menggunakan data dari ultralytics

```
#install library ultralytics
!pip install ultralytics
```

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from ultralytics import YOLO
Membuat model YOLO baru
model = YOLO('yolov8n.yaml')

	from	n	params	module	arguments
0	-1	1	464	ultralytics.nn.modules.conv.Conv	[3, 16, 3, 2]
1	-1	1	4672	ultralytics.nn.modules.conv.Conv	[16, 32, 3, 2]
2	-1	1	7360	ultralytics.nn.modules.block.C2f	[32, 32, 1, True]
3	-1	1	18560	ultralytics.nn.modules.conv.Conv	[32, 64, 3, 2]
4	-1	2	49664	ultralytics.nn.modules.block.C2f	[64, 64, 2, True]
5	-1	1	73984	ultralytics.nn.modules.conv.Conv	[64, 128, 3, 2]
6	-1	2	197632	ultralytics.nn.modules.block.C2f	[128, 128, 2, True]
7	-1	1	295424	ultralytics.nn.modules.conv.Conv	[128, 256, 3, 2]
8	-1	1	460288	ultralytics.nn.modules.block.C2f	[256, 256, 1, True]
9	-1	1	164608	ultralytics.nn.modules.block.SPPF	[256, 256, 5]
10	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
11	[-1, 6]	1	0	ultralytics.nn.modules.conv.Concat	[1]
12	-1	1	148224	ultralytics.nn.modules.block.C2f	[384, 128, 1]
13	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
14	[-1, 4]	1	0	ultralytics.nn.modules.conv.Concat	[1]
15	-1	1	37248	ultralytics.nn.modules.block.C2f	[192, 64, 1]
16	-1	1	36992	ultralytics.nn.modules.conv.Conv	[64, 64, 3, 2]
17	[-1, 12]	1	0	ultralytics.nn.modules.conv.Concat	[1]
18	-1	1	123648	ultralytics.nn.modules.block.C2f	[192, 128, 1]

```
-1 1 147712 ultralytics.nn.modules.conv.Conv
[-1, 9] 1 0 ultralytics.nn.modules.conv.Concat
 19
                                                                                                      [128, 128, 3, 2]
 20
                                                                                                      [1]
                         -1 1 493056 ultralytics.nn.modules.block.C2f
21] 1 897664 ultralytics.nn.modules.head.Detect
                                                                                                      [384, 256, 1]
 21
 22
             [15, 18, 21] 1
                                                                                                      [80, [64, 128, 256]]
YOLOv8n summary: 225 layers, 3157200 parameters, 3157184 gradients, 8.9 GFLOPs
```

Load model YOLO yang sebelumnya sudah dilatih oleh ulralytics yang sudah berbobot model = YOLO('yolov8n.pt')

Downloading https://github.com/ultralytics/assets/releases/download/v0.0.0/yolov8n.pt 'yolov8n.pt'... 100%| 6.23M/6.23M [00:00<00:00, 103MB/s]

Train model dengan data coco128.yml yang disediaakan oleh ultraytics results = model.train(data='coco128.yaml', epochs=3)

elephant	·						
Zebra 128	elephant	128	17	0.903	0.824	0.87	0.67
giraffe	bear	128		0.65	1	0.995	0.895
backpack umbrella 128 6 0.708 0.415 0.43 0.291 umbrella 128 18 0.76 0.527 0.65 0.438 0.291 tie 128 19 0.476 0.6561 0.161 0.0869 tie 128 7 0.7 0.714 0.685 0.48 suitcase 128 4 0.578 1 0.845 0.551 frisbee 128 5 0.657 0.77 0.732 0.633 skis 128 1 0.417 1 0.995 0.499 showboard 128 7 0.819 0.714 0.738 0.481 sports ball 128 6 0.724 0.448 0.53 0.291 kite 128 10 0.687 0.4 0.534 0.169 baseball bat 128 4 0.489 0.5 0.476 0.197 baseball glove 128 4 <t< td=""><td>zebra</td><td>128</td><td></td><td>0.861</td><td>1</td><td>0.995</td><td>0.965</td></t<>	zebra	128		0.861	1	0.995	0.965
umbrella 128 18 0.76 0.527 0.65 0.486 handbag 128 19 0.476 0.0501 0.161 0.0868 tie 128 7 0.77 0.714 0.685 0.48 suitcase 128 4 0.578 1 0.845 0.531 skis 128 1 0.417 1 0.995 0.409 snowboard 128 6 0.724 0.448 0.53 0.481 sports ball 128 6 0.724 0.448 0.53 0.291 kite 128 10 0.687 0.4 0.534 0.169 baseball bat 128 4 0.489 0.5 0.476 0.169 baseball bat 128 4 0.489 0.5 0.476 0.199 baseball bat 128 7 0.663 0.429 0.429 0.229 skateboard 128 6 0.757 <t< td=""><td>- U</td><td></td><td></td><td></td><td></td><td></td><td></td></t<>	- U						
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tie 128 7 0.7 0.714 0.685 0.48 suitcase 128 4 0.578 1 0.845 0.551 frisbee 128 5 0.657 0.77 0.732 0.639 skis 128 1 0.417 1 0.995 0.409 snowboard 128 7 0.819 0.714 0.738 0.481 sports ball 128 6 0.724 0.448 0.53 0.291 kite 128 10 0.687 0.4 0.534 0.169 baseball bat 128 4 0.489 0.5 0.476 0.197 baseball glove 128 7 0.663 0.429 0.429 0.294 skateboard 128 5 0.776 0.6 0.599 0.426 tennis racket 128 18 0.487 0.333 0.352 0.217 wine glass 128 16 0.75 0.375 0.551 0.314 wine glass 128 16 0.75 0.375 0.551 0.314 fork 128 6 1 0.293 0.372 0.256 knife 128 16 0.669 0.579 0.607 0.357 spoon 128 22 0.596 0.182 0.34 0.183 bowl 128 28 0.685 0.621 0.597 0.495 banana 128 1 0 0.499 0.579 0.607 0.357 spoon 128 22 0.596 0.182 0.34 0.183 bowl 128 28 0.685 0.621 0.597 0.495 broccoli 128 11 0.659 0.182 0.259 0.259 carrot 128 24 0.688 0.367 0.644 0.403 hot dog 128 2 0.567 1 0.828 0.795 pizza 128 5 0.674 1 0.995 0.669 pizza 128 3 0.663 0.367 0.644 0.403 hot dog 128 2 0.556 1.82 0.259 0.229 carrot 128 24 0.688 0.367 0.644 0.403 hot dog 128 2 0.556 1.82 0.595 0.882 donut 128 14 0.659 0.182 0.259 0.229 carrot 128 24 0.688 0.367 0.644 0.403 hot dog 128 2 0.567 1 0.828 0.796 pizza 128 5 0.674 1 0.995 0.669 pizza 128 5 0.674 1 0.995 0.882 donut 128 14 0.659 0.182 0.259 0.882 cake 128 4 0.819 1 0.995 0.882 couch 128 6 0.537 0.333 0.666 0.53 potted plant 128 13 0.518 0.579 0.513 0.411 toilet 128 13 0.518 0.579 0.513 0.411 toilet 128 13 0.518 0.579 0.513 0.412 fining table 128 13 0.518 0.579 0.513 0.412 microwave 128 8 0.849 0.5 0.576 0.497 cell phone 128 8 0.849 0.5 0.576 0.497 cell phone 128 8 0.899 0.590 0.591 coven 128 5 0.596 0.4 0.659 0.521 book 128 9 0.531 0.103 0.281 0.55 clock 128 9 0.531 0.103 0.281 0.15 clock 128 9 0.531 0.103 0.281 0.15 clock 128 9 0.511 0.079 0.0442 to 40.88 0.590 0.778 0.898 0.751 book 128 29 0.551 0.103 0.281 0.15 clock 128 9 0.511 0.079 0.0442 to 40.605 0.000 0.000 0.00000 0.0000 0.000		128		0.76	0.527		0.438
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frisbee 128 5 0.657 0.77 0.732 0.639 skis 128 1 0.417 1 0.995 0.4081 snowboard 128 7 0.819 0.714 0.448 0.53 0.291 kite 128 10 0.687 0.4 0.534 0.197 baseball bat 128 4 0.489 0.5 0.476 0.197 baseball glove 128 7 0.663 0.429 0.429 0.294 skateboard 128 5 0.776 0.6 0.599 0.426 tennis racket 128 7 0.663 0.429 0.429 0.242 tennis racket 128 16 0.75 0.357 0.551 0.314 0.297 bottle 128 18 0.487 0.333 0.352 0.217 wine glass 128 16 0.75 0.357 0.551 0.375 fork 128				0.7	0.714	0.685	0.48
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laptop 128 3 1 0 0.533 0.426 mouse 128 2 1 0 0 0 remote 128 8 0.849 0.5 0.576 0.497 cell phone 128 8 0 0 0.0427 0.0205 microwave 128 3 0.418 1 0.83 0.699 oven 128 5 0.351 0.4 0.357 0.281 sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442							
mouse 128 2 1 0 0 0 remote 128 8 0.849 0.5 0.576 0.497 cell phone 128 8 0 0 0.0427 0.0205 microwave 128 3 0.418 1 0.83 0.699 oven 128 5 0.351 0.4 0.357 0.281 sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
remote cell phone 128 8 0.849 0.5 0.576 0.497 cell phone microwave 128 8 0 0 0.0427 0.0205 microwave 128 3 0.418 1 0.83 0.699 oven 128 5 0.351 0.4 0.357 0.281 sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
cell phone 128 8 0 0 0.0427 0.0205 microwave 128 3 0.418 1 0.83 0.699 oven 128 5 0.351 0.4 0.357 0.281 sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
microwave 128 3 0.418 1 0.83 0.699 oven 128 5 0.351 0.4 0.357 0.281 sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421	remote	128	8	0.849	0.5	0.576	0.497
oven sink 128 sink 5 0.351 0.4 0.357 0.281 sink sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.369 0.63 0.421 teddy bear 128 21 1 0.369 0.63 0.421							
sink 128 6 0.351 0.167 0.242 0.166 refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
refrigerator 128 5 0.506 0.4 0.659 0.521 book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
book 128 29 0.531 0.103 0.281 0.15 clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
clock 128 9 0.819 0.778 0.898 0.751 vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421	•						
vase 128 2 0.44 1 0.828 0.745 scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
scissors 128 1 1 0 0.199 0.0442 teddy bear 128 21 1 0.369 0.63 0.421							
teddy bear 128 21 1 0.369 0.63 0.421							
toothbrush 128 5 1 0.484 0.828 0.528							
	toothbrush	128	5	1	0.484	0.828	0.528

Speed: 0.6ms preprocess, 3.2ms inference, 0.0ms loss, 3.6ms postprocess per image

Results saved to runs/detect/train

Evaluasi model dengan dataset validasi
results = model.val()

```
elephant
                128
                          17
                                0.904
                                        0.824
                                                 0.87
                                                          0.671
                                        1
1
       bear
                128
                          1
                                0.651
                                                 0.995
                                                          0.895
      zebra
                128
                                0.86
                                                0.995
                                                         0.953
                         9
     giraffe
                128
                               0.805
                                        0.921
                                               0.955
                                                          0.749
                                       0.333
    backpack
                128
                          6
                               0.577
                                                0.375
                                                         0.256
    umbrella
                128
                         18
                               0.758
                                        0.524
                                                 0.653
                                                          0.439
     handbag
                                0.46
                128
                         19
                                        0.0484
                                                 0.16
                                                        0.0871
                         7
                                               0.684
        tie
               128
                              0.705
                                       0.714
                                                         0.48
                                        1
                                               0.845
               128
                          4
                               0.582
                                                         0.551
    suitcase
     frisbee
                128
                          5
                               0.656
                                       0.769
                                             0./3∠
0.995
                                                0.732
                                                         0.639
                              0.424
               128
                         1
       skis
                                        1
                                                         0.41
   snowboard
               128
                         7
                               0.823
                                       0.714
                                               0.739
                                                         0.48
               128
                                        0.446
                                                         0.283
  sports ball
                         6
                               0.723
                                               0.514
       kite
                128
                         10
                               0.817
                                       0.451
                                                0.559
                                                         0.172
 baseball bat
                128
                          4
                               0.391
                                        0.25
                                                0.347
                                                         0.179
               128
                                       0.429
                                                0.429
baseball glove
                          7
                               0.645
                                                         0.33
                         5
   skateboard
               128
                               0.853
                                        0.6
                                               0.599
                                                         0.44
                               0.716
               128
                                                         0.331
tennis racket
                         7
                                        0.371
                                                0.46
                                                 0.354
     bottle
                128
                         18
                               0.444
                                        0.333
                                                          0.21
               128
                         16
                               0.749
                                       0.375
                                                0.526
                                                         0.319
  wine glass
        cup
               128
                        36
                                0.63
                                       0.285
                                               0.406
                                                        0.291
                                               0.273
       fork
                128
                               0.604
                                                         0.216
                         6
                                       0.167
       knife
                128
                         16
                               0.622
                                        0.562
                                                 0.614
                                                         0.368
      spoon
                128
                         22
                               0.693
                                        0.182
                                                0.349
                                                         0.195
                               0.748
               128
                        28
                                                0.63
                                                         0.519
       bowl
                                       0.635
                                 0
     banana
               128
                         1
                                        0
                                               0.142 0.0435
                                               0.497
                              0.405 0.715
                         2
4
    sandwich
                128
                                                         0.497
                128
                                1
                                        0.441
                                                0.995
                                                         0.669
      orange
                              0.673 0.182
               128
                        11
                                                0.277
                                                         0.221
    broccoli
                               0.65
                128
                        24
                                        0.375
                                               0.647
                                                         0.404
      carrot
                               0.555
     hot dog
                128
                         2
                                        1
                                                0.828
                                                         0.828
      pizza
                128
                          5
                               0.696
                                           1
                                                0.995
                                                          0.866
                        14
                              0.623
                                        1
1
                                               0.922
      donut
                128
                                                         0.838
                         4
       cake
               128
                               0.68
                                               0.995
                                                        0.834
               128
                                               0.421
                        35 0.448 0.514
      chair
                                                         0.255
                                      0.5
0.57
      couch
                128
                          6
                               0.643
                                                0.739
                                                         0.584
                              0.571
                        14
               128
                                               0.666
 potted plant
                                                         0.462
               128
                                1
       bed
                         3
                                       0.922
                                               0.995
                                                        0.777
               128
 dining table
                        13 0.464 0.462
                                               0.487
                                                         0.388
                        2
                                               0.995
                                                        0.946
                128
                                       0.853
                                1
     toilet
                               0.549
                                        0.5
                128
                          2
                                                0.638
                                                          0.61
                                        0 0.469 0.577
0 0.0388 0.00388
2 605 0.511
                         3 1 0 0.469
2 1 0 0.0388
8 0.851 0.5 0.605
8 0 0 0.0422
                         3
                128
     lanton
      mouse
               128
                128
      remote
   cell phone
                128
                                                         0.0205
                         3 0.325 0.667
5 0.348 0.4
               128
                                               0.789
   microwave
                                                        0.691
                                               0.359
       oven
               128
                                                        0.282
                128
                         6
                               0.381 0.167
       sink
                                                0.24
                                                         0.164
 refrigerator
                128
                          5
                               0.526
                                         0.4
                                                0.652
                                                         0.476
                               0.453
                        29
       book
                128
                                        0.103
                                                0.282
                                                         0.151
                         9 0.818
                                        0.778
                                                0.903
                128
                                                         0.735
      clock
                128
                         2
                              0.453
                                        1
                                                0.828
                                                         0.745
       vase
                          1
                                1
1
                128
                                           0
                                                 0.249
                                                         0.0546
    scissors
   teddy bear
                128
                          21
                                        0.368
                                                 0.627
                                                          0.419
                         5
  toothbrush
                128
                               0.683
                                        0.439
                                                0.815
                                                          0.53
```

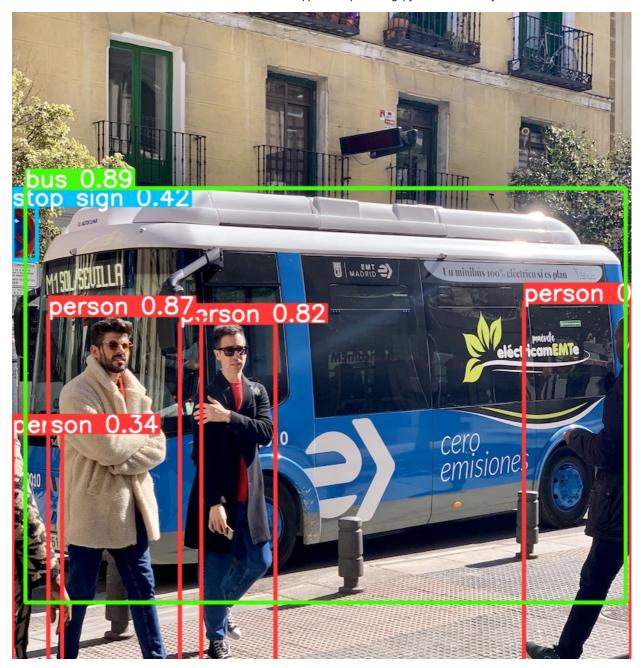
Speed: 0.2ms preprocess, 24.5ms inference, 0.0ms loss, 8.1ms postprocess per image Results saved to runs/detect/train2

```
# Melakukan deteksi objek dengan model yang sudah dibuat
results = model('https://ultralytics.com/images/bus.jpg')
```

```
retrigerator , /3: DOOK , /4: CLOCK , /5: Vase , /6: SCISSORS , //: teddy bear , /8: main unier , /9:
'toothbrush'}
orig_img: array([[[122, 148, 172],
        [120, 146, 170],
        [125, 153, 177],
        [157, 170, 184],
        [158, 171, 185],
        [158, 171, 185]],
        [[127, 153, 177],
        [124, 150, 174],
        [127, 155, 179],
        [158, 171, 185],
        [159, 172, 186],
        [159, 172, 186]],
        [[128, 154, 178],
        [126, 152, 176],
        [126, 154, 178],
        [158, 171, 185],
         [158, 171, 185],
        [158, 171, 185]],
       [[185, 185, 191],
        [182, 182, 188],
        [179, 179, 185],
        [114, 107, 112],
        [115, 105, 111],
        [116, 106, 112]],
        [[157, 157, 163],
        [180, 180, 186],
        [185, 186, 190],
         [107, 97, 103],
        [102, 92, 98],
        [108, 98, 104]],
       [[112, 112, 118],
[160, 160, 166],
        [169, 170, 174],
        [ 99, 89, 95],
[ 96, 86, 92],
        [102, 92, 98]]], dtype=uint8)
orig_shape: (1080, 810)
path: '/content/bus.jpg
probs: None
save dir: 'runs/detect/train3'
chood. Sinnonnococci. 6 608131561370307 'infononcoi. 13 800/156003383780 'nostnococci. 1 080066300/38/176613
```

```
import os
from IPython.display import display, Image
from IPython import display
display.clear_output()
```

Image(filename=f'/content/runs/detect/train3/bus.jpg',width=600)



Export the model to ONNX format
success = model.export(format='onnx')

PyTorch: starting from 'runs/detect/train/weights/best.pt' with input shape (1, 3, 640, 640) BCHW and output shap
requirements: Ultralytics requirement ['onnx>=1.12.0'] not found, attempting AutoUpdate...
Collecting onnx>=1.12.0

Downloading onnx-1.15.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (15.7 MB)

Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from onnx>=1.12.0) (1.23.5) Requirement already satisfied: protobuf>=3.20.2 in /usr/local/lib/python3.10/dist-packages (from onnx>=1.12.0) (3 Installing collected packages: onnx

Installing collected packages: onnx Successfully installed onnx-1.15.0

requirements: AutoUpdate success ✓ 8.5s, installed 1 package: ['onnx>=1.12.0'] requirements: ⚠ Restart runtime or rerun command for updates to take effect

https://colab.research.google.com/drive/1FcRvBYKelt3hvfeXel0hDF0Mrjrk5C_l#scrollTo=PrbrUtSgcpa-&printMode=true

ONNX: starting export with onnx 1.15.0 opset 17...

ONNX: export success 2 9.4s, saved as 'runs/detect/train/weights/best.onnx' (12.2 MB)

Export complete (11.1s)

Results saved to $\c / content/runs/detect/train/weights$

Predict: yolo predict task=detect model=runs/detect/train/weights/best.onnx imgsz=640

Validate: yolo val task=detect model=runs/detect/train/weights/best.onnx imgsz=640 data=/usr/local/lib/pyt

Visualize: https://netron.app

Menggunakan dataset sendiri

```
import locale
locale.getpreferredencoding = lambda: "UTF-8"
```

!pip install ultralytics

```
Requirement already satisfied: ultralytics in /usr/local/lib/python3.10/dist-packages (8.0.206)
Requirement already satisfied: matplotlib>=3.3.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (3
Requirement already satisfied: numpy>=1.22.2 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (1.23.
Requirement already satisfied: opencv-python>=4.6.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics)
Requirement already satisfied: pillow>=7.1.2 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (9.4.0
Requirement already satisfied: pyyaml>=5.3.1 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (6.0.1
Requirement already satisfied: requests>=2.23.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (2.
Requirement already satisfied: scipy>=1.4.1 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (1.11.3
Requirement already satisfied: torch>=1.8.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (2.1.0+
Requirement already satisfied: torchvision>=0.9.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (
Requirement already satisfied: tqdm>=4.64.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (4.66.1
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Requirement already satisfied: seaborn>=0.11.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (0.1
Requirement already satisfied: psutil in /usr/local/lib/python3.10/dist-packages (from ultralytics) (5.9.5)
Requirement already satisfied: py-cpuinfo in /usr/local/lib/python3.10/dist-packages (from ultralytics) (9.0.0)
Requirement already satisfied: thop>=0.1.1 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (0.1.1.p
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Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3
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Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=1.1.4->ultra
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests>=2.23.0->ul
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests>=2.23
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests>=2.23
Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytic
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Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytics)
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Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from cycler>=0.10->matplotlib>=3.3
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->torch>=1.
Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dist-packages (from sympy->torch>=1.8.0-
```

!pip install roboflow

```
Requirement already satisfied: roboflow in /usr/local/lib/python3.10/dist-packages (1.1.9)
Requirement already satisfied: certifi==2023.7.22 in /usr/local/lib/python3.10/dist-packages (from roboflow) (202
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.
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) (1.23.5)
Requirement already satisfied: opencv-python-headless==4.8.0.74 in /usr/local/lib/python3.10/dist-packages (from
Requirement already satisfied: Pillow>=7.1.2 in /usr/local/lib/python3.10/dist-packages (from roboflow) (9.4.0)
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) (2.31.0)
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.16.0)
Requirement already satisfied: supervision in /usr/local/lib/python3.10/dist-packages (from roboflow) (0.16.0)
Requirement already satisfied: urllib3>=1.26.6 in /usr/local/lib/python3.10/dist-packages (from roboflow) (2.0.7)
Requirement already satisfied: tqdm>=4.41.0 in /usr/local/lib/python3.10/dist-packages (from roboflow) (4.66.1)
Requirement already satisfied: PyYAML>=5.3.1 in /usr/local/lib/python3.10/dist-packages (from roboflow) (6.0.1)
Requirement already satisfied: requests-toolbelt in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.
Requirement already satisfied: python-magic in /usr/local/lib/python3.10/dist-packages (from roboflow) (0.4.27)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->robo
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->rob
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->robof
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests
Requirement already satisfied: scipy<2.0.0,>=1.9.0 in /usr/local/lib/python3.10/dist-packages (from supervision->
```

```
from roboflow import Roboflow
rf = Roboflow(api_key="nBWWcXSg2urfLx1WVJfL")
project = rf.workspace("hochschule-munchen").project("groceryclassification")
dataset = project.version(5).download("yolov8")

loading Roboflow workspace...
loading Roboflow project...
Dependency ultralytics==8.0.196 is required but found version=8.0.206, to fix: `pip install ultralytics==8.0.196`
Downloading Dataset Version Zip in GroceryClassification-5 to yolov8:: 100%| 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% |
```

from ultralytics import YOLO
Membuat model YOLO baru
model2 = YOLO('yolov8n.yaml')

	from	n	params	module	arguments
0	-1	1	464	ultralytics.nn.modules.conv.Conv	[3, 16, 3, 2]
1	-1	1	4672	ultralytics.nn.modules.conv.Conv	[16, 32, 3, 2]
2	-1	1	7360	ultralytics.nn.modules.block.C2f	[32, 32, 1, True]
3	-1	1	18560	ultralytics.nn.modules.conv.Conv	[32, 64, 3, 2]
4	-1	2	49664	ultralytics.nn.modules.block.C2f	[64, 64, 2, True]
5	-1	1	73984	ultralytics.nn.modules.conv.Conv	[64, 128, 3, 2]
6	-1	2	197632	ultralytics.nn.modules.block.C2f	[128, 128, 2, True]
7	-1	1	295424	ultralytics.nn.modules.conv.Conv	[128, 256, 3, 2]
8	-1	1	460288	ultralytics.nn.modules.block.C2f	[256, 256, 1, True]
9	-1	1	164608	ultralytics.nn.modules.block.SPPF	[256, 256, 5]
10	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
11	[-1, 6]	1	0	ultralytics.nn.modules.conv.Concat	[1]
12	-1	1	148224	ultralytics.nn.modules.block.C2f	[384, 128, 1]
13	-1	1	0	torch.nn.modules.upsampling.Upsample	[None, 2, 'nearest']
14	[-1, 4]	1	0	ultralytics.nn.modules.conv.Concat	[1]
15	-1	1	37248	ultralytics.nn.modules.block.C2f	[192, 64, 1]
16	-1	1	36992	ultralytics.nn.modules.conv.Conv	[64, 64, 3, 2]
17	[-1, 12]	1	0	ultralytics.nn.modules.conv.Concat	[1]
18	-1	1	123648	ultralytics.nn.modules.block.C2f	[192, 128, 1]
19	-1	1	147712	ultralytics.nn.modules.conv.Conv	[128, 128, 3, 2]
20	[-1, 9]	1	0	ultralytics.nn.modules.conv.Concat	[1]
21	-1	1	493056	ultralytics.nn.modules.block.C2f	[384, 256, 1]
22	[15, 18, 21]	1	897664	ultralytics.nn.modules.head.Detect	[80, [64, 128, 256]]
Y0L0v8n	summary: 225 lay	ers,	3157200	parameters, 3157184 gradients, 8.9 GFLOPs	

[#] Train the model using the 'data.yaml' dataset for 10 epochs
results = model2.train(data='/content/GroceryClassification-5/data.yaml', epochs=10)

```
Class
                   Images Instances
                                        BOX (P
                                                    К
                                                            mAP50 mAP50-95): 100%| 3/3
           all
                      88
                               182
                                        0.776
                                                            0.317
                                                   0.26
                                                                      0.136
                 box_loss
        GPU_mem
                           cls_loss
                                     dfl_loss Instances
Epoch
                                                             Size
 8/10
          2.3G
                    1.514
                              2.044
                                        2.314
                                                              640: 100% | 63/63 [00:19<00
                                                    39
                                        Box (P
                                                     R
                                                            mAP50 mAP50-95): 100%
          Class
                   Images
                          Instances
           all
                      88
                               182
                                        0.418
                                                  0.479
                                                            0.426
                                                                      0.173
Epoch
        GPU_mem
                 box_loss
                           cls_loss
                                     dfl_loss Instances
                                                             Size
                                                              640: 100% 63/63 [00:20<00
9/10
          2.27G
                    1.497
                              2.007
                                        2.278
                                                    46
          Class
                   Images
                          Instances
                                        Box(P
                                                     R
                                                            mAP50
                                                                  mAP50-95): 100%
           all
                      88
                               182
                                        0.411
                                                  0.491
                                                            0.501
                                                                      0.214
        GPU_mem
Epoch
                 box_loss
                           cls_loss
                                     dfl_loss Instances
                                                             Size
                                                                            | 63/63 [00:19<00
          2.28G
                    1,401
                              1.857
                                                              640: 100%
10/10
                                        2.19
                                                    18
          Class
                   Images
                          Instances
                                        Box(P
                                                     R
                                                            mAP50
                                                                  mAP50-95): 100%| 3/3
           a11
                      88
                                        0.528
                                                  0.519
                                                            0.455
                                                                      0.191
                               182
```

10 epochs completed in 0.064 hours.

Optimizer stripped from runs/detect/train6/weights/last.pt, 6.2MB Optimizer stripped from runs/detect/train6/weights/best.pt, 6.2MB

Validating runs/detect/train6/weights/best.pt...

Ultralytics YOLOv8.0.206 ♥ Python-3.10.12 torch-2.1.0+cu118 CUDA:0 (Tesla T4, 15102MiB)

YOLOv8n summary (fused): 168 layers, 3006623 parameters, 0 gradients, 8.1 GFLOPs

Class	Images	Instances	Box(P	R	mAP50	mAP50-95):	100% 3/3
all	88	182	0.408	0.491	0.5	0.215	
0	88	60	0.845	0.85	0.92	0.477	
oatMilk	88	6	0.323	0.167	0.488	0.16	
pasta	88	40	0.205	0.225	0.132	0.0443	
potato	88	76	0.258	0.724	0.46	0.177	

Speed: 0.3ms preprocess, 2.5ms inference, 0.0ms loss, 4.3ms postprocess per image

Results saved to runs/detect/train6

Evaluate the model's performance on the validation set results = model2.val()

```
Ultralytics YOLOv8.0.206 

✓ Python-3.10.12 torch-2.1.0+cu118 CUDA:0 (Tesla T4, 15102MiB)
YOLOv8n summary (fused): 168 layers, 3006623 parameters, 0 gradients, 8.1 GFLOPs
```

val: Scanning /content/GroceryClassification-5/valid/labels.cache... 88 images, 1 backgrounds, 0 corrupt: 100% WARNING ⚠ Box and segment counts should be equal, but got len(segments) = 7, len(boxes) = 182. To resolve this c : 100%| | 6/6 [0

C	lass	Images	Instances	Box(P	R	mAP50	mAP50-95):
	all	88	182	0.414	0.491	0.501	0.214
	0	88	60	0.848	0.85	0.92	0.477
oat	Milk	88	6	0.324	0.167	0.488	0.16
p	asta	88	40	0.212	0.225	0.132	0.0441
рс	tato	88	76	0.271	0.724	0.464	0.176

Speed: 0.9ms preprocess, 6.9ms inference, 0.0ms loss, 5.0ms postprocess per image Results saved to runs/detect/train62

Run inference

model2.predict('/content/tomato18.jpg', save=True, imgsz=320, conf=0.5)

```
boxes: ultralytics.engine.results.Boxes object
```

keypoints: None masks: None

names: {0: '0', 1: 'oatMilk', 2: 'pasta', 3: 'potato', 4: 'tomato'}

orig_img: array([[[30, 74, 45],

[30, 74, 45], [30, 74, 43],

[3, 74, 48],

[1, 69, 44],

[0, 66, 41]],

[[31, 75, 46],

[31, 75, 46],

[31, 75, 46],

[3, 74, 48],

```
[ 2, 73, 47],
        [ 0, 68, 43],
        [ 0, 66, 41]],
       ...,
       [[12, 72, 62],
        [11, 71, 61],
        [12, 69, 60],
        [ 2, 12, 12],
        [ 2, 12, 12],
[ 1, 11, 11]],
       [[11, 71, 61],
        [11, 71, 61],
        [12, 69, 60],
        [ 1, 13, 13],
        [ 3, 13, 13],
        [ 3, 13, 13]],
       [[11, 71, 61],
        [10, 70, 60],
        [12, 69, 60],
        ...,
[ 2, 14, 14],
        [ 3, 13, 13],
        [ 3, 13, 13]]], dtype=uint8)
orig_shape: (534, 800)
path: '/content/tomato18.jpg'
probs: None
save_dir: 'runs/detect/train65'
speed: {'preprocess': 6.340503692626953, 'inference': 28.33867073059082, 'postprocess': 2.141714096069336}]
```

import os
from IPython.display import display, Image
from IPython import display
display.clear_output()

Image(filename=f'/content/runs/detect/train63/tomato18.jpg',width=600)

