基于深度学习的目标检测

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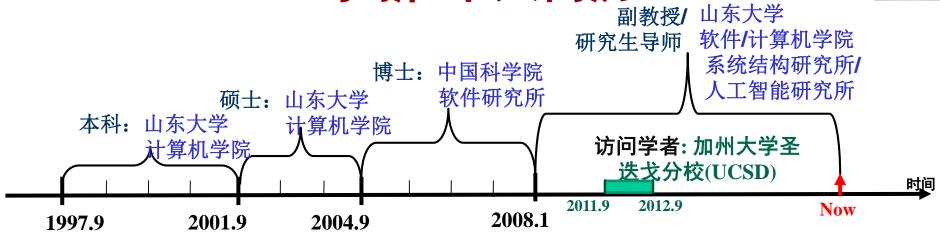






李新 个人简历





- > 研究方向
 - 目标检测与跟踪
 - 无人机智能巡检
 - 大数据处理

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电话: 138-531-23559

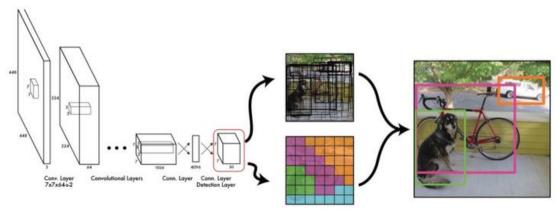




单阶段(1-stage)检测模型代表作: Yolo



单阶段模型没有中间的区域检出过程,直接从图片获得预测结果,也被成为Region-free方法。



- 1.准备数据:将图片缩放,划分为等分的网格,每个网格按跟Ground Truth的IoU分配到所要预测的样本。
- 2.卷积网络:由GoogleNet更改而来,每个网格对每个类别预测一个条件概率值,并在网格基础上生成B个box,每个box预测五个回归值,四个表征位置,第五个表征这个box含有物体(注意不是某一类物体)的概率和位置的准确程度(由IoU表示);
- 3.后处理: 使用NMS(Non-Maximum Suppression, 非极大抑制)过滤得到最后的预测框。

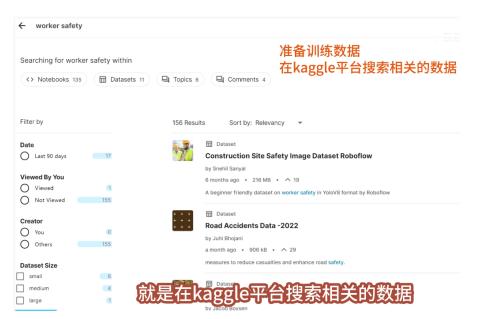
YOLO相比两阶段方法,其速度优势明显。但YOLO本身也存在一些问题,如划分网格较为粗糙,每个网格生成的box个数等限制了对小尺度物体和相近物体的检测。

论文链接: You Only Look Once: Unified, Real-Time Object Detection



➤在kaggle平台(需翻墙)搜相关数据

搜索Safety Helmet或者 Worker safety





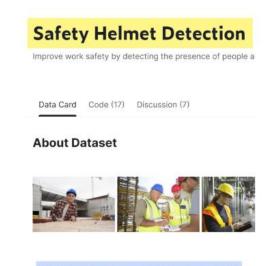


https://github.com/njvisionpower/Safety-Helmet-Wearing-Dataset



▶在kaggle平台(需翻墙)搜相关数据

搜索Safety Helmet或者 Worker safety



数据格式: PASCAL VOC

数据格式YoloV8,可直接使用

Construction Site Safety

Image Dataset Roboflow

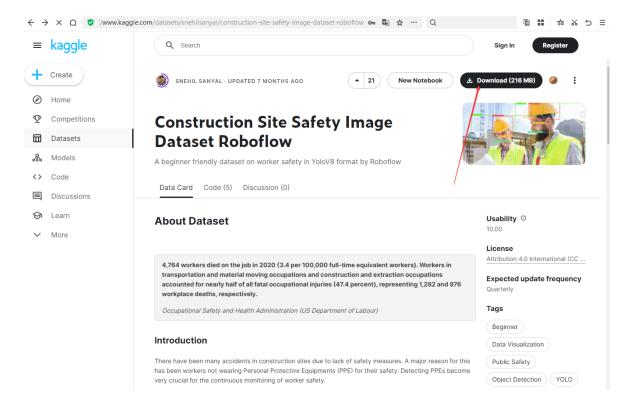
Code (3) Discussion (0)

Data Card

A beginner friendly dataset on worker safety in YoloV8 format by Roboflow

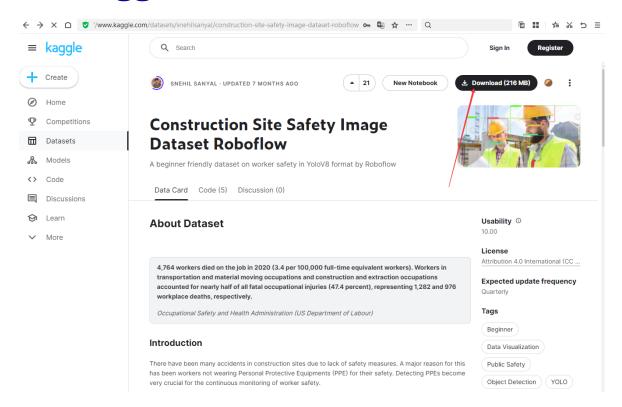


➤在kaggle平台(需翻墙)下载数据





➤在kaggle平台(需翻墙)下载数据

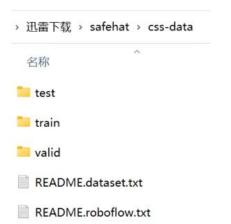




2. 训练数据



2605个训练数据





2 jpg.rf.2a2cafa3fd6db594c2f3a46f971b1c14.txt
2 jpg.rf.2eba3dc769a0689dda8f6eb3fdbd297e.txt
2 jpg.rf.5d1d2457d0de245aa673442ac06891df.txt
2 jpg.rf.7baa8d1dade31fde5df99a239a274fab.txt
2 jpg.rf.7c5f480de276d0dbd11edbcc4a81c19f.txt
2 jpg.rf.9ea7125de2c58de7327b2b061398d7f4.txt
2 jpg.rf.9ea6d55026ceb2e487c40a501b81f8.txt
2 jpg.rf.90ee66d55026ceb2e487c40a501b81f8.txt
2 jpg.rf.9360d1df64e52841007492e10100ffcd.txt

train/image保存了图片数据

train/labels保存了数据的标记结果

2. 训练数据





1	0	0.4984375 0.1109375 0.021875 0.0625
2	0	0.41640625 0.1328125 0.0578125 0.0546875
3	3	0.41640625 0.18671875 0.025 0.0296875
4	7	0.39609375 0.2953125 0.1125 0.2015625
5	7	0.48828125 0.2703125 0.0421875 0.184375
6	5	0.48515625 0.30546875 0.0484375 0.4484375
7	5	0.4078125 0.315625 0.1375 0.428125
8	8	0.6828125 0.14921875 0.346875 0.2984375
9	8	0.534375 0.21171875 0.05 0.1578125
10	3	0.07578125 0.9703125 0.0828125 0.059375
11	0	0.0859375 0.83671875 0.171875 0.2359375
12	5	0.1375 0.85703125 0.275 0.2859375
13	7	0.609375 0.76484375 0.2 0.4703125
14	5	0.68046875 0.76484375 0.3421875 0.4703125

数据图片

标记数据

> txt标记文件中每一行表示一个目标物体类别序号和位置

2. 训练数据



width = 0.021 * image_width height = 0.062 * mage height



1	0	0.4984375 0.1109375 0.021875 0.0625	
2	0	0.41640625 0.1328125 0.0578125 0.054687	5
3	3	0.41640625 0 18671875 0 025 0 0296875	
4	7	②- class_id: 类别编号	
5	7	0.4	
6	5	0.4 x_center: 边界框中心点x坐标 13	7
7		0 4	
8	8	o.e y_center: 边界框中心点y坐标 15	
9	8	0.5 width: 边界框的宽度 /5	
10	3	0.(WIGHT: 边乔性的免疫 /5	
11	0	0.0 height: 边界框的高度 ⁷⁵	
12	5	0.1 Height. 心が性的同皮	
13	7	0.609375 0.76484375 0.2 0.4703125	
14	5	0.68046875 0.76484375 0.3421875 0.47031	2

数据图片

标记数据

▶ 数据只有2605张图片,包括20000多个标记结果

3. 搭建环境



```
(base) C:\Users\49034>conda create -n safehat python=3.10
Collecting package metadata (current_repodata.json): done
Solving environment: done
```

```
(base) C:\Users\49034>activate safehat
(safehat) C:\Users\49034>pip install ultralytics
```

创建环境并安装python: conda create -n safehat python=3.10

激活环境: activate safehat

安装YOLO: pip install ultralytics

3.搭建环境



```
afehat. vaml
     # 模型训练时使用的yaml配置文件。该文件说明了数据的地址和待训练的类别
     # 配置三个路径,分别对应训练数据、验证数据、测试数据的地址
 2
 3
     # 训练数据用于模型的训练
 4
     train: C:\Users\49034\Desktop\yolo\safehat\data\train\images\
 5
     # 验证数据用于模型训练过程中的评估和参数调试
 6
 7
     val: C:\Users\49034\Desktop\yolo\safehat\data\valid\images\
     # 测试数据用于模型完成训练后的测试
 9
     test: C:\Users\49034\Desktop\yolo\safehat\data\test\images\
11
     # number of classes
12
     nc: 10
13
 14
     # class names
     #names: ['0', '1', '2', '3', '4', '5', '6', '7', '8', '9']
15
16
     #Hardhat - 安全帽
17
     #Mask - 口罩
     #NO-Hardhat - 无安全帽
18
     #NO-Mask - 无口罩
19
     #NO-Safety Vest - 无安全背心
21
     #Person - 人
     #Safety Cone - 安全锥
     #Safety Vest - 安全背心
23
     #machinery - 机械(挖掘机)
24
     #vehicle - 车辆
26
     names: [Hardhat, Mask, NO-Hardhat, NO-Mask, NO-Safety Vest, Person, Safety Cone, Safety Vest, machinery, vel
```

3. 训练



```
yolo_train.py 🏻
    from ultralytics import YOLO #导入Yolo模块
    # 加载yolov8的预训练模型,这个模型是yolov8使用了coco数据集训练的通用目标检测模型,
    # 我们将它作为基础模型,在该模型的基础上,训练安全帽模型
4
    model = YOLO('yolov8n.pt') # 用于加载模型
5
6
    # 训练用户自定义的数据集,数据的配置保存在safehat.yaml中,epochs等于100表示100轮迭代
    model.train(data='safehat.yaml', epochs=100)
9
10
    # 使用验证集验证效果
11
    model.val()
12
13
14
15
16
17
18
19
20
21
```

2. 训练文件



```
golo_train.py 🗵
    from ultralytics import YOLO #导入Yolo模块
 3
    # 加载yolov8的预训练模型,这个模型是yolov8使用了coco数据集训练的通用目标检测模型,
    # 我们将它作为基础模型,在该模型的基础上,训练安全帽模型
 4
    model = YOLO 'yolov8n.pt'
                          # 用于加载模型
 5
 6
 7
    # 训练用户自定义的数据集,数据的配置保存在safehat.yaml中,epochs等于100表示100轮迭代
    model.train(data= safehat.yaml , epochs=100)
 8
 9
    # 使用验证集验证效果
10
11
    model.val()
12
13
14
15
16
17
18
19
20
21
22
23
```

```
(safehat) C:\Users\49034\Desktop\yolo\safehat>python yolo_train.py
Ultralytics YOLOv8.0.170 Python-3.10.12 torch-2.0.1+cpu CPU (13th Gen Intel Core(小黑黑油各)(0)
engine\trainer: task=detect, mode=train, model=yolov8n.pt, data=safehat.yaml, epochs=100, patience=50
=False, device=None, workers=8, project=None, name=None, exist_ok=False, pretrained=True, optimizer=a
ls=False, rect=False, cos_lr=False, close_mosaic=10, resume=False, amp=True, fraction=1.0, profile=Fa
out=0.0, val=True, split=val, save_json=False, save_hybrid=False, conf=None, iou=0.7, max_det=300, ha
e, save_txt=False, save_conf=False, save_crop=False, show_labels=True, show_conf=True, vid_stride=1,
augment=False, agnostic_nms=False, classes=None, retina_masks=False, boxes=True, format=torchscript,
simplify=False, opset=None, workspace=4, nms=False, lr0=0.01, lrf=0.01, momentum=0.937, weight_decay
_bias_lr=0.1, box=7.5, cls=0.5, dfl=1.5, pose=12.0, kobj=1.0, label_smoothing=0.0, nbs=64, hsv_h=0.01
ale=0.5, shear=0.0, perspective=0.0, flipud=0.0, fliplr=0.5, mosaic=1.0, mixup=0.0, copy_paste=0.0, c
in3
Overriding model.yaml nc=80 with nc=10
                   from n
                                      module
                                                                                   arguments
                              params
                                      ultralytics.nn.modules.conv.Conv
                                                                                   [3, 16, 3, 2]
                                 464
                                      ultralytics.nn.modules.conv.Conv
                                                                                   [16, 32, 3, 2]
                     -1 1
                               4672
                                      ultralytics.nn.modules.block.C2f
                                                                                   [32, 32, 1, True]
                               7360
                              18560
                                      ultralytics.nn.modules.conv.Conv
                                                                                   [32, 64, 3, 2]
  4
                     -1 2
                              49664
                                      ultralytics.nn.modules.block.C2f
                                                                                   [64, 64, 2, True]
                                      ultralytics.nn.modules.conv.Conv
                              73984
                                                                                   [64, 128, 3, 2]
                                      ultralytics.nn.modules.block.C2f
  6
                     -1 2
                              197632
                                                                                   [128, 128, 2, True
                                      ultralytics.nn.modules.conv.Conv
                     -1 1
                              295424
                                                                                   [128, 256, 3, 2]
  8
                     -1
                              460288
                                      ultralytics.nn.modules.block.C2f
                                                                                   [256, 256, 1, True
  9
                                     ultralytics.nn.modules.block_SPPF
                     -1 1
                                                                                   [256, 256, 5]
 10
                     -1 1
                                                                                   [None, 2, 'nearest
```

11

[-1, 6]

4

[1]



```
yolo_test.py 🗵
    from ultralytics import YOLO #在代码中导入volo模块
    # 导入训练好的模型best.pt
4
    model = YOLO('best.pt')
5
 6
    # 随意找一些测试数据
    # 图片数据和视频数据都可以,直接将数据传入接口就可以了
8
    model.predict('01094 jpg.rf.15d2452cfbd2f25016ddb1bb90f0e19a.jpg', save=True)
 9
    model.predict('indianworkers.mp4', save=True)
10
    # 自己构造一些数据
11
    # 在识别自己构造的数据时, 传入了classes = [0, 2],
12
    # 代表只输出0和2, 也就是安全帽是否佩戴这两个类别
13
    # line width = 30表示指定识别框的字体大小为30
14
15
    model.predict('myselfl.jpg', save=True, classes = [0, 2], line width = 30)
16
    model.predict('myself2.jpg', save=True, classes = [0, 2], line width = 30)
17
    model.predict('myself1.MP4', save=True, classes = [0, 2], line width = 30)
18
    model.predict('myself2.MP4', save=True, classes = [0, 2], line width = 30)
```



```
(safehat) C:\Users\49034\Desktop\yolo\safehat>python yolo_test.py
image 1/1 C:\Users\49034\Desktop\volo\safehat\01094_ipg.rf.15d2452cfbd2f25016ddb1bb90f0e19a.ipg: 640x640 4 Hardhats, 2 NO-Masks, 5 Persons, 3 Safet
2 machinerys, 141.5ms
Speed: 2.5ms preprocess, 141.5ms inference, 6.0ms postprocess per image at shape (1, 3, 640, 640)
Results saved to runs\detect\predict
WARNING inference results will accumulate in RAM unless 'stream=True' is passed, causing potential out-of-memory
errors for large sources or long-running streams and videos. See https://docs.ultralytics.com/modes/predict/ for help.
Example:
    results = model(source=..., stream=True) # generator of Results objects
    for r in results:
        boxes = r.boxes # Boxes object for bbox outputs
        masks = r.masks # Masks object for segment masks outputs
        probs = r.probs # Class probabilities for classification outputs
video 1/1 (1/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 4 Safety Vests, 2 machinerys, 85.3ms
video 1/1 (2/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 4 Safety Vests, 2 machinerys, 86.0ms
video 1/1 (3/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 3 Safetv Vests, 2 machinerys, 78.1ms
video 1/1 (4/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 2 Safety Vests, 2 machinerys, 76.2ms
video 1/1 (5/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 2 Safety Vests, 2 machinerys, 73.7ms
video 1/1 (6/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 4 Safety Vests, 2 machinerys, 70.5ms
video 1/1 (7/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats. 2 Persons. 4 Safety Vests. 2 machinerys. 74.6ms
video 1/1 (8/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 3 Safety Vests, 2 machinerys, 71.4ms
video 1/1 (9/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 2 Hardhats, 2 Persons, 3 Safety Vests, 2 machinerys, 77.7ms
video 1/1 (10/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 2 Hardhats, 2 Persons, 3 Safety Vests, 3 machinerys, 85.3ms
video 1/1 (11/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 2 Hardhats, 2 Persons, 2 Safety Vests, 3 machinerys, 83.8ms
video 1/1 (12/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 2 Hardhats, 2 Persons, 3 Safety Vests, 2 machinerys, 1 vehicle, 81.6ms
video 1/1 (13/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 2 Hardhats, 2 Persons, 4 Safety Vests, 3 machinerys, 84.1ms
video 1/1 (14/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 3 Safety Vests, 3 machinerys, 84.0ms
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video 1/1 (2/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 4 Safety Vests, 2 machinerys, 86.0ms
video 1/1 (3/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats. 2 Persons, 3 Safety Vests, 2 machinerys, 78.1ms
video 1/1 (4/648) C:\Users\49034\Desktop\yolo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 2 Safety Vests, 2 machinerys, 76.2ms
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video 1/1 (13/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 2 Hardhats. 2 Persons. 4 Safety Vests. 3 machinerys. 84.1ms
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video 1/1 (15/648) C:\Users\49034\Desktop\volo\safehat\indianworkers.mp4: 384x640 3 Hardhats, 2 Persons, 3 Safety Vests, 3 machinerys, 78.5ms
```









谢 谢!











