#### DeepLab v1

Liang-Chieh Chen, George Papandreou, Iasonas Kokkinos, et al., Semantic image segmentation with deep convolutional nets and fully connected crfs. 2014

Input

DCNN

Aeroplane Coarse Score map

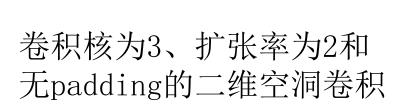
Atrous Convolution

Final Output

Fully Connected CRF

Bi-linear Interpolation

DeepLab v1处理流程



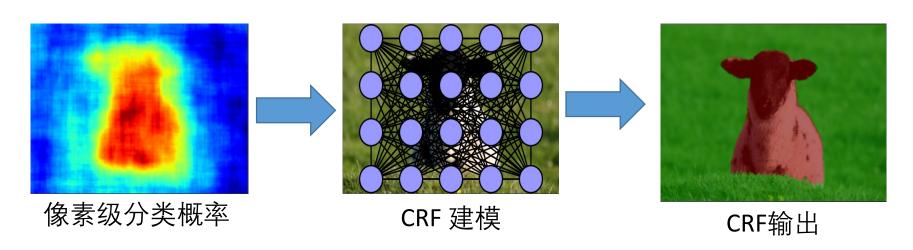
#### DeepLab v1

• DeepLabv1是在VGG16的基础上做了修改:

- ▶VGG16的全连接层转为卷积
- ▶最后的两个池化层去掉了下采样
- ▶后续卷积层的卷积核改为了空洞卷积
- ➤在ImageNet上预训练的VGG16权重上做finetune

## Conditional Random Fields(CRF)

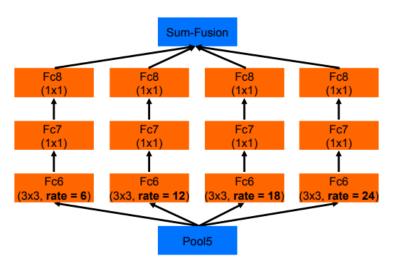
- FCNs, SegNet等基于CNN的分割算法是对每个像素进行相互独立的分类,分割时像素之间关系没有利用。
- CRF是一种概率图模型算法,此算法可以结合先验信息进行结构预测。

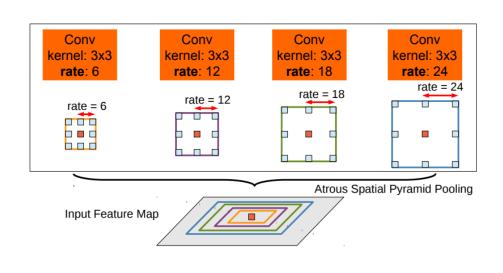


#### DeepLab v2

Liang-Chieh Chen, George Papandreou, et al., *Deeplab: Semantic image segmentation with deep convolutional nets, atrous convolution, and fully connected crfs.* IEEE transactions on PAMI, 2018.

### 多孔空间金字塔池化ASPP:





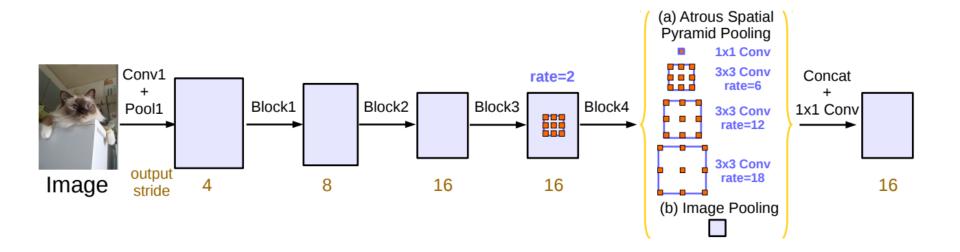
左图说明ASPP使用多个不同倍率的卷积核对池化结果进行卷积, 实现从多尺度捕捉目标和上下文。右图进行一个直观的说明,图 中黄色的点综合不同尺度的区域(不同颜色框)进行识别。

#### DeepLab v2

- ●DeepLab v2在v1的基础上进行改进
  - ▶用多尺度获得更好的分割效果(使用ASPP)
  - ➤ 基础层由VGG16转为ResNet
  - ➤使用不同的学习策略(poly)

#### DeepLab v3

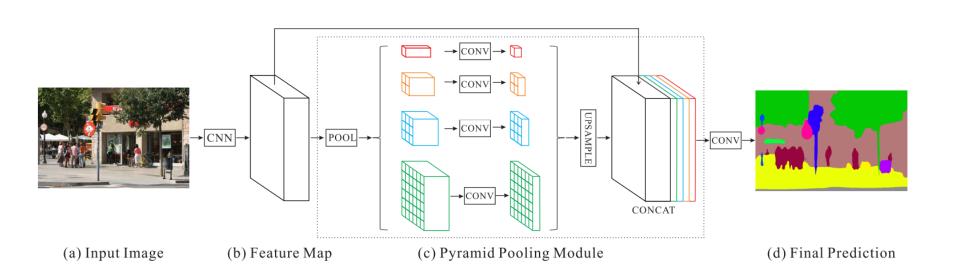
Liang-Chieh Chen, George Papandreou, Florian Schroff, et al., *Rethinking atrous convolution for semantic image segmentation.*, 2017.



相较于v2对ASPP模块进行了改进,多了一个1x1的卷积操作和一个全局平均池化操作

#### **PSPNet**

Zhao H., Shi J., Qi X., et al. *Pyramid scene parsing network*. in IEEE Conf. on CVPR. 2017.



PSPNet网络结构

主要是提出了一种金字塔池化(Global Pyramid Pooling),具有更好的多尺度信息

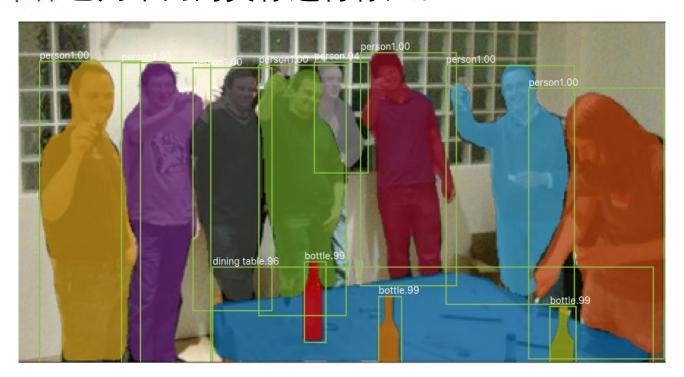
# 主要内容

- 1. 背景介绍
- 2. 传统图像分割
- 3. 基于深度学习的图像分割
  - 3.1 语义分割
  - 3.2 实例分割
- 4. 数据集和评价指标

## • 3.2 实例分割

#### 实例分割的定义:

图像的实例分割就是在<mark>语义分割</mark>的基础上,对同种物体的<mark>不同</mark> 实例个体也用不同的类标进行标注。



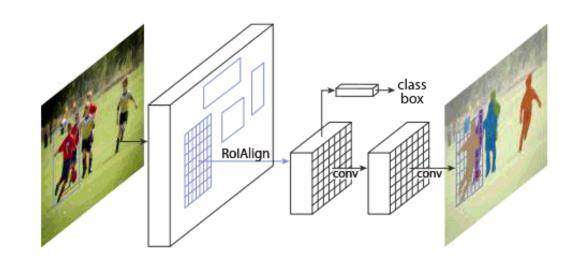
图中不仅将person和bottle进行分类,而且对同一类别(person或bottle)中的不同个体进行了区分,分别用不同颜色表示。

## • 3.1 实例分割

#### **Mask R-CNN**

He K., Gkioxari G., Dollar P., et al. *Mask R-CNN*. in ICCV, 2017.

Facebook AI Research (FAIR)



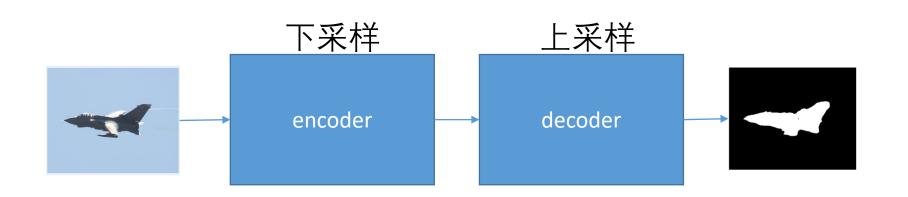
Mask R-CNN框架

使用<mark>多任务</mark>学习方法,结合了目标<mark>检测</mark>和语义<mark>分割</mark>任务,采 用多分支网络框架

## • 3. 基于深度学习的图像分割

图像分割网络总结

可以看出,这些网络结构都是非常相似的,都是基于encoderdecoder结构的,只不过说法不同。encoder一般是由卷积层、池化层构成的一个下采样过程,decoder则是一个上采样过程。这些网络的最大区别也就是上采样方法的不同。



# 主要内容

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## • 4. 数据集和评价指标

### 数据集

数据集中需要包括的信息:

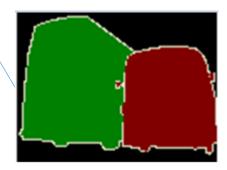
原图



=

语义分割标签

实例分割标签



常见数据集:

**Visual Object Classes Challenge 2012 (VOC2012)** 





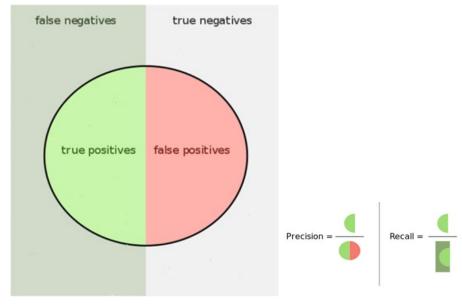


## • 4. 数据集和评价指标

### 评价指标

如何判断图像分割预测结果的好坏,有以下几个常用的指标:

- 1.AP (Average Precision)
- 2.IU/IoU (Intersection over Union)



圆圈里表示预测的前景; 圆圈外表示预测的背景; 矩形左表示真实的前景; 矩形右表示真实的背景;

在图像分割中,AP则是先对一张图片中的某一类物体的预测结果计算precision,然后对所有图片求平均。对于多类别的分割任务而言,一般采用mean AP,即将AP对所有类别求平均。

## • 4. 数据集和评价指标

评价指标

如何判断图像分割预测结果的好坏,有以下几个常用的指标:

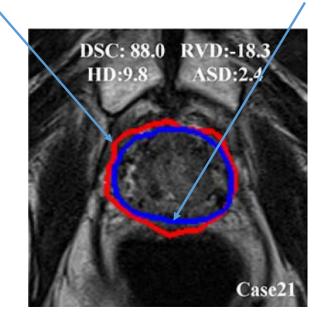
- 1.AP (Average Precision)
- 2. IU/IoU (Intersection over Union)

红色表示预测结果 蓝色色表示真实结果

#### IU/IoU:

对于两个区域R和R', 其overlap 计算如下:

$$O(R,R') = \frac{|R \cap R'|}{|R \cup R'|}$$



二者覆盖的越好,表面结果越精确

			aero plane	bicycle	bird	boat	bottle	bus	car	cat	chair	cow	dining table	dog	horse	motor bike	person	potted plant	sheep	sofa	train	tv mon
		<b>V</b>	$\nabla$	$\triangleright$	$\nabla$	abla	abla	$\nabla$	$\nabla$	abla	abla	abla	abla	abla	abla	abla	$\triangleright$	$\nabla$	abla	abla	abla	$\triangle$
$\triangleright$	DeepLabv3+_JFT [?]	89.0	97.5	77.9	96.2	80.4	90.8	98.3	95.5	97.6	58.8	96.1	79.2	95.0	97.3	94.1	93.8	78.5	95.5	74.4	93.8	
$\triangleright$	DeepLabv3+_AASPP [?]	88.5	97.4	80.3	97.1	80.1	89.3	97.4	94.1	96.9	61.9	95.1	77.2	94.2	97.5	94.4	93.0	72.4	93.8	72.6	93.3	
$\triangleright$	SRC-B-MachineLearningLab [?]	88.5	97.2	78.6	97.1	80.6	89.7	97.4	93.7	96.7	59.1	95.4	81.1	93.2	97.5	94.2	92.9	73.5	93.3	74.2	91.0	
$\triangleright$	MSCI [?]	88.0	96.8	76.8	97.0	80.6	89.3	97.4	93.8	97.1	56.7	94.3	78.3	93.5	97.1	94.0	92.8	72.3	92.6	73.6	90.8	1
$\triangleright$	ExFuse [?]	87.9	96.8	80.3	97.0	82.5	87.8	96.3	92.6	96.4	53.3	94.3	78.4	94.1	94.9	91.6	92.3	81.7	94.8	70.3	90.1	
$\triangleright$	DeepLabv3+ [?]	87.8	97.0	77.1	97.1	79.3	89.3	97.4	93.2	96.6	56.9	95.0	79.2	93.1	97.0	94.0	92.8	71.3	92.9	72.4	91.0	
$\triangleright$	DeepLabv3-JFT [?]	86.9	96.9	73.2	95.5	78.4	86.5	96.8	90.3	97.1	51.4	95.0	73.4	94.0	96.8	94.0	92.3	81.5	95.4	67.2	90.8	
$\triangleright$	DIS [?]	86.8	94.0	73.3	93.5	79.1	84.8	95.4	89.5	93.4	53.6	94.8	79.0	93.6	95.2	91.5	89.6	78.1	93.0	79.4	94.3	
•	** Gluon DeepLabV3 152 ** [?]	86.7	96.5	74.3	96.1	80.2	85.2	97.0	93.8	96.4	49.7	93.6	77.6	95.1	95.3	93.9	89.6	75.8	94.4	70.8	89.7	
$\triangleright$	CASIA_IVA_SDN [?]	86.6	96.9	78.6	96.0	79.6	84.1	97.1	91.9	96.6	48.5	94.3	78.9	93.6	95.5	92.1	91.1	75.0	93.8	64.8	89.0	
$\triangleright$	IDW-CNN [?]	86.3	94.8	67.3	93.4	74.8	84.6	95.3	89.6	93.6	54.1	94.9	79.0	93.3	95.5	91.7	89.2	77.5	93.7	79.2	94.0	
$\triangleright$	DFN [?]	86.2	96.4	78.6	95.5	79.1	86.4	97.1	91.4	95.0	47.7	92.9	77.2	91.0	96.7	92.2	91.7	76.5	93.1	64.4	88.3	
$\triangleright$	** GluonCV DeepLabV3 ** [?]	86.2	96.3	69.7	93.5	76.2	86.5	96.5	92.2	95.8	47.8	95.0	81.6	93.0	96.0	91.2	90.7	77.1	94.7	68.9	89.3	
$\triangleright$	EncNet [?]	85.9	95.3	76.9	94.2	80.2	85.3	96.5	90.8	96.3	47.9	93.9	80.0	92.4	96.6	90.5	91.5	70.9	93.6	66.5	87.7	
$\triangleright$	HPN [?]	85.8	94.1	67.0	95.2	81.9	88.3	95.5	90.4	95.9	40.0	92.7	82.5	91.7	95.3	92.6	91.6	73.6	94.1	69.4	91.1	
$\triangleright$	DeepLabv3 [?]	85.7	96.4	76.6	92.7	77.8	87.6	96.7	90.2	95.4	47.5	93.4	76.3	91.4	97.2	91.0	92.1	71.3	90.9	68.9	90.8	
$\triangleright$	** Auto-DeepLab-L ** [?]	85.6	96.5	77.3	94.8	74.1	84.0	97.1	88.7	94.5	53.5	91.6	79.2	88.4	94.2	90.2	91.2	75.1	90.1	70.7	89.1	
$\triangleright$	** DP-CAN_decoder ** [?]	85.5	95.9	77.8	91.6	75.0	81.7	96.6	92.4	97.1	42.7	93.5	74.1	93.9	95.0	91.4	91.2	78.1	94.6	66.5	89.8	
$\triangleright$	PSPNet [?]	85.4	95.8	72.7	95.0	78.9	84.4	94.7	92.0	95.7	43.1	91.0	80.3	91.3	96.3	92.3	90.1	71.5	94.4	66.9	88.8	
$\triangleright$	** GluonCV PSP ** [?]	85.1	95.7	70.9	92.8	75.6	85.0	96.5	91.7	95.0	41.8	92.3	78.8	90.4	95.6	93.4	90.6	76.1	93.5	66.7	89.5	
$\triangleright$	** ResNet-38_COCO ** [?]	84.9	96.2	75.2	95.4	74.4	81.7	93.7	89.9	92.5	48.2	92.0	79.9	90.1	95.5	91.8	91.2	73.0	90.5	65.4	88.7	
$\triangleright$	** DP-CAN ** [?]	84.6	96.5	77.7	87.6	73.9	79.9	96.8	92.9	95.7	40.8	92.9	74.0	91.7	95.0	92.5	89.7	77.2	94.6	64.6	90.2	
$\triangleright$	** resnet 101 + fast laddernet ** [?]	84.2	95.4	73.9	94.9	75.7	83.2	96.3	91.2	93.9	35.3	90.0	79.4	90.2	94.2	92.8	90.1	73.2	92.3	64.5	88.0	

http://host.robots.ox.ac.uk:8080/leaderboard/displaylb.php?cls=mean&challengeid=1 1&compid=6&submid=19760

## 参考资料

- 1.反卷积: https://blog.csdn.net/itleaks/article/details/80336825
- 2.空洞卷积: http://www.dataguru.cn/article-11860-1.html
- 3.深度学习语义分割发展史: https://www.sohu.com/a/155907339\_610300
- 4.图像分割综述: https://blog.csdn.net/Julialove102123/article/details/80493066
- 5.分水岭分割算法: https://www.cnblogs.com/mikewolf2002/p/3304118.html
- 6.实例分割图: https://blog.csdn.net/lanyuxuan100/article/details/70800246
- 7.CS231n课程视频: http://cs231n.stanford.edu/