Lecture 10:

Convolutional Neural Networks (CNN) – Part II



Markus Hohle
University California, Berkeley

Machine Learning Algorithms
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Berkeley Machine Learning Algorithms:





Outline

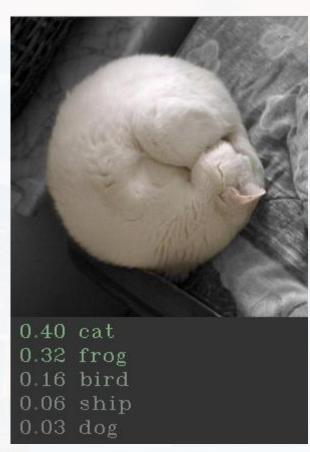
- Labeling Tools

- Calling a Pretrained CNN



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<u>"labelme"</u>

Description

Labelme is a graphical image annotation tool inspired by http://labelme.csail.mit.edu. It is written in Python and uses Qt for its graphical interface.



VOC dataset example of instance segmentation.

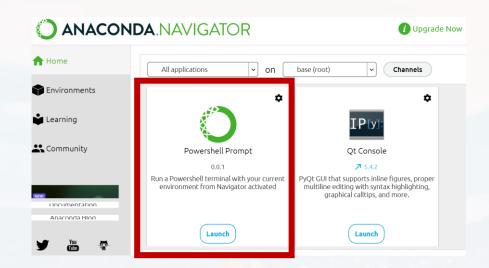


Other examples (semantic segmentation, bbox detection, and classification).



Various primitives (polygon, rectangle, circle, line, and point).



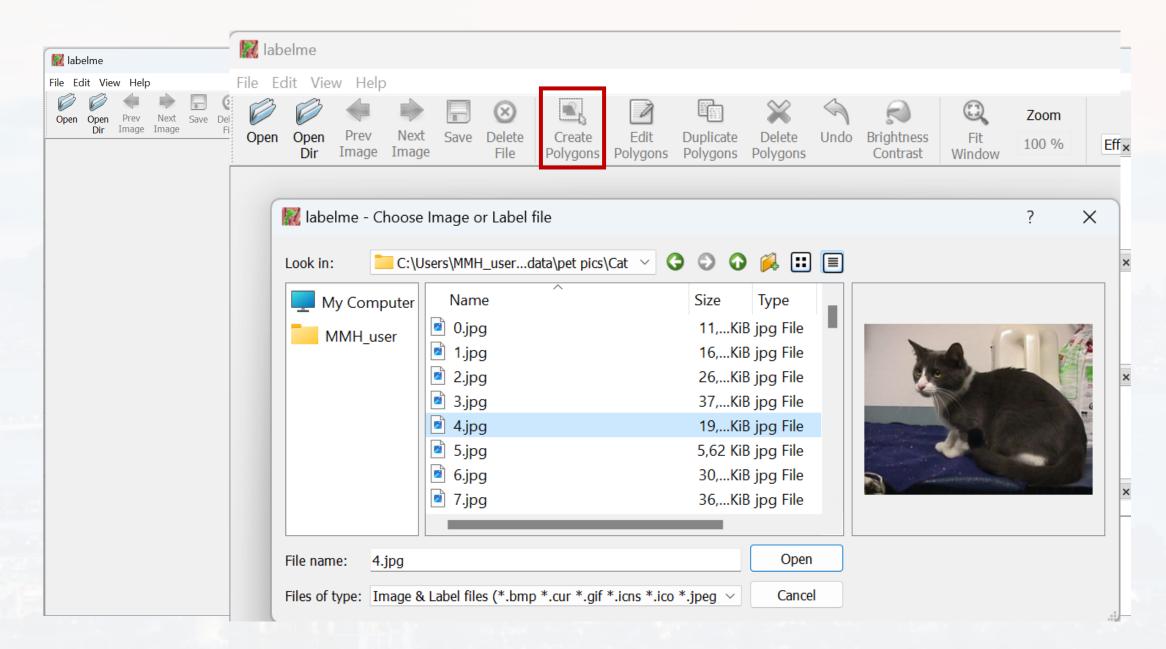


conda install labelme

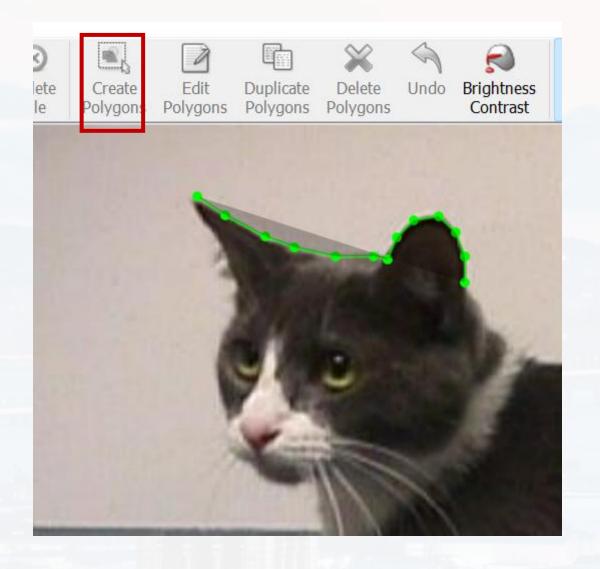
(base) PS C:\Users\MMH_user> conda activate labelme

(labelme) PS C:\Users\MMH_user> labelme

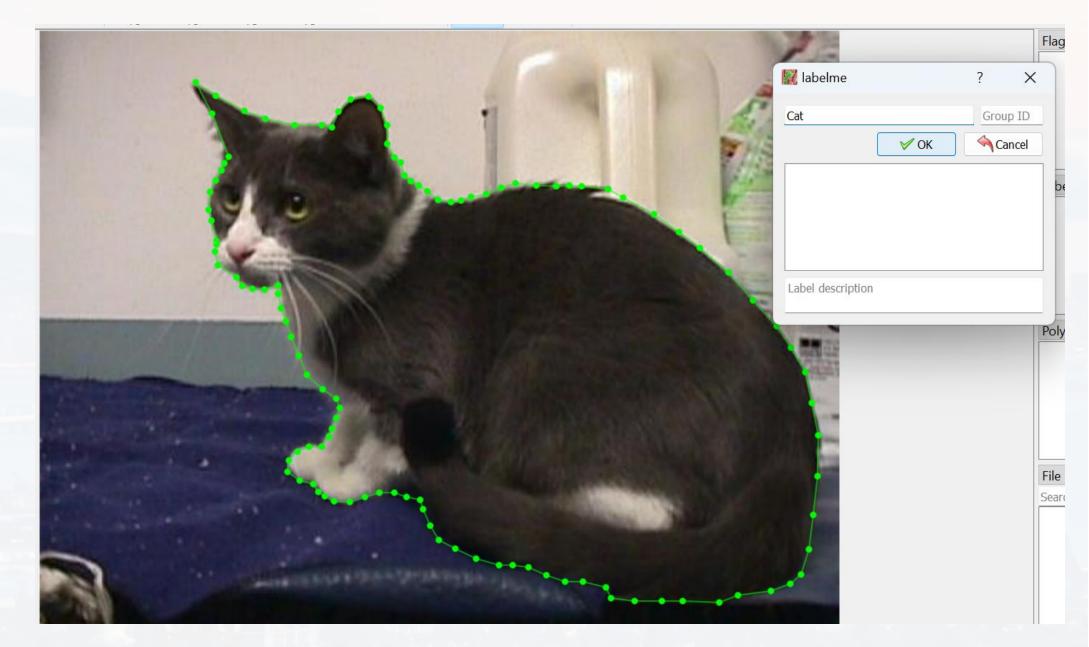














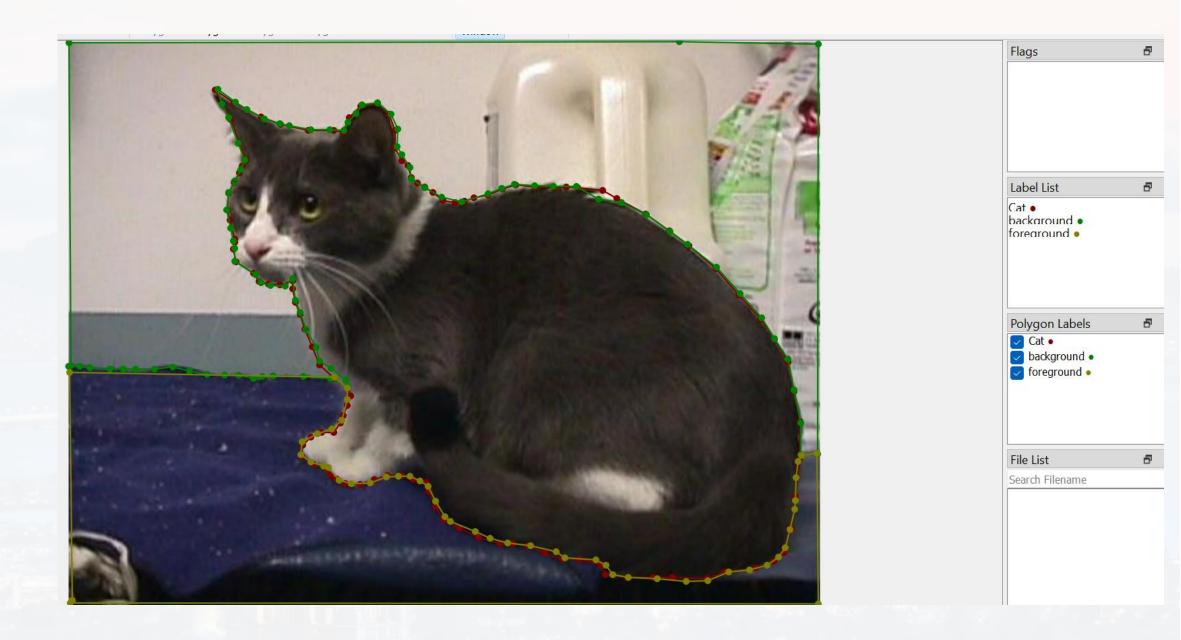




image is saved as .json

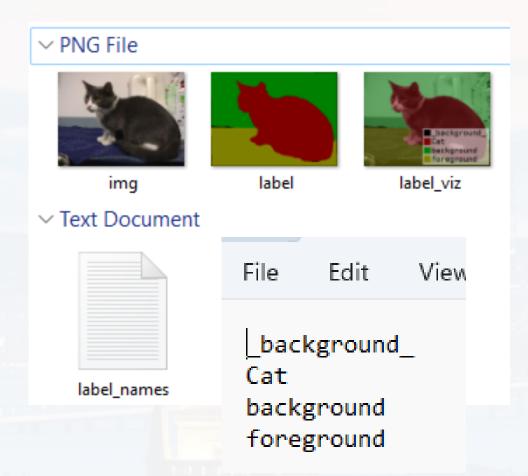


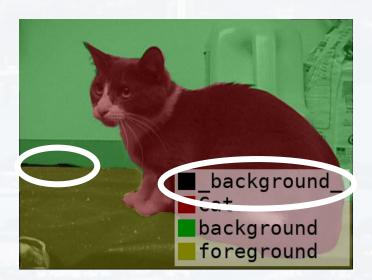
run within the labelme prompt:

labelme_export_json .\Cat4label.json -o Cat4label_json

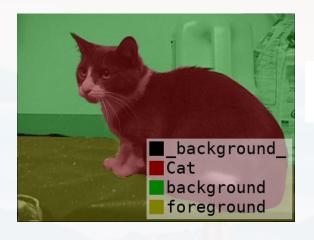




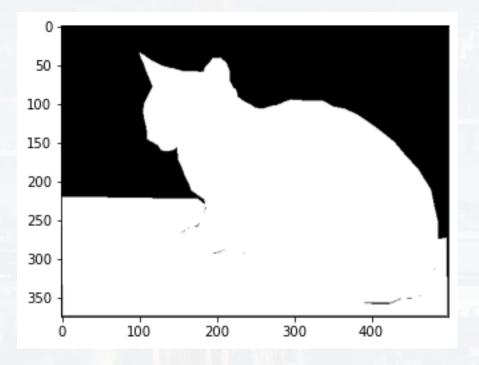


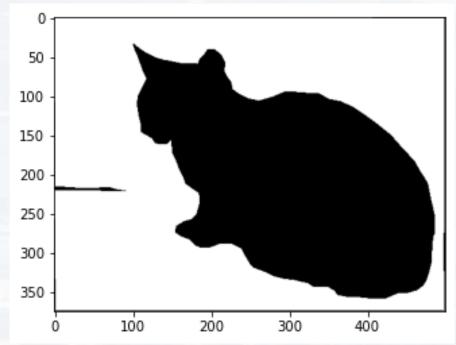




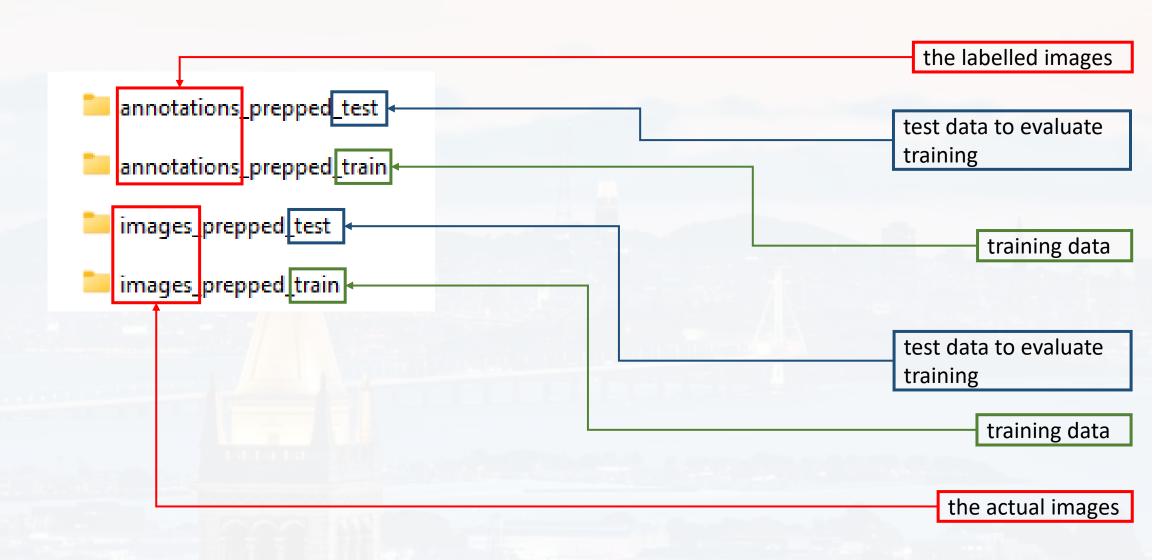


I = plt.imread('Cat/Cat4label_json/label.png')
I.shape
(375, 500, 4)







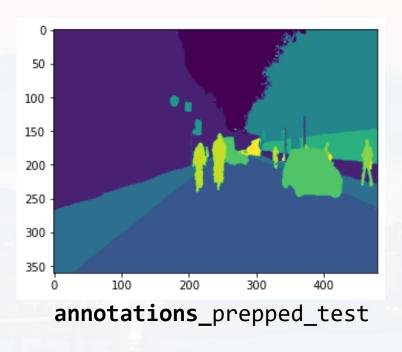




nice dataset



images_prepped_test

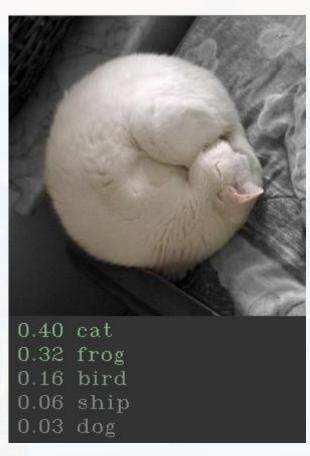


I = plt.imread('segmentation/pics/annotations prepped test/0016E5_07959.png')

I.shape (360, 480)

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demonstrating realistic segmentation with reasonable results will take a few hours

→ check out my code on GitHub

do augment

epochs = 5)

gen_use_multiprocessing = True,

auto resume checkpoint = True,

Segmentation

weights

Keras provides an

augmentation routine

```
from keras segmentation.models.unet import *
                                                                                  Type
                                                                                                               Names
                                                                                        'vgg16' 'vgg19'
                                                                                VGG
                                                                                ResNet
                                                                                        'resnet18' 'resnet34' 'resnet50' 'resnet101' 'resnet152
                                                                                SE-ResNet
                                                                                         'seresnet18' 'seresnet34' 'seresnet50' 'seresnet101' 'seresnet152'
                                                                                ResNeXt
                                                                                        'resnext50' 'resnext101'
                                                                                SE-
                                                                                         'seresnext50' 'seresnext101
                                                                                ResNeXt
                                                                                SENet154
                                                                                        'senet154'
                                                                                DenseNet
                                                                                         'densenet121' 'densenet169' 'densenet201'
                                                                                        'inceptionv3' 'inceptionresnetv2'
                                                                                Inception
                                                                                MobileNet
                                                                                         'mobilenet' 'mobilenetv2'
                                                                                         'efficientnetb0' 'efficientnetb1' 'efficientnetb2' 'efficientnetb3' 'efficientnetb4
                                                                                        'efficientnetb5' efficientnetb6' efficientnetb7'
model = unet(n_classes = n_classes,\
                                           input height = 416, input width = 608)
                                                                                                           calling the specific net-
model.train(
                                                                                                           work
            train images
                                           = my path + r"images prepped train//",
            train_annotations = my_path + r"annotations_prepped_train//",
            checkpoints path = my path + r"checkpoints/
                                                                                                           saves current
```

= True,



```
run:
S = SegmentMyImages()
S.Training()
```

S = SegmentMyImages()

S.Training()

```
Dataset verified!
Epoch 1/5
saved ../data/segmentation
pics/checkpoints//.0
Epoch 2/5
saved ../data/segmentation
pics/checkpoints//.1
Epoch 3/5
saved ../data/segmentation
pics/checkpoints//.2
Epoch 4/5
saved ../data/segmentation
pics/checkpoints//.3
Epoch 5/5
pics/checkpoints//.4
```

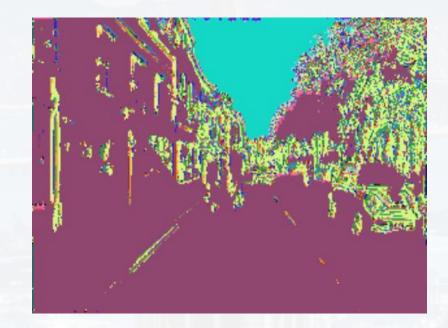
MyModel = S.TrainedModel

MyModel.summary()

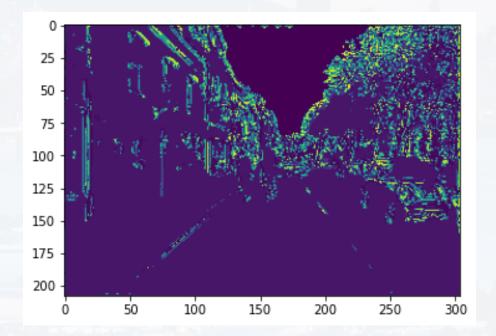
→ returns the structure of the CNN

applying the trained CNN to an image:

out = S.ApplyTrainedNetwork()



plt.imshow(out)



```
recovering model from checkpoints:
MyModel = S.TrainedModel
out
         = S.ApplyTrainedNetwork()
                                                applying the trained CNN to an image:
S.RecoverFromCheckpoint()
                                                                            untrained
                                                                           model (just
                                                                            CNN itself)
·#loading·untrained·CNN
 model = self.model
                                                       transfer the saved weights to untrained
 if not image_name:
                                                       network → now it starts from latest
     image_name = '0016E5_07965.png'
                                                       training state
#calling input from checkpoints
 latest = tf.train.<u>latest c</u>heckpoint(self.checkpoint_path)
 model.load_weights(latest)
```



visualizing weights:

→ see model.layers

nice example





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Thank you very much for your attention!



