



## 基于MXNET的图像检测 开发案例



- 个人简介
- 业务背景
- 业务流程
- 开发案例

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- 人工智能算法,计算机视觉算法
- 主要业务:各类发票和证件的检测、识别以及信息提取
- 主要开发语言: python
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- 存量文档电子化
  - 图书,发票等
- 快速理赔
  - 各类证件的在线识别
- 智能录入
  - 房产证、身份证等
- 文字翻译
  - 手机拍照图片文字的检测识别





- 通用模型
  - 文字方向检测
  - 文字位置检测
  - 通用文字识别
  - 关键信息提取
- 专用模型
  - 特定票据检测
  - 特定信息位置检测
  - 特定信息文字识别



## 业务流程:通用模型







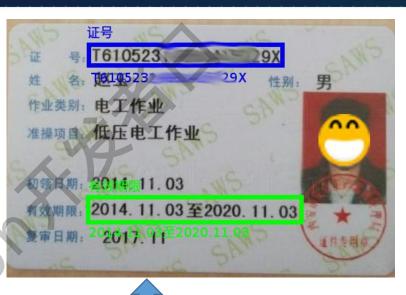
原始图片

目标检测





裁切校正



特定字段检测识别



## 开发案例



• 选取案例:特定种类的票据位置检测

• 原因:目标简单

• 选取框架: MXNET中的Gluon

• Gluon接口简单易上手

• 与计算机视觉配套的gluonCV库包含常用的检测网络

• 文档详细,方便对照相似案例实现定制化目标



- 相关参考
- <a href="https://mxnet.incubator.apache.org/">https://mxnet.incubator.apache.org/</a>
- <a href="https://gluon-cv.mxnet.io/#">https://gluon-cv.mxnet.io/#</a>





■ Secure | https://gluon-cv.mxnet.io/index.html

Model Zoo

Tutorials

Image Classification

Object Detection

- 01. Predict with pre-trained SSD models
- 02. Predict with pre-trained Faster RCNN models
- 03. Predict with pre-trained YOLO models
- 04. Train SSD on Pascal VOC dataset
- 05. Deep dive into SSD training: 3 tips to boost performance
- 06. Train Faster-RCNN endto-end on PASCAL VOC
- 07. Train YOLOv3 on PASCAL VOC

#### GluonCV: a Deep Learning Toolkit for Computer Vision

#### Supported Applications

Application

Image Classification:

recognize an object in an image.



Illustration





50+ models, including ResNet, MobileNet,

DenseNet, VGG, ...

Available Models

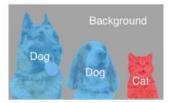
Object Detection

detect multiple objects with their bounding boxes in an image.

Faster RCNN, SSD, Yolo-v3

Semantic Segmentation:

associate each pixel of an image with a categorical label.



FCN, PSP, DeepLab v3

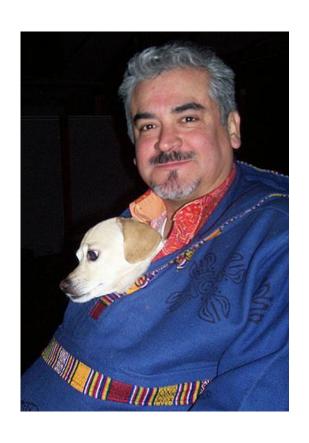
Instance Segmentation:

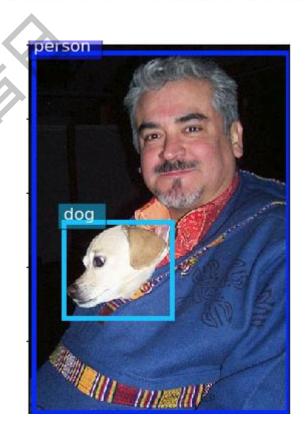
associate each pixel of an image with an instance label.



Mask RCNN







原图像

标注片段

画图



```
from gluoncv.data import VOCDetection
train dataset = VOCDetection(splits=[(2007, 'trainval'), (2012)
val dataset = VOCDetection(splits=[(2007, 'test')])
class VOCDetection(VisionDataset):
    def __init (self, ...):
    def len (self):
        return len(self. items)
    def getitem (self, idx):
        img id = self. items[idx]
        img_path = self._image_path.format(*img id)
        label = self. label cache[idx] if self. label cache else self. load label(idx)
        img = mx.image.imread(img path, 1)
        if self. transform is not None:
            return self. transform(img, label)
        return img, label
```



```
from gluoncv.data.transforms import presets
from gluoncv.data.batchify import Tuple
from mxnet.gluon.data import DataLoader
train transform = presets.ssd.SSDDefaultTrainTransform(width, height)
train transform = presets.ssd.SSDDefaultTrainTransform(width, height, anchors)
batchify fn = Tuple(Stack(), Stack(), Stack())
train loader = DataLoader(
    train_dataset.transform(train transform)
   batch size,
    shuffle=True,
    batchify fn=batchify fn,
    last batch='rollover',
    num workers=num workers)
```

```
from gluoncv import model_zoo
net = model_zoo.get_model('ssd_300_vgg16_atrous_voc', pretrained_base=False)
```



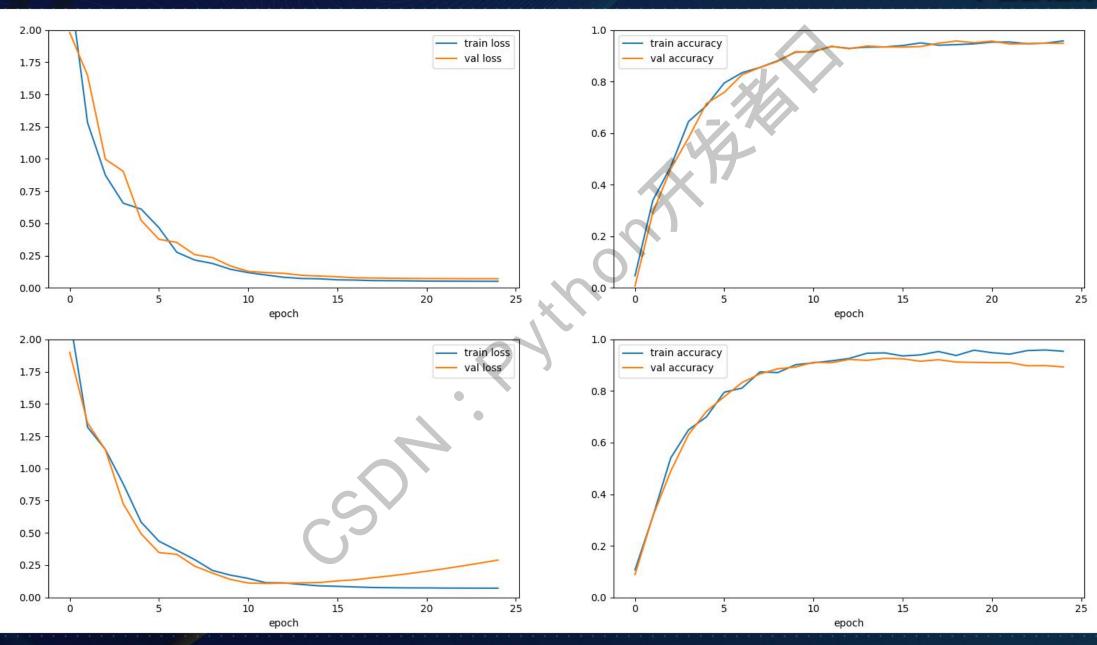
```
from gluoncv.loss import SSDMultiBoxLoss
mbox loss = SSDMultiBoxLoss()
trainer = gluon.Trainer(
   net.collect params(), 'sgd',
    {'learning rate': 0.001, 'wd': 0.0005, 'moment
for ib, batch in enumerate(train loader):
    if ib > 0:
       break
   print('data:', batch[0].shape)
   print('class targets:', batch[1].shape)
   print('box targets:', batch[2].shape)
   with autograd.record():
        cls_pred, box pred, anchors = net(batch[0])
        sum loss, cls loss, box loss = mbox loss(
            cls pred, box pred, batch[1], batch[2])
        # some standard gluon training steps:
        # autograd.backward(sum loss)
        # trainer.step(1)
```



- 损失函数的权重
- 学习率策略
- 多卡训练
- 状态监控
- 模型保存
- 终止条件
- 数据清洗



```
07:39:06 - Epoch 20 train, Batches 001815/001845: Total Loss: 0.596, Cls Loss: 0.326, Box Loss: 0.108, time: 30.43s, epoch left:
0.05h
07:39:18 - Epoch 20 train, Batches 001820/001845: Total Loss: 0.517, Cls Loss: 0.332, Box Loss: 0.074, time: 11.76s, epoch left:
0.02h
07:39:25 - Epoch 20 train, Batches 001825/001845: Total Loss: 0.512, Cls Loss: 0.312, Box Loss: 0.080, time: 7.19s, epoch left:
0.01h
07:39:34 - Epoch 20 train, Batches 001830/001845: Total Loss: 0.583, Cls Loss: 0.388, Box Loss: 0.078, time: 8.84s, epoch left:
0.01h
07:39:41 - Update[35037]: Change learning rate to 1.88677e-04
07:39:45 - Epoch 20 train, Batches 001835/001845: Total Loss: 0.529, Cls Loss: 0.324, Box Loss: 0.082, time: 10.69s, epoch left:
0.01h
07:40:10 - Epoch 20 train, Batches 001840/001845: Total Loss: 0.579, Cls Loss: 0.328, Box Loss: 0.100, time: 25.50s, epoch left:
200091/03-07 07:40:16,629 - INFO - Epoch 20, model saved ./models/ssd 512-20.params
2019-03-07 07:41:27,491 - INFO - epoch 20 validation result:
2019-03-07 07:41:27,492 - INFO - ID Kuang 1 0.9926505575019027
2019-03-07 07:41:27,492 - INFO - ID ZhaoPian 0.9918573615973989
2019-03-07 07:41:27,492 - INFO - ID Kuang 2 0.983387269507376
2019-03-07 07:41:27,493 - INFO - X Kuang 1 0.992835376812076
2019-03-07 07:41:27,493 - INFO - X YinZhang 0.9857278694205598
2019-03-07 07:41:27,493 - INFO - X Kuang 2 0.9918454945722626
2019-03-07 07:41:27,494 - INFO - X Kuang 3 0.9725970915795881
2019-03-07 07:41:27,494 - INFO - QiTa C 0.8413855138689824
2019-03-07 07:41:27,494 - INFO - MAP 0.9690358168575183
```



## 其它简单的例子



- 文字方向检测
- 分类模型
- Mobile net







# Python开发首日 让开发首紧跟 技术法統治

谢谢