Image-Object-Detection-PyTorch-YOLOR-GPL-Jupyter

One of Taiwan's proud recent new works, YOLOR, the most powerful object detection algorithm last year, greatly reduces the amount of computation, and increases the speed without reducing the accuracy. We organized the code so that we can use JupyterLab to perform the training and inference steps in sequence, which is easier to use, and produced an instruction slideshow.

Version 20230223

Applications

• The YOLOR solution can be applied to factory defect detection, medical image analysis, biological image analysis, industrial safety image analysis, mask image analysis, etc.



How to use

The main process is:

Annotate images -> Prepare files for training -> Training -> Inference

| Iupyter-Image-Object-Detection-YOLOR-GPL-1 |
|--|
| Name |
| 🖿 data |
| src |
| 1_annotation_pascal_voc_xml.ipynb |
| 2_convert_yolo_format.ipynb |
| 3_prepare_train_val_txt.ipynb |
| 4_delete_log.ipynb |
| 🖪 5_train.ipynb |
| 6_tensorboard.ipynb |
| 7_inference_image.ipynb |
| 8_inference_image_folder_1.ipynb |
| 9_inference_webcam.ipynb |

1_annotation_pascal_voc_xml.ipynb

Open the webpage for image annotation.

ipynb parameter:

- "port" is the port used by the webpage. If the port is occupied by the user, please change another port value by yourself.
- "dataset" is the dataset name
- "label_folder" is the image of the train folder, it can also be changed to "val" to label the image of the val folder.



See Annotation.pdf for how to use annotation pages.

2_convert_yolo_format.ipynb

Convert the voc xml label file to the yolo format. Before running, please confirm label.names under the label_file path in #parameters and whether the content filled in the category is correct.

supplement:

The content of label.names is the category name without background.

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| tumo | or | | | | | ~ |
| Ln 2, | Col 1 | | 100% | Unix (LF) | UTF-8 | |

3_prepare_train_val_txt.ipynb

Generate training and validation image path files train.txt and val.txt .

| train.tx | t - Notepad | iew Help | | | _ | | × |
|--|---|--|--|--|-----|----|---|
| data/tum data/tum data/tum data/tum data/tum data/tum data/tum data/tum data/tum | vormat v vor/train vor/train vor/train vor/train vor/train vor/train vor/train | /images/ /images/ /images/ /images/ /images/ /images/ /images/ /images/ /images/ | tumor-1 tumor-1 tumor-2 tumor-2 tumor-3 tumor-4 tumor-6 tumor-7 tumor-8 tumor-9 | .jpg 0.jpg 1.jpg .jpg .jpg .jpg .jpg .jpg .jpg | | | ~ |
| | Ln 1, Col 1 | | 100% | Windows (CRLF) | UTF | -8 | |



4_delete_log.ipynb

Delete the log files left over from previous training.

Set the training voc.yaml related parameters

Set the content of the voc.yaml file in the dataset, set the name of the data set, the number of categories and the name.



Set training yolor_csp_x.cfg related parameters

Set the content of the yolor_csp_x.cfg file in the dataset, set the number of categories.



5_train.ipynb

Start training.

ipynb parameter:

- dataset is the dataset name.
- weights_file is the pretrained model path used, None means not to use the pretrained model for training.
- devices is the GPU id used.
- epochs is the number of training epochs.

| Scanni Scanni Image Using Loggir Starti | Scanning labels data\tumor\train\labels.cache3 (10 found, 0 missing, 0 empty, 0 duplicate, for 10 images): 10it [00:00, 27.27it/s] Scanning labels data\tumor\val\labels.cache3 (3 found, 0 missing, 0 empty, 0 duplicate, for 3 images): 3it [00:00, ?it/s] Image sizes 512 train, 512 test Using 1 dataloader workers Logging results to data\tumor\model Starting training for 1000 epochs | | | | | | | | | | |
|--|--|---|--|---|--------------------------------|---|---|--|--|--|--|
| E | Epoch | gpu_mem | box | obj | cls | total | targets | img_size | | | |
| 6 | 0/999 | 5.98G | 0.03105 | 0.01745 | 0 | 0.0485 | 5 | 512: | 100% 2/2 | 2 [00:19<00:00, | 9.63s/it] |
| E | Epoch | gpu_mem | box | obj | cls | total | targets | img_size | | | |
| 1 | 1/999 | 7.39G | 0.02041 | 0.008021 | 0 | 0.02843 | 1 | 512: | 100% | 2 [00:00<00:00, | 2.78it/s] |
| E | Epoch | gpu_mem | box | obj | cls | total | targets | img_size | | | |
| 2 | 2/999 | 7.39G | 0.01482 | 0.008739 | 0 | 0.02356 | 0 | 512: | 100% | [00:00<00:00, | 2.37it/s] |
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| 3 | 3/999 | 7.39G | 0.02104 | 0.008882 | Ø | 0.02992 | 2 | 512: | 100% | [00:00<00:00, | 2.43it/s] |
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| on\lik | b\site-p | backages\t | :orch\func | tional.py:56 | 8: User | larning: to | rch.meshgr | id: in an | upcoming rele | ase, it will be | required to pass the indexing a |
| gument | t. (Irig upp VE | gered int | ternally a | t C:\action | t type, i | <pre>work\pyt</pre> | orch\pytor .definedl | ch\builder | \windows\pyto | orch\aten\src\Ale | en\native\lensorShape.cpp:2228.) |
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| | | all | | 3 | 3 | 0.439 | 1 | 0.913 | 0.73 | | |
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| E | Epoch | gpu mem | box | obi | cls | total | targets | img size | | | |
| 4 | 4/999 | 7.23G | 0.02456 | 0.01079 | Ø | 0.03535 | 3 | 512: | 100% 2/2 | [00:00<00:00, | 2.46it/s] |
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| | | all | : | 3 | 3 | 0.441 | 4 | | | | |
| | | | | | | 01112 | 1 | 0.913 | 0.73 | | |
| | | | | | | 01112 | 1 | 0.913 | 0.73 | | |
| E | Epoch | gpu_mem | box | obj | cls | total | targets | 0.913 | 0.73 | | |
| 5 | Epoch 5/999 | gpu_mem 7.24G | box 0.02236 | obj 0.01873 | cls Ø | total 0.04109 | targets | 0.913 img_size 512: | 0.73 | 2 [00:00<00:00, | 2.36it/s] |
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| 5 | Epoch 5/999 | gpu_mem 7.24G Class all | box 0.02236 Images | obj 0.01873 s Target 3 | cls Ø 3 | total 0.04109 P 0.386 | targets 2 R 1 | 0.913 img_size 512: mAP@.5 0.913 | 0.73 100% 2/2 mAP@.5:.95: 0.73 | 2 [00:00<00:00, 100% ∎ 1/1 [00: | 2.36it/s] 00<00:00, |
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| E | Epoch 5/999 Epoch 6/999 | gpu_mem 7.24G Class all gpu_mem 7.24G Class | box 0.02236 Image: box 0.02884 Image: | obj 0.01873 s Target 3 0.009161 s Target | cls 0 3 cls 0 5 | total 0.04109 P 0.386 total 0.038 P | targets 2 R 1 targets 2 R | 0.913 img_size 512: mAP@.5 0.913 img_size 512: mAP@.5 | 0.73 100% [] 2/2 mAP@.5:.95: 0.73 100% [] 2/2 mAP@.5:.95: | 2 [00:00<00:00, 100% 1/1 [00: 2 [00:00<00:00, 100% 1/1 [00: | 2.36it/s] 00<00:00, 2.45it/s] 00<00:00, |

6_tensorboard.ipynb

You can view the training loss curve and other related information through TensorBoard.



7_inference_image.ipynb

Infer a single image.

ipynb parameter:

- dataset is the dataset name.
- source is the inferred image path.
- weights_file is the inference model path.

dataset = "tumor" source = "data/%s/test/images/tumor-10.jpg" %(dataset) image_size = 512 yaml_file = "data/%s/label.names"%(dataset) cfg_file = "data/%s/yolor_csp_x.cfg"%(dataset) weights_file = "data/%s/model/best.pt" %(dataset) device = "0" threshold = 0.6

[5]: %run src/detect.py --source \$source --img-size \$image_size --names \$yaml_file --cfg \$cfg_file --weights \$weights_file --conf \$threshold --devi

Namespace(weights=['data/tumor/model/best.pt'], source='data/tumor/test/images/tumor-10.jpg', output='inference/output', img_size=512, conf_th
res=0.6, iou_thres=0.5, device='0', view_img=True, save_txt=False, classes=None, agnostic_nms=False, augment=False, update=False, cfg='data/tu
mor/yolor_csp_x.cfg', names='data/tumor/label.names', show_rate=False, save_img=False)

D:\App4AI-2222\gpu\python\lib\site-packages\torch\functional.py:568: UserWarning: torch.meshgrid: in an upcoming release, it will be required to pass the indexing argument. (Triggered internally at C:\actions-runner_work\pytorch\pytorch\builder\windows\pytorch\aten\src\ATen\native \TensorShape.cpp:2228.)

return _VF.meshgrid(tensors, **kwargs) # type: ignore[attr-defined]



8_inference_image_folder_1.ipynb

Infer all images in the folder.

ipynb parameter:

- dataset is the dataset name.
- source is the inferred image path.
- weights_file is the inference model path.

tumor-1

tumor 0.716797

Underkill Rate: 0(0.00%), Overkill Rate: 0(0.00%), Right Rate: 1(100.00%), Total: 1



9_inference_webcam.ipynb

Infer the image of the webcam. Press "q" on the display to turn the webcam off.

Reference

- Please refer to the readme.txt in the SDK folder.
- LEADERG AppForAI: https://www.leaderg.com/appforai-windows
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