

10张图带你认识图像分割的前世今生

汇报人 张 伟(Charmve)

M 运微 CUD 迈微AI研习社

Maiwei Al Lab

TERM, TERM POWER





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AI+农业,下一场"工业革命"即将爆发

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Machine Learning, Computer Vision

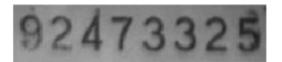


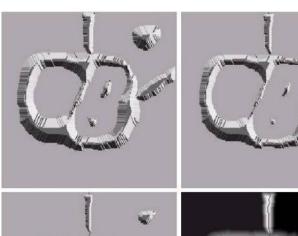
图 4. 原始图像

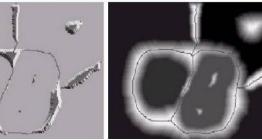
9247332

图 5. 阈值低,对亮区效果好,则暗区差

5247332**5**

图 6. 阈值高,对暗区效果好,则亮区差



















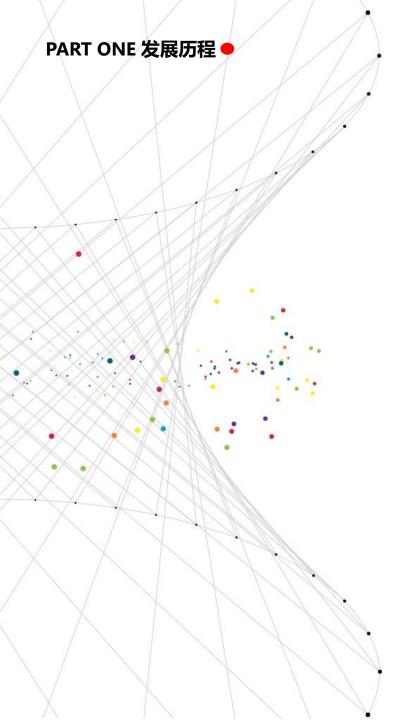


(d) Prewitt 算法

(e) Kirsch 算法

(f) Laplacian 算法

概念区分

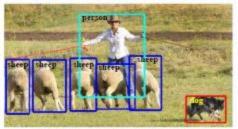




(a) Object Classification



(c) Semantic Segmentation



(b) Generic Object Detection (Bounding Box)

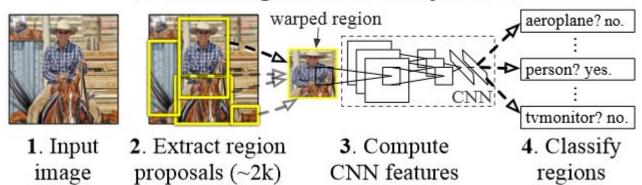


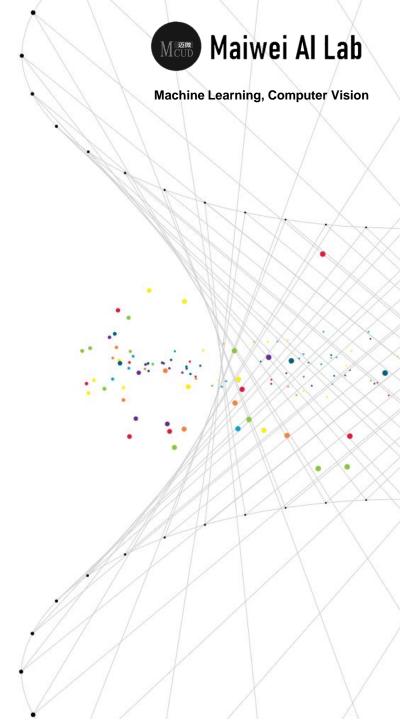
(d) Object Instance Segmetation





R-CNN: Regions with CNN features

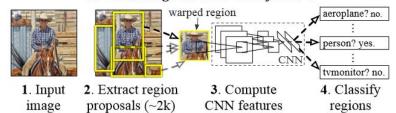




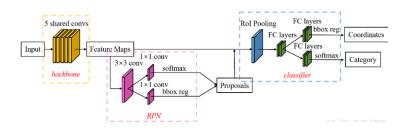
PART TWO 研究现状 **R-CNN** 2013.11 bbox softmax regressor Rol feature feature map vector For each Rol Faster-RCNN NIPS15

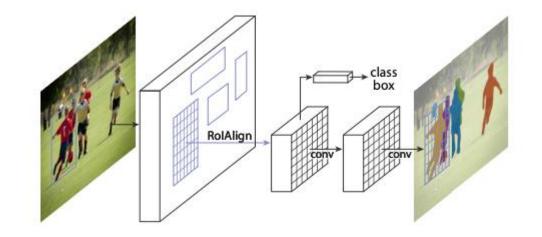
传统分割方法

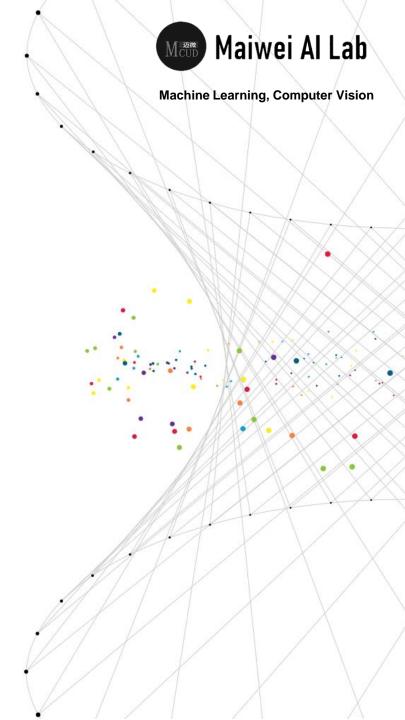
R-CNN: Regions with CNN features



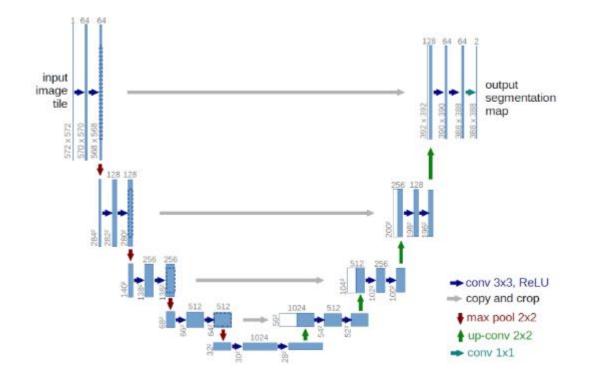
Fast-RCNN ICCV15

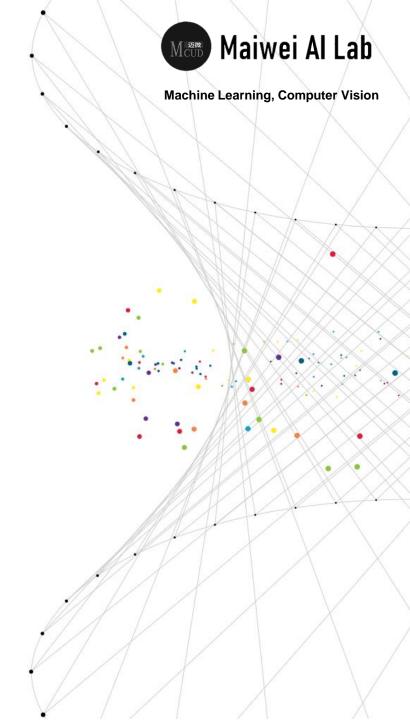






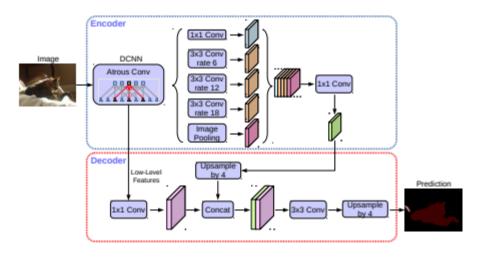
PART TWO 研究现状 传统分割方法 SegNet Convolutional Encoder-Decoder Output Input Pooling Indices **RGB** Image Segmentation Conv + Batch Normalisation + ReLU SegNet: A Deep Convolutional Encoder-Decoder Architecture for Image Segmentation

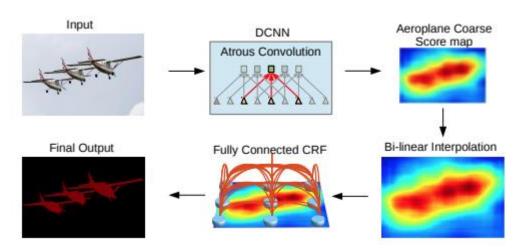




PART TWO 研究现状

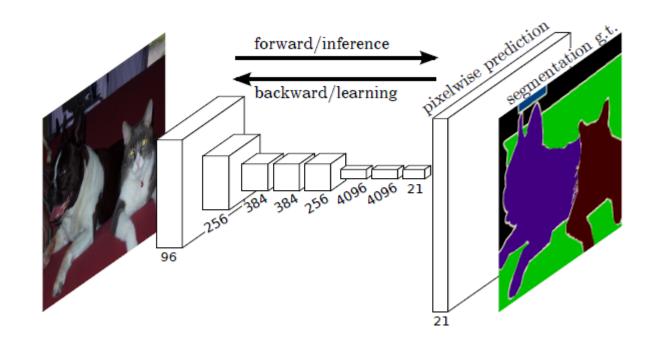
传统分割方法







Machine Learning, Computer Vision



* J. Long, E. Shelhamer, and T. Darrell, "Fully convolutional networks for semantic segmentation," in CVPR, pp. 3431–3440, 2015.

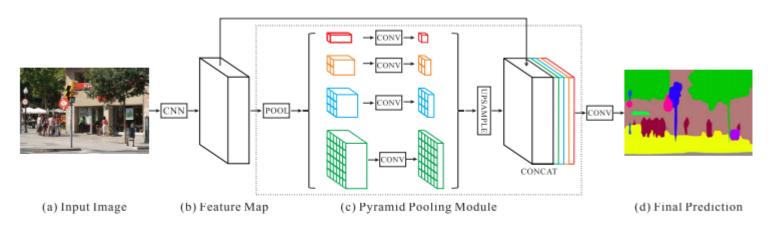
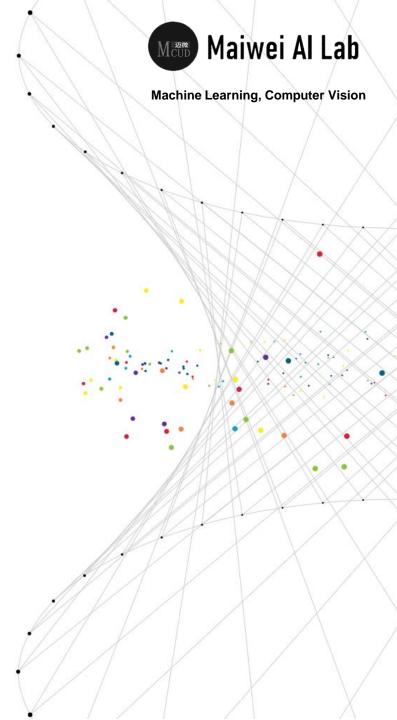


Figure 3. Overview of our proposed PSPNet. Given an input image (a), we first use CNN to get the feature map of the last convolutional layer (b), then a pyramid parsing module is applied to harvest different sub-region representations, followed by upsampling and concatenation layers to form the final feature representation, which carries both local and global context information in (c). Finally, the representation is fed into a convolution layer to get the final per-pixel prediction (d).



PART TWO 研究现状

传统分割方法

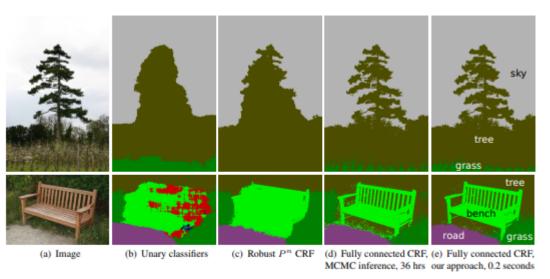


Figure 1: Pixel-level classification with a fully connected CRF. (a) Input image from the MSRC-21 dataset. (b) The response of unary classifiers used by our models. (c) Classification produced by the Robust P^n CRF [9]. (d) Classification produced by MCMC inference [17] in a fully connected pixel-level CRF model; the algorithm was run for 36 hours and only partially converged for the bottom image. (e) Classification produced by our inference algorithm in the fully connected model in 0.2 seconds.



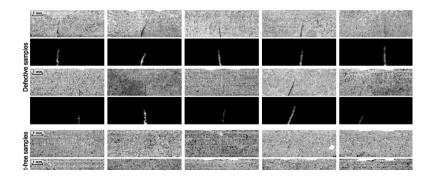


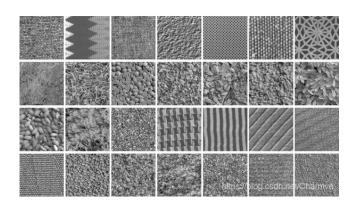
PART THREE 项目推荐 •

Surface Defect Detection: Dataset & Papers

Constantly summarizing open source data sets in the field of surface defect research is very important. Important critical papers from year 2017 have been collected and compiled, which can be viewed in the [Papers] folder.

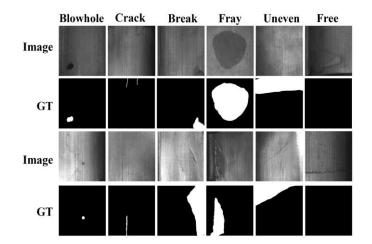


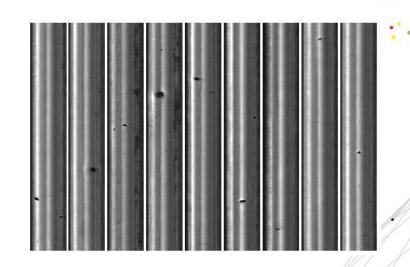












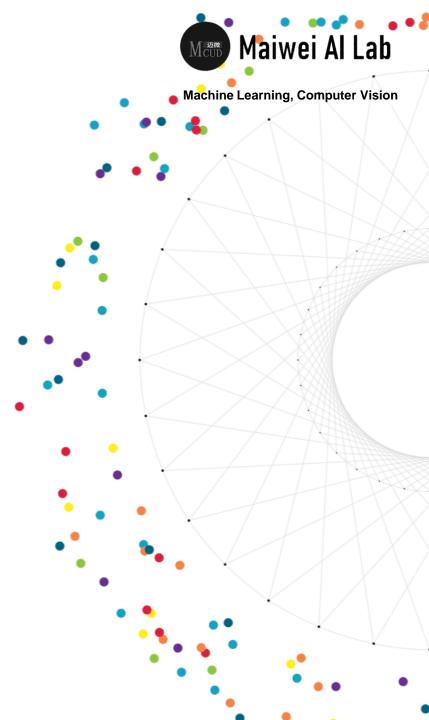
Mirror & Glass Detection in Real-world Scenes

Charmve | English | Chinese



Mirror and Glass Detection/Segmentation





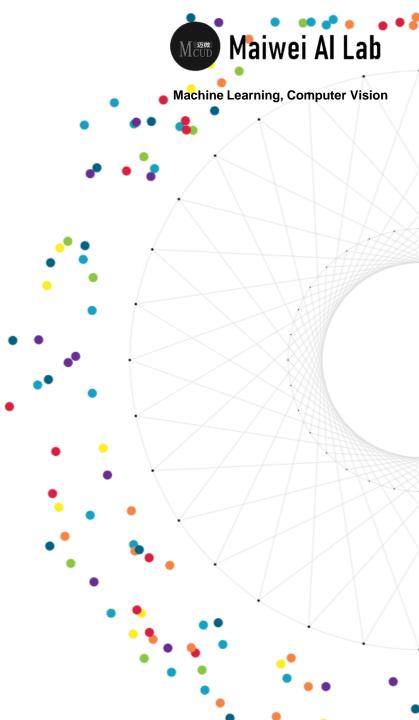
PyTorch for Semantic Segmentation

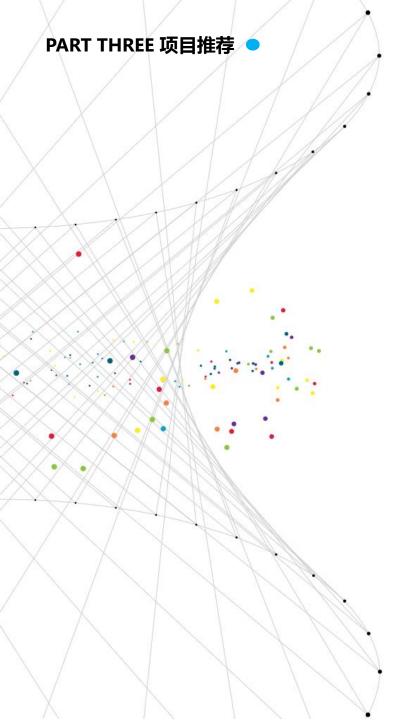
This repository contains some models for semantic segmentation and the pipeline of training and testing models, implemented in PyTorch

Models

- 1. Vanilla FCN: FCN32, FCN16, FCN8, in the versions of VGG, ResNet and DenseNet respectively (Fully convolutional networks for semantic segmentation)
- 2. U-Net (U-net: Convolutional networks for biomedical image segmentation)
- 3. SegNet (Segnet: A deep convolutional encoder-decoder architecture for image segmentation)
- 4. PSPNet (Pyramid scene parsing network)
- 5. GCN (Large Kernel Matters)
- 6. DUC, HDC (understanding convolution for semantic segmentation)
- 7. Mask-RCNN (paper, code from FAIR, code PyTorch)







Scene-Text-Detection

Tracking the latest progress in Scene Text Detection and Recognition: Must-read papers well organized with code and dataset.

Author: Wei ZHANG

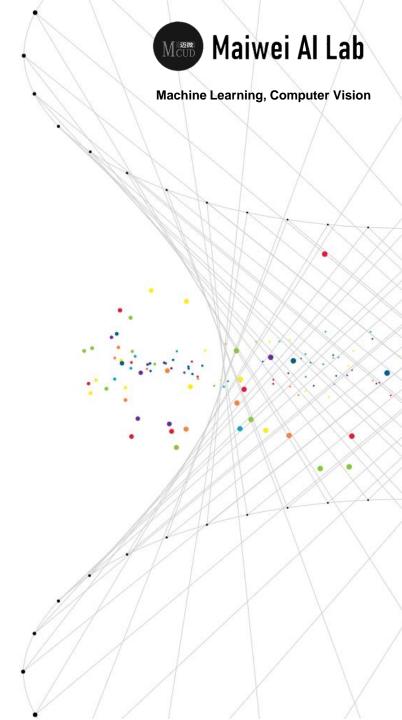
- 1.Datasets
 - o 1.1 Horizontal-Text Datasets
 - 1.2 Arbitrary-Quadrilateral-Text Datasets
 - o 1.3 Irregular-Text Datasets
 - o 1.4 Synthetic Datasets
 - 1.5 Comparison of Datasets
- 2. Survey
- 3. Evaluation
- 4. OCR Service
- 5. References and Code
- GtiHub https://github.com/Charmve/Scene-Text-Detection

PART THREE 项目推荐 •



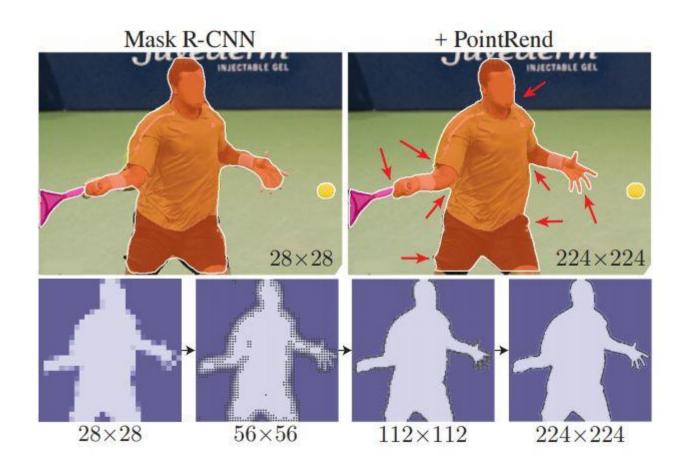


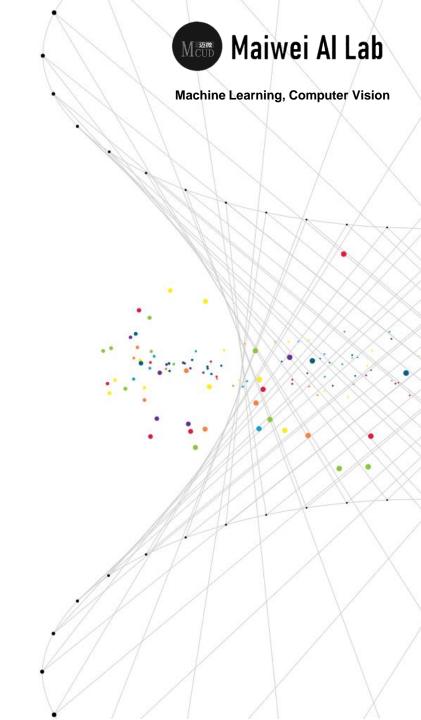






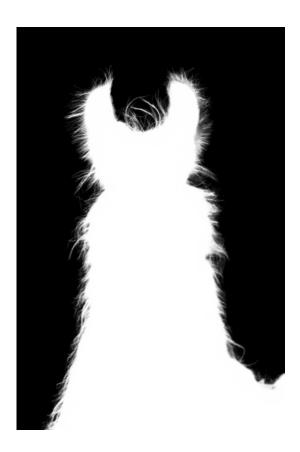






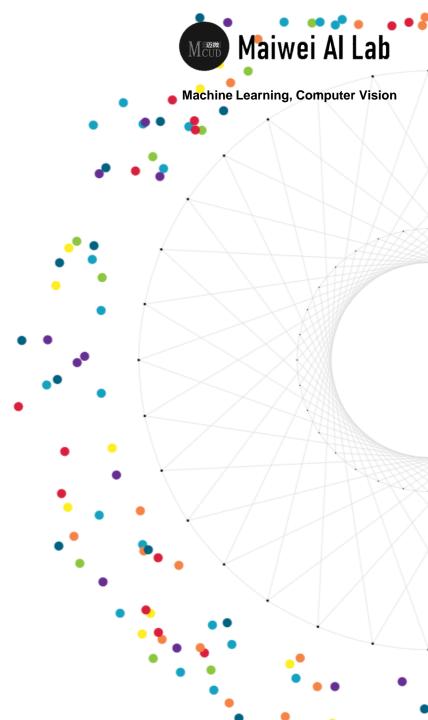
End-to-end Animal Image Matting



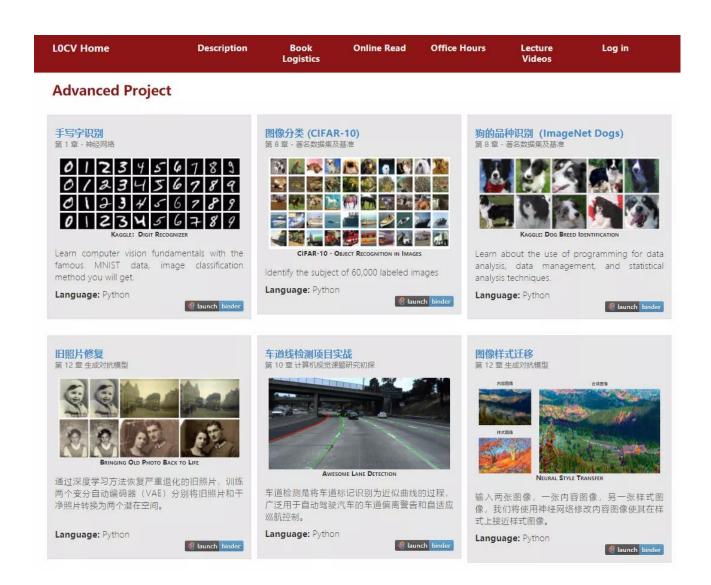


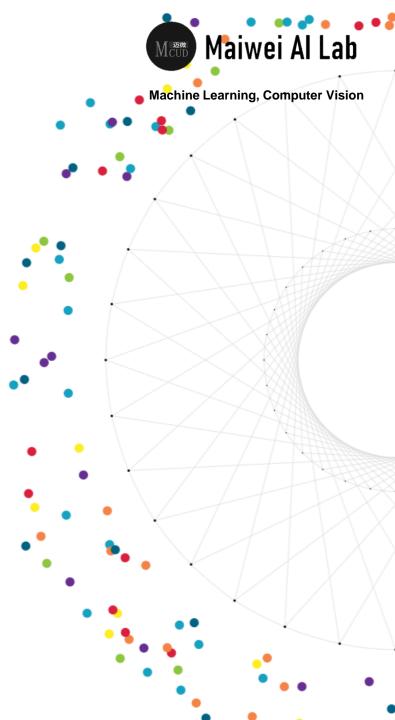
CVPR 2020
[arXiv | Project Page | Video | Code | Related Work]

GtiHub https://github.com/JizhiziLi/animal-matting



PART FOUR 创新应用









PART FIVE 参考文献 ●

- [1] Wu H, Wiesner-Hanks T, Stewart E L, et al. Autonomous End-to-end Detection of Northern Leaf Blight directly from Aerial Imagery[J].
- [2] Doudkin A A, Inyutin A V, Petrovsky A I, et al. Three-level neural network for data clusterization on images of infected crop field[J]. Journal of Research and Applications in Agricultural Engineering, 2007, 52(1): 5-7.
- [3]净浩泽, 图像分割综述. 2019.07.

https://blog.csdn.net/electech6/article/details/95242875 [4]张伟, 江户川柯壮. 从R-CNN到YOLO, 2020 图像目标检测算法综述. 2020.10.

https://blog.csdn.net/Charmve/article/details/109252834







https://github.com/Charmve/



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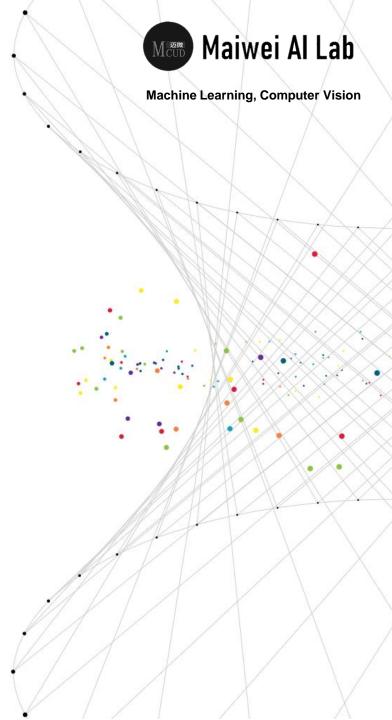
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机器学习+计算机视觉

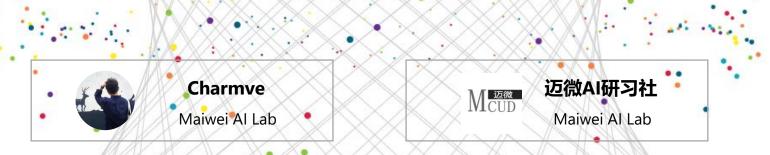


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