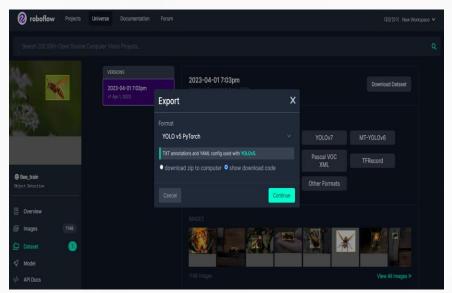
수집된 이미지를 바탕으로 객체인식 실습

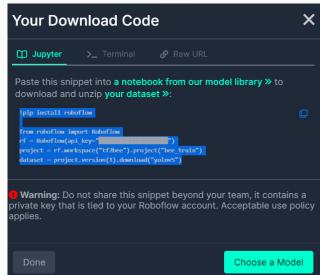
이 준 혁 (2023250419) 산 업 인 공 지 능 학 과

진행순서

- 1. Datasets 가져오기
- 2. Yolo 클론 설치 및 프로그램 패키지 설치
- 3. 이미지 저장 리스트생성 및 이미지 개수확인
- 4. 트레이닝셋과 밸리데이션 셋 나누기
- 5. 위치 변경
- 6. 트레이닝 시키기
- 7. 트레이닝 완료하고 확인하기
- 8. 이미지 넣고 추론하기
- 9. 내가 원하는 이미지를 넣고 확인하기

1. 데이터셋 가져오기 (1)





로보플로우사이트에서 데이터셋 구하기

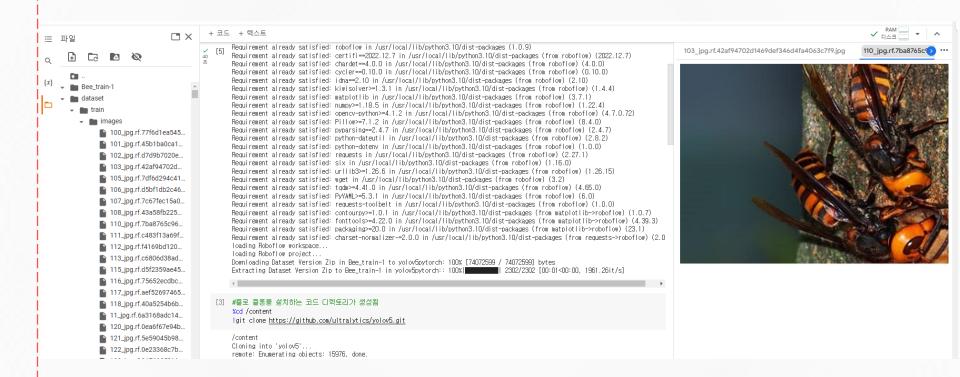
1. 데이터셋 가져오기 (2)

```
[5] #말벌에 데이터셋을 가져오기 위한 코드(roboflow에서 dataset을 제공하고 있음)
    !pip install roboflow
    from roboflow import Roboflow
    rf = Roboflow(api_key="fNmB1Zp1Cdu4W0ENOrmQ")
    project = rf.workspace("tf2bee").project("bee_train")
    dataset = project.version(1).download("volov5")
    Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
    Requirement already satisfied: roboflow in /usr/local/lib/python3.10/dist-packages (1.0.9)
    Requirement already satisfied: certifi==2022.12.7 in /usr/local/lib/python3.10/dist-packages (from roboflow) (2022.12.7)
    Requirement already satisfied: chardet == 4.0.0 in /usr/local/lib/python3.10/dist-packages (from roboflow) (4.0.0)
    Requirement already satisfied: cycler==0.10.0 in /usr/local/lib/python3.10/dist-packages (from roboflow) (0.10.0)
    Requirement already satisfied: idna==2.10 in /usr/local/lib/python3.10/dist-packages (from roboflow) (2.10)
    Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.4.4)
    Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (from roboflow) (3.7.1)
    Requirement already satisfied: numpy>=1.18.5 in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.22.4)
    Requirement already satisfied: opency-python>=4.1.2 in /usr/local/lib/python3.10/dist-packages (from roboflow) (4.7.0.72)
    Requirement already satisfied: Pillow>=7.1.2 in /usr/local/lib/python3.10/dist-packages (from roboflow) (8.4.0)
    Requirement already satisfied: pyparsing==2.4.7 in /usr/local/lib/python3.10/dist-packages (from roboflow) (2.4.7)
    Requirement already satisfied: python-dateutil in /usr/local/lib/python3.10/dist-packages (from roboflow) (2.8.2)
    Requirement already satisfied: python-dotenv in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.0)
    Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from roboflow) (2.27.1)
    Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.16.0)
    Requirement already satisfied: urllib3>=1.26.6 in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.26.15)
    Requirement already satisfied: wget in /usr/local/lib/python3.10/dist-packages (from roboflow) (3.2)
    Requirement already satisfied: tgdm>=4.41.0 in /usr/local/lib/python3.10/dist-packages (from roboflow) (4.65.0)
    Requirement already satisfied: PyYAML>=5.3.1 in /usr/local/lib/python3.10/dist-packages (from roboflow) (6.0)
    Requirement already satisfied: requests-toolbelt in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.0)
    Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (1.0.7)
    Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (4.39.3)
    Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (23.1)
    Requirement already satisfied: charset-normalizer~=2.0.0 in /usr/local/lib/python3.10/dist-packages (from requests->roboflow) (2.0
    loading Roboflow workspace...
    loading Roboflow project...
    Downloading Dataset Version Zip in Bee_train-1 to yolov5pytorch: 100% [74072599 / 74072599] bytes
    Extracting Dataset Version Zip to Bee_train-1 in yolov5pytorch:: 100%| 2302/2302 [00:01<00:00, 1961.26it/s]
```



데이터셋 주소 입력하고 이미지파일 확인

1. 데이터셋 가져오기 (3)



데이터셋 주소 입력하고 이미지파일 확인

2. 욜로 환경 설치



욜로 클론 설치, 패키지 설치

3. 이미지저장 리스트생성 및 이미지 개수확인

```
[6] #data.vaml 파일 확인
    %cat /content/dataset/data.yaml
    names:
    - hornet
    nc: 1
    roboflow:
      license: CC BY 4.0
      project: bee_train
      url: https://universe.roboflow.com/tf2bee/bee_train/dataset/1
      version: 1
      workspace: tf2bee
    test: ../test/images
    train: Bee_train-1/train/images
    val: Bee_train-1/valid/images
[8] #이미지 저장한 리스트 생성 및 이미지개수확인
    %cd /
    from alob import alob
     img_list = glob('content/dataset/train/images/*.jpg')
    print(len(img_list))
     1148
```

이미지 리스트 및 개수확인

4. 트레이닝셋과 밸리데이션 셋나누기 및 텍스트 파일로 데이터셋 저장



Train, val 값 TXT로 저장

5. 위치변경

```
#yaml의 위치에 트레이닝셋과 밸리데이션 셋을 넣어주기
import yaml

with open('content/dataset/data.yaml', 'r') as f:
    data = yaml.safe_load(f)

print(data)

data['train'] = '/content/dataset/train.txt'
data['val'] = '/content/dataset/val.txt'

with open('/content/dataset/data.yaml', 'w') as f:
    yaml.dump(data, f)
print(data)

on': 1, 'workspace': 'tf2bee'}, 'test': '../test/images', 'train': 'Bee_train-1/train/images', 'val': 'Bee_train-1/valid/images'}
on': 1, 'workspace': 'tf2bee'}, 'test': '../test/images', 'train': '/content/dataset/train.txt', 'val': '/content/dataset/val.txt'}
```

트레이닝 시키기 위해 위치지정

6. 트레이닝 시키기

#트레이닝 시키기

%cd /content/yolov5/

Pipython train.py --img 416 --batch 16 --epochs 50 --data /content/dataset/data.vaml --cfg ./models/yolov5m.yaml --weights yolov5m.pt --name bee_yolov5m_results

... /content/yolov5

train: weights=yolov5m.pt, cfg=./models/yolov5m.yaml, data=/content/dataset/data.yaml, hyp=data/hyps/hyp.scratch-low.yaml, epochs=50, batch_size=16, imgsz=416, rect=Fal github: up to date with https://github.com/ultralytics/yolov5 YOLOV5 💋 v7.0-178-ga199480 Python-3.10.11 torch-2.0.1+cu118 CPU

hyperparameters: Ir0=0.01, Irf=0.01, momentum=0.937, weight_decay=0.0005, warmup_epochs=3.0, warmup_momentum=0.8, warmup_bias_Ir=0.1, box=0.05, cls=0.5, cls_pw=1.0, ok ClearML: run 'pip install clearml' to automatically track, visualize and remotely train YOLOV5 💋 in ClearML

Comet: run 'pip install comet_ml' to automatically track and visualize YOLOv5 💋 runs in Comet

TensorBoard: Start with 'tensorboard --logdir runs/train', view at http://localhost:6006/

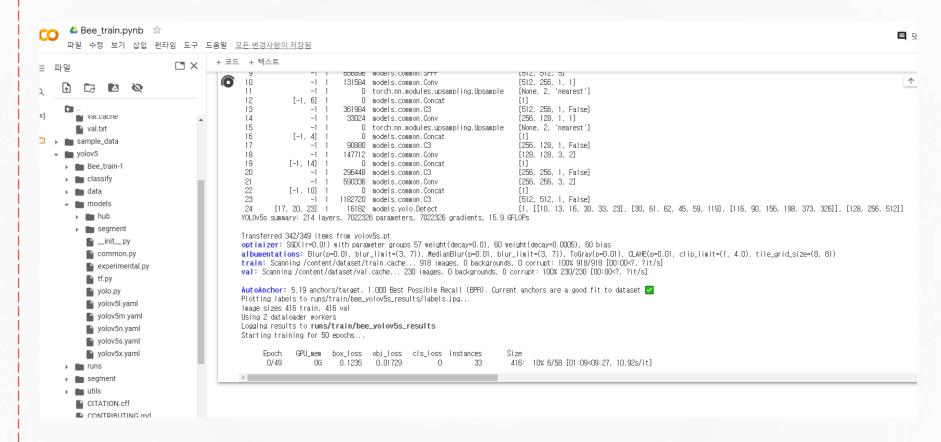
Overriding model.yaml nc=80 with nc=1

	from	Π	params	module	arguments
0	-1	1	5280	models.common.Conv	[3, 48, 6, 2, 2]
1	-1	1	41664	models.common.Conv	[48, 96, 3, 2]
2	-1	2	65280	models.common.C3	[96, 96, 2]
3	-1	1	166272	models.common.Conv	[96, 192, 3, 2]
4	-1	4	444672	models.common.C3	[192, 192, 4]
5	-1	1	664320	models.common.Conv	[192, 384, 3, 2]
6	-1	6	251 2896	models.common.C3	[384, 384, 6]
7	-1	1	2655744	models.common.Conv	[384, 768, 3, 2]
8	-1	2	4134912	models.common.C3	[768, 768, 2]
9	-1	1	1476864	models.common.SPPF	[768, 768, 5]
10	-1	1	295680	models.common.Conv	[768, 384, 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	2	1182720	models.common.C3	[768, 384, 2, False]
14	-1	1	74112	models.common.Conv	[384, 192, 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	2	296448	models.common.C3	[384, 192, 2, False]
18	-1	1	332160	models.common.Conv	[192, 192, 3, 2]
19	[-1, 14]	1	0	models.common.Concat	[1]
20	-1	2	1035264	models.common.C3	[384, 384, 2, False]
21	-1	1	1327872	models.common.Conv	[384, 384, 3, 2]
22	[-1 10]	1	Ω	models common Concat	[1]

6. 트레이닝 시키기

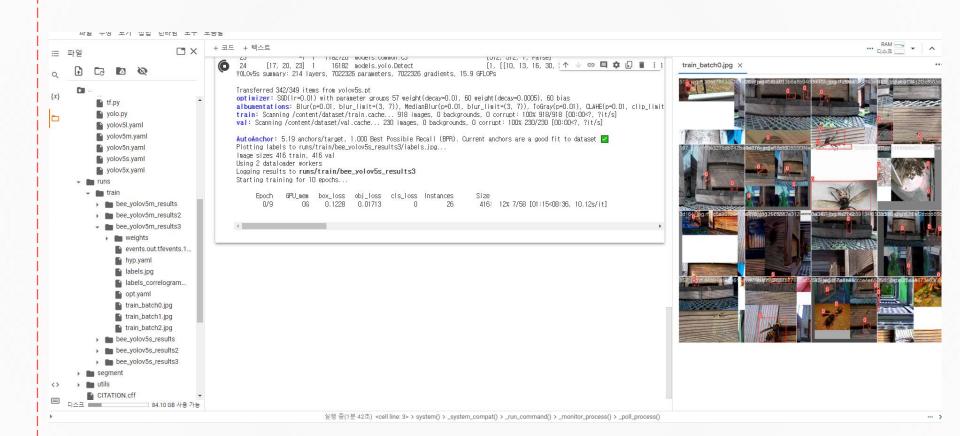
```
2512896 models.common.C3
                                                                         1384, 384, 61
                                                                         [384, 768, 3, 2]
                         2655744 models.common.Conv
                        4134912 models.common.C3
                                                                         [768, 768, 2]
                  -1 1 1476864 models.common.SPPF
                                                                         [768, 768, 5]
10
                          295680 models.common.Conv
                                                                         [768, 384, 1, 1]
11
                  -1 1
                               O torch.nn.modules.upsampling.Upsample
                                                                         [None, 2, 'nearest']
12
             [-1, 6] 1
                               O models.common.Concat
13
                  -1 2 1182720 models.common.C3
                                                                         [768, 384, 2, False]
14
                  -1 1
                           74112 models.common.Conv
                                                                         [384, 192, 1, 1]
                                                                         [None, 2, 'nearest']
 15
                  -1 1
                               O torch.nn.modules.upsampling.Upsample
16
             [-1, 4] 1
                               O models.common.Concat
                                                                         [1]
17
                  -1 2
                          296448 models.common.C3
                                                                         [384, 192, 2, False]
                                                                         [192, 192, 3, 2]
18
                  -1 1 332160 models.common.Conv
19
            [-1, 14] 1
                               O models.common.Concat
20
                  -1 2 1035264 models.common.C3
                                                                         [384, 384, 2, False]
21
                  -1 1 1327872 models.common.Conv
                                                                         [384, 384, 3, 2]
22
            [-1, 10] 1
                               O models.common.Concat
                                                                         [1]
                  -1 2 4134912 models.common.C3
                                                                         [768, 768, 2, False]
        [17, 20, 23] 1 24246 models.volo.Detect
                                                                         [1, [[10, 13, 16, 30, 33, 23], [30, 61, 62, 45, 59, 119], [116, 90, 156, 198, 373, 326]], [192, 384, 768]]
YOLOv5m summary: 291 layers, 20871318 parameters, 20871318 gradients, 48.2 GFLOPs
Transferred 474/481 items from yolov5m.pt
optimizer: SGD(Ir=0.01) with parameter groups 79 weight(decay=0.0), 82 weight(decay=0.0005), 82 bias
albumentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01, blur_limit=(3, 7)), ToGrav(p=0.01), CLAHE(p=0.01, clip_limit=(1, 4.0), tile_grid_size=(8, 8))
train: Scanning /content/dataset/train... 918 images, 0 backgrounds, 0 corrupt: 100% 918/918 [00:03<00:00, 271.83it/s]
train: New cache created: /content/dataset/train.cache
val: Scanning /content/dataset/val... 230 images, 0 backgrounds, 0 corrupt: 100% 230/230 [00:00<00:00, 269.37it/s]
val: New cache created: /content/dataset/val.cache
AutoAnchor: 5.19 anchors/target, 1.000 Best Possible Recall (BPR), Current anchors are a good fit to dataset 🛂
Plotting labels to runs/train/bee_yolov5m_results3/labels.jpg...
Image sizes 416 train, 416 val
Using 2 dataloader workers
Logging results to runs/train/bee volov5m results3
Starting training for 50 epochs...
     Epoch
              GPU_mem box_loss obj_loss cls_loss Instances
      0/49
                         0.1144
                                  0.01823
                                                    Π
                                                             38
                                                                      416: 2% 1/58 [00:37<35:54, 37.80s/it]
```

6. 트레이닝 시키기 모델 s로 변경, 학습 수 10번으로 변경



속도향상을 위해 코랩 GPU 활용 및 모델사이즈 s로 변경, 학습 수 10번으로 변경

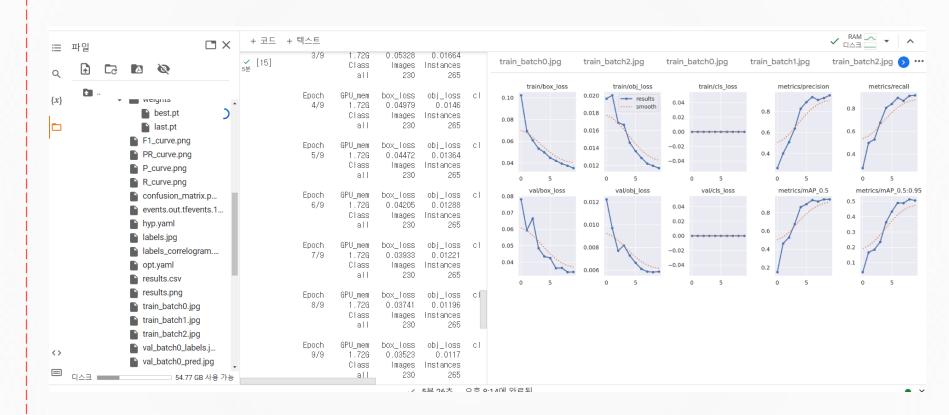
6. 트레이닝 모델 배치활동 확인



7. 트레이닝 완료하고 확인하기

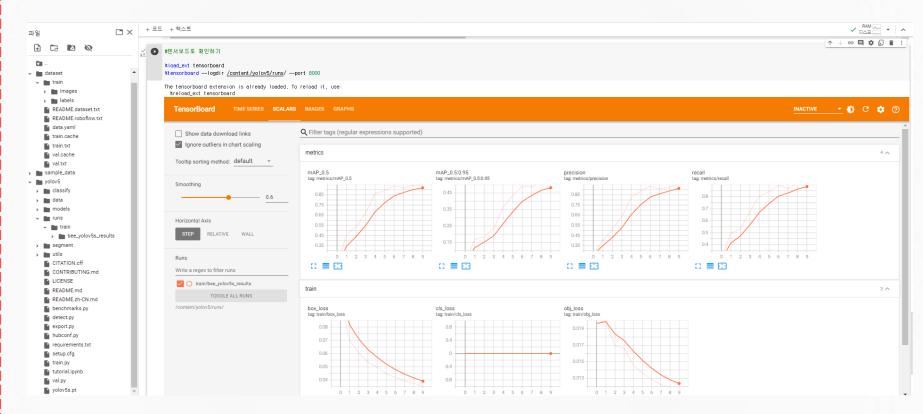
타일		+ 코드 + 텍스트	1,720	0,04313	0,0140	U	13	410.		[00:20\00:00, 2.02]()			
		✓ [15] 5분	Class all	Images 230	Instances 265	P 0.826	R 0.806	mAP50 0.861	mAP50-95: 0.362	100% 8/8 [00:03<00:00,	2.61it/s]		
sample_data		Epoch 5/9	GPU_mem 1.72G	box_loss 0.04472	obj_loss 0.01364	cls_loss O	Instances 20	Size	100% 58/58	[00:22<00:00, 2.53it/:	e]		
yolov5	•	3,3	Class	Images	Instances	P	R	mAP50	mAP50-95:	100% 8/8 [00:03<00:00,			
classify			all	230	265	0.888	0.841	0.893	0.433				
data		Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Size	100 50./50		1		
▶ m models		6/9	1.72G Class	0.04205 Images	0.01288 Instances	0 P		416: mAP50		58 [00:23<00:00, 2.51it/s] 5: 100% 8/8 [00:03<00:00, 2.49it/s]			
→ m runs			all	230			0.936 0.894		0.489				
→ train		Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Size					
→ bee_yolov5s_results		7/9	1.72G Class	0.03933 Images	0.01221 Instances	0 P		416: mAP50		8 [00:22<00:00, 2.57it/s] : 100% 8/8 [00:03<00:00, 2.25it/s]			
weights			all	230	265	0.909	0.868	0.922	0.487	100% 0/0 [00.03<00.00,	2.2011/5]		
best.pt		Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Size					
F1_curve.	na	8/9	1.72G	0.03741	0.01196	0	8	416:		[00:22<00:00, 2.57it/s			
PR_curve.			Class all	lmages 230	Instances 265	P 0.927	R 0.911	mAP50 0.945	mAP50-95: 0.512	100% 8/8 [00:03<00:00,	2.21it/s]		
P_curve.p	ng								0.012				
R_curve.p	ng	Epoch 9/9	GPU_mem 1.72G	box_loss 0.03523	obj_loss O.O117	cls_loss O	Instances 11	Size 416:	100% 58/58	[00:22<00:00, 2.58it/	sl		
confusion	_matrix.p	3,3	Class	Images	Instances	P	R	mAP50	mAP50-95:	100% 8/8 [00:03<00:00,			
events.out.tfevents.1			all	230	265	0.953	0.916	0.945	0.505				
hyp.yaml		10 epochs cor											
labels.jpg labels_correlogram		Optimizer stripped from runs/train/bee_yolov5s_results/weights/last.pt, 14.3MB Optimizer stripped from runs/train/bee_yolov5s_results/weights/best.pt, 14.3MB											
_	reiogram												
opt.yaml results.cs	Validating runs/train/bee_yolov5s_results/weights/best.pt Fusing layers												
results.pn	YOLOv5s summary: 157 layers, 7012822 parameters, 0 gradients, 15.8 GFLOPs Class Images Instances P R mAP50-95: 100% 8/8 [00:04<00:00, 1.62it/s] all 230 265 0.927 0.911 0.945 0.512 Results saved to runs/train/bee_yolov5s_results												
train_batch0.jpg													
train_bate													
train_batc	31 0	4									,		
val_batch	_labels.j									↑ ↓ ⊖ E			

7. 트레이닝 완료하고 확인하기



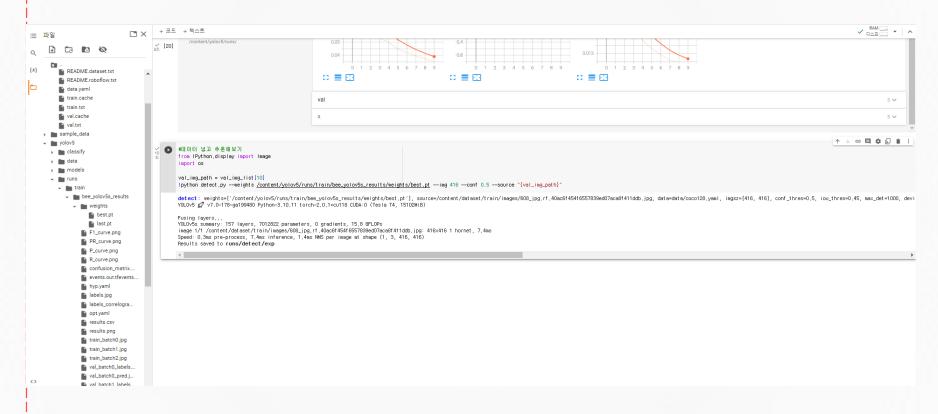
결과값 확인

7. 트레이닝 완료하고 확인하기



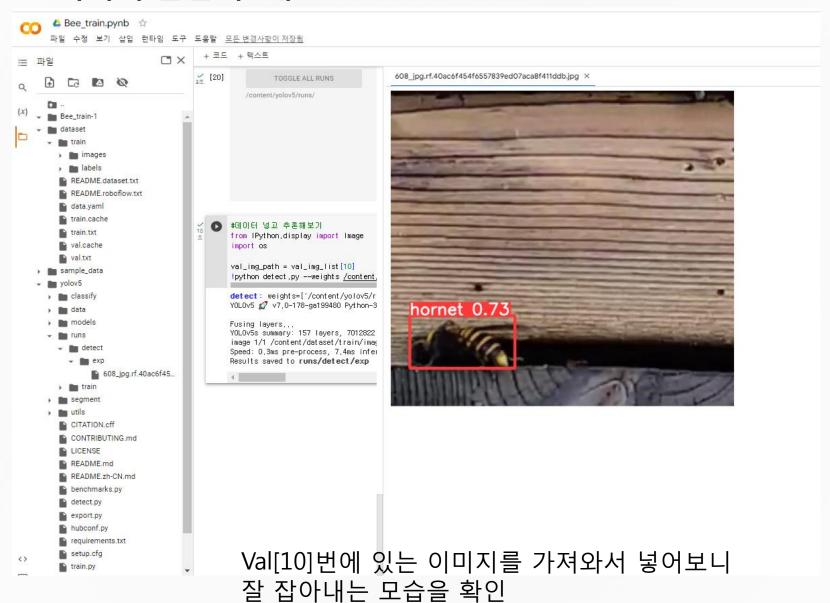
텐서플로우로 결과값 확인

8. 이미지 삽입 후 테스트

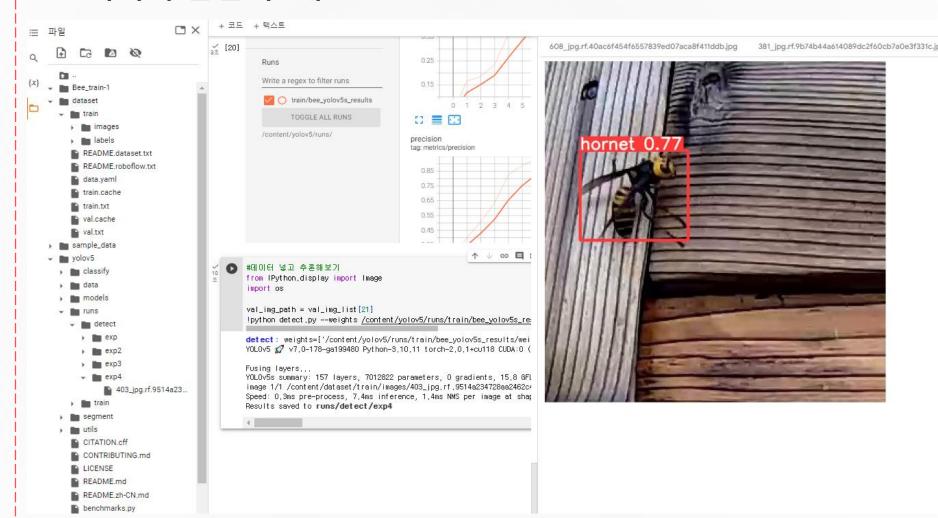


데이터 넣고 추론하기 runs/detect/exp로 저장됨

8. 이미지 삽입 후 테스트



8. 이미지 삽입 후 테스트

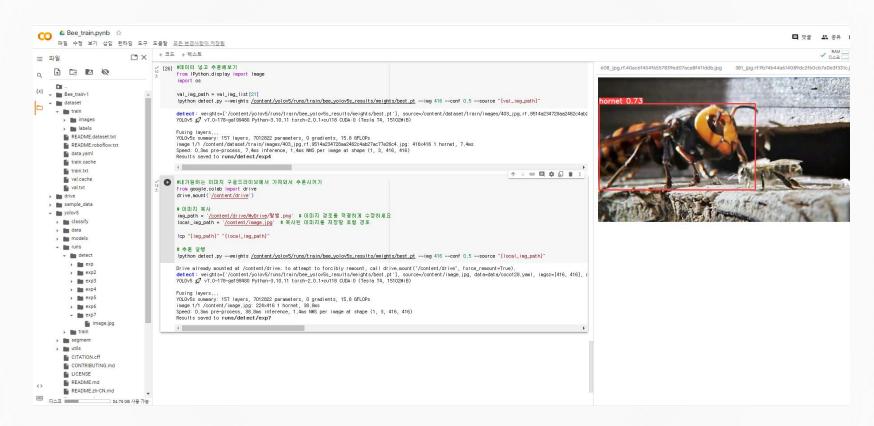


Val[21]번에 있는 이미지를 가져와서 넣어보니 잘 잡아내는 모습을 확인

9. 원하는 이미지 업로드 후 테스트



9. 원하는 이미지 업로드 후 테스트



구글드라이브에 업로드한 나의 이미지로 추론하기 정상적 인식 확인

감사합니다