

TENSORBOARD & OBJECT DETECTION

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OVERVIEW

- The Individual Pieces
 - Object detection
 - TensorFlow
 - TensorBoard
- TensorBoard + Object Detection
- TensorBoard on HPC
- Understanding the TensorBoard interface

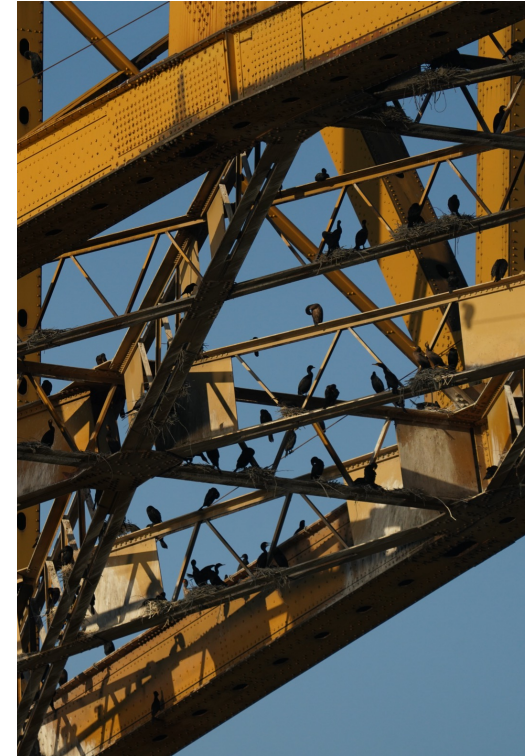
OBJECT DETECTION

OBJECT DETECTION

- Computer vision task where a trained model identify objects within an image or video
- Applications include self-driving vehicles, remote sensing, wildlife monitoring



WILDLIFE MONITORING EXAMPLE



TENSORFLOW

TENSORFLOW

- A free & open source library for machine learning, focused on deep neural networks
- The Alliance [TensorFlow documentation](#)

* TF is available in other languages (e.g. JavaScript, C++, Java) but is most commonly used with Python

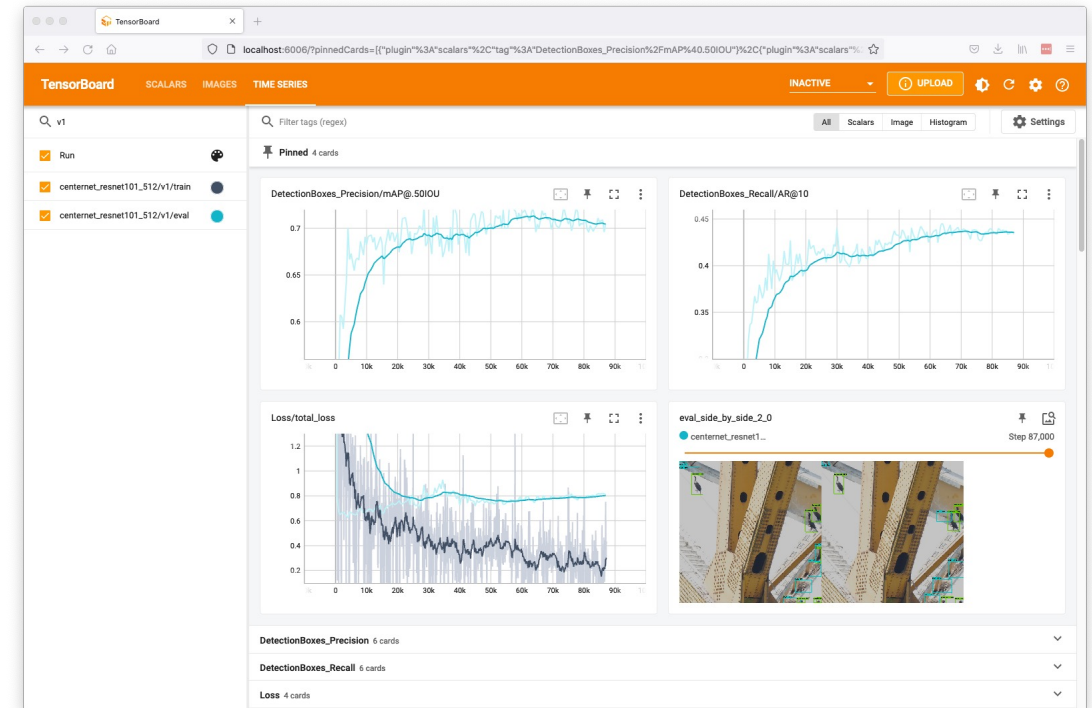
OBJECT DETECTION & TENSORFLOW

- The TensorFlow Object Detection API is one of the research models/implementations provided in the [TF Model Garden](#)
- Learn more in TF's [object detection tutorial](#) & the Object Detection API's [Training & Evaluation with TensorFlow 2](#) guide

TENSORBOARD

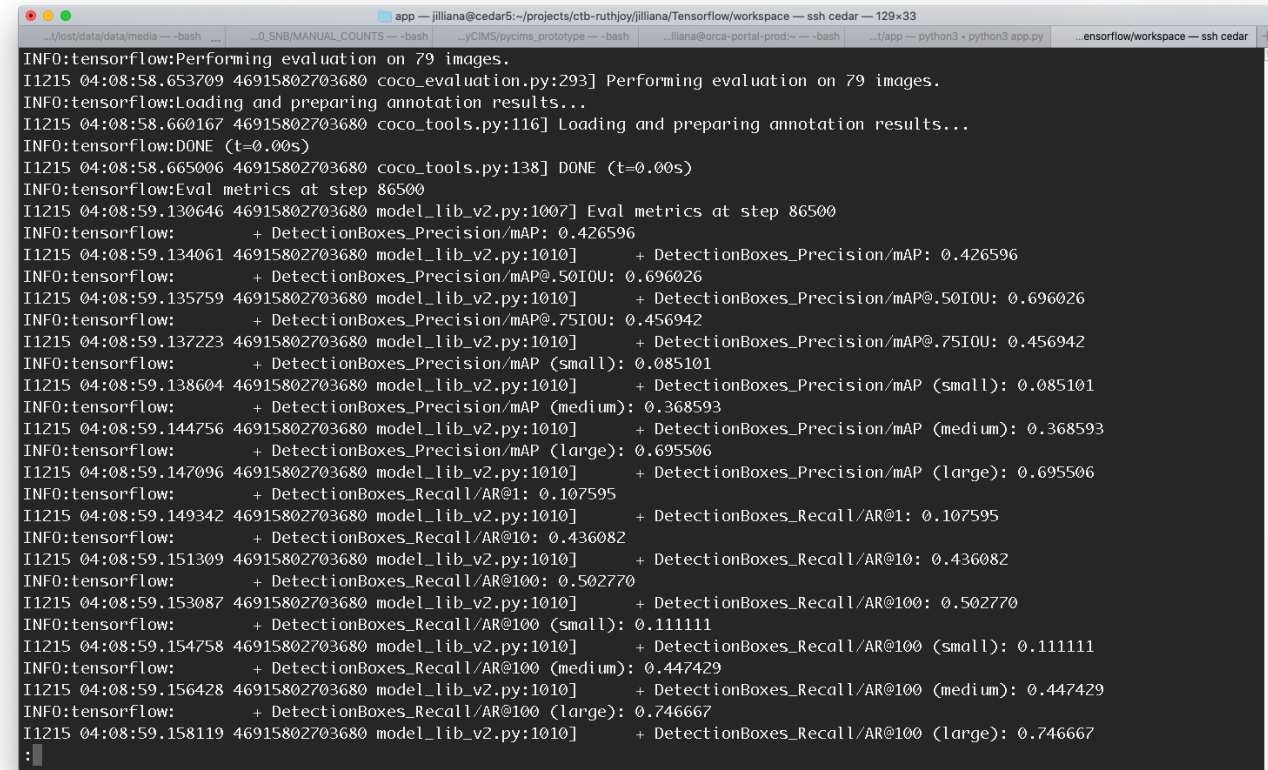
WHAT IS TENSORBOARD?

- TF's visual toolkit for machine learning experimentation
- Visually track model information (e.g. model structure, mAP, etc)



WHY USE TENSORBOARD?

- Machine Learning usually requires experimentation
- We need a way to compare models (aka runs)
- Ideally we want to avoid manually inspecting logs



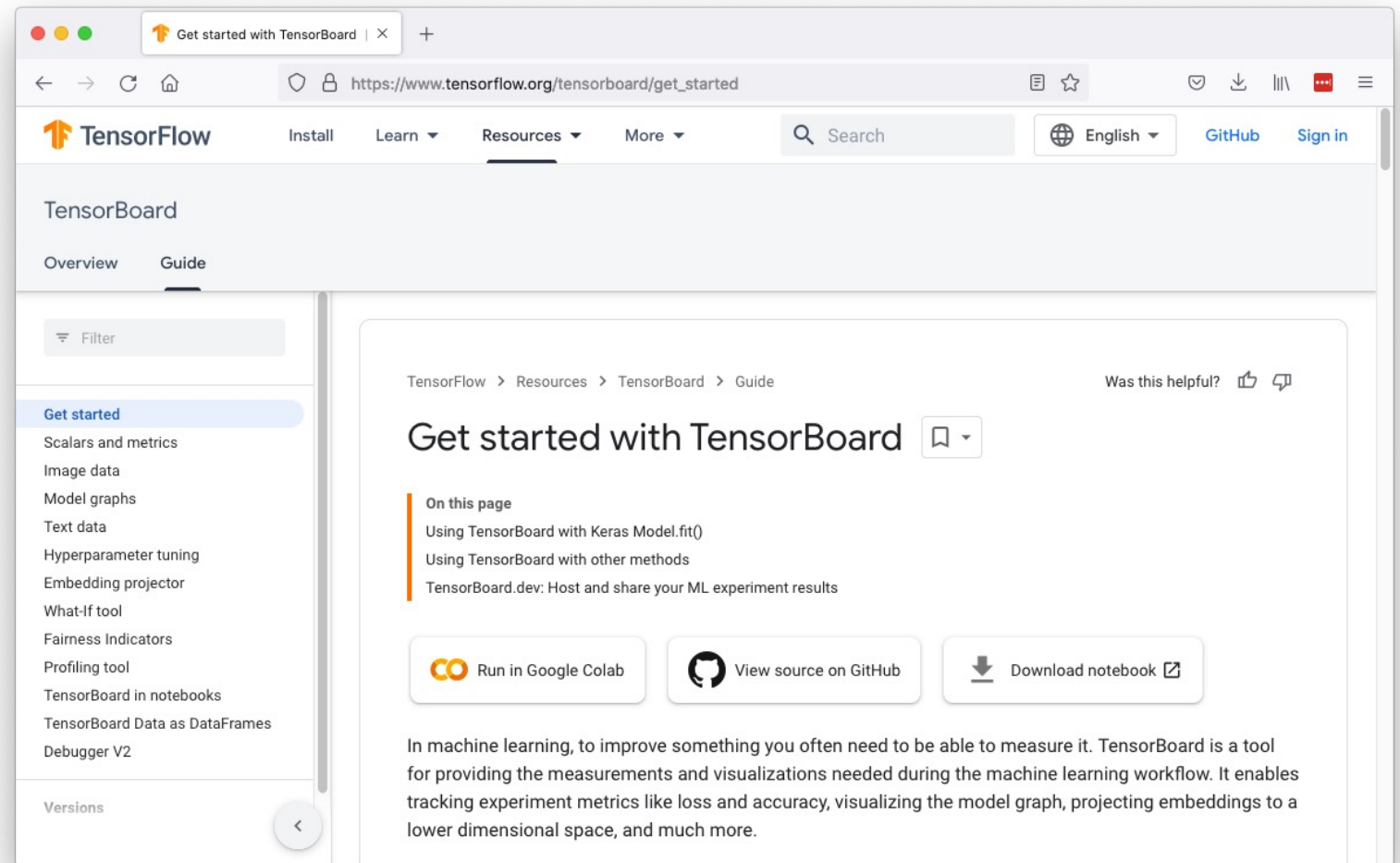
```
app — jilliana@cedar5:~/projects/ctb-ruthjoy/jilliana/Tensorflow/workspace — ssh cedar — 129x33
~/float/data/data/media — bash ... 0_SNB/MANUAL_COUNTS — bash ... yCIMS/pycims_prototype — bash ... jilliana@orca-portal-prod — bash ... t/app — python3 + python3 app.py ... ensorflow/workspace — ssh cedar
INFO:tensorflow:Performing evaluation on 79 images.
I1215 04:08:58.653709 46915802703680 coco_evaluation.py:293] Performing evaluation on 79 images.
INFO:tensorflow:Loading and preparing annotation results...
I1215 04:08:58.666167 46915802703680 coco_tools.py:116] Loading and preparing annotation results...
INFO:tensorflow:DONE (t=0.00s)
I1215 04:08:58.665006 46915802703680 coco_tools.py:138] DONE (t=0.00s)
INFO:tensorflow:Eval metrics at step 86500
I1215 04:08:59.130646 46915802703680 model_lib_v2.py:1007] Eval metrics at step 86500
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP: 0.426596
I1215 04:08:59.134061 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP: 0.426596
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP@.50IOU: 0.696026
I1215 04:08:59.135759 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP@.50IOU: 0.696026
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP@.75IOU: 0.456942
I1215 04:08:59.137223 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP@.75IOU: 0.456942
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP (small): 0.085101
I1215 04:08:59.138604 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP (small): 0.085101
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP (medium): 0.368593
I1215 04:08:59.144756 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP (medium): 0.368593
INFO:tensorflow:
+ DetectionBoxes_Precision/mAP (large): 0.695506
I1215 04:08:59.147096 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Precision/mAP (large): 0.695506
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@1: 0.107595
I1215 04:08:59.149342 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@1: 0.107595
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@10: 0.436082
I1215 04:08:59.151309 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@10: 0.436082
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@100: 0.502770
I1215 04:08:59.153087 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@100: 0.502770
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@100 (small): 0.111111
I1215 04:08:59.154758 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@100 (small): 0.111111
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@100 (medium): 0.447429
I1215 04:08:59.156428 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@100 (medium): 0.447429
INFO:tensorflow:
+ DetectionBoxes_Recall/AR@100 (large): 0.746667
I1215 04:08:59.158119 46915802703680 model_lib_v2.py:1010] + DetectionBoxes_Recall/AR@100 (large): 0.746667
:
```

HOW DOES TENSORBOARD WORK?

- TensorFlow writes event files during model training
- TensorBoard parses these files & generates visualizations
- You can control what events are logged using `FileWriter` instance(s)

MORE ON TENSORBOARD

→ [TensorBoard guide](#)



TENSORBOARD + OBJECT DETECTION

SETTING UP EVENT FILES

- Setup pipeline.config to generate event files for **scalars*** and **images***
- eval_config – how should evaluation be done?
- eval_input_reader – what data should be used for evaluation?

```
eval_config: {  
  metrics_set: "coco_detection_metrics"  
  use_moving_averages: false  
  batch_size: 1  
  num_visualizations: 50  
  min_score_threshold: 0.1  
}  
  
eval_input_reader: {  
  label_map_path: "models/snb3/label_map.pbtxt"  
  shuffle: false  
  num_epochs: 1  
  tf_record_input_reader {  
    input_path: "models/snb3/validation.tfrecord-*"  
  }  
}
```

* Other info (e.g. graph structure, Tensor histograms) can be tracked too, but we won't cover this today.

SCALARS

→ By default evaluation uses COCO detection metrics

2. Metrics

The following 12 metrics are used for characterizing the performance of an object detector on COCO:

```

Average Precision (AP):
AP                                     % AP at IoU=.50:.05:.95 (primary challenge metric)
APIoU=.50                           % AP at IoU=.50 (PASCAL VOC metric)
APIoU=.75                           % AP at IoU=.75 (strict metric)

AP Across Scales:
APsmall                             % AP for small objects: area < 322
APmedium                           % AP for medium objects: 322 < area < 962
APlarge                             % AP for large objects: area > 962

Average Recall (AR):
ARmax=1                             % AR given 1 detection per image
ARmax=10                            % AR given 10 detections per image
ARmax=100                           % AR given 100 detections per image

AR Across Scales:
ARsmall                             % AR for small objects: area < 322
ARmedium                           % AR for medium objects: 322 < area < 962
ARlarge                             % AR for large objects: area > 962

```

1. Unless otherwise specified, *AP* and *AR* are averaged over multiple Intersection over Union (*IoU*) values. Specifically we use 10 *IoU* thresholds of .50:.05:.95. This is a break from tradition, where *AP* is computed at a single *IoU* of .50 (which corresponds to our metric $AP_{IoU=.50}$). Averaging over *IoUs* rewards detectors with better localization.
2. *AP* is averaged over all categories. Traditionally, this is called "mean average precision" (*mAP*). We make no distinction between *AP* and *mAP* (and likewise *AR* and *mAR*) and assume the difference is clear from context.
3. *AP* (averaged across all 10 *IoU* thresholds and all 80 categories) will determine the challenge winner. This should be considered the single most important metric when considering performance on COCO.
4. In COCO, there are more small objects than large objects. Specifically: approximately 41% of objects are small (area < 32²), 34% are medium (32² < area < 96²), and 24% are large (area > 96²). Area is measured as the number of pixels in the segmentation mask.
5. *AR* is the maximum recall given a fixed number of detections per image, averaged over categories and *IoUs*. *AR* is related to the metric of the same name used in [proposal evaluation](#) but is computed on a per-category basis.
6. All metrics are computed allowing for at most 100 top-scoring detections per image (across all categories).
7. The evaluation metrics for detection with bounding boxes and segmentation masks are identical in all respects except for the *IoU* computation (which is performed over boxes or masks, respectively).

SCALARS

- Specified using `metrics_set`
- See the [docs](#) for the other available metrics

```
eval_confia: {  
  metrics_set: "coco_detection_metrics"  
  use_moving_averages: false  
  batch_size: 1  
  num_visualizations: 50  
  min_score_threshold: 0.1  
}  
  
eval_input_reader: {  
  label_map_path: "models/snb3/label_map.pbtxt"  
  shuffle: false  
  num_epochs: 1  
  tf_record_input_reader {  
    input_path: "models/snb3/validation.tfrecord-*"  
  }  
}
```

IMAGES

- Visualize model predictions on evaluation data
- Compare ground truth with predictions

```
eval_config: {  
  metrics_set: "coco_detection_metrics"  
  use_moving_averages: false  
  batch_size: 1  
  num_visualizations: 50  
  min_score_threshold: 0.1  
}  
  
eval_input_reader: {  
  label_map_path: "models/snb3/label_map.pbtxt"  
  shuffle: false  
  num_epochs: 1  
  tf_record_input_reader {  
    input_path: "models/snb3/validation.tfrecord-*"  
  }  
}
```

IMAGES

- Specified with `num_visualizations`
- `min_score_threshold` controls which detections are visualized
- To explore other options, checkout the [eval.proto](#) file

```
eval_config: {  
  metrics_set: "coco_detection_metrics"  
  use_moving_averages: false  
  batch_size: 1  
  num_visualizations: 50  
  min_score_threshold: 0.1  
}  
  
eval_input_reader: {  
  label_map_path: "models/snb3/label_map.pbtxt"  
  shuffle: false  
  num_epochs: 1  
  tf_record_input_reader {  
    input_path: "models/snb3/validation.tfrecord-*"  
  }  
}
```

TENSORBOARD ON HPC

TENSORBOARD ON HPC

- During model training
- After model training is complete

DURING MODEL TRAINING

- Adjust your job script
- Submit your job as normal
- Connect to the node running the job
- Visit TensorBoard in your browser

DURING MODEL TRAINING

ADJUST → **SUBMIT** → **CONNECT** → **VISIT**

→ Activate evaluation in your job script

```
python model_main_tf2.py \  
  --pipeline_config_path=$MODELDIR/pipeline.config \  
  --model_dir=$MODELDIR \  
  --checkpoint_dir=$MODELDIR \  
  --num_workers=1 \  
  --sample_1_of_n_eval_examples=1 &
```

→ Activate TensorBoard in your job script

```
tensorboard --logdir=$MODELDIR --host 0.0.0.0 --load_fast false &
```


DURING MODEL TRAINING

ADJUST → **SUBMIT** → CONNECT → VISIT

- Submit your job to the cluster as normal
`sbatch train_model.sh`
- Wait for your job to begin

DURING MODEL TRAINING

ADJUST → SUBMIT → **CONNECT** → VISIT

→ Find what node your model is running on
`squeue -u user`

JOBID	USER	ACCOUNT	NAME	ST	TIME_LEFT	NODES	CPUS	TRES_PER_N	MIN_MEM	NODELIST	(REASON)
47623590	jilliana	def-jilliana	train.sh	R	9:43	1	1	gres:gpu:1	32G	cdr250	(None)

↑
Job ID

↑
HPC
node

DURING MODEL TRAINING

ADJUST → SUBMIT → **CONNECT** → VISIT

→ Find the node your model is running on

→ Find the port TensorBoard is running on
`grep "TensorBoard" slurm-JOBID.out`

```
TensorBoard 2.6.0 at http://0.0.0.0:6006/ (Press CTRL+C to quit)
```

↑
Port
on
HPC
node

DURING MODEL TRAINING

ADJUST → SUBMIT → **CONNECT** → VISIT

- Find what node your model is running on
- Find what port TensorBoard is running on
- Connect to the node where training is happening

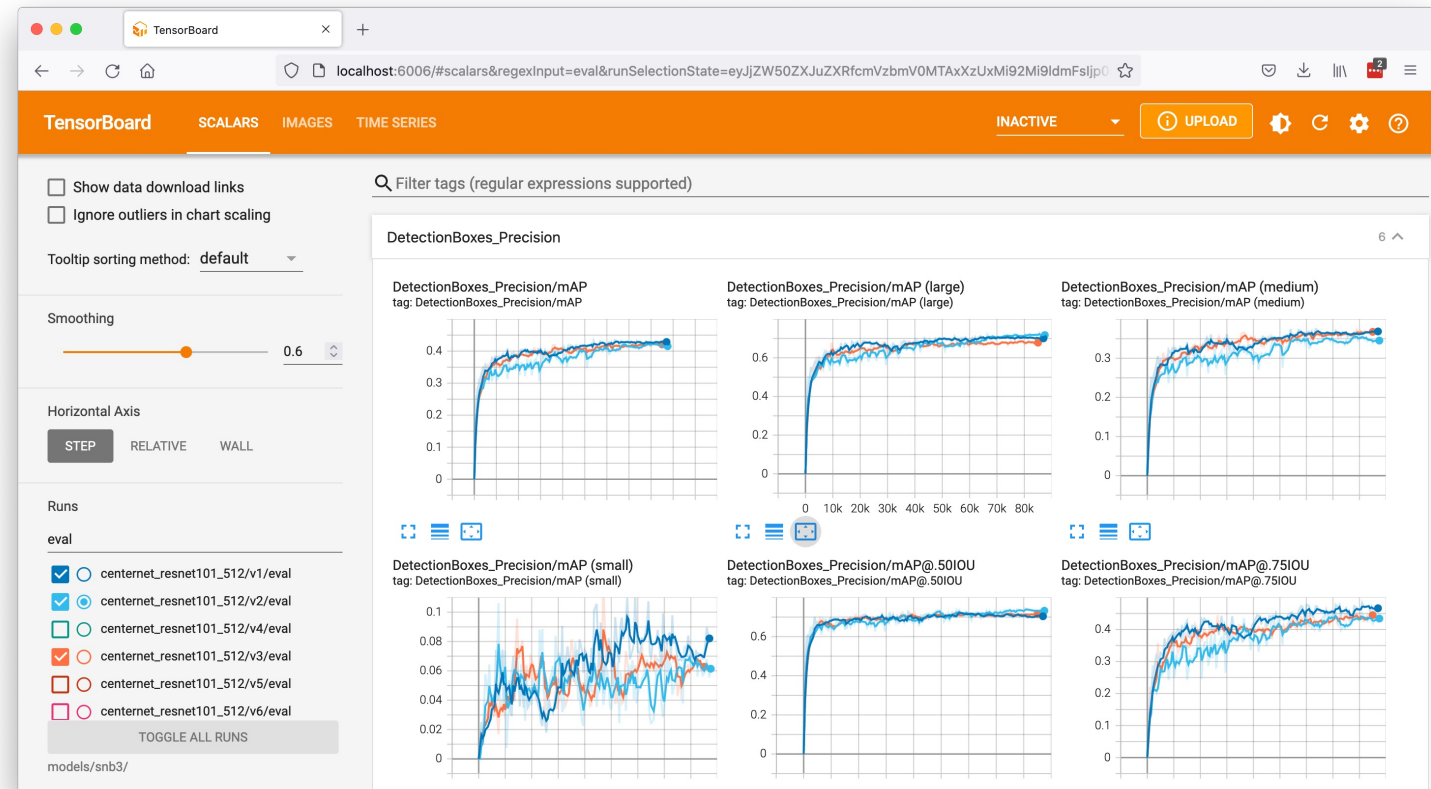
```
ssh -N -f -L localhost:6006:cdr917:6006 user@cedar.computecanada.ca
```

HPC node Port on HPC node Your user Your HPC cluster

DURING MODEL TRAINING

ADJUST → SUBMIT → CONNECT → VISIT

→ Visit localhost:6006
on your favourite
browser



AFTER MODEL TRAINING*

- Request an interactive job
- Start TensorBoard
- Connect to the node running the job
- Visit TensorBoard in your browser

* Requires event files to have been generated during training

AFTER MODEL TRAINING

REQUEST → **START TENSORBOARD** → **CONNECT** → **VISIT**

→ Request an interactive job

```
salloc --mem=16GB --time=1:00:00
```

→ Wait for the interactive session to begin

```
[jilliana@cedar1 Tensorflow]$ salloc --account=def-jilliana --time=1:00:00
salloc: Pending job allocation 47519849
salloc: job 47519849 queued and waiting for resources
salloc: job 47519849 has been allocated resources
salloc: Granted job allocation 47519849
[jilliana@cdr544 Tensorflow]$
```

↑
HPC
node

AFTER MODEL TRAINING

REQUEST → **START TENSORBOARD** → CONNECT → VISIT

→ Run TensorBoard

```
tensorboard --logdir=models/dir/ --host 0.0.0.0 --load_fast false
```

→ Wait for confirmation that TensorBoard is running

```
[jilliana@cdr783 workspace]$ tensorboard --logdir=models/snb3/ --host 0.0.0.0 --load_fast false
2022-10-10 15:49:15.936765: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic library 'libcud
a.so.1'; dLError: libcuda.so.1: cannot open shared object file: No such file or directory
2022-10-10 15:49:15.937499: W tensorflow/stream_executor/cuda/cuda_driver.cc:269] failed call to cuInit: UNKNOWN ERROR (303)
2022-10-10 15:49:15.937903: I tensorflow/stream_executor/cuda/cuda_diagnostics.cc:156] kernel driver does not appear to be running
on this host (cdr783.int.cedar.computecanada.ca): /proc/driver/nvidia/version does not exist
TensorBoard 2.9.1 at http://0.0.0.0:6006/ (Press CTRL+C to quit)
```

↑
Port on
HPC node

AFTER MODEL TRAINING

REQUEST → START TENSORBOARD → **CONNECT** → VISIT

→ Find what node & port your job is running on

→ Connect to the node where training is happening

```
ssh -N -f -L localhost:6006:cdr917:6006 user@cedar.computecanada.ca
```

↑
HPC
node

↑
Port on
HPC node

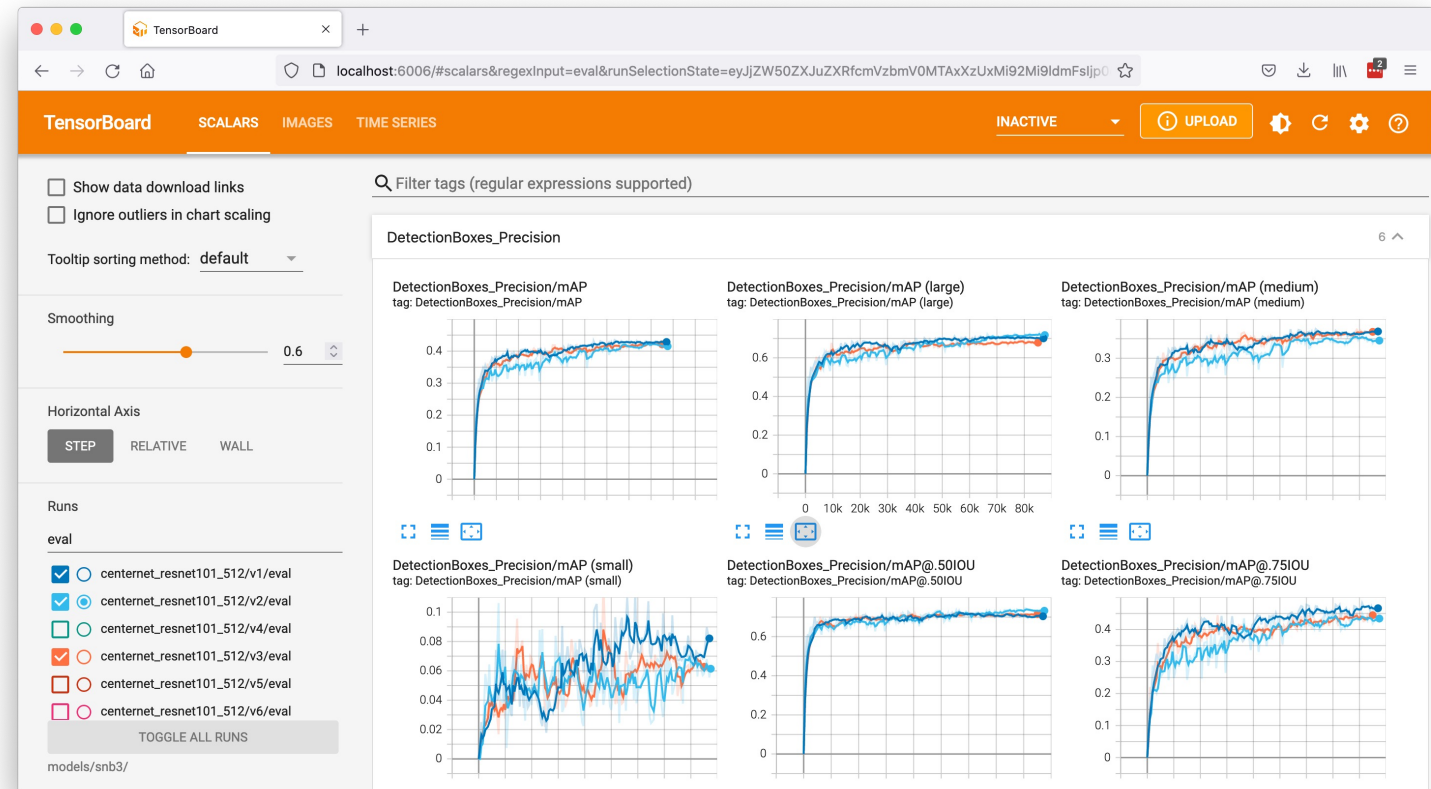
↑
Your
user

↑
Your HPC
cluster

AFTER MODEL TRAINING

REQUEST → START TENSORBOARD → CONNECT → VISIT

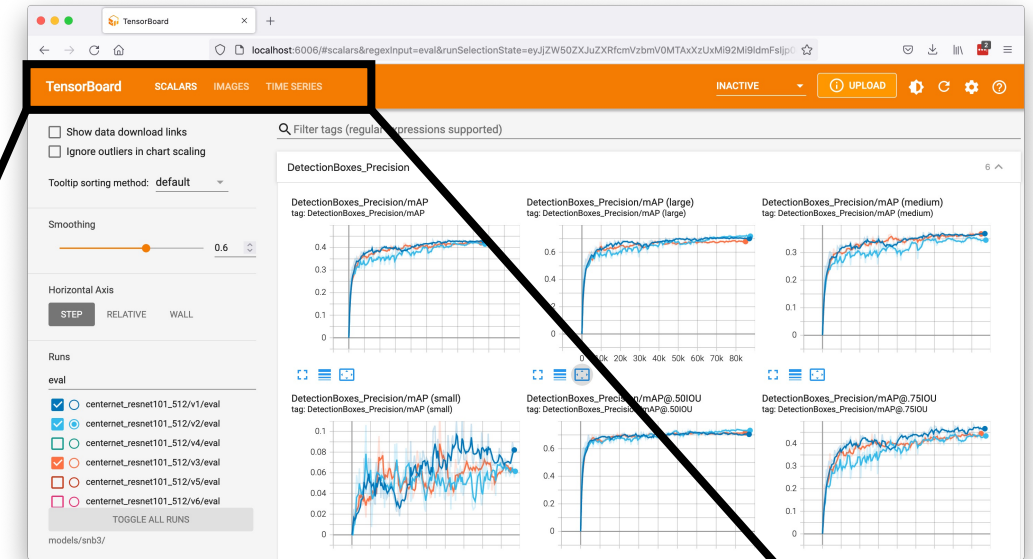
→ Visit localhost:6006
on your favourite
browser



EXPLORING TENSORBOARD

EXPLORING TENSORBOARD

- Scalars
- Images
- Time Series



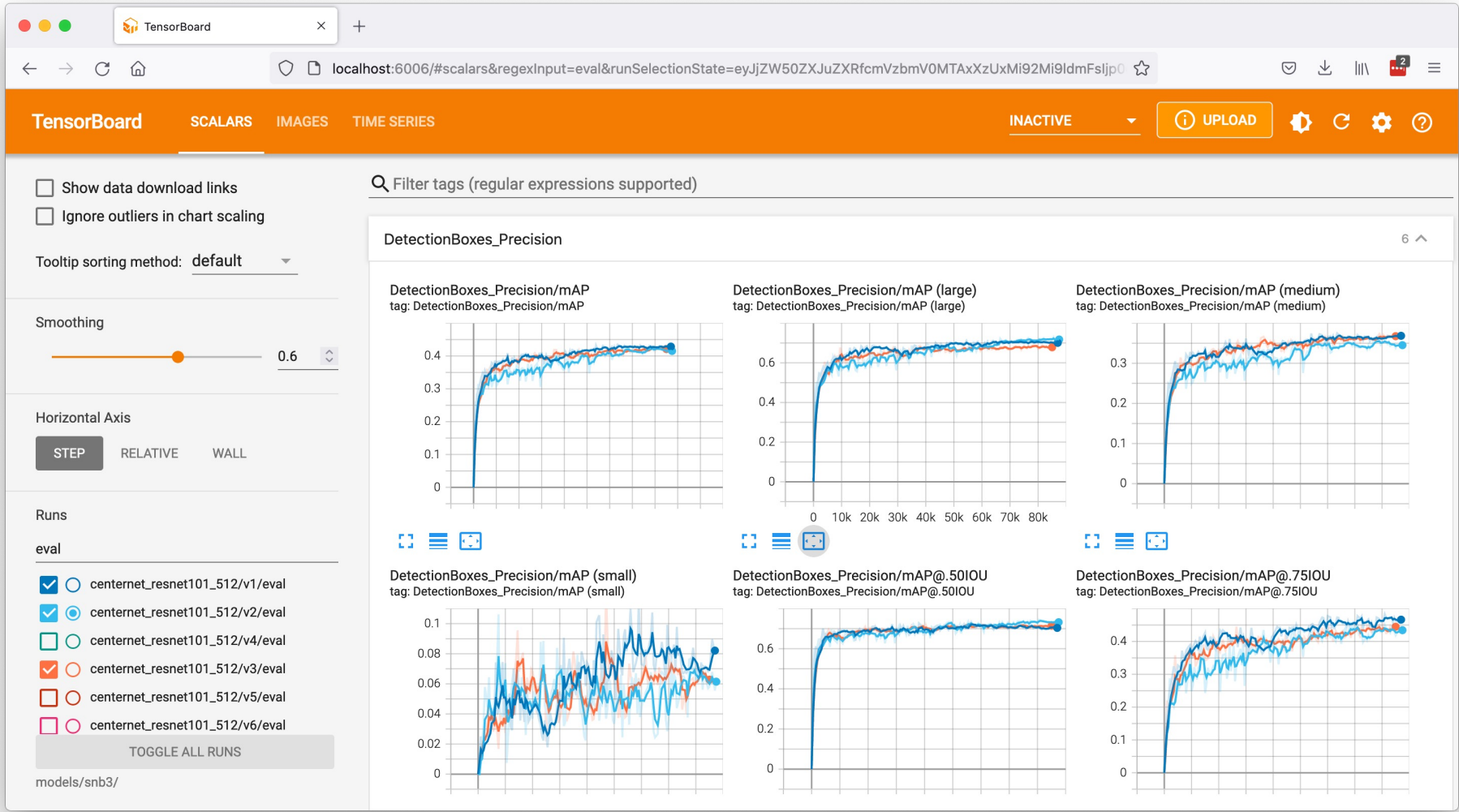
TensorBoard

SCALARS

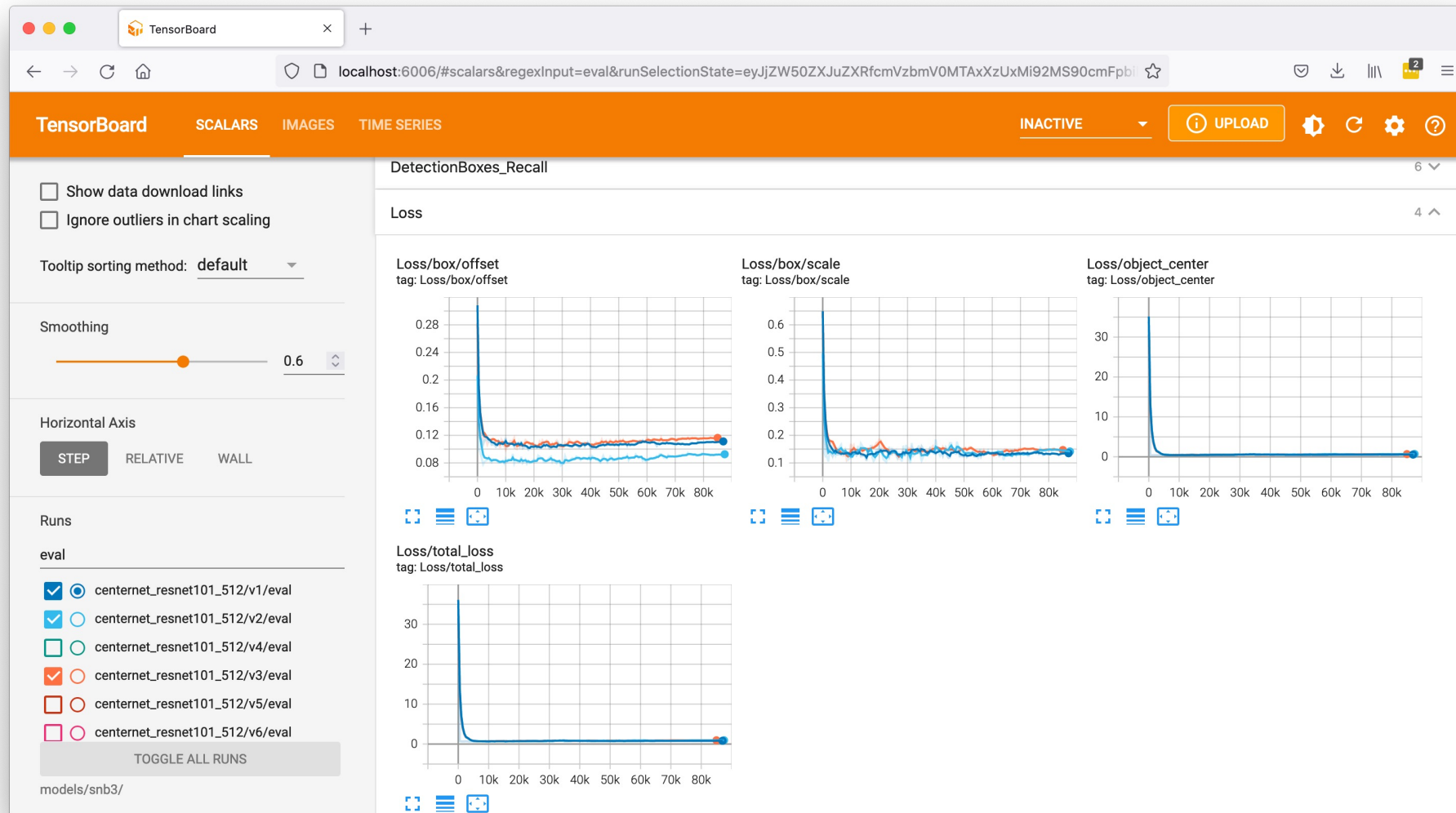
IMAGES

TIME SERIES

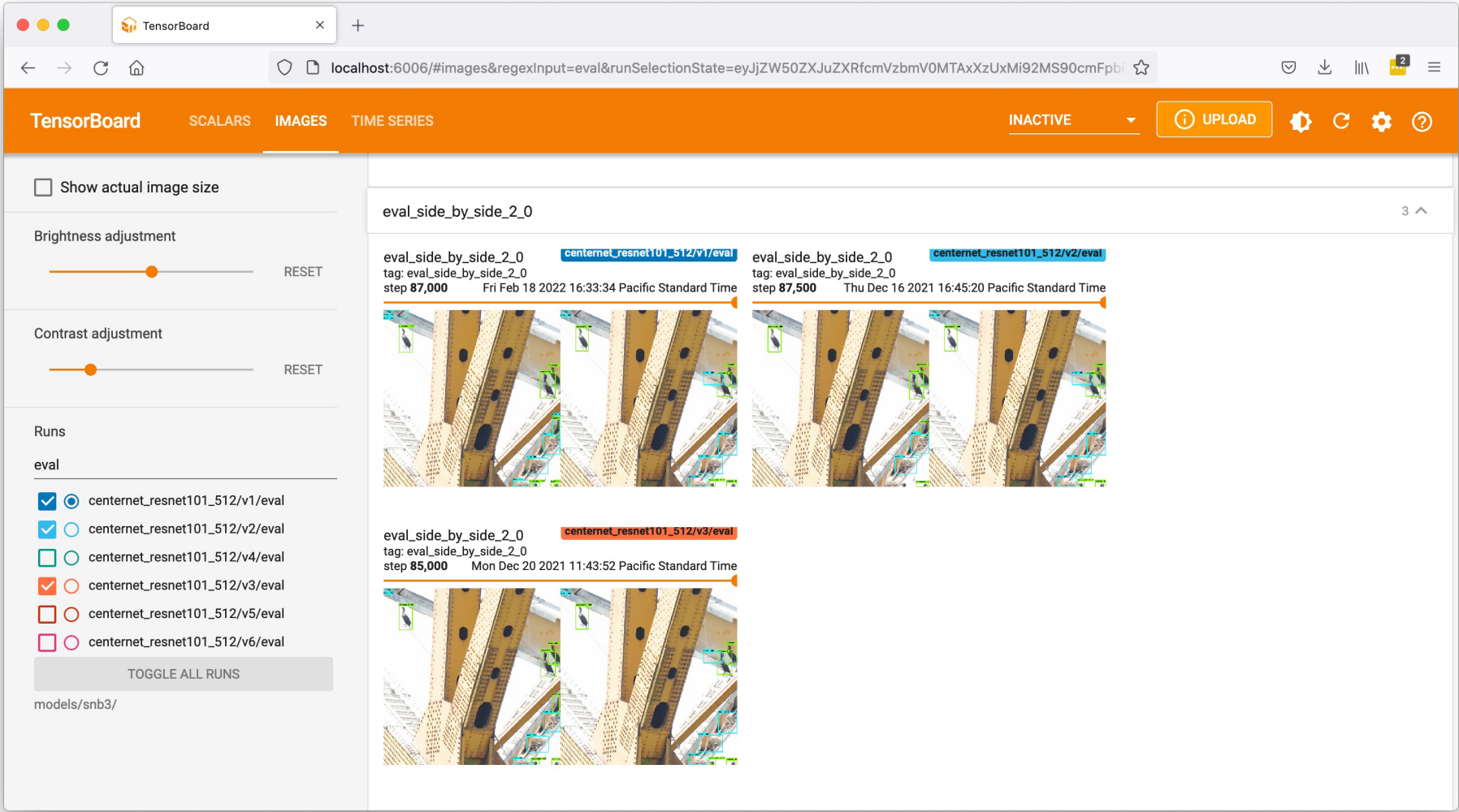
SCALARS



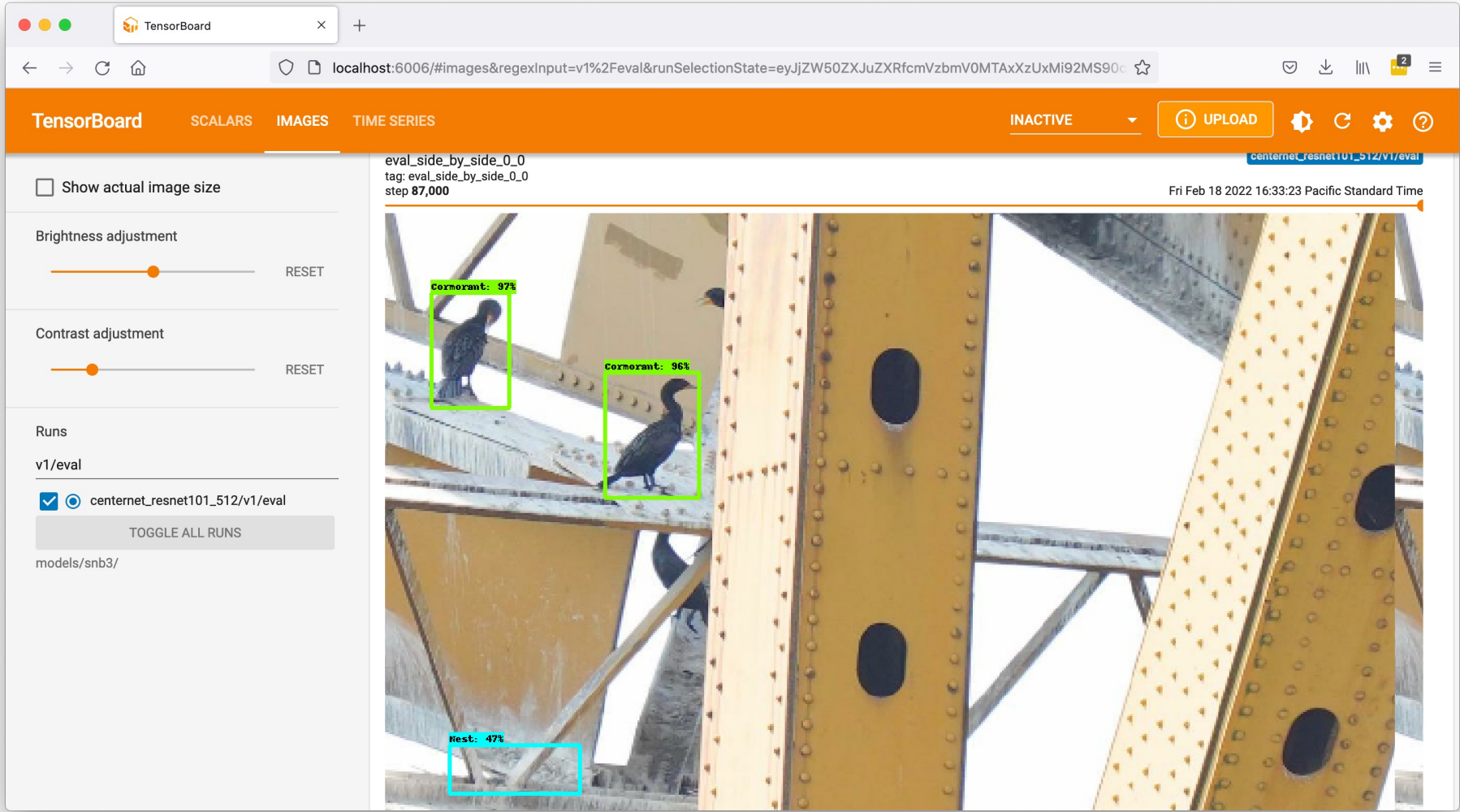
SCALARS



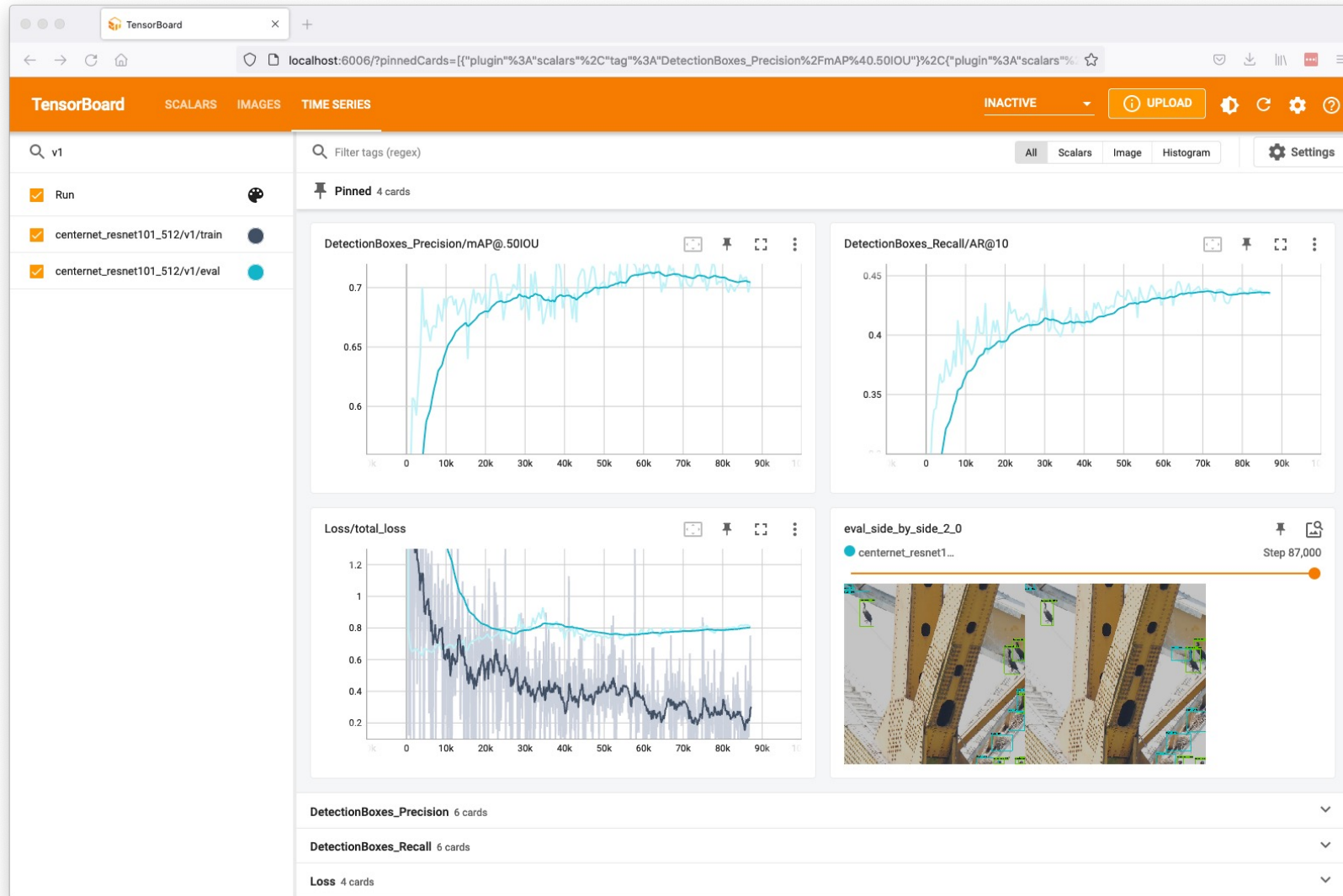
IMAGES



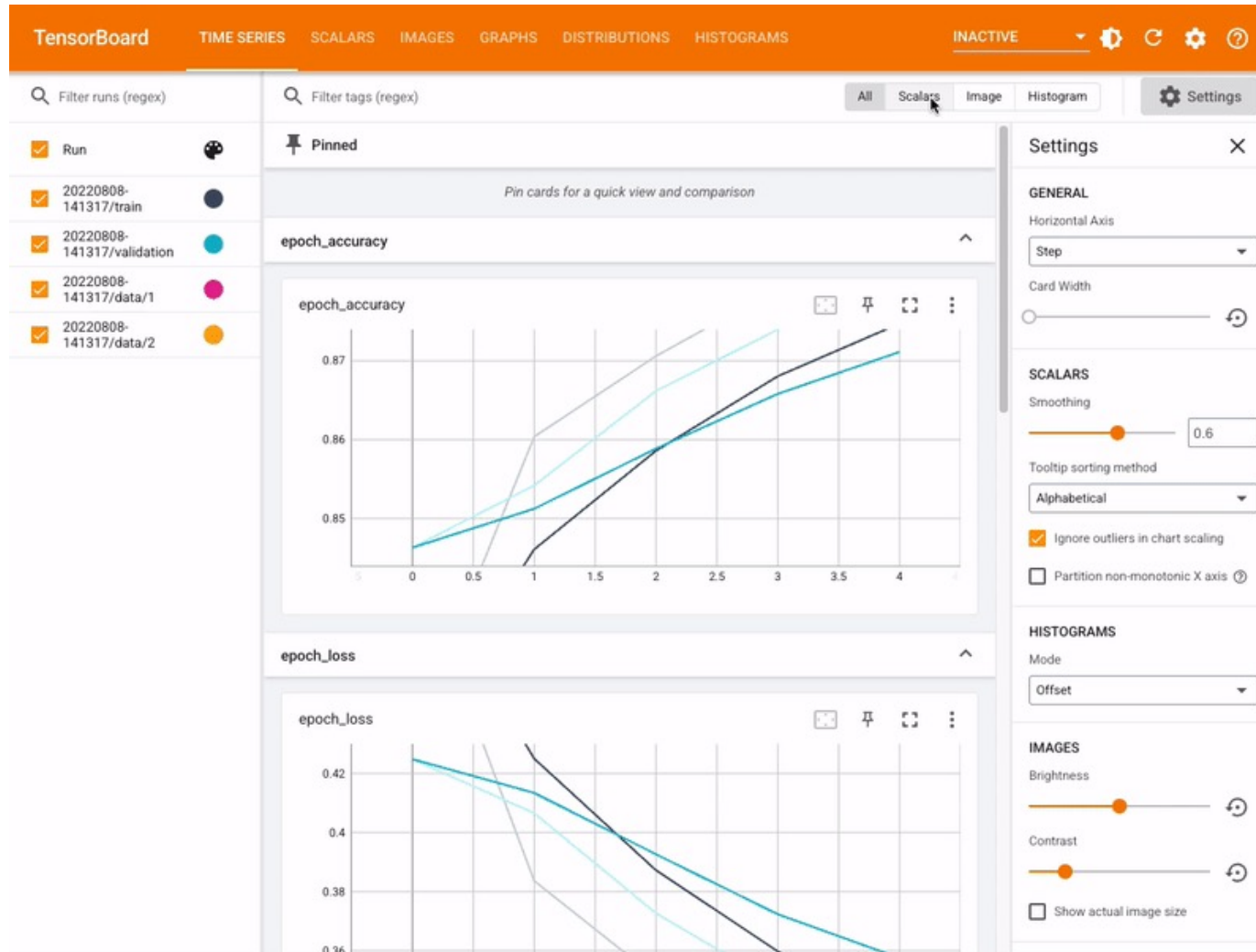
IMAGES



TIME SERIES



OTHER TENSORBOARD DASHBOARDS



THANK YOU

QUESTIONS?