# **Run an Instance Segmentation Model**

For some applications it isn't adequate enough to localize an object with a simple bounding box. For instance, you might want to segment an object region once it is detected. This class of problems is called **instance segmentation**.

### Materializing data for instance segmentation {#materializing-instance-seg}

Instance segmentation is an extension of object detection, where a binary mask (i.e. object vs. background) is associated with every bounding box. This allows for more fine-grained information about the extent of the object within the box. To train an instance segmentation model, a groundtruth mask must be supplied for every groundtruth bounding box. In additional to the proto fields listed in the section titled <a href="Using-your own dataset">Using-your own dataset</a>, one must also supply <a href="image/object/mask">image/object/mask</a>, which can either be a repeated list of single-channel encoded PNG strings, or a single dense 3D binary tensor where masks corresponding to each object are stacked along the first dimension. Each is described in more detail below.

# **PNG Instance Segmentation Masks**

Instance segmentation masks can be supplied as serialized PNG images.

```
image/object/mask = ["\x89PNG\r\n\x1A\n\x00\x00\rIHDR\...", ...]
```

These masks are whole-image masks, one for each object instance. The spatial dimensions of each mask must agree with the image. Each mask has only a single channel, and the pixel values are either 0 (background) or 1 (object mask). **PNG masks are the preferred** parameterization since they offer considerable space savings compared to dense numerical masks.

# **Dense Numerical Instance Segmentation Masks**

Masks can also be specified via a dense numerical tensor.

```
image/object/mask = [0.0, 0.0, 1.0, 1.0, 0.0, ...]
```

For an image with dimensions  $\[mum\_boxes\]$  and  $\[mum\_boxes\]$  groundtruth boxes, the mask corresponds to a [  $\[mum\_boxes\]$ ,  $\[mum\_boxes\]$ ,  $\[mum\_boxes\]$ ,  $\[mum\_boxes\]$  are read in row-major format, so the elements are organized as:

```
... mask 0 row 0 ... mask 0 row 1 ... // ... mask 0 row H-1 ... mask 1 row 0 ...
```

where each row has W contiguous binary values.

To see an example tf-records with mask labels, see the examples under the <u>Preparing Inputs</u> section.

#### Pre-existing config files

We provide four instance segmentation config files that you can use to train your own models:

- 1. mask\_rcnn\_inception\_resnet\_v2\_atrous\_coco
- 2. mask\_rcnn\_resnet101\_atrous\_coco
- 3. mask rcnn resnet50 atrous coco
- 4. mask\_rcnn\_inception\_v2\_coco

For more details see the detection model zoo.

# **Updating a Faster R-CNN config file**

Currently, the only supported instance segmentation model is Mask R-CNN, which requires Faster R-CNN as the backbone object detector.

Once you have a baseline Faster R-CNN pipeline configuration, you can make the following modifications in order to convert it into a Mask R-CNN model.

- 1. Within train\_input\_reader and eval\_input\_reader, set load\_instance\_masks to True . If using PNG masks, set mask\_type to PNG\_MASKS, otherwise you can leave it as the default 'NUMERICAL\_MASKS'.
- $2. \ Within the \ \texttt{faster\_rcnn} \ \ \texttt{config, use a} \ \ \texttt{MaskRCNNBoxPredictor} \ \ \textbf{as the} \ \ \texttt{second\_stage\_box\_predictor} \ .$
- 3. Within the MaskRCNNBoxPredictor message, set predict\_instance\_masks to True . You must also define conv hyperparams .
- 4. Within the faster rcnn message, set number of stages to 3.
- 5. Add instance segmentation metrics to the set of metrics: 'coco mask metrics'.
- 6. Update the <code>input\_path</code> s to point at your data.

Please refer to the section on Running the pets dataset for additional details.

Note: The mask prediction branch consists of a sequence of convolution layers. You can set the number of convolution layers and their depth as follows:

- 1. Within the MaskRCNNBoxPredictor message, set the mask\_prediction\_conv\_depth to your value of interest. The default value is 256. If you set it to 0 (recommended), the depth is computed automatically based on the number of classes in the dataset.
- 2. Within the MaskRCNNBoxPredictor message, set the mask\_prediction\_num\_conv\_layers to your value of interest.

  The default value is 2.