 실습 환경 셋업
print(f"Setup complete. Using torch {torch.__version__} ({torch.cuda.get_device_properties(0).name if torch.cuda.is_available() else 'CPU'})")
```

```
Setup complete. Using torch 1.12.0+cu113 (Tesla P100-PCIE-16GB)
```

colab

모델링

Install Dependencies

https://github.com/ultralytics/yolov5

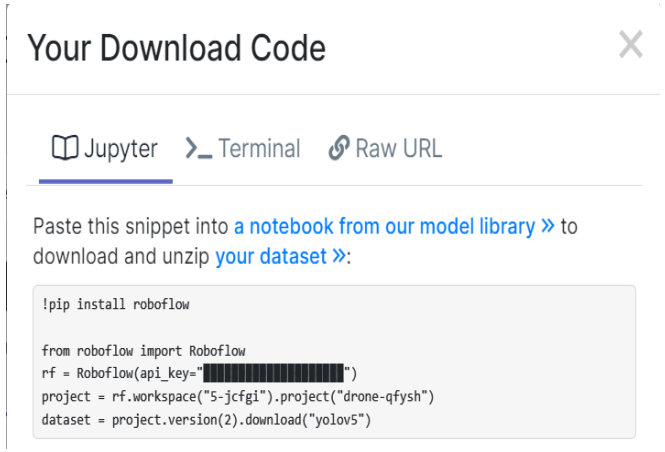
```
#clone YOLOv5 repository
!git clone https://github.com/ultralytics/yolov5  # clone repo
%cd yolov5
%pip install -qr requirements.txt # install dependencies
```

Yolov5 repository 다운로드 및 필요한 라이브러리 설치

모델 학습

Drone Detection

데이터셋 로드하기




```
%pip install -q roboflow
from roboflow import Roboflow

# dataset format : yolov5-Pytorch
rf = Roboflow(api_key="fkGLXDbaDekHbC0t6y7y")
project = rf.workspace("5-jcfgi").project("drone-qfysh")
dataset = project.version(1).download("yolov5")
```

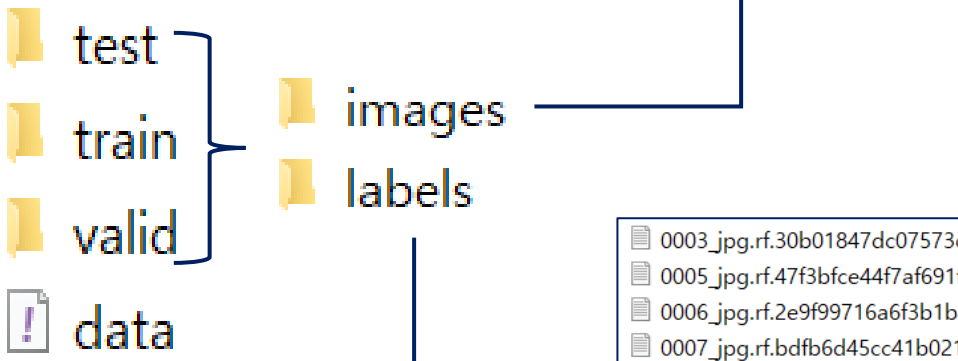
```
loading Roboflow workspace...
loading Roboflow project...
Downloading Dataset Version Zip in /content/datasets/drone-1 to yolov5pytorch: 100% [677072690 / 677072690] bytes
Extracting Dataset Version Zip to /content/datasets/drone-1 in yolov5pytorch:: 100%|██████████| 40574/40574 [00:05<00:00, 7548.00it/s]
```

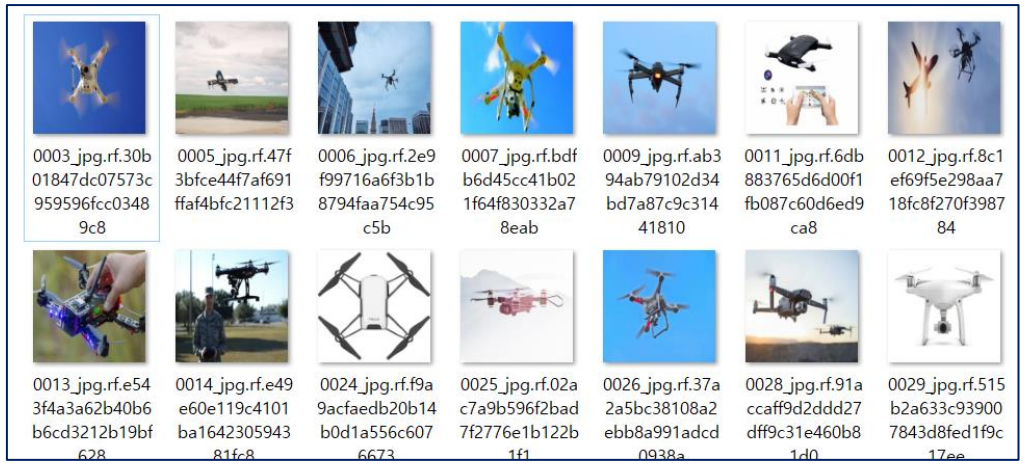
Roboflow에서 생성한 데이터셋 다운로드

모델 학습

Drone Detection

데이터셋 로드하기

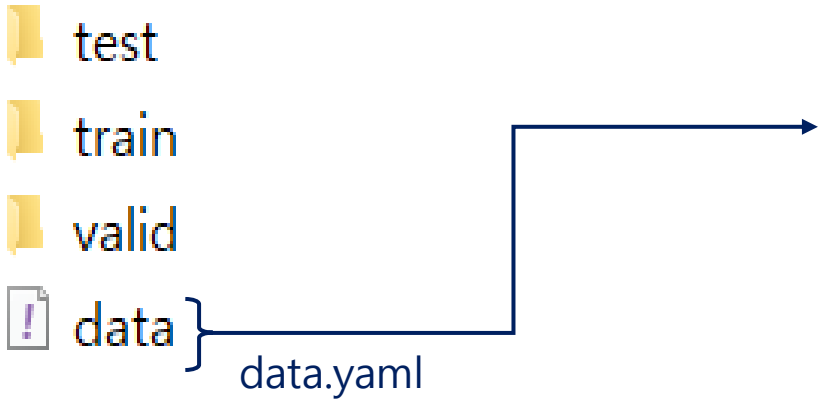




모델 학습

Drone Detection

데이터셋 로드하기




```
names:
- drone
nc: 1
test: ../datasets/test/images
train: ../datsets/train/images
val: ../datasets/valid/images
```

클래스 이름

클래스 개수

데이터 경로

모델 학습

Drone Detection

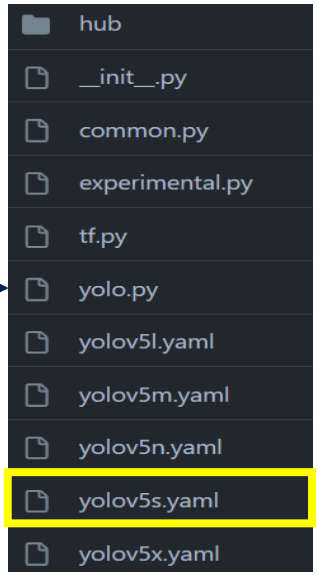
모델로 학습하기

yolov5s 모델로 train data 학습시키기

```
!python train.py --img 416 --batch 64 --epochs 150 --data {dataset.location}/data.yaml --weights yolov5s.pt --cache
```

```
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: 0.05
cls: 0.5
cls_pw: 1.0
obj: 1.0
obj_pw: 1.0
iou_t: 0.2
anchor_t: 4.0
fl_gamma: 0.0
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: 1.0
mixup: 0.0
copy_paste: 0.0
```

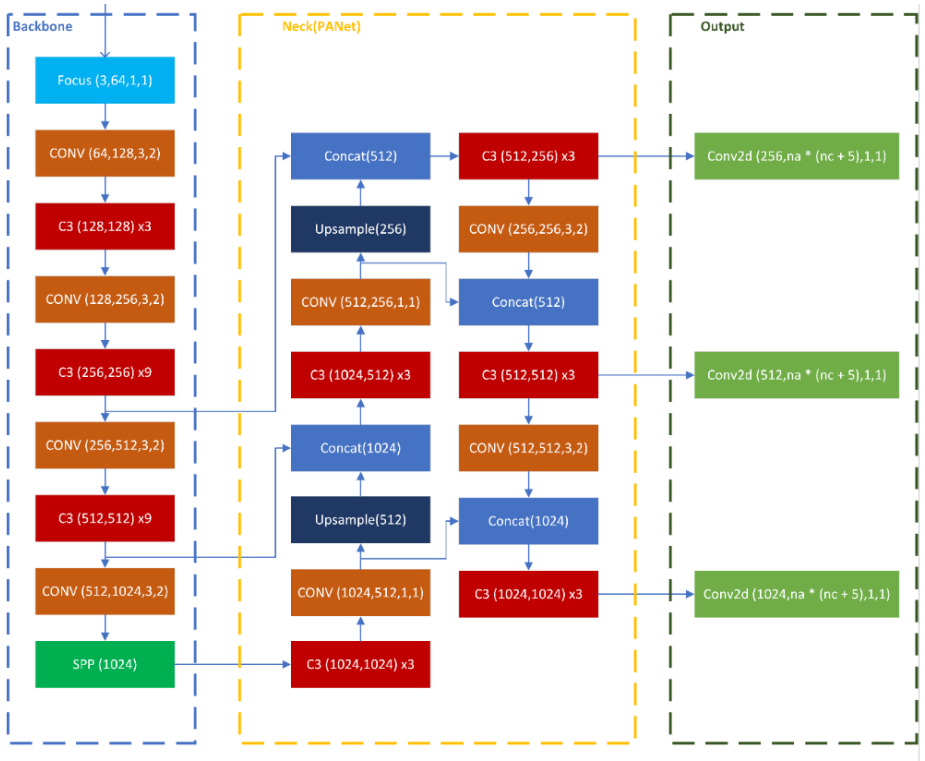
```
epochs: 150
batch_size: 64
imgsz: 416
rect: false
resume: false
nosave: false
noval: false
noautoanchor: false
noplots: false
evolve: null
bucket: ''
cache: ram
image_weights: false
device: ''
multi_scale: false
single_cls: false
optimizer: SGD
sync_bn: false
workers: 8
project: runs/train
name: exp
exist_ok: false
quad: false
cos_lr: false
label_smoothing: 0.0
patience: 100
freeze:
- 0
save_period: -1
seed: 0
local_rank: -1
entity: null
upload_dataset: false
bbox_interval: -1
artifact_alias: latest
save_dir: runs/train/exp
```



모델 학습

Drone Detection

모델로 학습하기



	from	n	params	module	arguments
0	-1	1	3520	models.common.Conv	[3, 32, 6, 2, 2]
1	-1	1	18560	models.common.Conv	[32, 64, 3, 2]
2	-1	1	18816	models.common.C3	[64, 64, 1]
3	-1	1	73984	models.common.Conv	[64, 128, 3, 2]
4	-1	2	115712	models.common.C3	[128, 128, 2]
5	-1	1	295424	models.common.Conv	[128, 256, 3, 2]
6	-1	3	625152	models.common.C3	[256, 256, 3]
7	-1	1	1180672	models.common.Conv	[256, 512, 3, 2]
8	-1	1	1182720	models.common.C3	[512, 512, 1]
9	-1	1	656896	models.common.SPPF	[512, 512, 5]
10	-1	1	131584	models.common.Conv	[512, 256, 1, 1]
11	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
12	[-1, 6]	1	0	models.common.Concat	[1]
13	-1	1	361984	models.common.C3	[512, 256, 1, False]
14	-1	1	33024	models.common.Conv	[256, 128, 1, 1]
15	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
16	[-1, 4]	1	0	models.common.Concat	[1]
17	-1	1	90880	models.common.C3	[256, 128, 1, False]
18	-1	1	147712	models.common.Conv	[128, 128, 3, 2]
19	[-1, 14]	1	0	models.common.Concat	[1]
20	-1	1	296448	models.common.C3	[256, 256, 1, False]
21	-1	1	590336	models.common.Conv	[256, 256, 3, 2]
22	[-1, 10]	1	0	models.common.Concat	[1]
23	-1	1	1[illegible]20	models.common.C3	[512, 512, 1, False]
24	[17, 20, 23]	1	16182	models.yolo.Detect	[1, [[10, 13, 16, 30, 33, 23], [

512]]

Backbone

Neck

Head

모델 학습

Drone Detection

모델로 학습하기

```
Model summary: 270 layers, 7022326 parameters, 7022326 gradients, 15.9 GFLOPs
```

```
150 epochs completed in 4.766 hours.
Optimizer stripped from runs/train/exp/weights/last.pt, 14.3MB
Optimizer stripped from runs/train/exp/weights/best.pt, 14.3MB
```

```
# export model's weights
from google.colab import files
files.download('./runs/train/exp/weights/last.pt')
files.download('./runs/train/exp/weights/best.pt')
```

train의 마지막 epoch 가중치(last.pt)를 저장
train 중에 가장 좋게 기록된 epoch 가중치(best.pt)를 저장

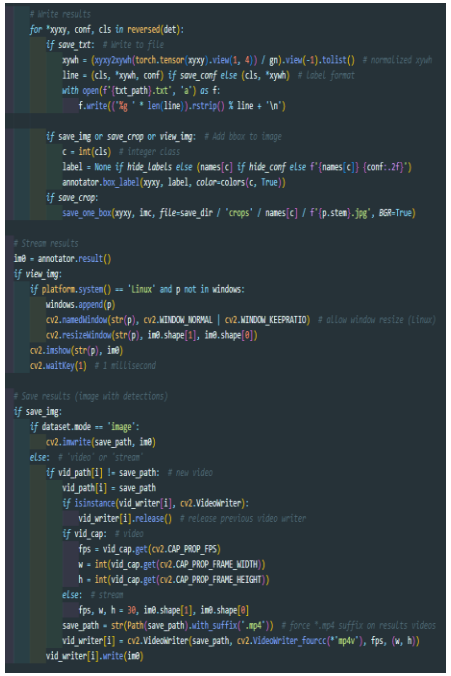
모델 학습

Drone Detection

학습결과로 추론하기

best.pt

```
!python detect.py --weights runs/train/exp/weights/best.pt --img 416 --conf 0.1 --data=/content/datasets/drone-1/data.yaml --source {dataset.location}/test/images
```



학습한 결과(best.pt)로 test image 데이터 사용하여 detection 추론하기

```
1/848 /content/datasets/drone-1/test/images/0004_jpg.rf.7f647a5b8084c3f4c08b7e4660aa8ef5.jpg: 416x416 1 drone, Done. (0.007s)
2/848 /content/datasets/drone-1/test/images/0010_jpg.rf.6fefb73c856b491ca3bc59de7ea5f1ab.jpg: 416x416 2 drones, Done. (0.007s)
3/848 /content/datasets/drone-1/test/images/0015_jpg.rf.b6e4220a2f2d5769a367931c9d3a6ba0.jpg: 416x416 1 drone, Done. (0.006s)
4/848 /content/datasets/drone-1/test/images/0016_jpg.rf.9a616d894666af100c11a589a9c7a11b.jpg: 416x416 1 drone, Done. (0.006s)
5/848 /content/datasets/drone-1/test/images/0020_jpg.rf.39b504214d68b00ef113af5ae9b4564a.jpg: 416x416 1 drone, Done. (0.007s)
6/848 /content/datasets/drone-1/test/images/0032_jpg.rf.8a763fbc0f40fdf65fa87cacfcd97686.jpg: 416x416 1 drone, Done. (0.006s)
7/848 /content/datasets/drone-1/test/images/0044_jpg.rf.de91ec0dfac8ebc1cc35b768f0b0cb16.jpg: 416x416 1 drone, Done. (0.007s)
8/848 /content/datasets/drone-1/test/images/0046_jpg.rf.bda603adf98403d7392bbbfba4b2a0a3.jpg: 416x416 1 drone, Don
9/848 /content/datasets/drone-1/test/images/0048_jpg.rf.18705db337313ebec7a076df1ff94f0c.jpg: 416x416 1 drone, Don
10/848 /content/datasets/drone-1/test/images/0049_jpg.rf.fc57badd07ffb85fa2c286a1a80399a3.jpg: 416x416 2 drones D
11/848 /content/datasets/drone-1/test/images/0050_jpg.rf.d94b38b4e8505a8689532527278b84e1.jpg: 416x416 1 drone, Done. (0.006s)
12/848 /content/datasets/drone-1/test/images/0054_jpg.rf.86d40e147323166e81b7bdad2baf90a3.jpg: 416x416 1 drone, Done. (0.006s)
13/848 /content/datasets/drone-1/test/images/0059_jpg.rf.6d201a081412a8525e21cfcb8399dabe.jpg: 416x416 1 drone, Done. (0.007s)
14/848 /content/datasets/drone-1/test/images/0081_jpg.rf.3bae22052b46875fa59fe760b864e362.jpg: 416x416 2 drones Done. (0.008s)
15/848 /content/datasets/drone-1/test/images/0081_jpg.rf.e8394d7a12abbd098099bfdb08bffbc6.jpg: 416x416 2 drones Done. (0.007s)
16/848 /content/datasets/drone-1/test/images/0082_jpg.rf.7ae8e754f9eeb58dcdfc6df8ac2ddb5c.jpg: 416x416 1 drone, Done. (0.006s)
17/848 /content/datasets/drone-1/test/images/0082_jpg.rf.f4c3c1f90e067b6a071aac8c3e2af535.jpg: 416x416 1 drone, Done. (0.006s)
18/848 /content/datasets/drone-1/test/images/0085_jpg.rf.7c1cfcadc68ebba7f110e68fdedaf1b3.jpg: 416x416 1 drone, Done. (0.006s)
19/848 /content/datasets/drone-1/test/images/0086_jpg.rf.0434723d335204a029d8dd82c582e744.jpg: 416x416 1 drone, Done. (0.006s)
```

data.yaml

```
names:
- drone
```

'drone'으로 추론

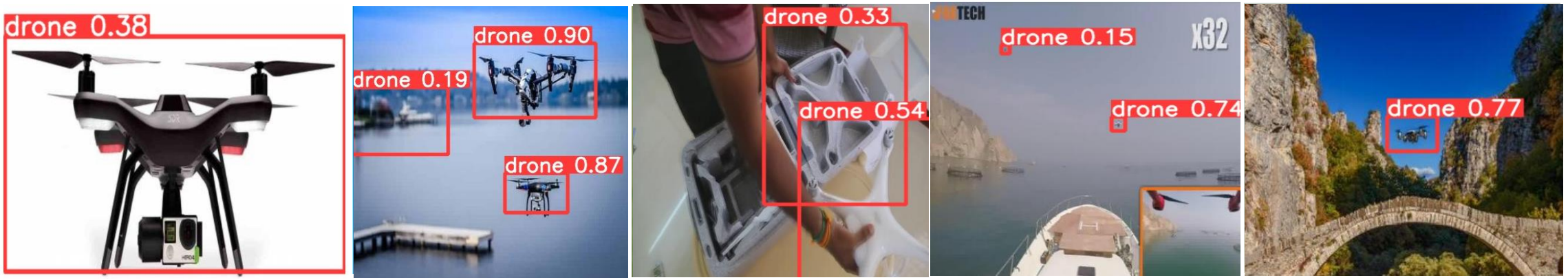
모델 학습

Drone Detection

추론 결과 확인

```
import glob
from IPython.display import Image, display

for imageName in glob.glob('/content/yolov5/runs/detect/exp/*.jpg'): #assuming JPG
    display(Image(filename=imageName))
    print("₩n")
```

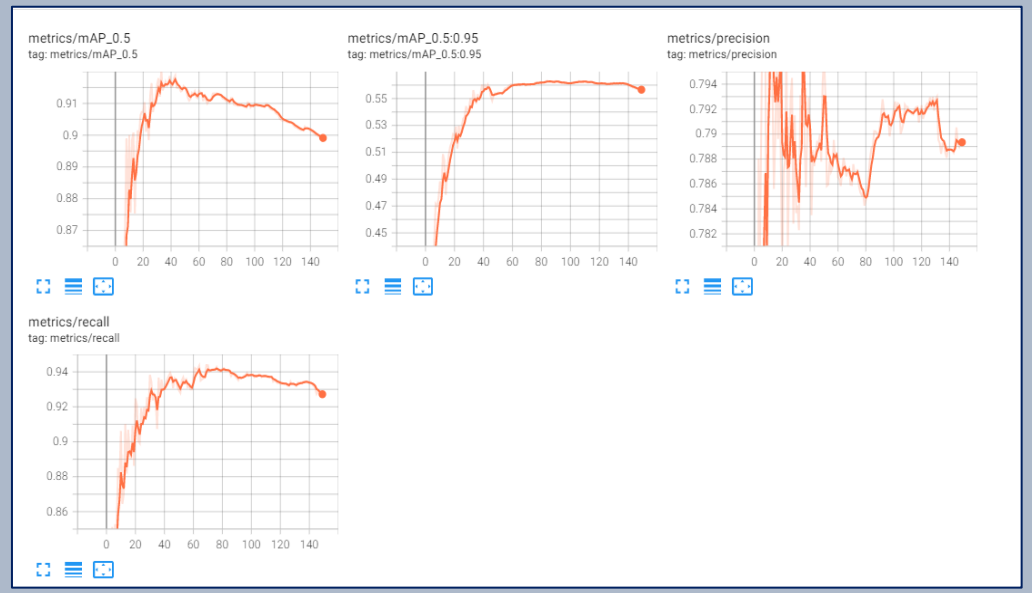
추론 결과 시각화



완벽하진 않지만, 그래도 **drone detection**이 가능해짐!!!

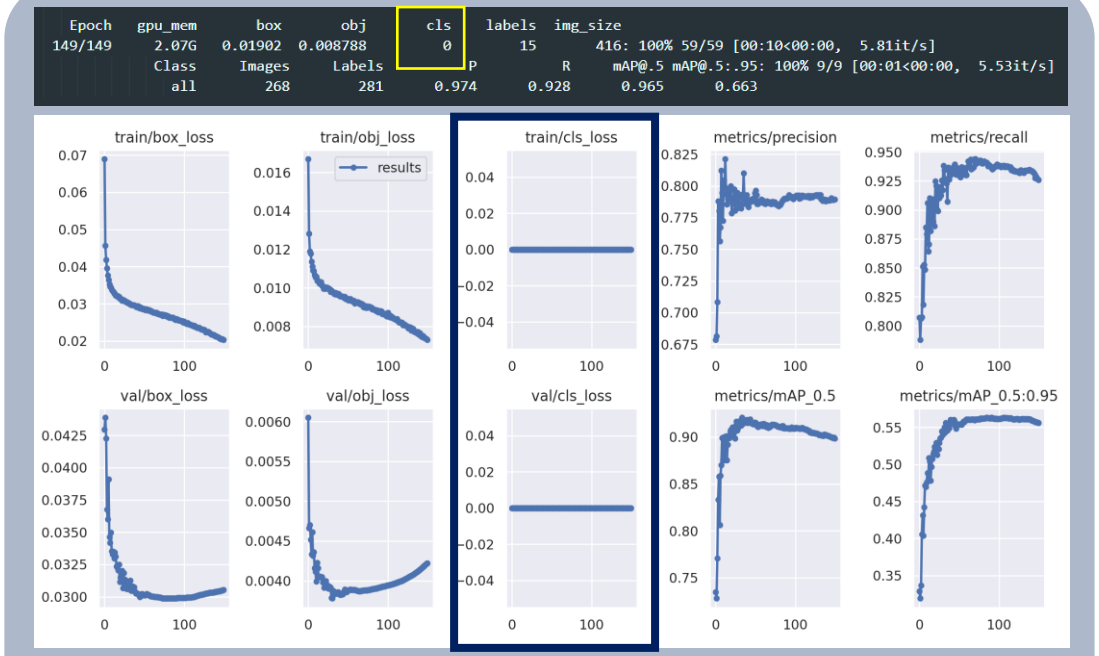
결과

Drone Detection



Tensorboard 시각화

mAP : 객체 탐지 **정확도** 평가 지표

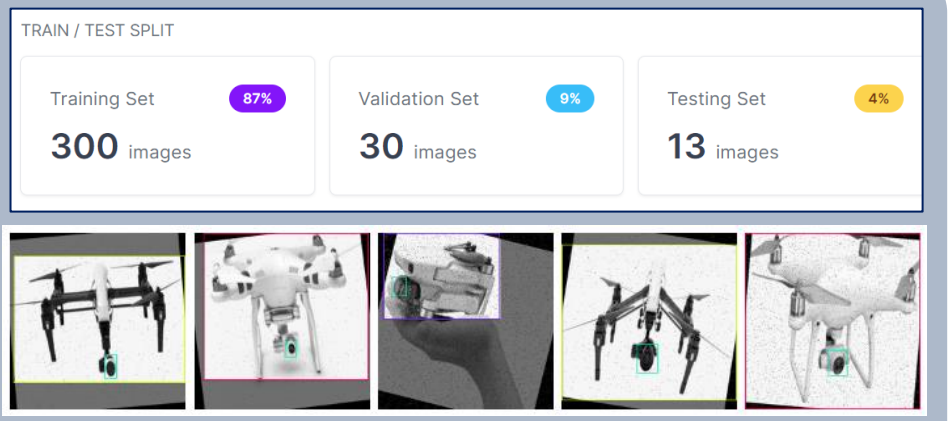


학습함에 따라 전체적으로 loss는 감소하고 정확도가 증가

모델 학습

Drone Classificatioin

데이터 로드하기



class : 4
[DJI Phantom, DJI Mavic, DJI Inspire, camera]

EXPORT

```
# roboflow에서 데이터셋 가져오기
rf = Roboflow(api_key="fkGLXDbaDekHbCOt6y7y")
project = rf.workspace("5-jcfgi").project("dji-fw3wu")
dataset = project.version(1).download("yolov5")
```

```
loading Roboflow workspace...
loading Roboflow project...
Downloading Dataset Version Zip in /content/datasets/DJI-1 to yolov5pytorch: 100% [14700352 / 14700352] bytes

Extracting Dataset Version Zip to /content/datasets/DJI-1 in yolov5pytorch:: 100%|██████████| 698/698 [00:00<00:00, 4753.66it/s]
```

Roboflow에서 생성한 classification 데이터셋 다운로드

모델 학습

Drone Classificatioin

모델 학습하기

```
!python train.py --img 416 --batch 16 --epochs 150 --data /content/datasets/drone-1/data.yaml --weights yolov5s.pt --cache
```

```yaml
names:
- drone
- DJI Phantom
- DJI Mavic
- DJI Inspire
- camera

nc: 5 # 0부터 4까지

train:
  - ../datasets/train/images
  - ../datasets/EAI-1/train/images

val:
  - ../datasets/valid/images
  - ../datasets/EAI-1/valid/images
```

Detection에 사용한 데이터셋과 merge하여 train data 학습시키기

모델 학습

Drone Classificatioin

모델로 추론하기

```
!python detect.py --weights runs/train/exp3/weights/best.pt --img 416 --data=/content/datasets/drone-1/data.yaml --source /content/datasets/drone-1/test/images
```

```
image 24/149 /content/datasets/drone-1/test/images/121_jpg.rf.3ff3329f936cb4273450c62284b222f4.jpg: 416x416 1 drone, 8.3ms
image 25/149 /content/datasets/drone-1/test/images/123_jpg.rf.87b695b26b2ca39517f072ad8dcee6c1.jpg: 416x416 (no detections), 7.6ms
image 26/149 /content/datasets/drone-1/test/images/15_jpg.rf.00c4f1c2987ef8c438859ac5161c81b5.jpg: 416x416 1 DJI Inspire, 1 camera, 8.0ms
image 27/149 /content/datasets/drone-1/test/images/1_jpg.rf.444746208eb676b874c31544417398bb.jpg: 416x416 1 DJI Inspire, 1 camera, 8.1ms
image 28/149 /content/datasets/drone-1/test/images/56_jpg.rf.21705cbc5b10ee676b4fd9f388938cd8.jpg: 416x416 1 DJI Phantom, 1 drone, 8.2ms
image 29/149 /content/datasets/drone-1/test/images/68_jpg.rf.229a70006091b8903cb7ccc4f817a645.jpg: 416x416 1 DJI Phantom, 8.6ms
image 30/149 /content/datasets/drone-1/test/images/69_jpg.rf.e8c1936e4c3ecc4b8aaebdd7a7594e03.jpg: 416x416 1 DJI Phantom, 1 camera, 8.1ms
image 31/149 /content/datasets/drone-1/test/images/79_jpg.rf.a2eb7306232a200b73cd88ef42e472ff.jpg: 416x416 1 DJI Phantom, 1 camera, 7.5ms
image 32/149 /content/datasets/drone-1/test/images/81_jpg.rf.98fd39c6614366424e6e8a918edcfcd6.jpg: 416x416 1 DJI Phantom, 7.8ms
image 33/149 /content/datasets/drone-1/test/images/8_jpg.rf.753c5a71ea37d762032efbb9c4e1c9d7.jpg: 416x416 1 DJI Phantom, 7.8ms
image 34/149 /content/datasets/drone-1/test/images/foto00233_jpg.rf.6fa26ec7649ed9a71ed7cf3f56d1a47f.jpg: 416x416 1 drone, 7.8ms
image 35/149 /content/datasets/drone-1/test/images/foto00378_jpg.rf.1fbc4bf22ceaccd85c39a65d940469eb.jpg: 416x416 1 drone, 8.2ms
image 36/149 /content/datasets/drone-1/test/images/foto01016_jpg.rf.99412a8bc19ff1a5d6e7d540db741077.jpg: 416x416 1 drone, 8.7ms
image 37/149 /content/datasets/drone-1/test/images/foto01161_jpg.rf.af5d4b6abd1ec1973ff536f849781b8b.jpg: 416x416 1 drone, 8.0ms
image 38/149 /content/datasets/drone-1/test/images/foto01190_jpg.rf.1f6bc6436634616543b10ec4276ae9ab.jpg: 416x416 1 drone, 7.6ms
image 39/149 /content/datasets/drone-1/test/images/foto01451_jpg.rf.61b243fad2ed6af26d236c692388294f.jpg: 416x416 1 drone, 7.8ms
image 40/149 /content/datasets/drone-1/test/images/foto01944_jpg.rf.e4ef94e4149e8cfac5a8ff35c66a8916.jpg: 416x416 1 drone, 7.8ms
image 41/149 /content/datasets/drone-1/test/images/foto02756_jpg.rf.68dafd51d1353fcb929cb9d6118d6036.jpg: 416x416 (no detections), 7.6ms
image 42/149 /content/datasets/drone-1/test/images/foto04496_jpg.rf.6994d29427ab1a6c19c1e3e13304272c.jpg: 416x416 1 drone, 8.0ms
```

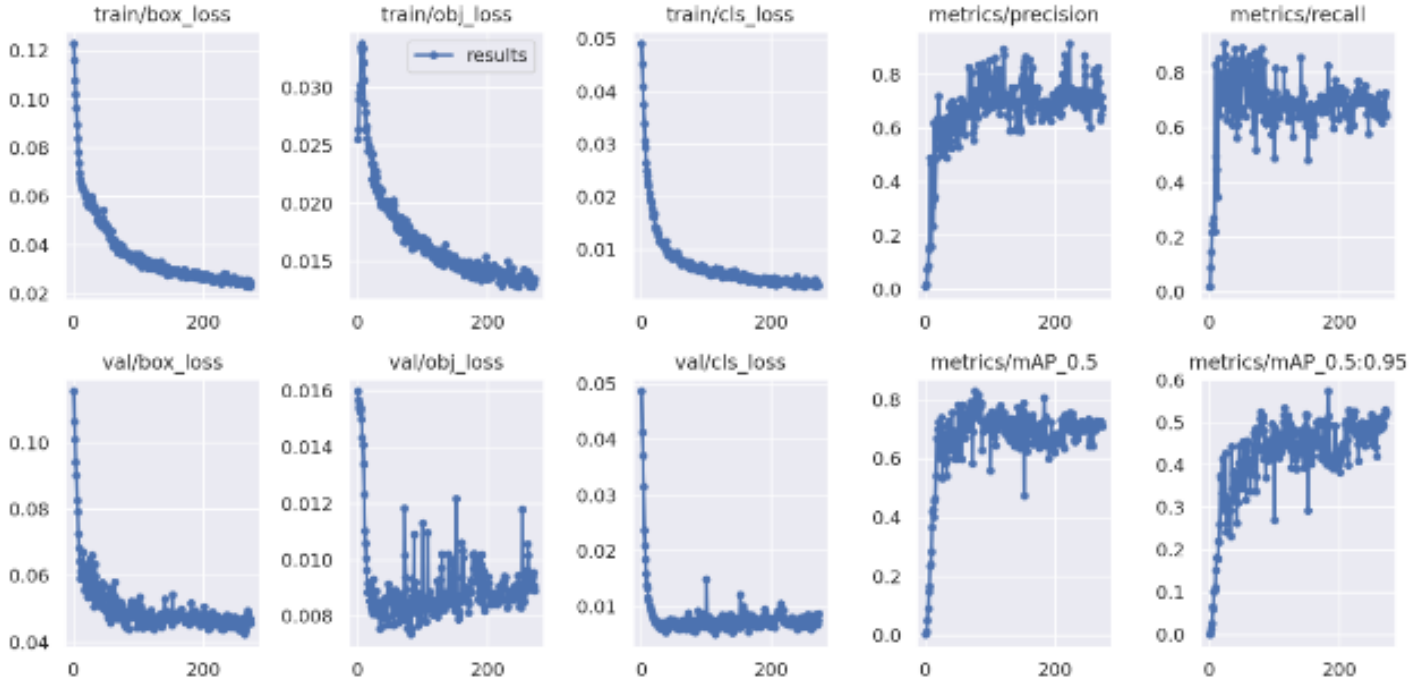
모델 학습

Drone Classificatioin

모델로 추론하기

결과

Drone Classificatioin

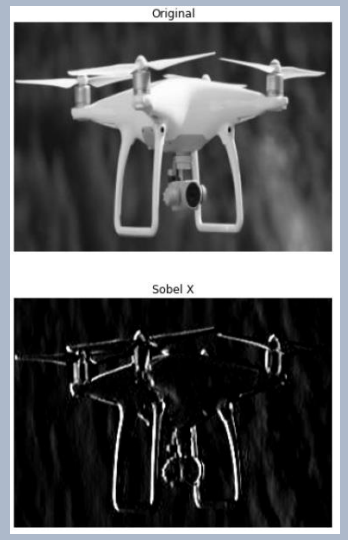


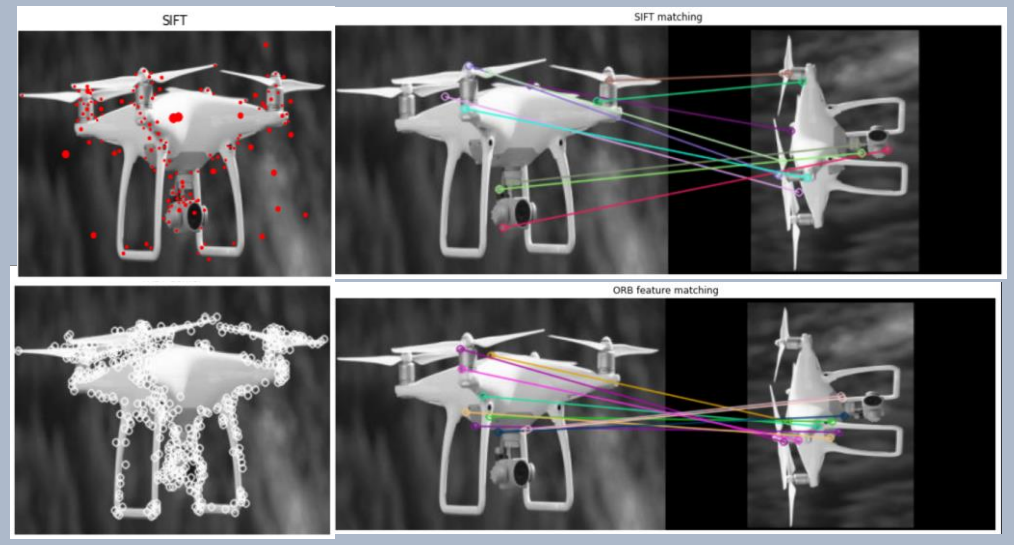
Drone detection에 비해 성능이 떨어짐

Object Tracking

Drone Feature Mapping

- Harris, SIFT, FAST, ORB : 드론 이미지에서 객체 특징 추출 및 매칭하는 Image Processing



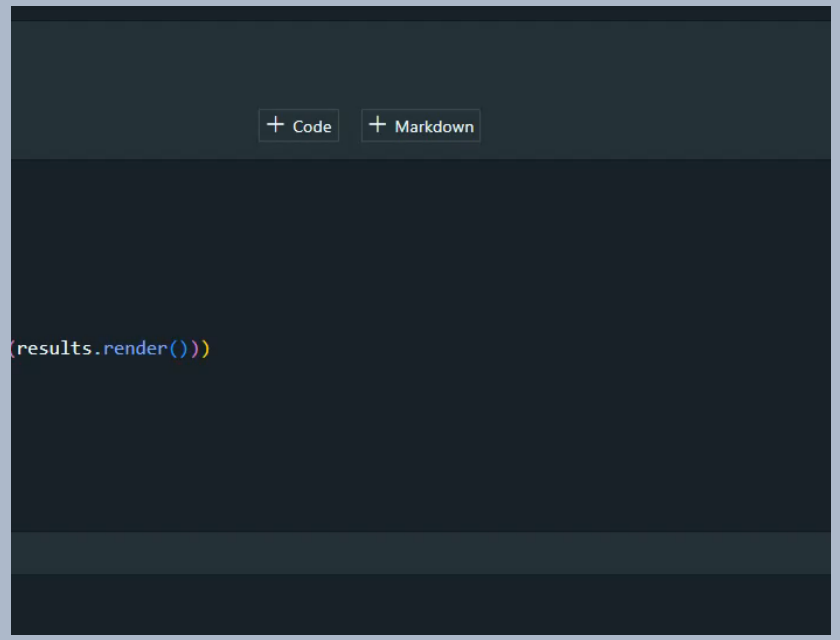


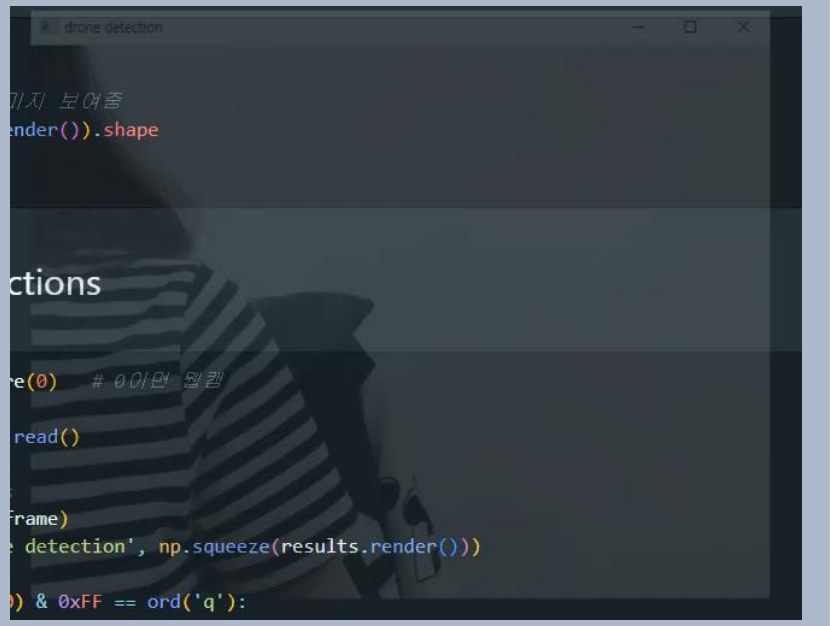
결과

Real-time detection

Using webcam

```
# download pre-trained model(yolov5) from pytorch hub
# model = torch.hub.load('ultralytics/yolov5', 'yolov5s')    # 0.50 airplane
# model = torch.hub.load('ultralytics/yolov5', 'custom', path='yolov5/best.pt', force_reload=True)

model = torch.hub.load('ultralytics/yolov5', 'custom',path='yolov5/best.pt', force_reload=True)

img = "https://assets.weforum.org/article/image/-SEFZUUSxWOm74QkKw80M7_6dvG3c6bMB6wAHvb9YVA.jpg"
results = model(img)

results.print()
results.show()
%matplotlib inline
plt.imshow(np.squeeze(results.render()))
plt.axis('off')
plt.savefig('detection01.jpg', dpi=300, bbox_inches='tight')
plt.show()

results.xyxy
results.render()
results.show()  # 이미지 보여줌
np.squeeze(results.render()).shape

cap = cv2.VideoCapture(0)   # 0이면 웹캠
while cap.isOpened():
    ret, frame = cap.read()

    # Make detections
    results = model(frame)
    cv2.imshow('drone detection', np.squeeze(results.render()))

    if cv2.waitKey(10) & 0xFF == ord('q'):
        break
cap.release()
cv2.destroyAllWindows()
```





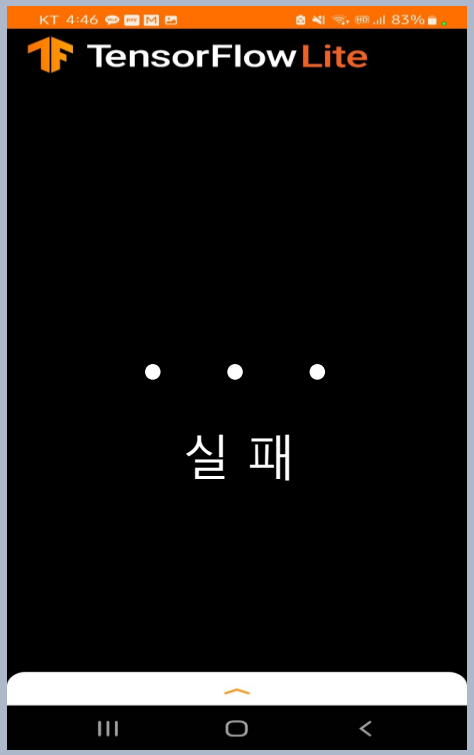
Application

tflite- web

```
import tensorflow as tf
!python export.py - -weights ./best.pt - -imgsz 416 - -include tflite
```

best-fp16.tflite

```
package org.tensorflow.lite.examples.detection.tflite;

import android.content.res.AssetManager;

import java.io.IOException;

public class DetectorFactory {
    public static YoloV5Classifier getDetector(
            final AssetManager assetManager,
            final String modelFilename)
            throws IOException {
        String labelFilename = null;
        boolean isQuantized = false;
        int inputSize = 0;
        int[] output_width = new int[]{0};
        int[][] masks = new int[][]{{0}};
        int[] anchors = new int[]{0};

        if (modelFilename.equals("best-fp16.tflite")) {
            labelFilename = "file:///android_asset/customclasses.txt";
            isQuantized = false;
            inputSize = 416;
            output_width = new int[]{40, 20, 10};
            masks = new int[][]{{0, 1, 2}, {3, 4, 5}, {6, 7, 8}};
            anchors = new int[]{
                    10,13, 16,30, 33,23, 30,61, 62,45, 59,119, 116,90, 156,198, 373,326
            };

        return YoloV5Classifier.create(assetManager, modelFilename, labelFilename, isQuantized,
                inputSize);
    }

}
```



Application

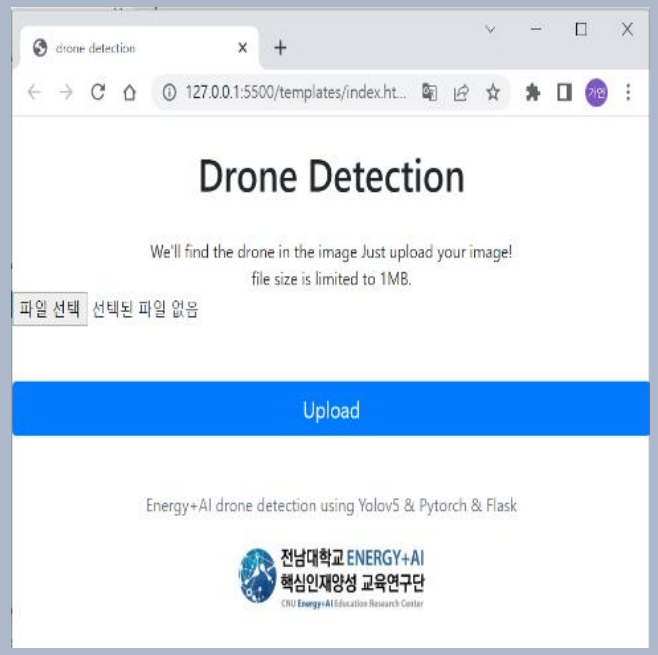
Flask - web

```python
mport argparse
mport io
mport os
rom PIL import Image
mport torch
rom flask import Flask, render_template, request, redirect

pp = Flask(__name__)
app.route("/", methods=["GET", "POST"])
ef predict():
    if request.method == "POST":
        if "file" not in request.files:
            return redirect(request.url)
        file = request.files["file"]
        if not file:
            return 'None'

        img_bytes = file.read()
        img = Image.open(io.BytesIO(img_bytes))
        results = model(img, size=640)

        # for debugging
        # data = results.pandas().xyxy[0].to_json(orient="records")
        # return data

        results.render()  # updates results.imgs with boxes and labels
        for img in results.imgs:
            img_base64 = Image.fromarray(img)
            img_base64.save("static/image0.jpg", format="JPEG")
        return redirect("static/image0.jpg")

    return render_template("index.html", image_file="templates/logo.png")


f __name__ == "__main__":
    parser = argparse.ArgumentParser(description="Flask app exposing yolov5 models")
    parser.add_argument("--port", default=5000, type=int, help="port number")
    args = parser.parse_args()

    model = torch.hub.load('ultralytics/yolov5', 'custom', path='DJI.pt', force_reload=True, autoshape=True)
    model.eval()
    app.run(host="0.0.0.0", port=args.port)  # debug=True causes Restarting with stat
```

```
(drone) C:\Users\Enc\Desktop\E+AI\AIM\Flask_webapp>python app.py
Downloading: "https://github.com/ultralytics/yolov5/zipball/master" to C:\Users\Enc/.cache\torch\hub\master.zip
YOLOv5  2022-8-22 Python-3.10.4 torch-1.12.0+cpu CPU

Fusing layers...
Model summary: 213 layers, 7023610 parameters, 0 gradients, 15.8 GFLOPs
Adding AutoShape...
 * Serving Flask app 'app' (lazy loading)
 * Environment: production
   WARNING: This is a development server. Do not use it in a production deployment.
   Use a production WSGI server instead.
 * Debug mode: off
 * Running on all addresses (0.0.0.0)
   WARNING: This is a development server. Do not use it in a production deployment.
 * Running on http://127.0.0.1:5000
 * Running on http://10.10.101.47:5000 (Press CTRL+C to quit)
```

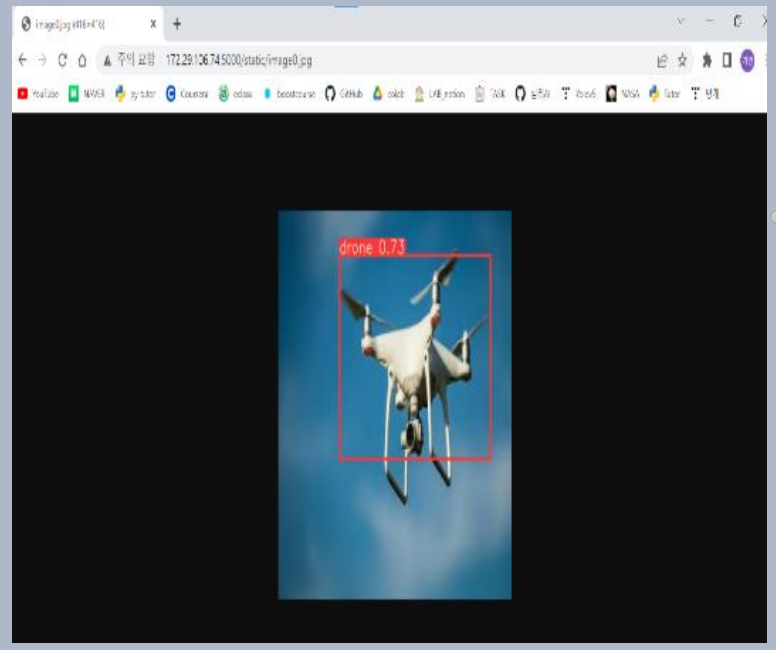
Application

Flask - web



Main 화면

Image 파일 업로드



Detection 결과 출력

결론

YOLO 모델로 드론 이미지 및 영상데이터를 학습시켜
객체를 감지하고 종류 구분이 가능한 프로토타입 AI모델을 제작함

학습한 모델(best.pt)을 바탕으로
flask를 활용한 detection web-애플리케이션에 적용함

충분하지 않는 데이터셋, GPU 용량 문제, 부족한 지식 등으로
인한 다양한 모델 적용 및 학습 시 튜닝시도 부족

Reference

- https://github.com/ultralytics/yolov5
- Unauthorized Unmanned Aerial Vehicle Detection using YOLOv5 and Transfer Learning, doi:10.20944/preprints202202.0185.v1
- https://github.com/AarohiSingla/TFLite-Object-Detection-Android-App-Tutorial-Using-YOLOv5
- https://github.com/ViAsmit/YOLOv5-Flask
- YOLOv4: Optimal Speed and Accuracy of Object Detection
- https://www.youtube.com/watch?v=yqkISICHH-U
- https://www.youtube.com/watch?v=WQeoO7MI0Bs
- https://www.youtube.com/results?search_query=yolov5+tflite
- https://www.tensorflow.org/lite/guide?hl=ko

감사합니다

전남대학교 ENERGY+AI
핵심인재양성 교육연구단
CNU Energy+AI Education Research Center

김주완(인공지능학부)
서가연(인공지능학부)
조하영(인공지능학부)