Image-Object-Detection-YOLOv4-CPP-Jupyter

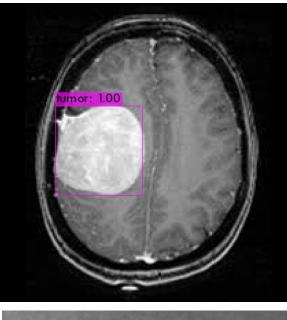
Russian Alexey Bochkovskiy, the maintainer of YOLO Darknet, found that the CSPNet detector developed by Wang Jianyao, the post-doc of the Chinese Academy of Sciences and director Liao Hongyuan, was fast and good, so he invited the Chinese Academy of Sciences to develop YOLOv4 for the backbone, and did various parts of the previous generation of YOLOv3. The improvement can not only maintain a certain detection speed, but also greatly improve the detection accuracy and reduce the usage of hardware.

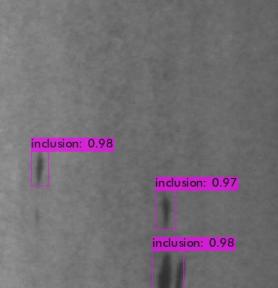
Version 20230223

Applications

 YOLOv4 solutions can be applied to factory defect detection, medical image analysis, biological image analysis, industrial security image analysis, mask image analysis, luxury car brand image analysis, etc.







How to use

The main process is:

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

Ħ	/ Jupyter-Image-Object-Detection-	YOLOv4-CPP-1	4 /
Nai	me		
	bin		
	data		
	src		
	1_annotation_pascal_voc_xml.ipynb		
	2_calculate_anchors_GPU.ipynb		
	3_convert_yolo_format.ipynb		
	4_prepare_train_txt.ipynb		
	5_prepare_val_txt.ipynb		
	6_prepare_config_file.ipynb		
	7_train_CPU.ipynb		
	7_train_GPU.ipynb		
	8_inference_CPU.ipynb		
	8_inference_GPU.ipynb		
	9_inference_webcam_CPU.ipynb		
	9_inference_webcam_GPU.ipynb		
	10_inference_folder_1_CPU.ipynb		
	10_inference_folder_1_GPU.ipynb		
	11_inference_folder_demo_CPU.ipynb		
	11_inference_folder_demo_GPU.ipynb		
	12_YOLOv4_auto_labeling_GPU.ipynb		

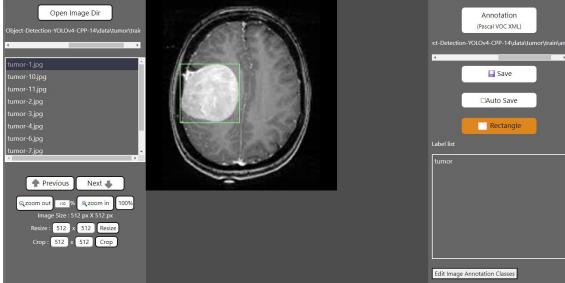
1_annotation_pascal_voc_xml.ipynb

→ C ① 127.0.0.1:8801

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.
- "image_folder" is the image path.
- "annotation_path" is the path to the annotation archive.



See Annotation.pdf for how to use annotation pages.

2_calculate_anchors_GPU.ipynb

Calculate the anchor value suitable for your dataset. Before running, please confirm the label.names in your dataset and whether the category filled in the content is correct.

supplement:

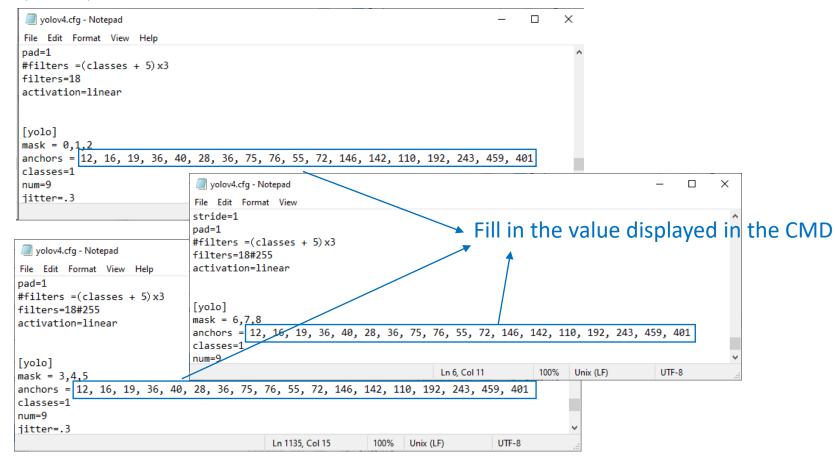
The content of label.names is the category name, excluding background, and the format distinguishes multiple categories by wrapping lines.

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CUDA-version: 11030 (11060), cuDNN: 8.2.1, CUDNN_HALF=1, GPU count: 1 CUDNN_HALF=1 OpenCV version: 4.5.5 num_of_clusters = 9, width = 512, height = 512 read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt anchors = 40, 30, 51, 57, 75, 45, 67, 81, 91, 86, 83,117, 120,115, 128,152, 159,157	CUDNN_HALF=1 OpenCV version: 4.5.5 num_of_clusters = 9, width = 512, height = 512 read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt	C:\Users\ai\Desktop\App4AI-2225\sdk\Jupyter-Image-Object-Detection-YOLOv4-CPP-14\src\darknet\build\darknet\x64\darknet.exe	-	×
<pre>num_of_clusters = 9, width = 512, height = 512 read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt</pre>	<pre>num_of_clusters = 9, width = 512, height = 512 read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt</pre>	CUDNN_HALF=1		
<pre>read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt</pre>	<pre>read labels from 10 images loaded image: 10 box: 11 all loaded. calculating k-means++ iterations = 1 counters_per_class = 11 avg IoU = 97.03 % Saving anchors to the file: anchors.txt</pre>			
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- Saving anchors to the file: anchors.txt	- Saving anchors to the file: anchors.txt	counters_per_class = 11		
		avg IoU = 97.03 %		
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		anchors = 40, 30, 51, 57, 75, 45, 67, 81, 91, 86, 83,117, 120,115, 128,152, 159,157		

Set yolov4.cfg anchor parameters

Fill in the value generated after running 2_calculate_anchors_GPU.ipynb into the yolov4.cfg anchor point parameter in the dataset dataset.



3_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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4_prepare_train_txt.ipynb

Generate training image path files "train.txt".

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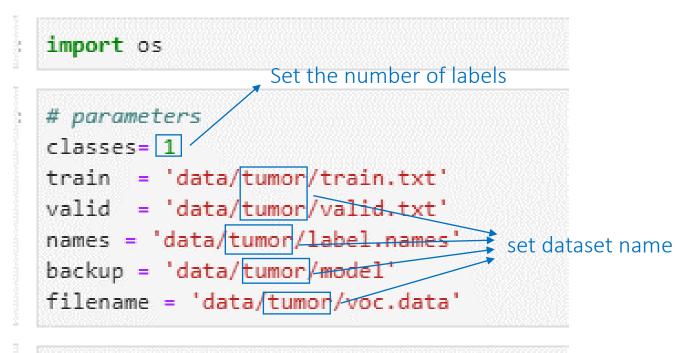
5_prepare_val_txt.ipynb

Generate validation image path files "val.txt".

// val.txt -	Notepad				_		×
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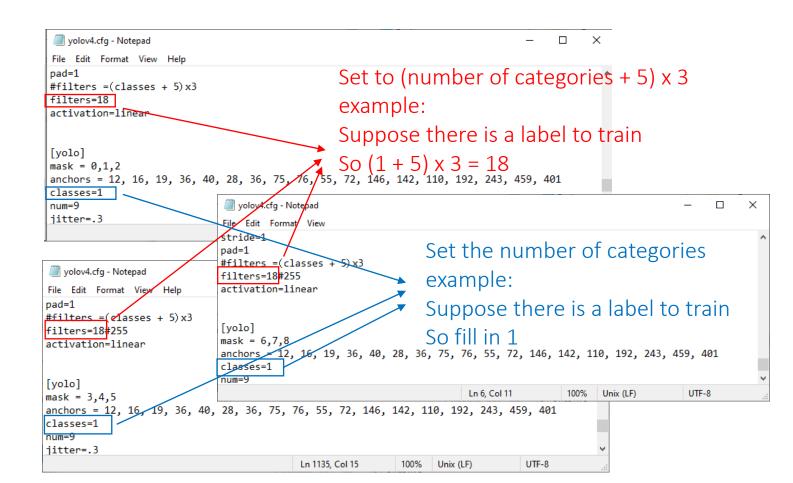
6_prepare_config_file.ipynb

Set ipynb parameters, such as dataset name, number of categories.



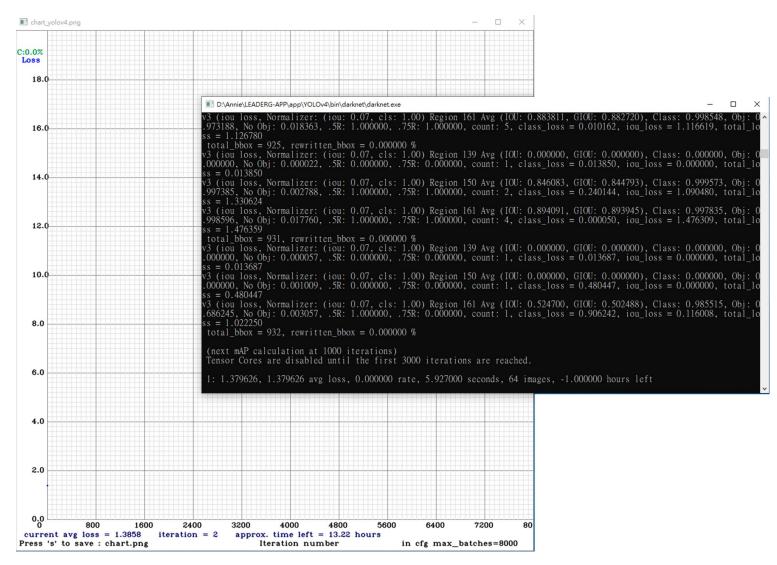
Set training yolov4.cfg related parameters

Set the parameters related to the number of file types in the yolov4.cfg file in the dataset.



7_train_GPU.ipynb

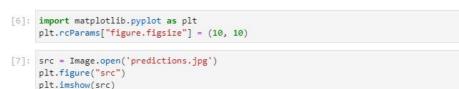
Start training. The training is divided into two types: GPU and CPU. You can choose GPU or CPU mode according to your own hardware.



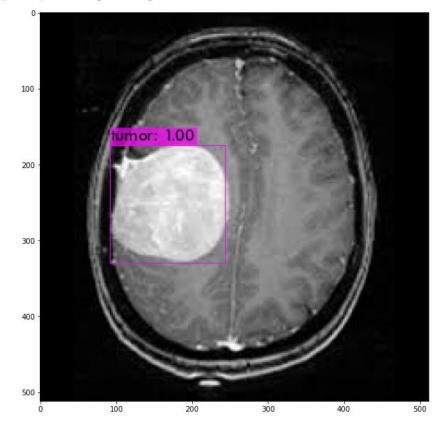
8_inference_GPU.ipynb

Infer a single image.

You can choose GPU or CPU mode according to your own hardware.



[7]: <matplotlib.image.AxesImage at 0x28bbfa33bb0>

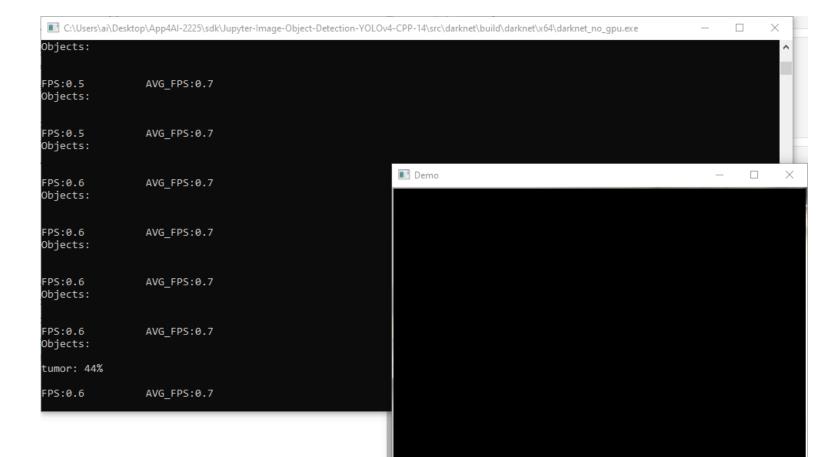


9_inference_webcam_GPU.ipynb

Infer the image of the webcam.

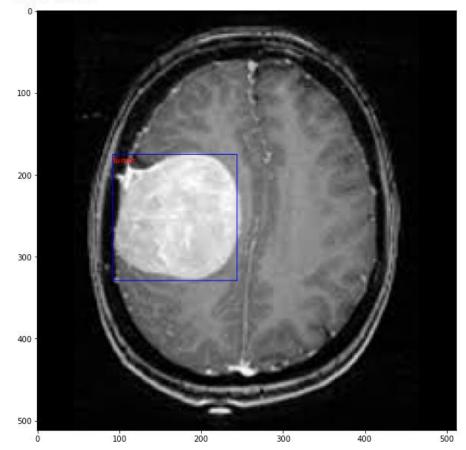
You can choose GPU or CPU mode according to your own hardware.

Press the "Esc" key on the display to turn the webcam off.



10_inference_folder_1_GPU.ipynb

Infer all images in the folder. You can choose GPU or CPU mode according to your own hardware. tumor-1.jpg tumor 168 252 92 175 152 154



Underkill Rate: 0.00%, Overkill Rate: 0.00%, Right Rate: 100.00%, Total: 1

11_inference_folder_demo_GPU.ipynb

Continue to infer all images in the folder

You can choose GPU or CPU mode according to your own hardware.

Press the "Esc" key on the display to turn off continuous inference.

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ta/tumor/test/images/tumor-10.jpg sultString = [{"type": "tumor", "score": "0.999947", "x": "285", "y": "209", "width": "108", "height": s = 1	"118"}]	^
:	"101"}]		
ia/Tumor/test/images/tumor-2.jpg sultString = [{"type": "tumor", "score": "0.999688", "x": "149", "v": "113", "width": "68", "height": tumor", "score": "0.999685", "x": "181", "y": "189", "width" s = 1	"92"},	{"type - [*: 1 ×
ca/tumor/test/images/tumor-3.jpg sultString = [{"type": "tumor", "score": "0.999319", "x": " s = 1			
<pre>ta/tumor/test/images/tumor-4.jpg sultString = [{"type": "tumor", "score": "0.999958", "x": " s = 1</pre>	1		
ta/tumor/test/images/tumor-6.jpg sultString = [{"type": "tumor", "score": "0.997691", "x": " s = 1	3	1	
ta/tumor/test/images/tumor-7.jpg sultString = [{"type": "tumor", "score": "0.999919", "x": " s = 1		1	
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ca/tumor/test/images/tumor-9.jpg sultString = [{"type": "tumor", "score": "0.999070", "x": " s = 1			
ta/tumor/test/images/tumor-1.jpg			

12_YOLOv4_auto_labeling_GPU.ipynb

According to the trained model, the images in the test folder in the dataset are automatically labeled, and the labeled images and labeled voc files are stored in the auto_labeling folder of the dataset.

Reference

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