

Instance Segmentation on the Wood Debris Problem

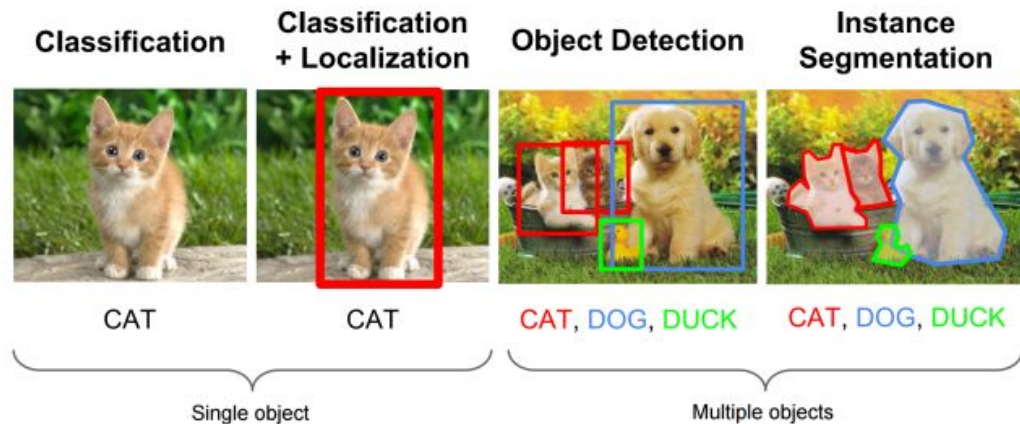
A study focused on the application of instance segmentation for the wood debris problem, as proposed by Dr. Nikolai Strigul.

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EN-250 Fall 2019
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A brief overview of Instance Segmentation...



What is Instance Segmentation?



Classification: is there a cat in this picture?

Localization: roughly, where is the cat? (Bounding box)

Object detection: classification + localization with multiple classes

What's the shape of the objects? Where exactly are they? **Instance Segmentation** answers these questions.

How does Instance Segmentation work?

**Object
Detection**



**Instance
Segmentation**



Instance segmentation creates a pixel-wise mask for each object in the image.

This technique gives us a far more granular understanding of the object(s) in the image.

History

Pre-deepLearning era

Used techniques that compared the pixel colors to some thresholds, and separated into classes just by pixel color.

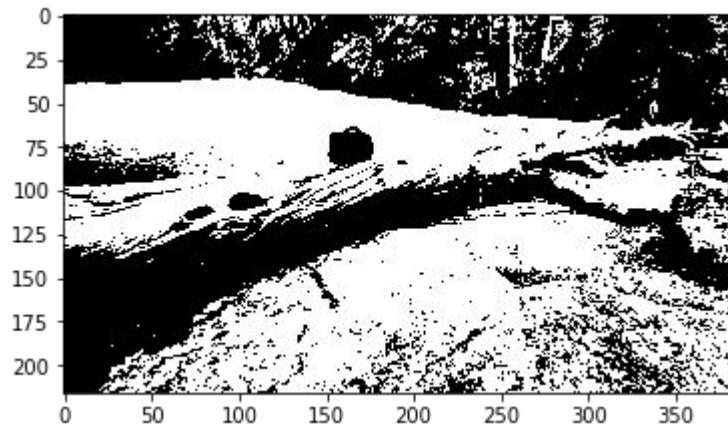
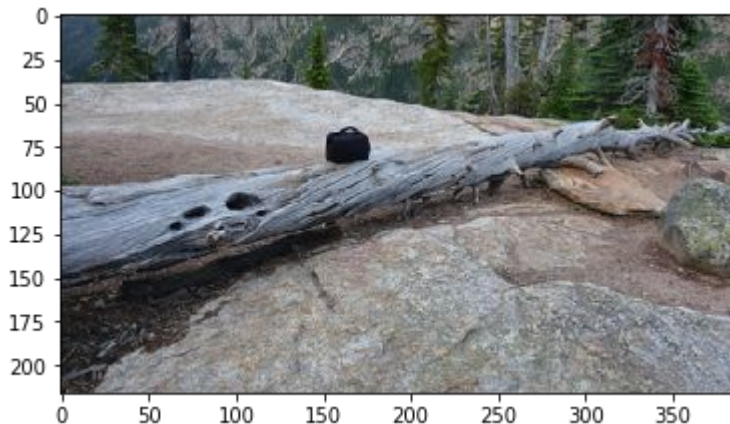
No consideration of shape, or continuity of an object.

Model 1: Distance from mean color

1. Convert the image to grayscale
2. Flatten an image
3. Compare each pixel value to the mean
4. If above the mean, predict class 0.
5. If below the mean, predict class 1.

Results:

Simple approach doesn't provide any meaningful segmentation.

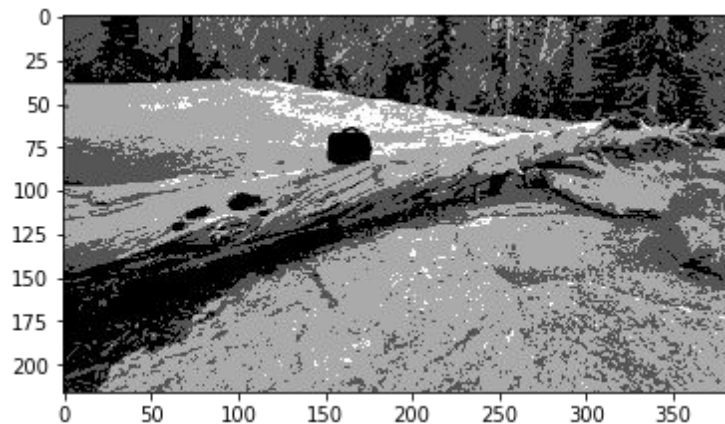
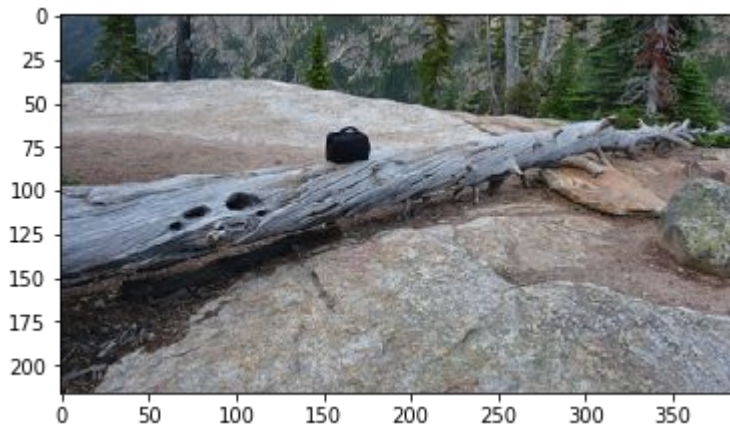


Model 2: Distance from mean color, multiclass

1. Convert to grayscale
2. Flatten an image
3. Compare each pixel value to the mean
4. If above A and below B, predict class 0.
5. If above B and below C, predict class 1.
6. If above C and below D, predict class 2.

Results:

Even by using 4 classes, this simple approach doesn't provide any meaningful segmentation.



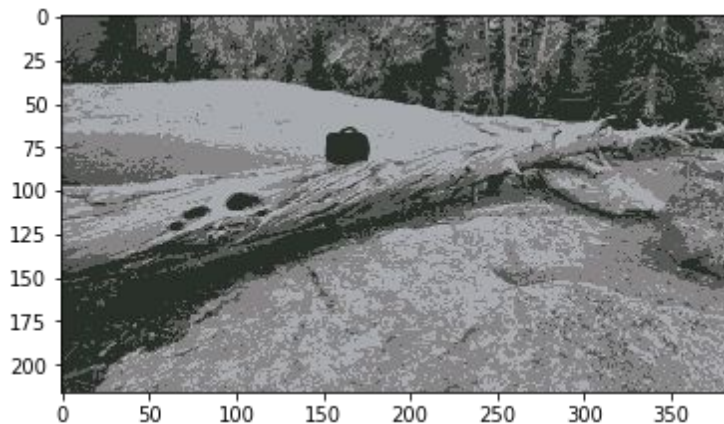
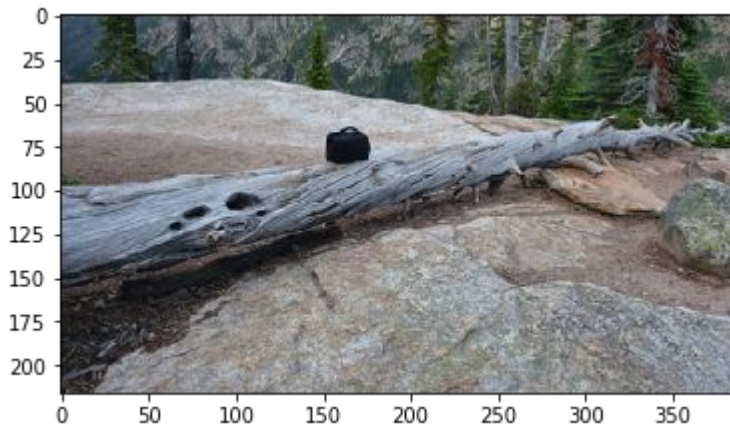
Model 3: Using k-means clustering

1. Flatten an image
2. Use sklearn Kmeans method
3. Select a number of clusters to split unto
4. Reshape the image back to its original shape

Results:

This method does not convert to grayscale, so we can use the colors as additional information.

Even so, it's too simple. This 4 clusters provide no meaningful segmentation

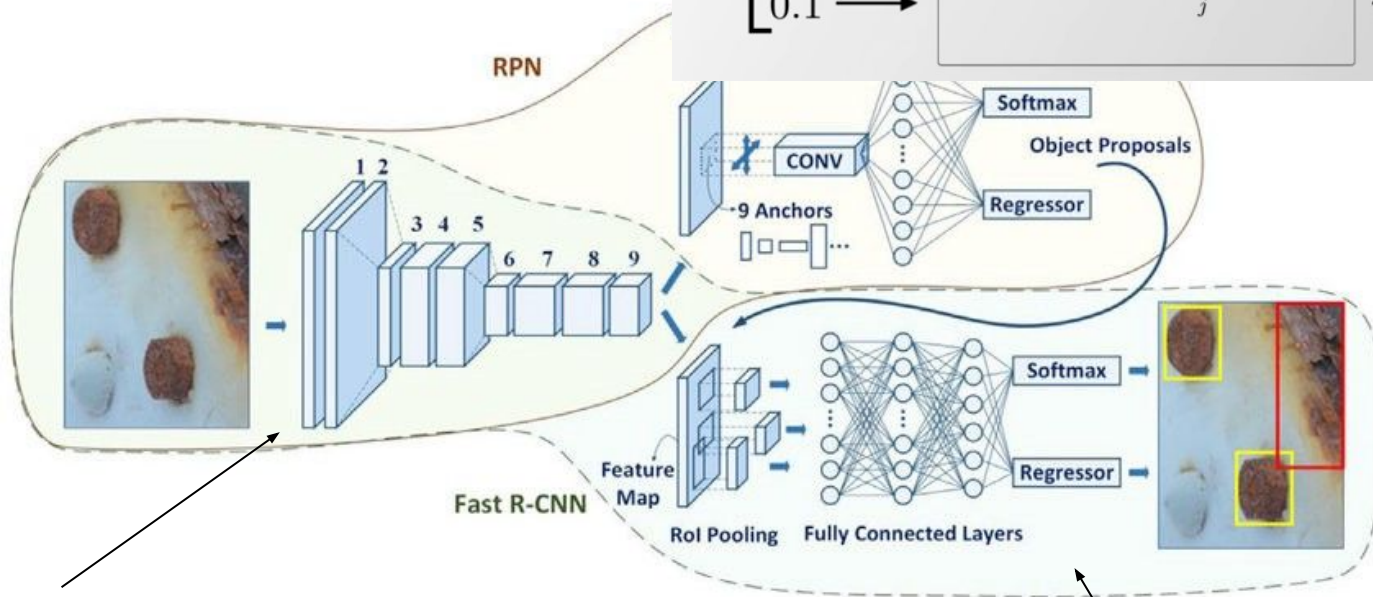


Mask R-CNN

State of the art instance segmentation technique

History

Built on top of Faster R-CNN



Part 1: Convolutional Layers
for feature extraction

Part 3: Detect objects

LOGITS
SCORES

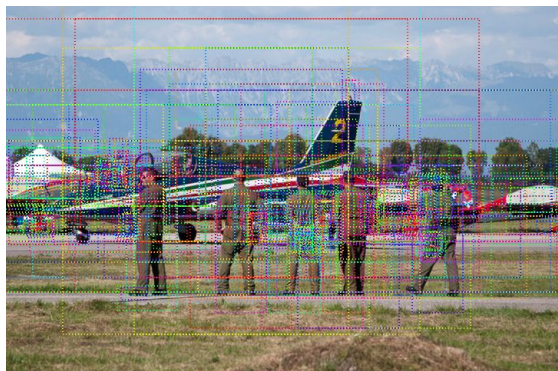
○ SOFTMAX

PROBABILITIES

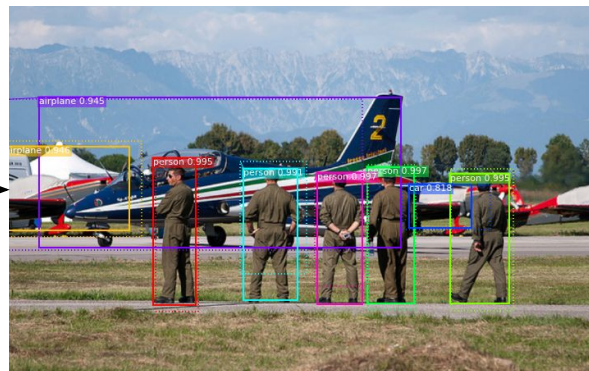
$$y \begin{cases} 2.0 \longrightarrow \\ 1.0 \longrightarrow \\ 0.1 \longrightarrow \end{cases} \begin{matrix} \longrightarrow \\ \longrightarrow \\ \longrightarrow \end{matrix} \begin{matrix} p = 0.7 \\ p = 0.2 \\ p = 0.1 \end{matrix}$$
$$S(y_i) = \frac{e^{y_i}}{\sum_j e^{y_j}}$$

Model 4: Mask R-CNN

1. Anchor sorting and filtering
2. Bounding Box Refinement
3. Generating masks



1



2



3

Loss function

Standard Cross Entropy

Each pixel of the output of the network is compared with the corresponding pixel in the ground truth segmentation image. We will apply standard cross-entropy loss on each pixel.

Other loss functions: DeepMask's CE

$$\mathcal{L}(\theta) = \sum_k \left(\frac{1+y_k}{2w^o h^o} \sum_{ij} \log(1 + e^{-m_k^{ij} f_{segm}^{ij}(x_k)}) + \lambda \log(1 + e^{-y_k f_{score}(x_k)}) \right)$$

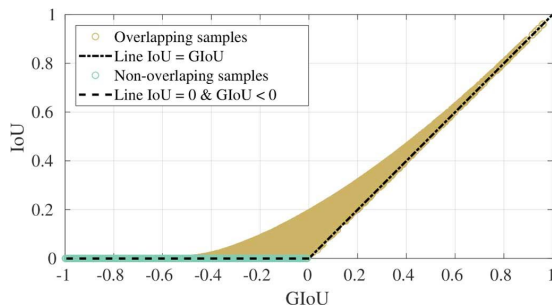
the loss function is a sum of binary logistic regression losses, one for each location of the segmentation network and one for the object score.

Metric: Intersection over Union

$$IoU = \frac{|A \cap B|}{|A \cup B|} = \frac{|I|}{|U|}$$

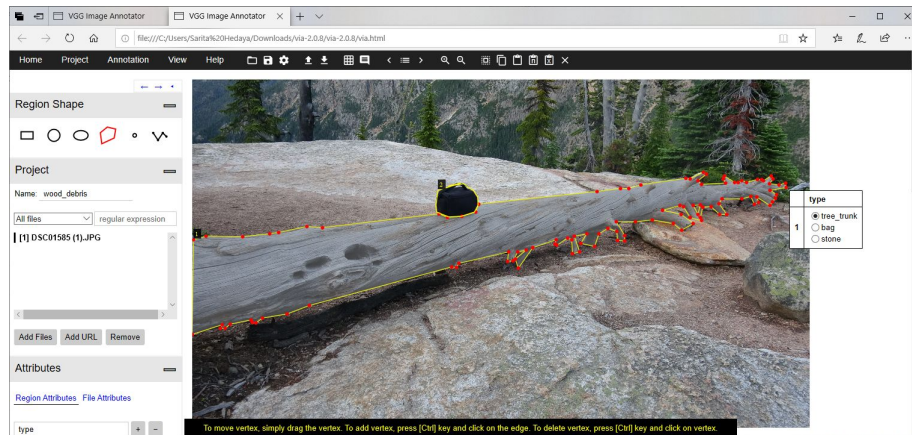
In practice we use GloU because in a neural network, any given loss function must be differentiable to allow for backpropagation.

In cases where there is no intersection, IoU has no value and therefore no gradient. GloU however, is always differentiable



How to train a Mask R-CNN model?

1. Select a dataset of images containing the objects to segment
2. Annotate the images by drawing boundaries around the objects (~3 minutes per image)
3. Use transfer learning from the COCO dataset. (~120,000 images)
4. Train using a modern GPU. Mask R-CNN uses ResNet101, a fairly large model.
5. Inspect your results



Using VIA (VGG Image Annotator) to draw boundaries

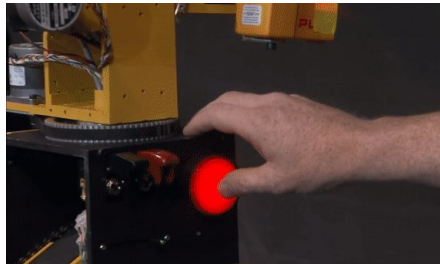
Using a GPU

GPU

- hundreds of simpler cores
- thousand of concurrent hardware threads
- maximize floating-point throughput
- most die surface for integer and fp units

CPU

- few very complex cores
- single-thread performance optimization
- transistor space dedicated to complex ILP
- few die surface for integer and fp units



ir



My progress training a Mask R-CNN

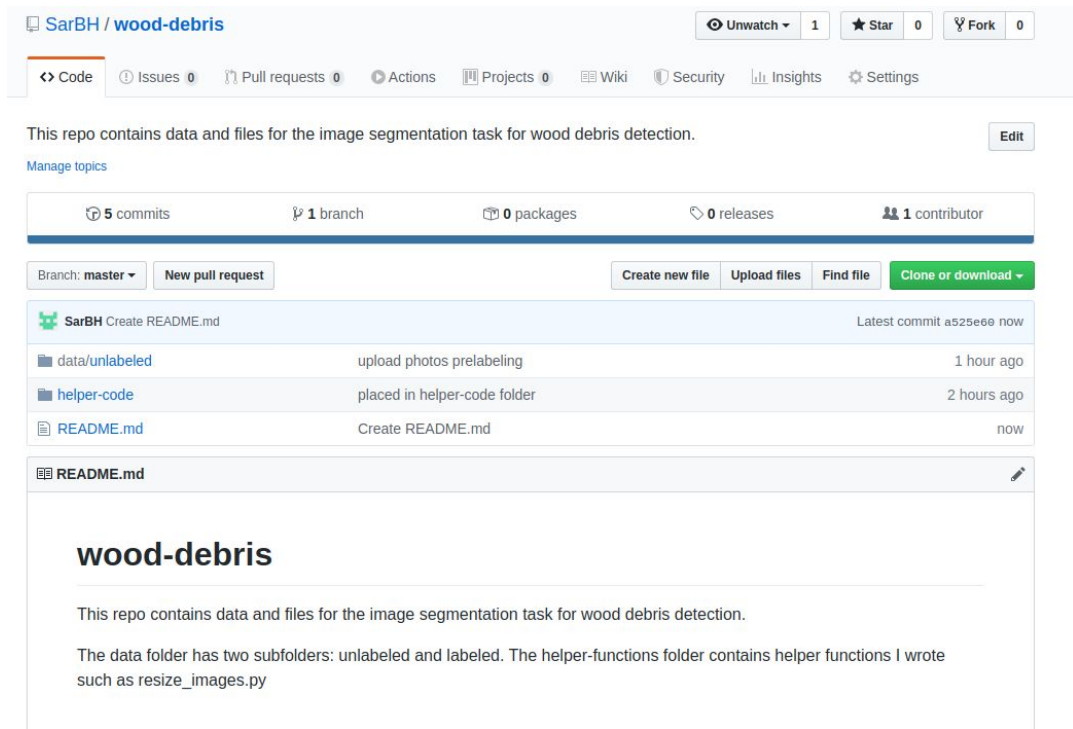


1. Create GitHub repository

A GitHub repository is used to store files, help with version control, manage changes, aid with teamwork, and backup data.

I created a GitHub repo that you can access with this link:

<https://github.com/SarBH/wood-debris>



The screenshot shows the GitHub interface for the repository 'SarBH / wood-debris'. At the top, there are buttons for 'Unwatch', 'Star' (0), and 'Fork' (0). Below this is a navigation bar with tabs for 'Code', 'Issues' (0), 'Pull requests' (0), 'Actions', 'Projects' (0), 'Wiki', 'Security', 'Insights', and 'Settings'. The main content area states 'This repo contains data and files for the image segmentation task for wood debris detection.' and includes an 'Edit' button. Below this, there's a section for repository statistics: '5 commits', '1 branch', '0 packages', '0 releases', and '1 contributor'. A table lists recent commits:

Commit Message	Time
upload photos prelabeling	1 hour ago
placed in helper-code folder	2 hours ago
Create README.md	now

The 'README.md' file is expanded, showing the repository's purpose and folder structure:

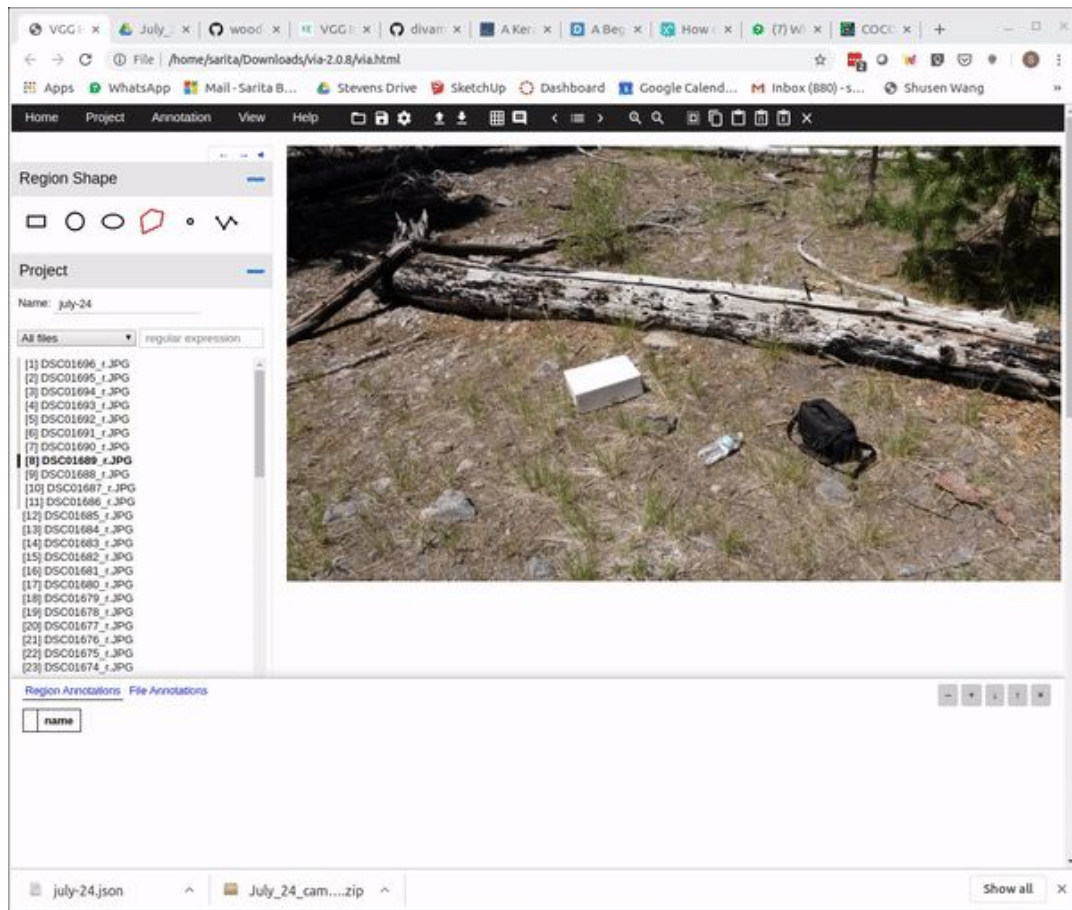
wood-debris

This repo contains data and files for the image segmentation task for wood debris detection.

The data folder has two subfolders: unlabeled and labeled. The helper-functions folder contains helper functions I wrote such as `resize_images.py`

2. Prepare data

- Select images to train on
- Resize images to a more manageable size
- Manually annotate them to create masks
- Create image mask from polygon points



2. Prepare data



Original Image



VIA annotation

```
"DSC01695_r.JPG360793": {  
  "filename": "DSC01695_r.JPG",  
  "size": 360793,  
  "regions": [  
    {  
      "shape_attributes": {  
        "name": "polygon",  
        "all_points_x": [  
          511,  
          474,  
          479,  
          550,  
          584,  
          580  
        ],  
        "all_points_y": [  
          292,  
          315,  
          343,  
          355,  
          326,  
          303  
        ]  
      },  
      "region_attributes": {  
        "name": "box"  
      }  
    }  
  ]  
}
```

VIA output as a
.json file



Mask

3. Baseline to measure progress

Machine learning is an **iterative process** that relies on **experimentation**. Because of this, it's important to **set baselines to measure improvement**.

Since my idea is to repurpose a pre-trained model, I baselined by **evaluating the original model on our data**.



info@cocodataset.org

Home People Dataset Tasks Evaluate

COCO Explorer

COCO 2017 train/val browser (123,287 images, 886,284 instances). Crowd labels not shown.

Future improvements to our approach...

01

Even if we get segmentation to work reliably, camera bag can take multiple “sizes” because of the loose carrying belt, therefore should not be used to calibrate volume of trees.

Use reference objects that don't change shape due to positioning, a non malleable solid object should work better.

02

Images are not taken from the “top view”, so depending on the angle and how far the reference objects are will change the mask size. Therefore cannot be reliably used to estimate debris size.

Take pictures from a birds eye view, possibly using a drone.

03

Current method of taking photos does not account for possible debris duplicates, i.e. multiple photos taken of the same branch will be added twice to the total debris calculation.

Replace current capturing system by an automated system that keeps a map of already captured areas.

Sources

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https://github.com/matterport/Mask_RCNN

Additional Slides

Go to <https://github.com/SarBH/wood-debris>

Transfer Learning

In practice, very few people train an entire Convolutional Network from scratch (with random initialization), because it is relatively rare to have a dataset of sufficient size.

Instead, it is common to pretrain a ConvNet on a very large dataset (e.g. ImageNet, which contains 1.2 million images with 1000 categories), and then use the ConvNet either as an initialization or a fixed feature extractor for the task of interest.

Implementation

Each pixel of the output of the network is compared with the corresponding pixel in the ground truth segmentation image. We will apply standard cross-entropy loss on each pixel.