

드론 감지 및 분류 AI모델



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ENERGY+AI 2022 하계 마이크로 캡스톤 디자인



목 차

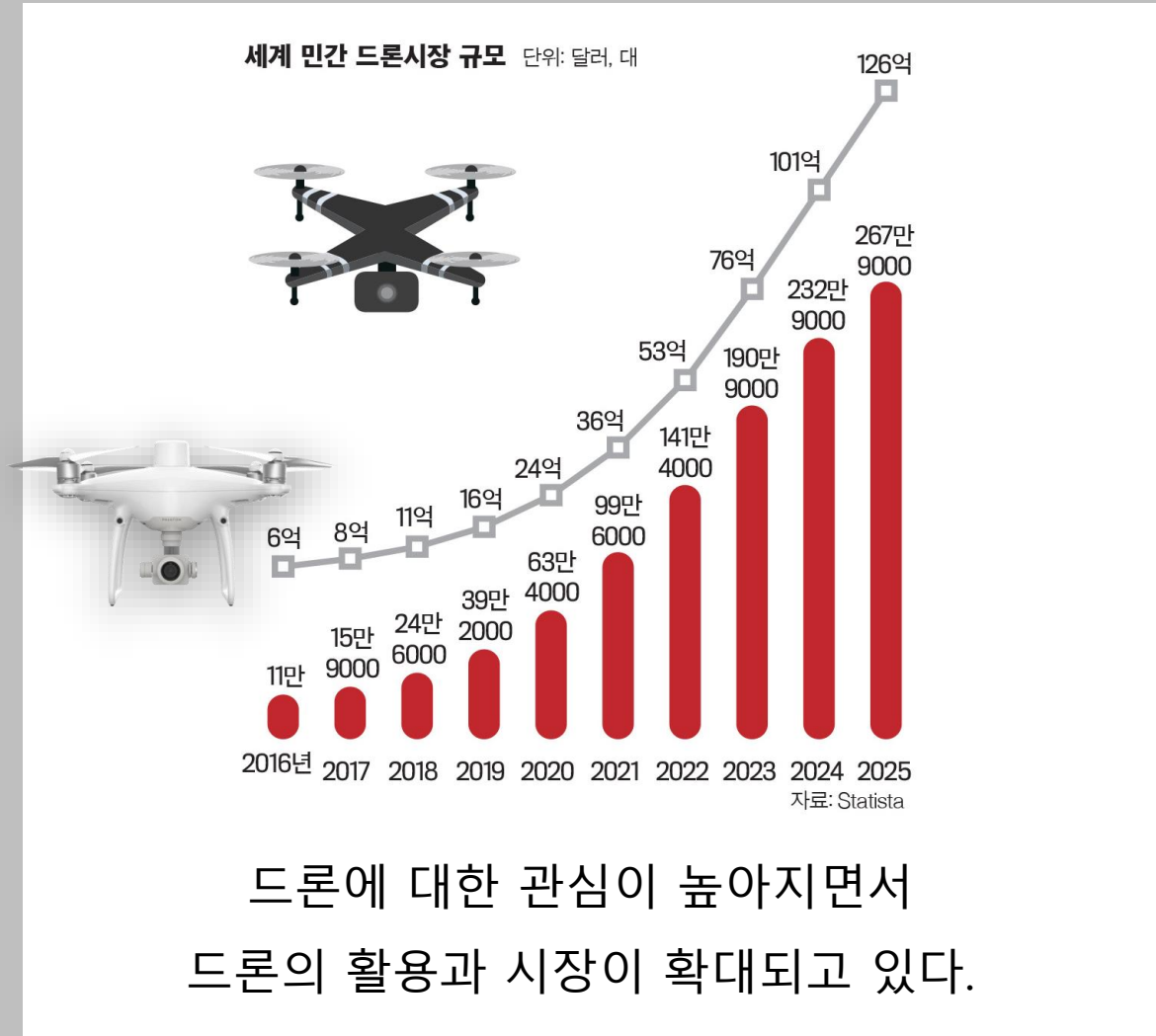


연구 배경과 목표

데이터 수집과 전처리 과정

모델링과 결과

연구 배경



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

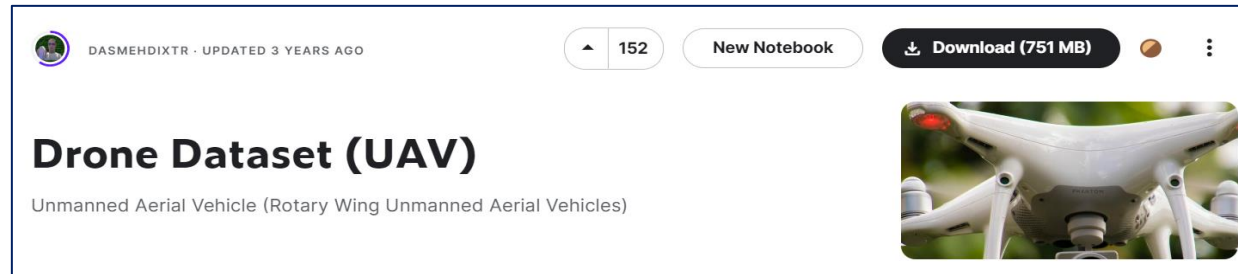


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CNU Energy+AI Education Research Center

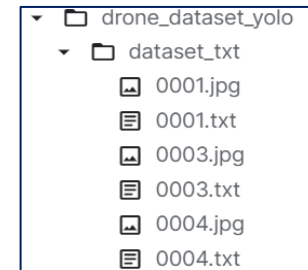
드론 데이터 수집 방법

플랫폼 활용

kaggle

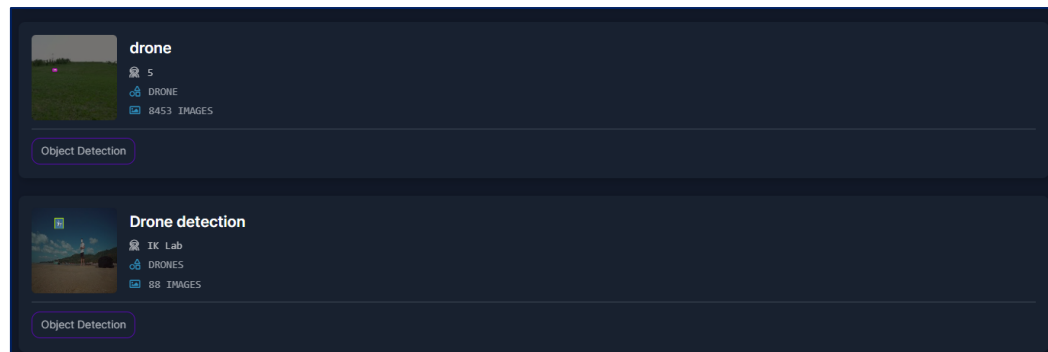


<https://www.kaggle.com/datasets/dasmehdixtr/drone-dataset-uav>



라벨링 된 데이터 다운로드

roboflow

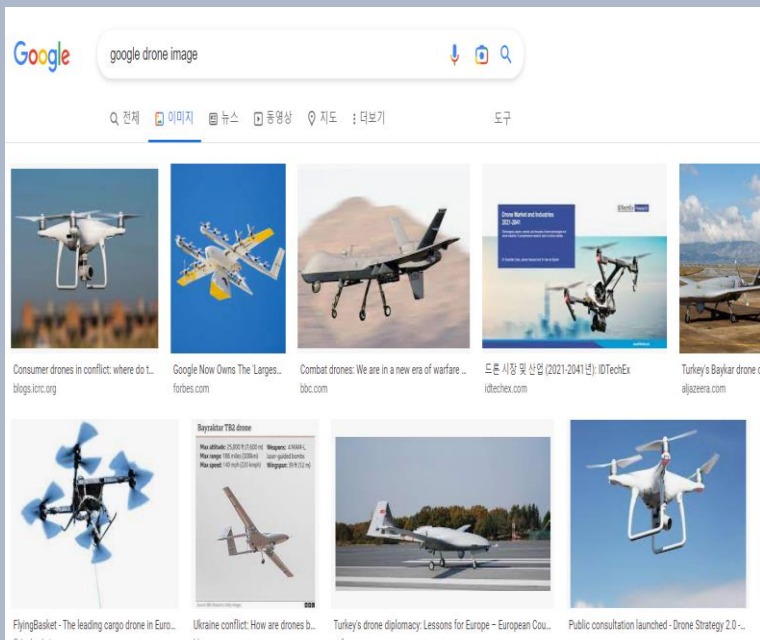


<https://universe.roboflow.com/search?q=drone>

데이터 수집

드론 데이터 수집 방법

워크롤링



이미지 크롤링으로 드론 데이터 수집

```
import urllib.request
from http import HTTPStatus

# 获取url 3. 获取url
url = "https://www.kinoo.com/free/git/1/"

req = urllib.request.Request(url)
res = urllib.request.urlopen(req).read()

# 4. 从url 中获取 json 数据 5. 从url 中获取 json 数据
soup = BeautifulSoup(res, "html.parser")

# 6. 遍历 json 数据
img_urls = soup.findAll("div", {"class": "list"})

# 7. 遍历 json 数据
img_urls = soup.findAll("div", {"class": "list"})

# 8. 遍历 json 数据
img_urls = soup.findAll("div", {"class": "list"})
```

```
import os
import sys
import urllib.request
from selenium import webdriver
from time import sleep
from bs4 import BeautifulSoup

keyword = input('검색어 : ')
maxImages = int(input('다운로드 시도할 최대 이미지 수 : '))

# 로컬컴퓨터의 디렉토리 생성을 위해서 crawled_img폴더 안에 폴더 하나 더 생성
# 폴더이름에는 '검색어' 키워드, '이미지' 개수, '폴더' 종류
path = 'crawled_img/' + keyword + '_' + str(maxImages)

try:
    # 폴더에는 이미 생성되었는지 확인
    if not os.path.exists(path):
        os.makedirs(path)
    # 로컬컴퓨터의 경로 출력
    print('다운로드 시도할 디렉토리 : ' + path)
except OSError:
    print('os error')
    sys.exit(0)

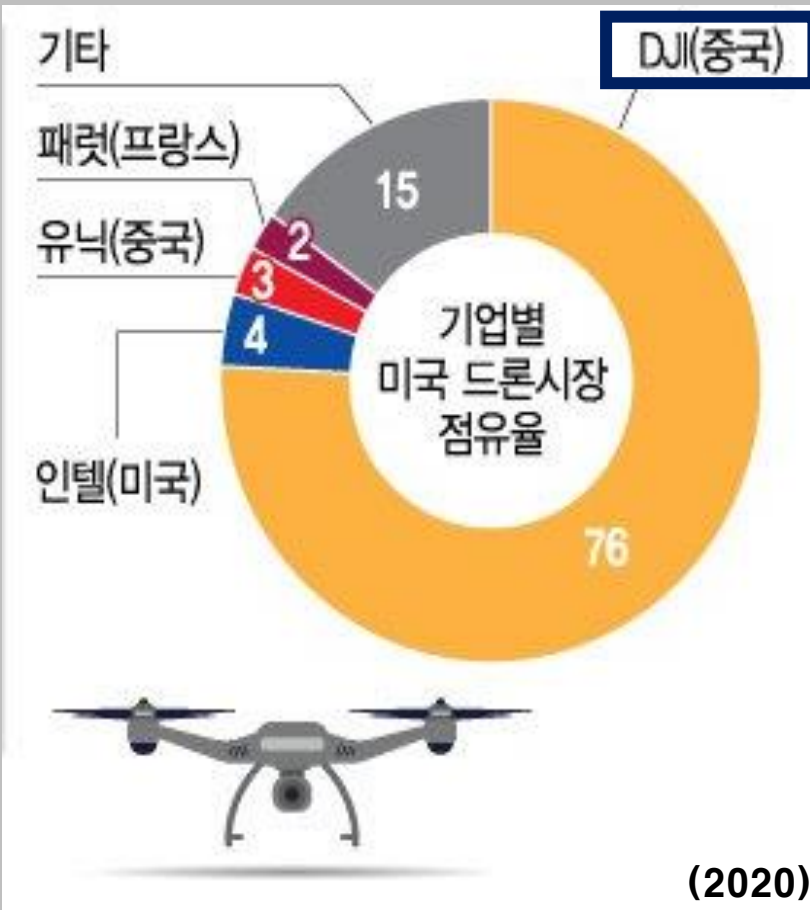
pages = int((maxImages-1)/100)+1 # 1번 ~ 100번 까지되는 이미지 수(100을 넘지않도록 제한)
imgCount = 0 # 이미지를 다운로드 할 이미지 수
success = 0 # 성공한 이미지 수
finish = False # 이미지가 다운로드 되었는지 여부

# 크롬 드라이버 설정
# 크롬 브라우저를 실행하고 이미지 주소에 get을 사용 하면, 기본값으로 user-agent: 불려가게 된다
chrome_options = webdriver.ChromeOptions()
```

[illegible]

드론 데이터 종류

미국 드론시장 내 점유율(%)



DJI mavic



DJI Phantom



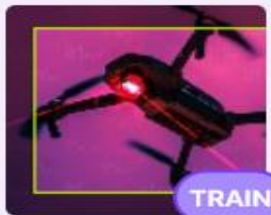
DJI Inspire

데이터 라벨링

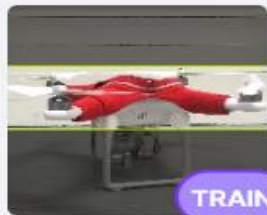
- **roboflow** 플랫폼 이용
- 수집한 데이터 셋: yolov5-pytorch로 export하여 모델에 적용

Detection

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



pic_158.jpg



pic_852.jpg



pic_226.jpg

Classification

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



데이터 전처리 및 증강



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전처리 : `resize(416*416)`

original



resized



416*416

train : valid : test = 70 : 20 : 10

dataset

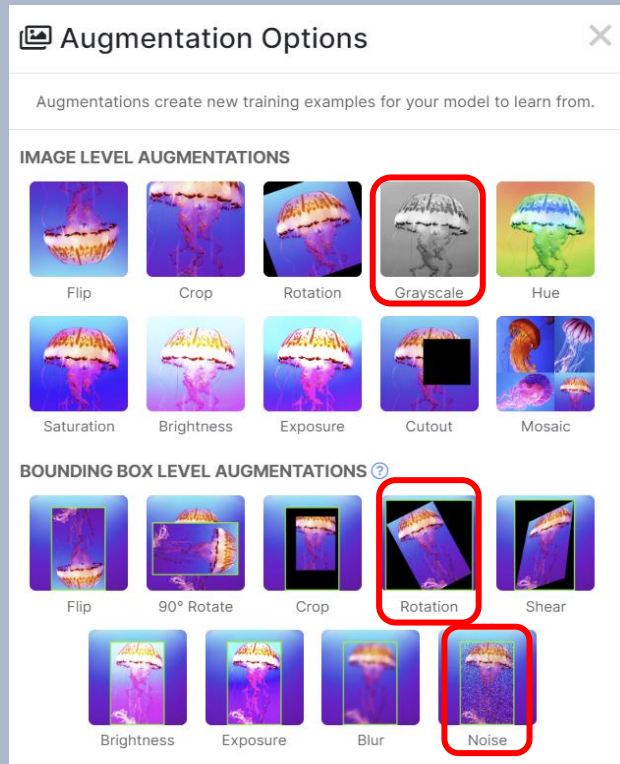


```
!pip install roboflow

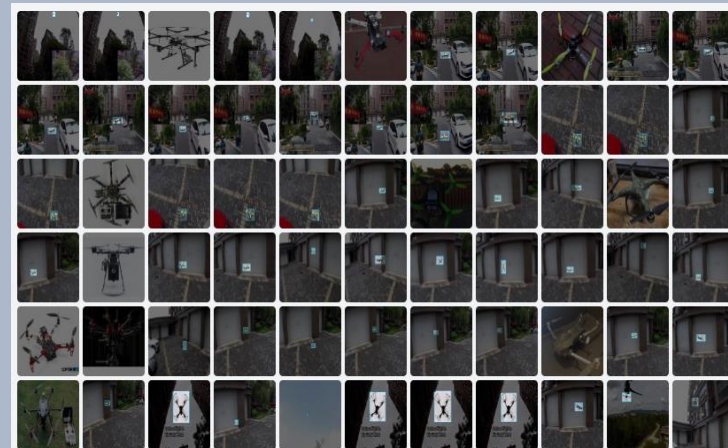
from roboflow import Roboflow
rf = Roboflow(api_key="████████████████████")
project = rf.workspace("5-jcfigi").project("drone-qfysh")
dataset = project.version(2).download("yolov5")
```

데이터 전처리 및 증강

데이터 증강(x3)



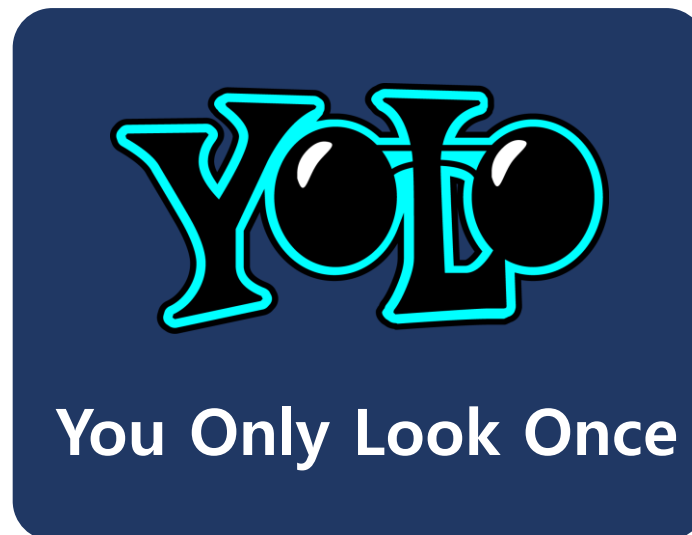
- Rotation ($-15^{\circ} \sim +15^{\circ}$)
- Grayscale (Apply to 30% of images)
- Noise (5% of pixels)



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

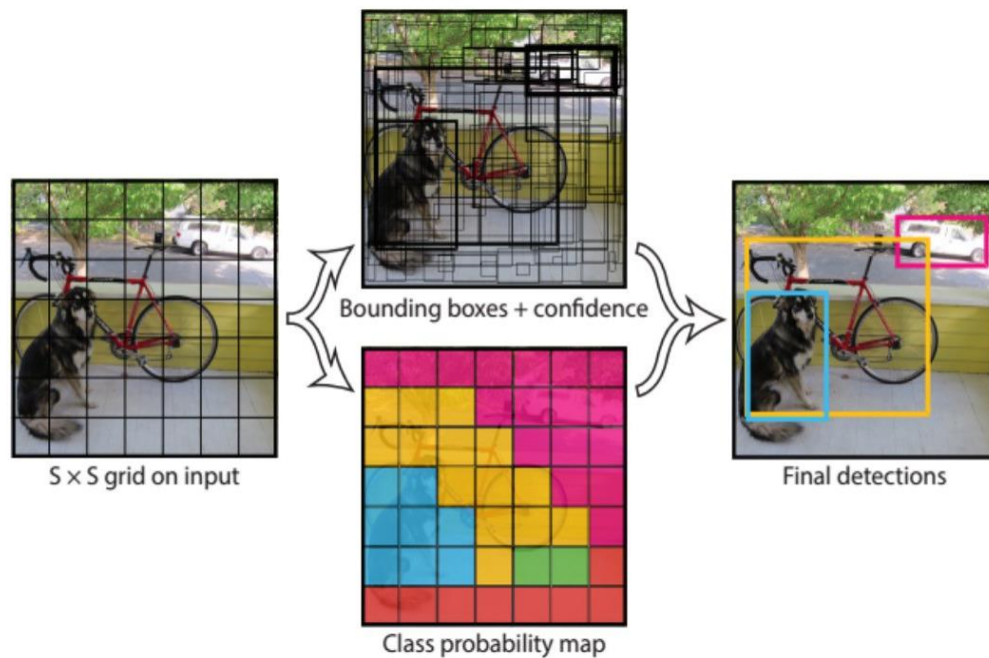
기존의 Object detection 알고리즘은 real-time으로 사용하기에는 속도가 느리다는 한계점

High performance & Fast detection speed



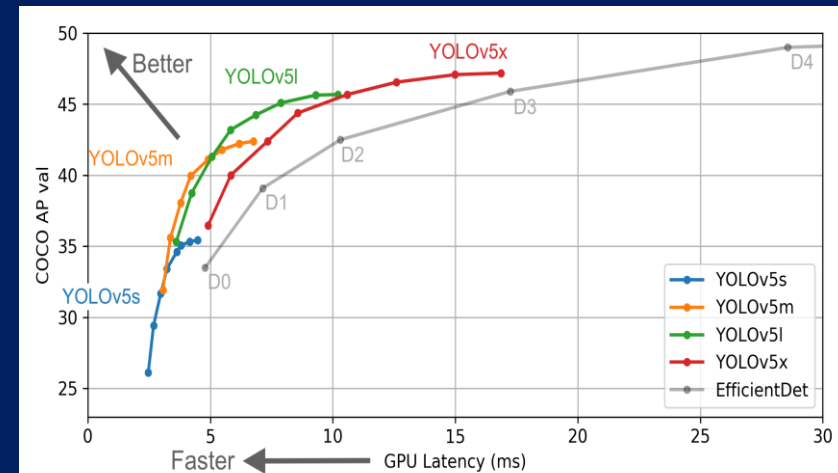
모델 선택

YOLO



YOLOv5 PyTorch

Small YOLOv5s	Medium YOLOv5m	Large YOLOv5l	XLarge YOLOv5x
14 MB _{FP16} 2.2 ms _{V100} 36.8 mAP _{COCO}	41 MB _{FP16} 2.9 ms _{V100} 44.5 mAP _{COCO}	90 MB _{FP16} 3.8 ms _{V100} 48.1 mAP _{COCO}	168 MB _{FP16} 6.0 ms _{V100} 50.1 mAP _{COCO}

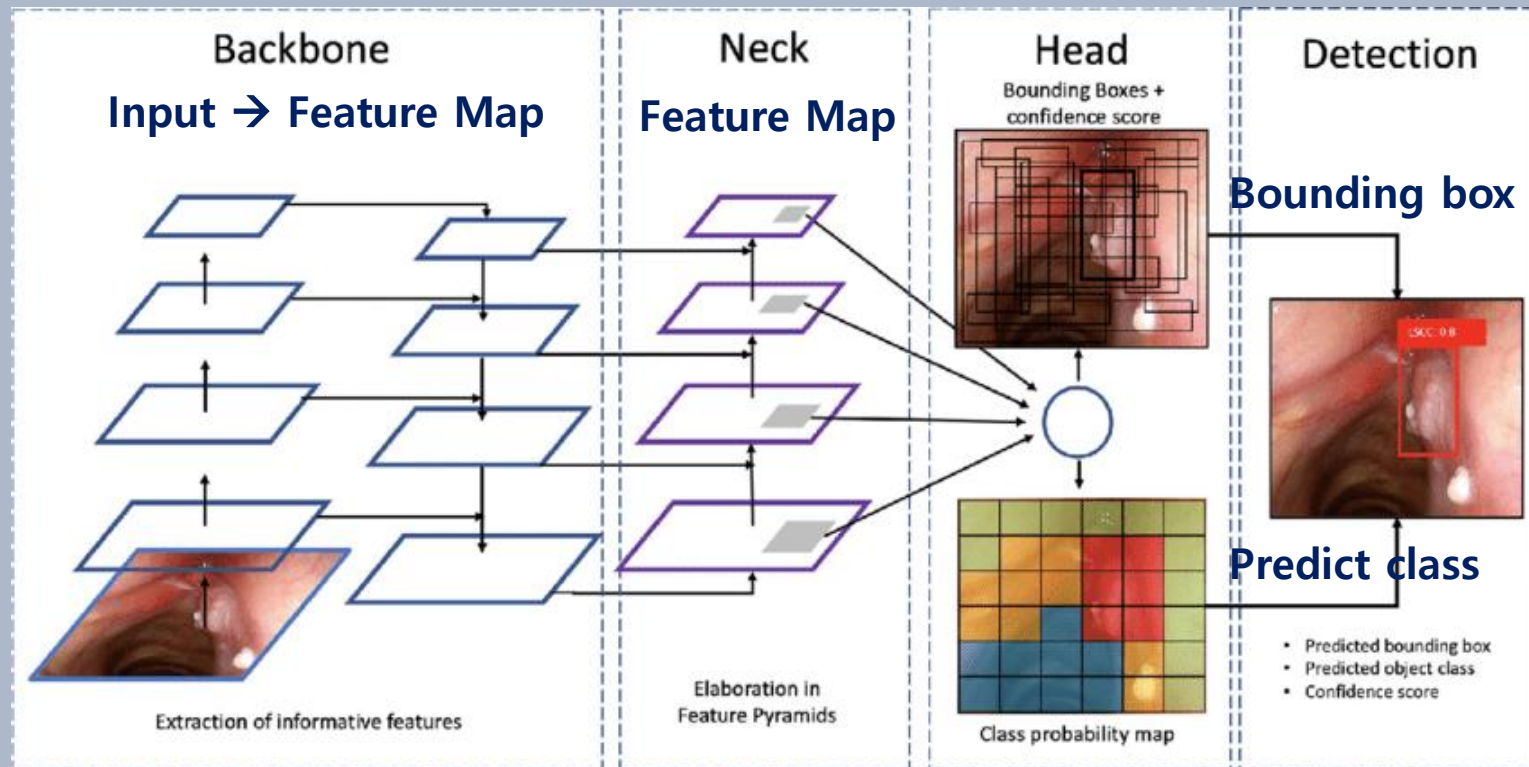


성능이 좋진 않지만, detection speed(fps)에 강점이 있는 YOLOv5s 선택!

모델 구조



Architecture



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-PCI-E-16GB)

colab

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

Drone Detection

데이터셋 로드하기

Your Download Code

 Jupyter  Terminal  Raw URL

Paste this snippet into [a notebook from our model library](#) » to download and unzip [your dataset](#) »:

```
!pip install roboflow

from roboflow import Roboflow
rf = Roboflow(api_key="XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX")
project = rf.workspace("5-jcfig1").project("drone-qfysh")
dataset = project.version(2).download("yolov5")
```



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

# dataset format : yolov5-Pytorch
rf = Roboflow(api_key="fkGLXDBaDekHbC0t6y7y")
project = rf.workspace("5-jcfigi").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

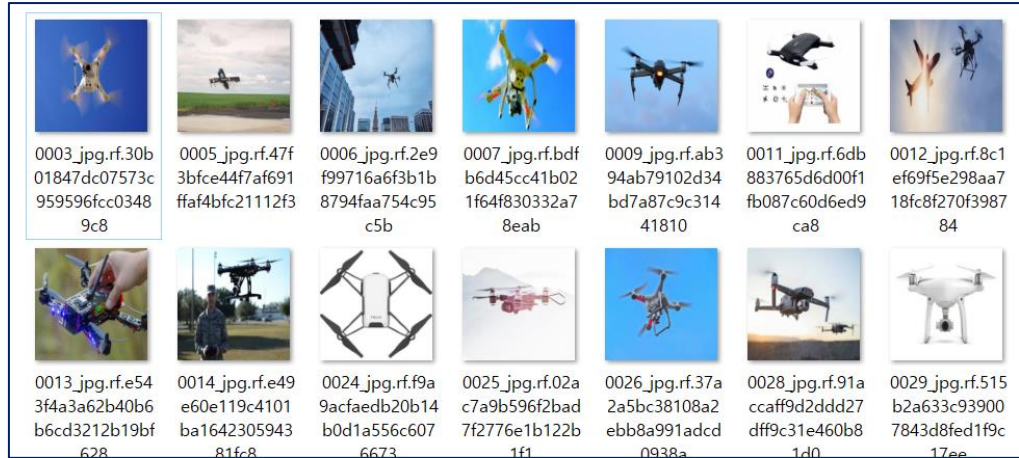
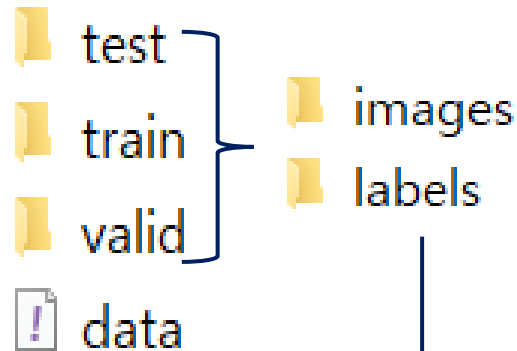
```
Extracting Dataset Version Zip to /content/datasets/drone-1 in yolov5pytorch:: 100%|██████████| 40574/40574 [00:05<00:00, 7548.00it/s]
```

Roboflow에서 생성한 데이터셋 다운로드

모델 학습

Drone Detection

데이터셋 로드하기

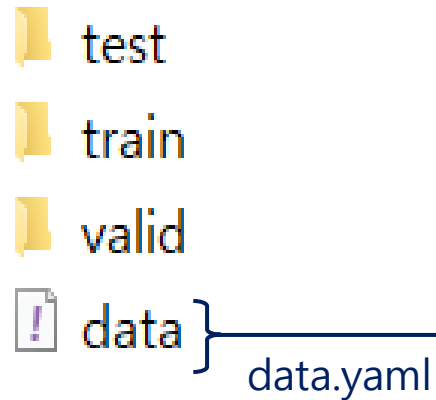


0003_jpg.rf.30b01847dc07573c959596fcc03489c8
0005_jpg.rf.47f3bfce44f7af691ffa4b4b628
0006_jpg.rf.2e9f99716a6f3b1b8794faa754c95c5b
0007_jpg.rf.bdfb6d45cc41b021f64f830332a71f1
0009_jpg.rf.ab394ab79102d34bd7a...ca8
0011_jpg.rf.6db883765d6d00f1fb08...ca8
0012_jpg.rf.8c1ef69f5e298aa718fc8f270f398784
0013_jpg.rf.e543f4a3a62b40b6b6cd3212b19bf628
0014_jpg.rf.e49e60e119c4101ba164...81fc8
0024_jpg.rf.f9a9acfaedb20b14b0d1a556c6076673
0025_jpg.rf.02ac7a9b596f2bad7f2776e1b122b1f1
0026_jpg.rf.37a2a5bc38108a2ebb8a...0938a
0028_jpg.rf.91accaff9d2ddd27dff9c31e460b81d0
0029_jpg.rf.515b2a633c939007843d...17ee

0003_jpg.rf.30b01847dc07573c959596fcc03489c8 - Windows 메모장
파일(F) 편집(E) 서식(O) 보기(V) 도움말(H)
0 0.5024038461538461 0.4735576923076923 0.6502403846153846 0.6887019230769231

Drone Detection

데이터셋 로드하기



names: 클래스 이름
- drone

nc: 1 클래스 개수

데이터 경로

```
test: ../datasets/test/images  
train: ../datasets/train/images  
val: ../datasets/valid/images
```


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
```

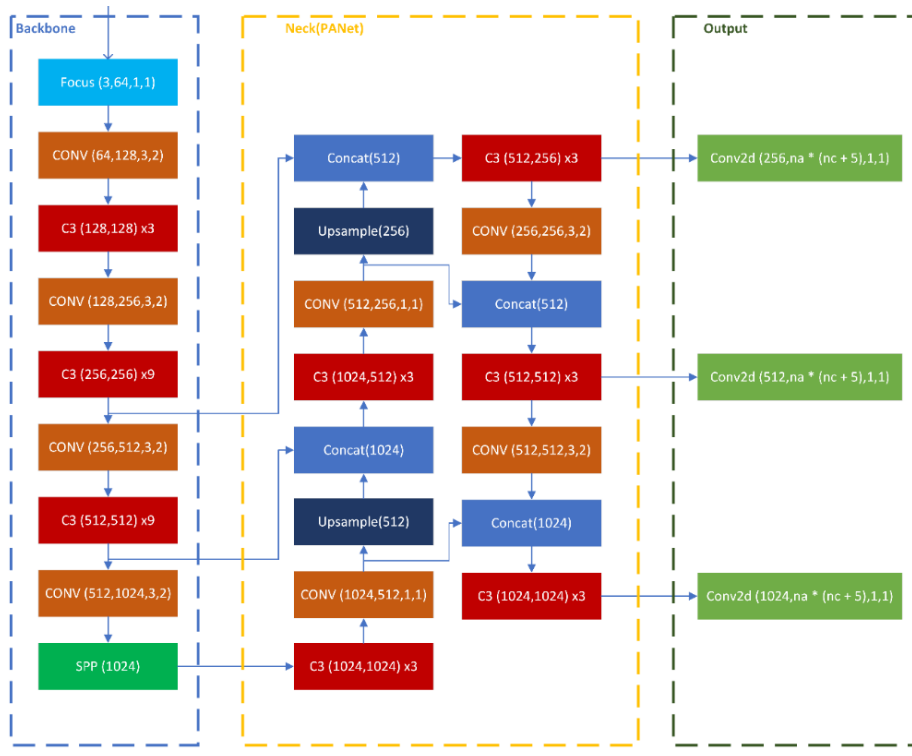


- hub
- __init__.py
- common.py
- experimental.py
- tf.py
- yolo.py
- yolov5l.yaml
- yolov5m.yaml
- yolov5n.yaml
- yolov5s.yaml**
- yolov5x.yaml

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	1182720	models.common.C3	[512, 512, 1, False]
24	[17, 20, 23]	1	16182	models.yolo.Detect	[1, [[10, 13, 16, 30, 33, 23], [3, 7, 8, 11, 14, 15, 17, 19, 20, 22, 24, 25, 26, 27, 28, 29, 31, 32, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65, 66, 67, 68, 69, 70, 71, 72, 73, 74, 75, 76, 77, 78, 79, 80, 81, 82, 83, 84, 85, 86, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 100, 101, 102, 103, 104, 105, 106, 107, 108, 109, 110, 111, 112, 113, 114, 115, 116, 117, 118, 119, 120, 121, 122, 123, 124, 125, 126, 127, 128, 129, 130, 131, 132, 133, 134, 135, 136, 137, 138, 139, 140, 141, 142, 143, 144, 145, 146, 147, 148, 149, 150, 151, 152, 153, 154, 155, 156, 157, 158, 159, 160, 161, 162, 163, 164, 165, 166, 167, 168, 169, 170, 171, 172, 173, 174, 175, 176, 177, 178, 179, 180, 181, 182, 183, 184, 185, 186, 187, 188, 189, 190, 191, 192, 193, 194, 195, 196, 197, 198, 199, 200, 201, 202, 203, 204, 205, 206, 207, 208, 209, 210, 211, 212, 213, 214, 215, 216, 217, 218, 219, 220, 221, 222, 223, 224, 225, 226, 227, 228, 229, 230, 231, 232, 233, 234, 235, 236, 237, 238, 239, 240, 241, 242, 243, 244, 245, 246, 247, 248, 249, 250, 251, 252, 253, 254, 255, 256, 257, 258, 259, 260, 261, 262, 263, 264, 265, 266, 267, 268, 269, 270, 271, 272, 273, 274, 275, 276, 277, 278, 279, 280, 281, 282, 283, 284, 285, 286, 287, 288, 289, 290, 291, 292, 293, 294, 295, 296, 297, 298, 299, 300, 301, 302, 303, 304, 305, 306, 307, 308, 309, 310, 311, 312, 313, 314, 315, 316, 317, 318, 319, 320, 321, 322, 323, 324, 325, 326, 327, 328, 329, 330, 331, 332, 333, 334, 335, 336, 337, 338, 339, 340, 341, 342, 343, 344, 345, 346, 347, 348, 349, 350, 351, 352, 353, 354, 355, 356, 357, 358, 359, 360, 361, 362, 363, 364, 365, 366, 367, 368, 369, 370, 371, 372, 373, 374, 375, 376, 377, 378, 379, 380, 381, 382, 383, 384, 385, 386, 387, 388, 389, 390, 391, 392, 393, 394, 395, 396, 397, 398, 399, 400, 401, 402, 403, 404, 405, 406, 407, 408, 409, 410, 411, 412, 413, 414, 415, 416, 417, 418, 419, 420, 421, 422, 423, 424, 425, 426, 427, 428, 429, 430, 431, 432, 433, 434, 435, 436, 437, 438, 439, 440, 441, 442, 443, 444, 445, 446, 447, 448, 449, 450, 451, 452, 453, 454, 455, 456, 457, 458, 459, 460, 461, 462, 463, 464, 465, 466, 467, 468, 469, 470, 471, 472, 473, 474, 475, 476, 477, 478, 479, 480, 481, 482, 483, 484, 485, 486, 487, 488, 489, 490, 491, 492, 493, 494, 495, 496, 497, 498, 499, 500, 501, 502, 503, 504, 505, 506, 507, 508, 509, 510, 511, 512]]]

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

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
```

```
# Write results
for *xyxy, conf, cls in reversed(det):
    if save_txt: # Write to file
        xywh = (xyxy2xywh(torch.tensor(xyxy).view(1, 4)) / gn).view(-1).tolist() # normalized xywh
        line = (cls, *xywh, conf) if save_conf else (cls, *xywh) # label format
        with open(f'{txt_path}.txt', 'a') as f:
            f.write('%g ' * len(line).rstrip() % line + '\n')

    if save_img or save_crop or view_img: # Add box to image
        c = int(cls) # integer class
        label = None if hide_labels else (names[c] if hide_conf else f'{names[c]} {conf:.2f}')
        annotator.box_label(xyxy, label, color=colors(c, True))
    if save_crop:
        save_one_box(xyxy, imc, file_save_dir / 'crops' / names[c] / f'{p.stem}.jpg', BGR=True)

# Stream results
im0 = annotator.result()
if view_img:
    if platform.system() == 'Linux' and p not in windows:
        windows.append(p)
        cv2.namedWindow(str(p), cv2.WINDOW_NORMAL | cv2.WINDOW_KEEPRATIO) # allow window resize (Linux)
        cv2.resizeWindow(str(p), im0.shape[1], im0.shape[0])
    cv2.imshow(str(p), im0)
    cv2.waitKey(1) # 1 millisecond

# Save results (image with detections)
if save_img:
    if dataset.mode == 'image':
        cv2.imwrite(save_path, im0)
    else: # video: save stream
        if vid_path[0] != save_path: # new video
            vid_path[0] = save_path
        if isinstance(vid_writer[0], cv2.VideoWriter):
            vid_writer[0].release() # release previous video writer
        if vid_cap: # save as video
            fps = vid_cap.get(cv2.CAP_PROP_FPS)
            w = int(vid_cap.get(cv2.CAP_PROP_FRAME_WIDTH))
            h = int(vid_cap.get(cv2.CAP_PROP_FRAME_HEIGHT))
        else: # stream
            fps, w, h = 30, im0.shape[1], im0.shape[0]
        save_path = str(Path(save_path).with_suffix('.mp4')) # force *.mp4 suffix on results videos
        vid_writer[0] = cv2.VideoWriter(save_path, cv2.VideoWriter_fourcc(*'mp4v'), fps, (w, h))
        vid_writer[0].write(im0)
```

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.8a763fbc0f40fd65fa87cacfd97686.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.bda603adf98403d7392bbbfb4b2a0a3.jpg: 416x416 1 drone, Done. (0.006s)
9/848 /content/datasets/drone-1/test/images/0048_jpg.rf.18705db337313ebec7a076df1ff94f0c.jpg: 416x416 1 drone, Done. (0.006s)
10/848 /content/datasets/drone-1/test/images/0049_jpg.rf.fc57badd07ffb85fa2c286a1a80399a3.jpg: 416x416 2 drones, Done. (0.006s)
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.6d201a081412a8525e21cfc8399dabe.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.e8394d7a12abbd098099bfbdb08bffbcb6.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.7c1cfcadc68ebba7f110e68fdeaf1b3.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

Drone Detection

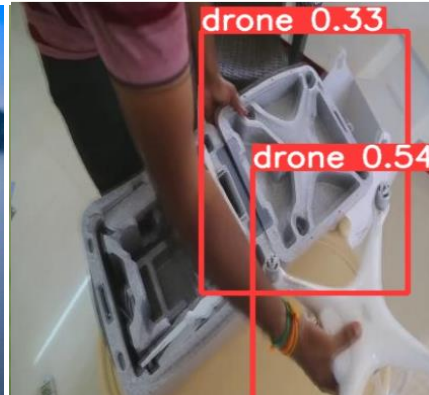
추론 결과 확인

```
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")
```

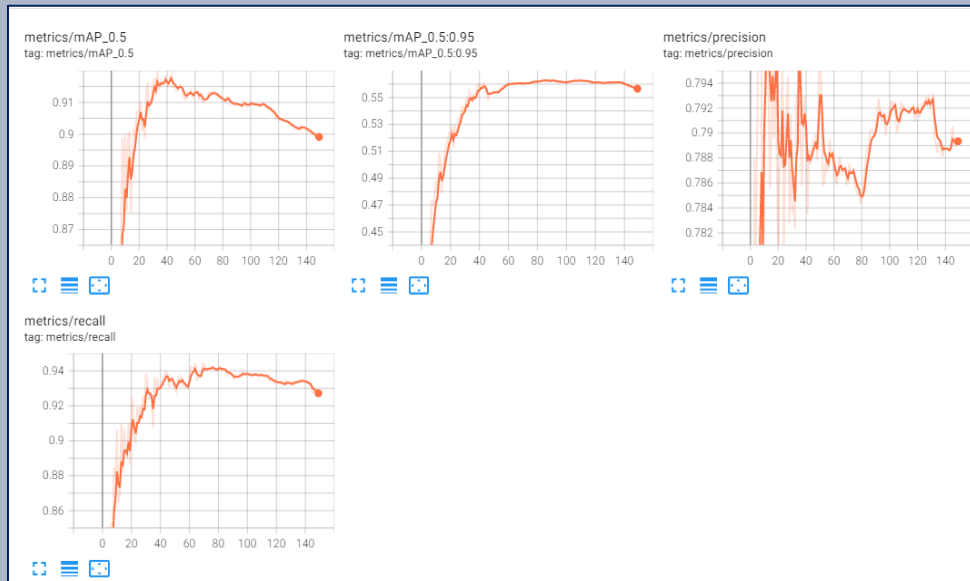


추론 결과 시각화



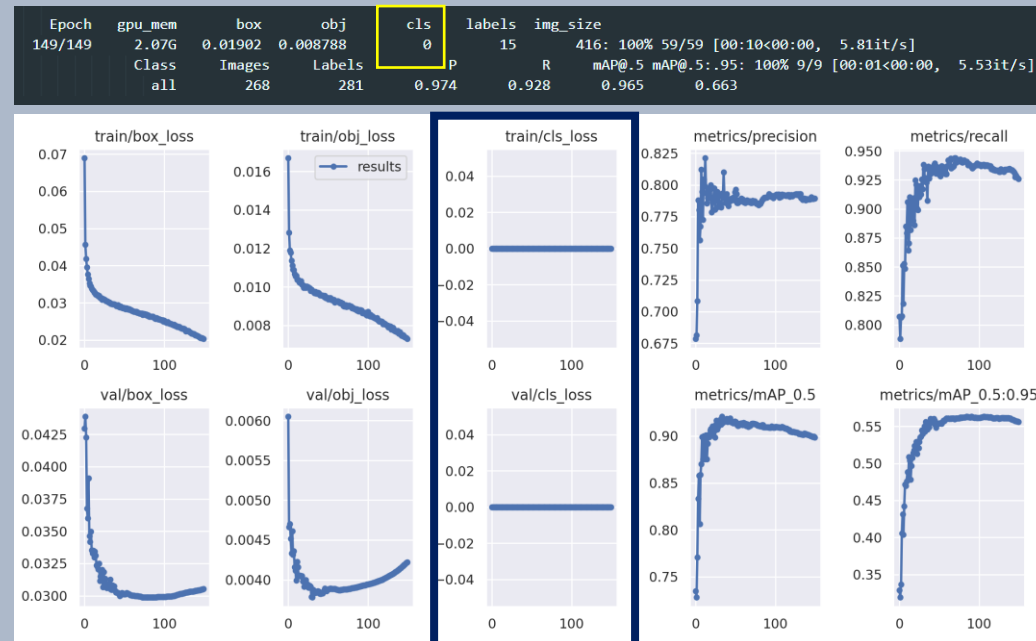
완벽하진 않지만, 그래도 **drone detection**이 가능해짐!!!

Drone Detection



Tensorboard 시각화

mAP : 객체 탐지 정확도 평가 지표



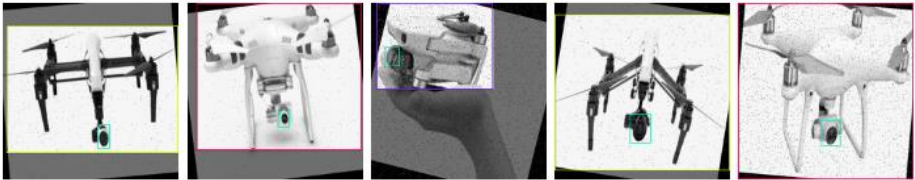
학습함에 따라 전체적으로 loss는 감소하고 정확도가 증가

Drone Classification

데이터 로드하기

TRAIN / TEST SPLIT

Training Set 300 images 87%	Validation Set 30 images 9%	Testing Set 13 images 4%
-----------------------------------	-----------------------------------	--------------------------------



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

EXPORT

```
# roboflow에서 데이터셋 가져오기
rf = Roboflow(api_key="fkGLXDbaDekHbC0t6y7y")
project = rf.workspace("5-jcfigi").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

Drone Classification

모델 학습하기

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

```
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

Drone Classification

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
```

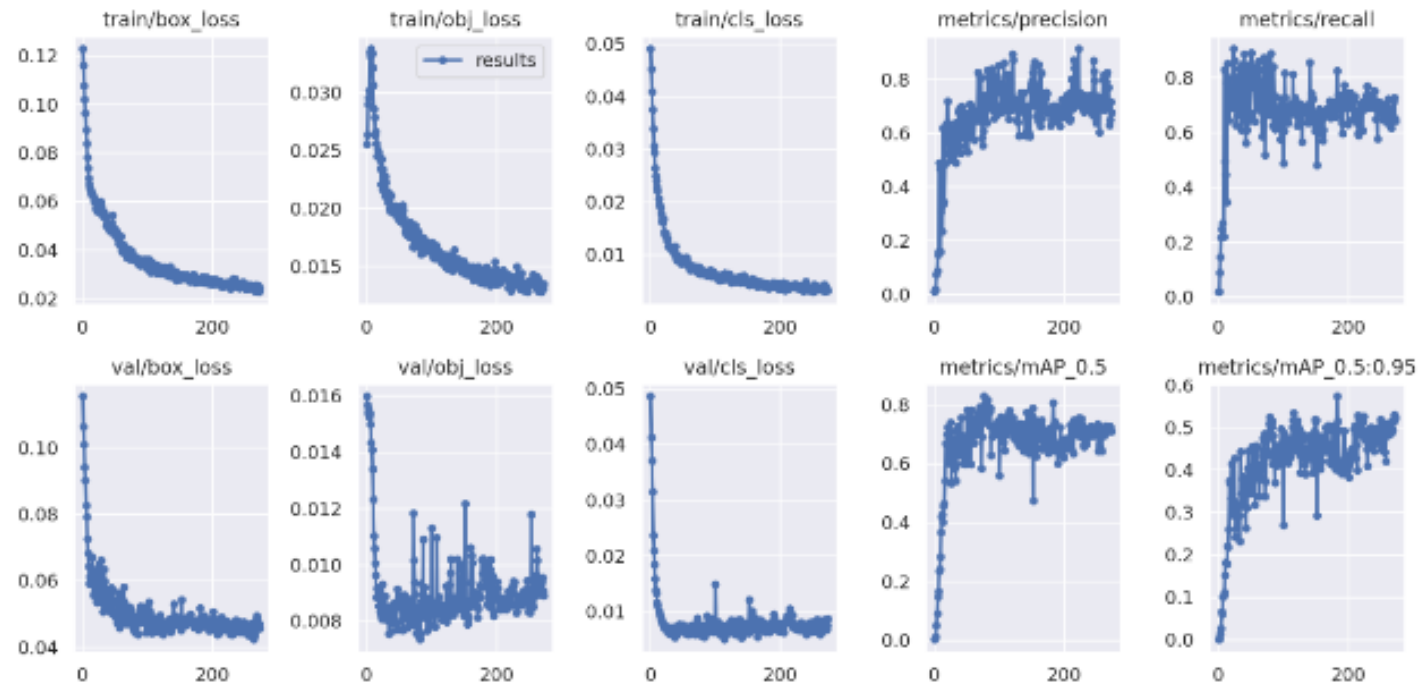
```
image 24/149 /content/datasets/drone-1/test/images/121.jpg.rf.3ff3329f936cb4273450c62284b222f4.jpg: 416x416 1 drone, 8.3ms
image 25/149 /content/datasets/drone-1/test/images/123.jpg.rf.87b695b26b2ca39517f072ad8dcee6c1.jpg: 416x416 (no detections), 7.6ms
image 26/149 /content/datasets/drone-1/test/images/15.jpg.rf.00c4f1c2987ef8c438859ac5161c81b5.jpg: 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, 8.3ms
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
```


Drone Classification

모델로 추론하기



Drone Classification

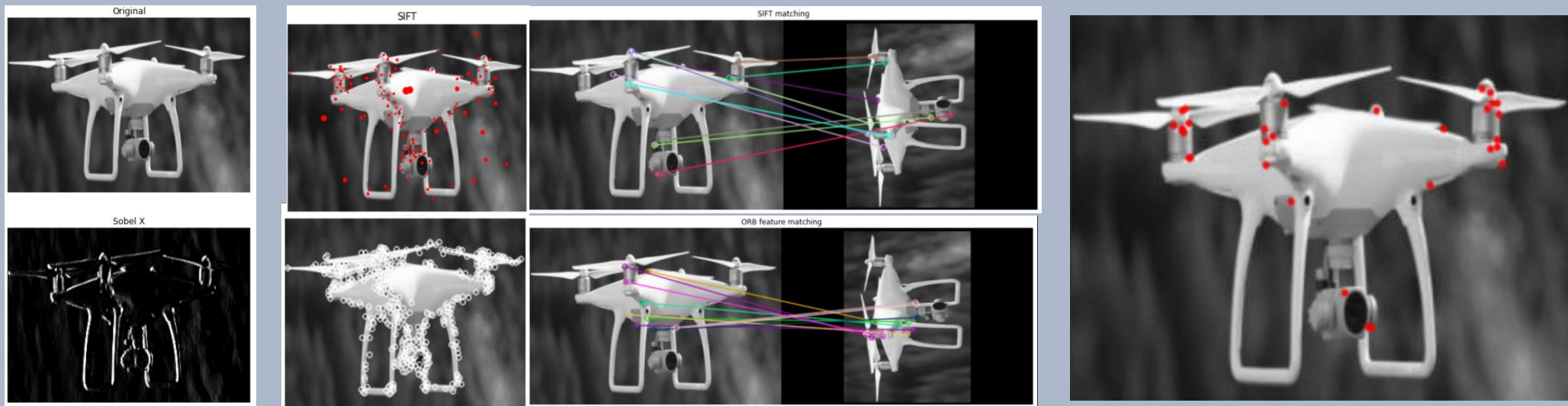


Drone detection에 비해 성능이 떨어짐

Object Tracking

Drone Feature Mapping

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



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/SEF2U5x0m74Qkxw80M7_6dvG3c68MB6wAHvB9YVA.jpg"
results = model(img)

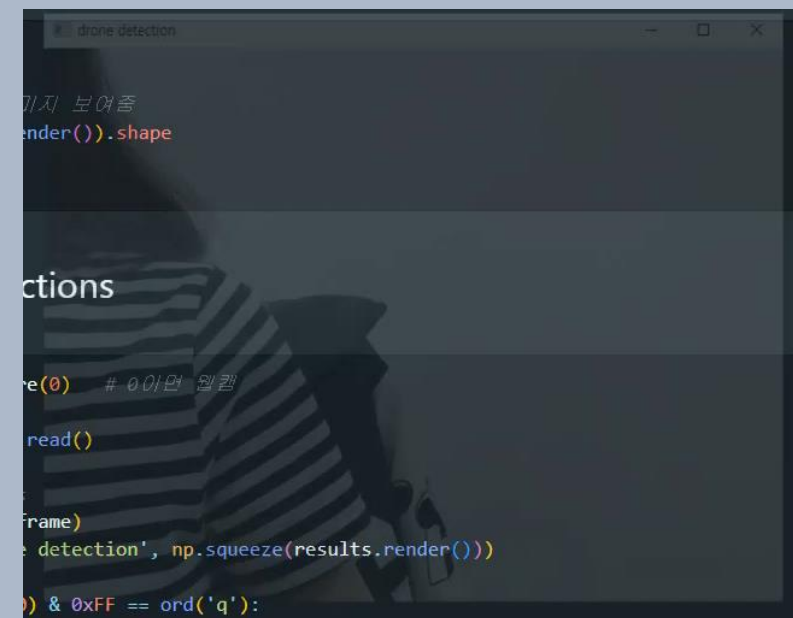
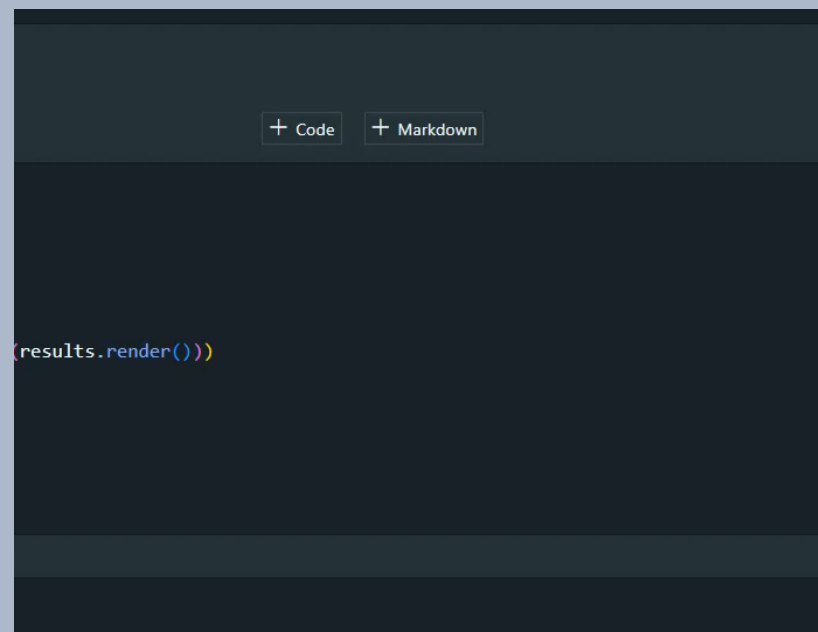
results.print()
results.show()
%matplotlib inline
plt.imshow(np.squeeze(results.render()))
plt.axis('off')
plt.savefig('detection01.jpg', dpi=300, bbox_inches='tight')
plt.show()

results.xyxy
results.render()
results.show() # 0.007 0.007
np.squeeze(results.render()).shape

cap = cv2.VideoCapture(0) # 0.007 0.007
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;

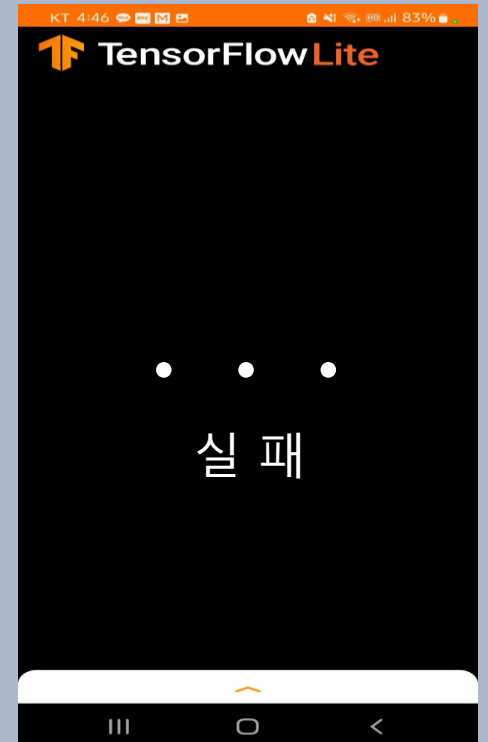
import android.content.res.AssetManager;

import java.io.IOException;

public class DetectorFactory {
    public static VoloV5Classifier getDetector(
        final AssetManager assetManager,
        final String modelFilename)
        throws IOException {
        String labelFilename = null;
        boolean isQuantized = false;
        int inputSize = 0;
        int[] outputWidth = 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;
            outputWidth = 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 VoloV5Classifier.create(assetManager, modelFilename, labelFilename, isQuantized,
            inputSize);
    }
}
```



Application



Flask - web

```
import argparse
import io
import os
from PIL import Image
import torch
from flask import Flask, render_template, request, redirect

app = Flask(__name__)
app.route("/", methods=["GET", "POST"])
def 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")

if __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...

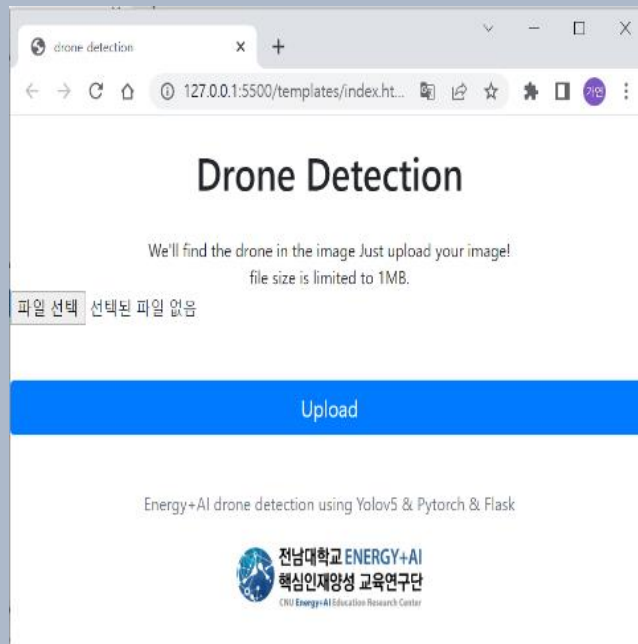
Model summary: 213 layers, 7023610 parameters, 0 gradients, 15.8 GFLOPs

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

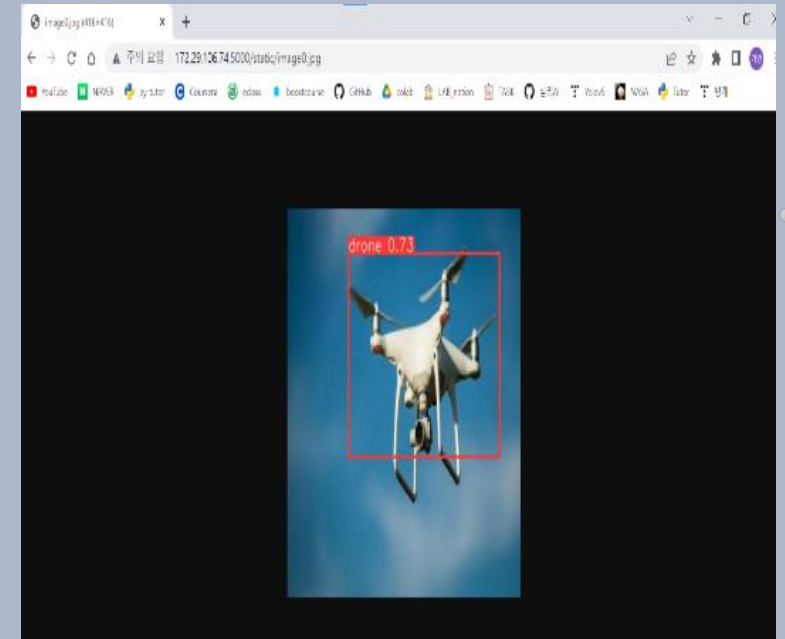
Flask – web



Main 화면



Image 파일 업로드



Detection 결과 출력

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



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