

roboflow

Axis-Aligned Bounding Box (AABB) - txt file

class_id coordinates (x_center, y_center) (width) (height)

0 0.51328125 0.67265625 0.20859375 0.20546875



Oriented Bounding Box (OBB) - txt file

class_id coordinates (x_center, y_center) (width) (height) (Rotation angle (θ))







YOLOv8.2

Unleashing Next-Gen AI Capabilities

Discover more



YOLOv9 training and deployment



Advanced tracking with YOLOv8-OB



Zero-shot promptable YOLO-Worldv2 models



40% faster ultralytics import speed



YOLOv8.2 with Raspberry Pi 5 CI and tutorials

Download the App


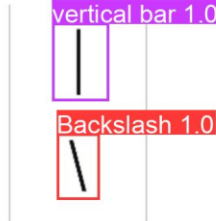








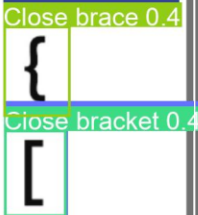





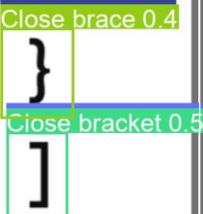





```
1 from ultralytics import YOLO
2
3 def main():
4     # Load a model
5     model = YOLO("yolov8n.yaml") # build a new model from scratch
6     model = YOLO("yolov8n.pt") # load a pretrained model (recommended for training)
7
8     # Use the model
9     model.train(data="momo640.yaml", epochs=100) # train the model
10    metrics = model.val() # evaluate model performance on the validation set
11    results = model("https://attach.setn.com/newsimages/2019/07/09/2010347-XXL.jpg")
12    path = model.export(format="onnx") # export the model to ONNX format
13
14 if __name__ == '__main__':
15     main()
```

```
1 train: ./momo640/train # training dataset
2 val: ./momo640/valid # validation dataset
3 nc: 1
4 names: ['Pothole']
```




- Multi Class
- Inhouse training
- Customized test program
- YOLO v9與v8用法
- Downlaod Ultralytics, my code
- Use more threshold to analyze efficientnet backbone

key_44_png.rf.197b06b30aeb51043549345458fba6c57 O 1.0 	key_43_png.rf.144ba635531ebc19919a932b3defd81e3130 vertical bar 1.0 Backslash 1.0 	key_136_png.rf.902f84318e608470b3986b23a7e65b8dbec S 1.0 	key_135_png.rf.1e65b8dbec H 0.9 
key_43_png.rf.144ba635531ebc19919a932b3defd81e3130 P 1.0 	key_512_png.rf.44df26b956684980b55294b209299c81dac Action 2 1.0 Page Down key 0.3 end  	key_512_png.rf.44df26b956684980b55294b209299c81dac D 1.0 	key_135_png.rf.1e65b8dbec J 0.9  
key_42_png.rf.395138ec4ec0413351a9a8c7a81bc2df37 Close brace 0.4 Close bracket 0.4 	key_335_png.rf.71108c90 Caps Lock 1.0 caps lock 	key_514_png.rf.1268a2abec85f6a00e91e4d72d790f F 1.0  	key_135_png.rf.1e65b8dbec start 0.8 start  option 1.0 opt 
key_41_png.rf.b82a05f320456498a5c07285d171108c90 Close brace 0.4 Close bracket 0.5 	key_137_png.rf.1268edccc99842041d5766a725f2deac6 A 0.9 	key_137_png.rf.1268edccc99842041d5766a725f2deac6 G 1.0 	key_135_png.rf.1e65b8dbec Fn 1.0 fn 

```
model = YOLO('best.pt')

image_dir = 'word_key_yolo_used'
image_paths = [os.path.join(image_dir, file) for file in os.listdir(image_dir) if file.endswith(('png', 'jpg', 'jpeg'))]

results = model(image_paths)

save_dir = 'detected_images'
os.makedirs(save_dir, exist_ok=True)

for i, (result, img_path) in enumerate(zip(results, image_paths)):
    image_name = Path(img_path).name
    detections = []

    if result.bboxes:
        boxes = result.bboxes.data
        for box in boxes:
            class_id = int(box[5])
            confidence = box[4]
            class_name = model.names[class_id]
            detections.append(f"{class_name} ({confidence:.2f})")

    print(f"Image: {image_name}, Detections: {'', '.join(detections)}")

    save_path = os.path.join(save_dir, f'key_{i}.jpg')
    result.save(save_path)
```




Volume Up 0.93



F12 0.97

F12

Screen Capture 0.98



F8 0.93

F8

Alt 0.98

alt

Command 0.81

cmd ⌘

Action 2 0.79



end 0.96

end

YOLO V8n

```
1 from ultralytics import YOLO
2
3 def main():
4     # Load a model
5     model = YOLO("yolov8n.yaml") # build a new model from scratch
6     model = YOLO("yolov8n.pt") # load a pretrained model (recommended for training)
7
8     # Use the model
9     model.train(data="momo640.yaml", epochs=100) # train the model
10    metrics = model.val() # evaluate model performance on the validation set
11    results = model("https://attach.setn.com/newsimages/2019/07/09/2010347-XXL.jpg")
12    path = model.export(format="onnx") # export the model to ONNX format
13
14 if __name__ == '__main__':
15     main()
```

```
# YOLOv8.0n backbone
backbone:
  # [from, repeats, module, args]
  - [-1, 1, Conv, [64, 3, 2]] # 0-P1/2
  - [-1, 1, Conv, [128, 3, 2]] # 1-P2/4
  - [-1, 3, C2f, [128, True]]
  - [-1, 1, Conv, [256, 3, 2]] # 3-P3/8
  - [-1, 6, C2f, [256, True]]
  - [-1, 1, Conv, [512, 3, 2]] # 5-P4/16
  - [-1, 6, C2f, [512, True]]
  - [-1, 1, Conv, [1024, 3, 2]] # 7-P5/32
  - [-1, 3, C2f, [1024, True]]
  - [-1, 1, SPPF, [1024, 5]] # 9

# YOLOv8.0n head
head:
  - [-1, 1, nn.Upsample, [None, 2, "nearest"]]
  - [[-1, 6], 1, Concat, [1]] # cat backbone P4
  - [-1, 3, C2f, [512]] # 12

  - [-1, 1, nn.Upsample, [None, 2, "nearest"]]
  - [[-1, 4], 1, Concat, [1]] # cat backbone P3
  - [-1, 3, C2f, [256]] # 15 (P3/8-small)

  - [-1, 1, Conv, [256, 3, 2]]
  - [[-1, 12], 1, Concat, [1]] # cat head P4
  - [-1, 3, C2f, [512]] # 18 (P4/16-medium)

  - [-1, 1, Conv, [512, 3, 2]]
  - [[-1, 9], 1, Concat, [1]] # cat head P5
  - [-1, 3, C2f, [1024]] # 21 (P5/32-large)

  - [[15, 18, 21], 1, Detect, [nc]] # Detect(P3, P4, P5)
```

YOLO V9c

```
from ultralytics import YOLO

def main():
    # Load a model
    model = YOLO("yolov9c.yaml") # build a new model from scratch
    model = YOLO("yolov9c.pt") # load a pretrained model (recommended for training)

    # Use the model
    model.train(data="momo640.yaml", epochs=100) # train the model
    metrics = model.val() # evaluate model performance on the validation set
    results = model("https://attach.setn.com/newsimages/2019/07/09/2010347-XXL.jpg")
    path = model.export(format="onnx") # export the model to ONNX format

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

```
# gelan backbone
backbone:
  - [-1, 1, Conv, [64, 3, 2]] # 0-P1/2
  - [-1, 1, Conv, [128, 3, 2]] # 1-P2/4
  - [-1, 1, RepNCSPeLan4, [256, 128, 64, 1]] # 2
  - [-1, 1, ADown, [256]] # 3-P3/8
  - [-1, 1, RepNCSPeLan4, [512, 256, 128, 1]] # 4
  - [-1, 1, ADown, [512]] # 5-P4/16
  - [-1, 1, RepNCSPeLan4, [512, 512, 256, 1]] # 6
  - [-1, 1, ADown, [512]] # 7-P5/32
  - [-1, 1, RepNCSPeLan4, [512, 512, 256, 1]] # 8
  - [-1, 1, SPpELan, [512, 256]] # 9

head:
  - [-1, 1, nn.Upsample, [None, 2, 'nearest']]
  - [[-1, 6], 1, Concat, [1]] # cat backbone P4
  - [-1, 1, RepNCSPeLan4, [512, 512, 256, 1]] # 12

  - [-1, 1, nn.Upsample, [None, 2, 'nearest']]
  - [[-1, 4], 1, Concat, [1]] # cat backbone P3
  - [-1, 1, RepNCSPeLan4, [256, 256, 128, 1]] # 15 (P3/8-small)

  - [-1, 1, ADown, [256]]
  - [[-1, 12], 1, Concat, [1]] # cat head P4
  - [-1, 1, RepNCSPeLan4, [512, 512, 256, 1]] # 18 (P4/16-medium)

  - [-1, 1, ADown, [512]]
  - [[-1, 9], 1, Concat, [1]] # cat head P5
  - [-1, 1, RepNCSPeLan4, [512, 512, 256, 1]] # 21 (P5/32-large)

  - [[15, 18, 21], 1, Detect, [nc]] # Detect(P3, P4, P5)
```


YOLO V8n



YOLO V9c

