



limhponer / computervision-final-prep

[Code](#)[Issues](#)[Pull requests](#)[Actions](#)[Projects](#)[Wiki](#)[Security](#)[In](#)[computervision-final-prep / lab / Lab 06 \(ML Foundations\)-20251128 / Annotations.ipynb](#)

limhponer ML

fef7e4e · 2 hours ago



312 lines (312 loc) · 8.05 KB

[Preview](#)[Code](#)[Blame](#)

Raw



Tutorial: Image Annotation for Computer Vision

- Annotation is the process of labeling images so machines can learn to “see” like humans.
- It’s the ground truth used to train AI models in tasks such as object detection and segmentation.

Types of annotation

- Bounding box
- Segmentation

Bounding Box

- Definition: The simplest form of annotation. Draw a rectangle around the object of interest.
- Use case: Object detection (e.g., detecting cats, cars, faces).
- How it’s stored: Coordinates of top-left corner (x, y), plus width and height.
- Advantages: Simple, fast, small annotation files.
- Limitations: Doesn’t capture the exact shape of the object (lots of background included).

Example:

- A box drawn around a cat in an image.
- Supported by formats: YOLO, COCO, Pascal VOC.

Segmentation

- Definition: More precise labeling, where each object is outlined by its exact shape.

Types:

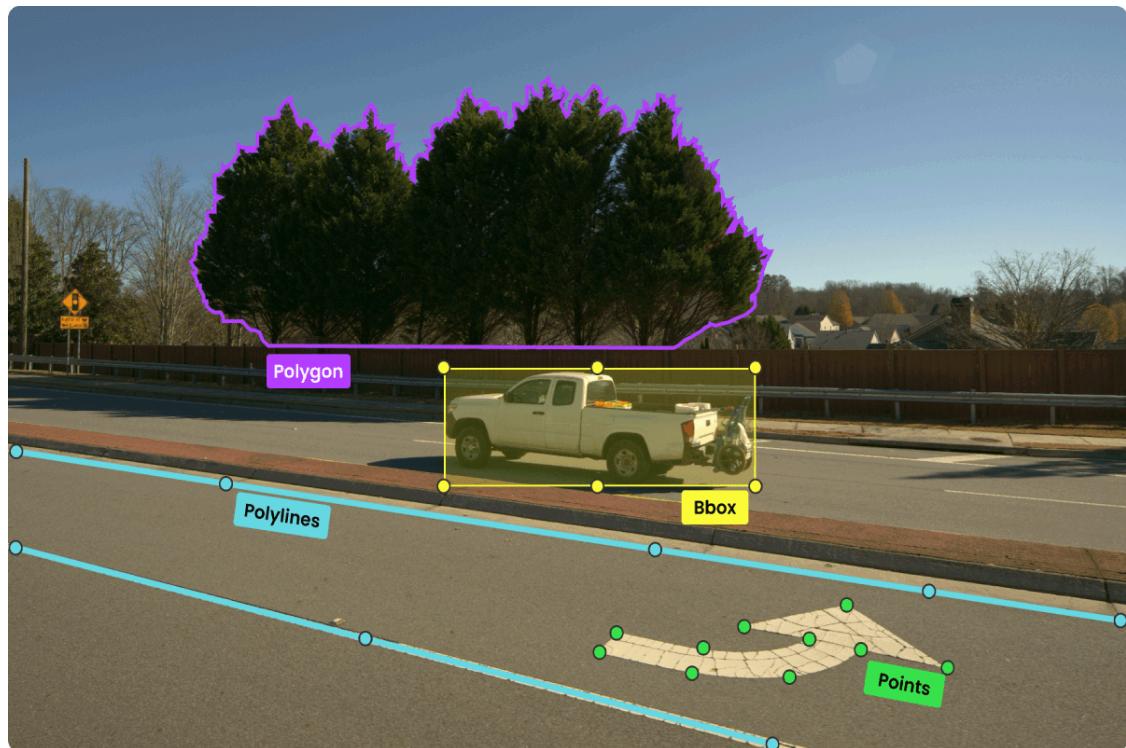
- Polygonal segmentation: draw polygons around objects.
- Pixel-wise segmentation (masking): each pixel is assigned a class label.
- Use case: Semantic segmentation (road/lane markings), instance segmentation

(detecting multiple objects of the same class).

- How it's stored: Polygons (list of points) or binary masks.
- Advantages: Very accurate, useful for detailed tasks.
- Limitations: Time-consuming, large annotation files.

Example:

- Carefully outlining the cat's ears, tail, and body instead of just using a rectangle.
- Supported by formats: COCO, Pascal VOC (mask), LabelMe JSON.



Annotation Tools

- For bounding boxes: [labelImg](#), [VGG Image Annotator \(VIA\)](#), [CVAT](#).
- For segmentation: [LabelMe](#), [VGG Image Annotator \(VIA\)](#), [CVAT](#).

Annotation Formats

Different datasets and frameworks use different annotation formats. Let's look at the **three most common ones** with examples.

COCO Format (JSON)

- **Supports:** Bounding boxes, segmentation, keypoints

- **Bounding box format:** [x_min, y_min, width, height]
- **Widely used in:** Detectron2, MMDetection, COCO dataset

Example:

```
{
  "images": [
    {
      "id": 1,
      "file_name": "cat.jpg",
      "width": 800,
      "height": 600
    }
  ],
  "annotations": [
    {
      "id": 1,
      "image_id": 1,
      "category_id": 1,
      "bbox": [100, 150, 200, 300],
      "area": 60000,
      "iscrowd": 0
    }
  ],
  "categories": [
    {
      "id": 1,
      "name": "cat",
      "supercategory": "animal"
    }
  ]
}
```

Pascal VOC Format (XML)

- **Supports:** Bounding boxes, segmentation masks
- **Bounding box format:** (xmin, ymin, xmax, ymax)
- **Widely used in:** TensorFlow Object Detection API, Pascal VOC dataset

Example

```
<annotation>
  <folder>images</folder>
  <filename>cat.jpg</filename>
  <size>
    <width>800</width>
    <height>600</height>
    <depth>3</depth>
  </size>
  <object>
    <name>cat</name>
```

```

<pose>Unspecified</pose>
<truncated>0</truncated>
<difficult>0</difficult>
<bndbox>
    <xmin>100</xmin>
    <ymin>150</ymin>
    <xmax>300</xmax>
    <ymax>450</ymax>
</bndbox>
</object>
</annotation>

```

YOLO Format (TXT)

Supports: Bounding boxes only

Bounding box format: class_id x_center y_center width height

(all values normalized between 0-1)

Example 0 0.25 0.5 0.25 0.5

Explanation (for an image of width 800 and height 600):

- 0 → class ID for "cat"
- x_center = $(100 + 200/2) / 800 = 0.25$
- y_center = $(150 + 300/2) / 600 = 0.5$
- width = $200 / 800 = 0.25$
- height = $300 / 600 = 0.5$

VGG IMAGE ANNOTATOR

In []:

```
!git clone https://github.com/nearkyh/via-1.0.5.git
```

1. Open the folder via-1.0.5
2. Run via.html

Follow the demonstration in the lab for simple bbox and segmentation.

labelimg

If activation scripts are blocked, run the process-scope bypass first:

```
Set-ExecutionPolicy -Scope Process -ExecutionPolicy Bypass
```

Register the venv as Jupyter Kernel

```
python -m ipykernel install --user --name=cv-lab-venv --display-name  
'Python (cv-lab-venv)'
```

Install pip

```
python -m ensurepip --upgrade
```

```
python -m pip install --upgrade pip setuptools wheel
```

Install package labelimg

```
python -m pip install labelimg
```

Run labelimg

```
labelimg
```

Check more on labelimg here: <https://pypi.org/project/labelImg/>