Adding Machine Learning to your developer toolbox

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Google Al Developer Advocate



gusthema 🎔



In this talk

01

Turnkey APIs to help you

All developers can leverage Machine Learning From Mobile, to Web and Backend 02

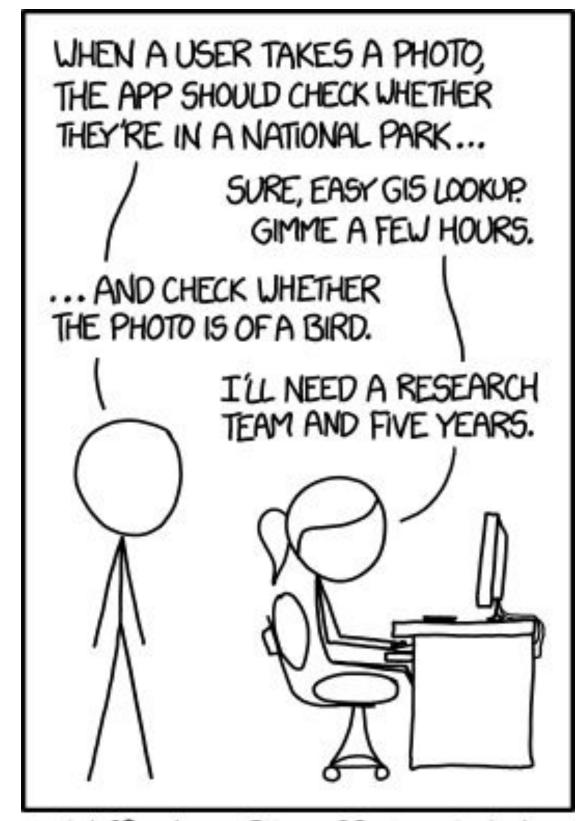
Where to find models

What should you do when you need more than Turnkey APIs?

03

Customizing a model

If no model solves your problem, customizing will help you



IN CS, IT CAN BE HARD TO EXPLAIN THE DIFFERENCE BETWEEN THE EASY AND THE VIRTUALLY IMPOSSIBLE.

xkcd.com/1425/





TensorFlow



Mobile Developers

Android, iOS

Pros

Easy API

Run on-device

Available offline

Vision and Text

Models included



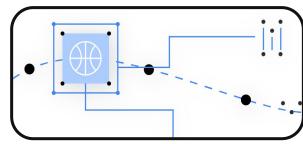
developers.google.com/ml-kit



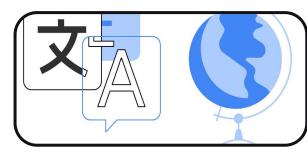




Text recognition



Object detection and tracking



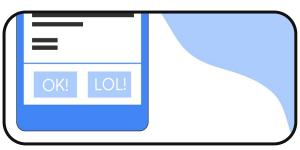
On-device translation



Barcode scanning



lmage labeling



Smart reply



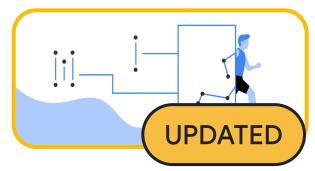
Face detection



Digital ink recognition



Language identification



Pose detection



Selfie segmentation



Entity extraction

```
val image = InputImage.fromBitmap(bitmap, rotation)
val labeler =
ImageLabeling.getClient(ImageLabelerOptions.DEFAULT_OPTIONS)
labeler.process(image)
                .addOnSuccessListener { labels ->
                    for (label in labels) {
                        val text = label.text
                        val confidence = label.confidence
                        val index = label.index
```

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                        val text = label.text
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```



+98% confidence

Bird



www.tensorflow.org/lite

TensorFlow Hub

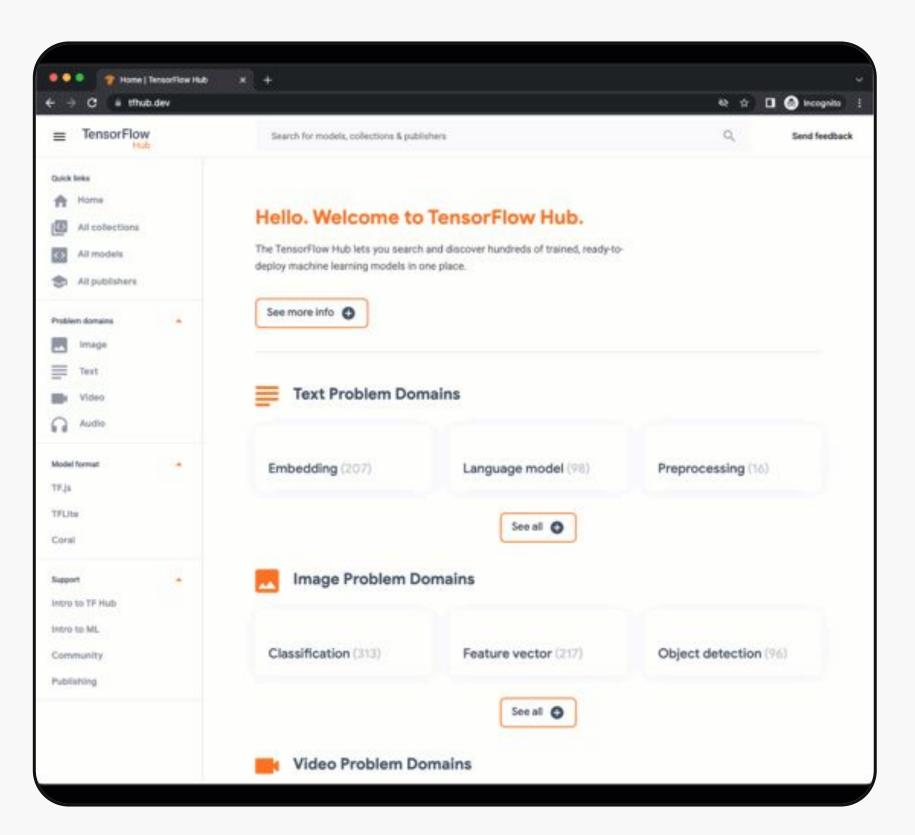
The models repository for all your TensorFlow needs

Models for all kinds of tasks



tfhub.dev

TensorFlow Lite



```
val image = InputImage.fromBitmap(bitmap, rotation)
val localModel = LocalModel.Builder()
        .setAssetFilePath("birds.tflite")
        .build()
val customOptions = CustomImageLabelerOptions
        .Builder(localClassifier)
        .build()
val labeler = ImageLabeling.getClient(customOptions)
labeler.process(image)
                .addOnSuccessListener { labels ->
```

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val image = InputImage.fromBitmap(bitmap, rotation)
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```



+70% confidence

Berylline Hummingbird

The first recorded travels by Europeans to China and back date from this time. The most famous traveler of the period was the Venetian Marco Polo, whose account of his trip to "Cambaluc," the capital of the Great Khan, and of life there astounded the people of Europe. The account of his travels, Il milione (or, The Million, known in English as the Travels of Marco Polo), appeared about the year 1299. Some argue over the accuracy of Marco Polo's accounts due to the lack of mentioning the Great Wall of China, tea houses, which would have been a prominent sight since Europeans had yet to adopt a tea culture, as well the practice of foot binding by the women in capital of the Great Khan. Some suggest that Marco Polo acquired much of his knowledge through contact with Persian traders since many of the places he named were in Persian.

How did some suspect that Polo learned about China instead of by actually visiting it?

Answer: through contact with Persian traders



TensorFlow Task Library

https://goo.gle/36T1PuW

TensorFlow Task Library

Pros

Customizable

Data processing

High Performance

Extensible

Open sourced

https://goo.gle/36T1PuW

TensorFlow Task Library

https://goo.gle/36T1PuW

Tasks

Image Classification

Object Detection

Image Segmentation

Text Classification

Text Question and Answer

Audio Classification

Similarity Search







```
// Initialize the audio classifier
val classifier = AudioClassifier.createFromFile(context, modelPath)
val audioTensor = classifier.createInputTensorAudio()

// Initialize the audio recorder
val record = classifier.createAudioRecord()
record.startRecording()

// Load the latest audio sample
audioTensor.load(record)
val output = classifier.classify(audioTensor)
```

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https://goo.gle/3uv4a7k



Web Developers

Install

API W

Resources *

More *

Q Search

English -

For JavaScript

Overview **Tutorials** Guide Models

Demos

API

Go from zero to hero with web ML in a new online course from TensorFlow.js.

Register now

TensorFlow.js is a library for machine learning in JavaScript

Develop ML models in JavaScript, and use ML directly in the browser or in Node.js.

See tutorials

See models

See demos

Tutorials show you how to use TensorFlow.js with

Pre-trained, out-of-the-box models for common use

Live demos and examples run in your browser using TensorFlow.js.

www.tensorflow.org/js



TensorFlow.js models

Out of the box

Image classification

Pose Detection

Body segmentation

Object Detection

Hand pose detection

Text toxicity

Face landmark detection

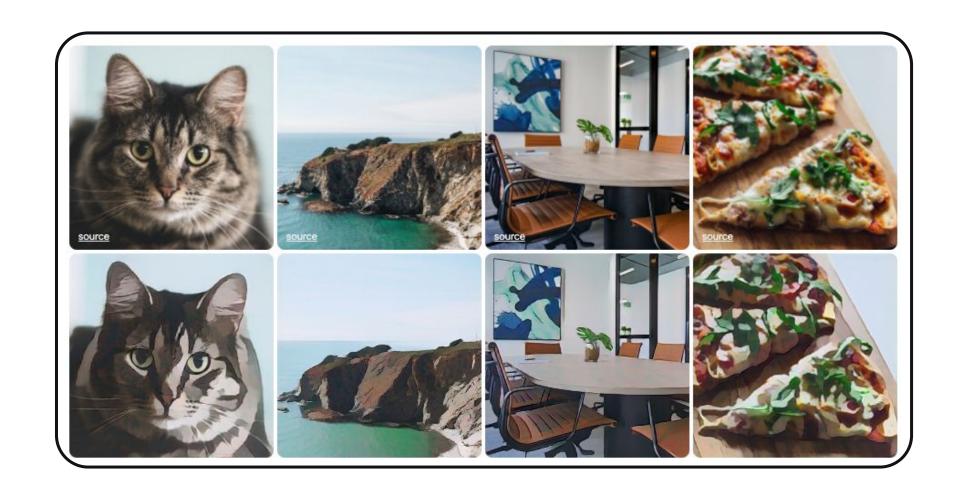
Image segmentation

Text Question and Answering

www.tensorflow.org/js/models

```
<script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs@1.0.1"> </script>
<script src="https://cdn.jsdelivr.net/npm/@tensorflow-models/mobilenet@1.0.0">
</script>
<img id="img" src="bird.jpg" crossorigin="anonymous" />
<script>
  const IMG = document.getElementById('img');
  async function classify() {
      const model = await mobilenet.load();
      // Classify the image.
      const predictions = await model.classify(IMG);
      console.log('Predictions: ');
      console.log(predictions);
 classify();
</script>
```

TensorFlow.js + TFLite models



https://goo.gle/3iGihkH

```
<script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs"></script>
<script
src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs-tflite@0.0.1-alpha.7/dist/tf-tflite.min.js"></sc</pre>
ript>
// Load the model.
let tfliteModel = await tflite.loadTFLiteModel()
  "https://tfhub.dev/google/lite-model/aiy/vision/classifier/birds_V1/3");
// Prepare input.
const EXPECTED_IMAGE_SIZE = 224;
const IMG = document.querySelector("img");
let inputTensor = tf.cast(
 tf.image
    .resizeBilinear(tf.browser.fromPixels(IMG),
                    [EXPECTED_IMAGE_SIZE, EXPECTED_IMAGE_SIZE]) // Resize
    .expandDims(),
  "int32");
// Run the inference and get the output tensor.
let outputTensor = tfliteModel.predict(inputTensor);
console.log(outputTensor.dataSync());
```

```
<script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs"></script>
<script
src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs-tflite@0.0.1-alpha.7/dist/tf-tflite.min.js"></sc</pre>
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```

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Backend Developers

TensorFlow Serving

www.tensorflow.or g/tfx/guide/serving

Can serve multiple models and versions simultaneously

gRPC and REST api

Low latency, high throughput

Supports A/B testing experimental models

Out-of-the-box support for accelerators (GPU)

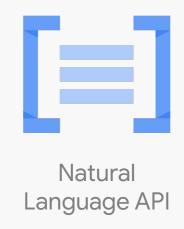
```
# Download the TensorFlow Serving Docker image and repo
docker pull tensorflow/serving
# Start TensorFlow Serving container and open the REST API port
docker run -t --rm -p 8501:8501 \
    -v "/my_models/saved_model/model_half_plus_two:/models/half_plus_two" \
    -e MODEL_NAME=half_plus_two \
    tensorflow/serving &
# Query the model using the predict API
curl -d '{"instances": [1.0, 2.0, 5.0]}' \
    -X POST http://localhost:8501/v1/models/half_plus_two:predict
# Returns => { "predictions": [2.5, 3.0, 4.5] }
```

Google Cloud ML APIs

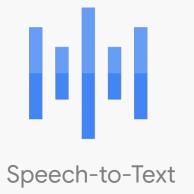
Powerful ML models over REST and RPC

https://goo.gle/3wOen1o





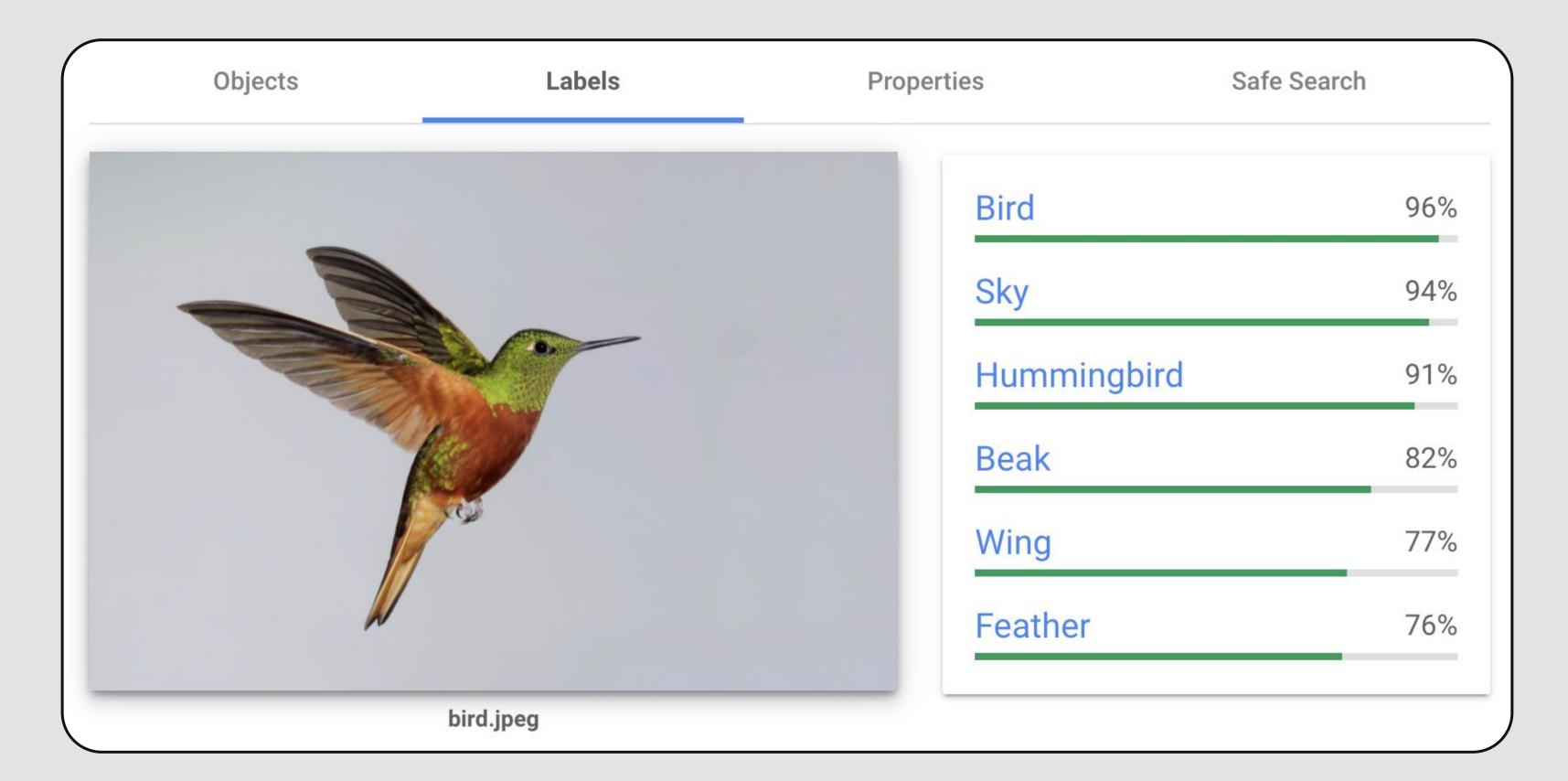












What is next?

Customizing a model

And when there is no model that solve my problem?

What can you do?

How to customize?

When there is no model that solve my problem?



Training Models

TensorFlow Lite Model Maker

Simplify the process of training a model using custom data

Model types include:

Image-based

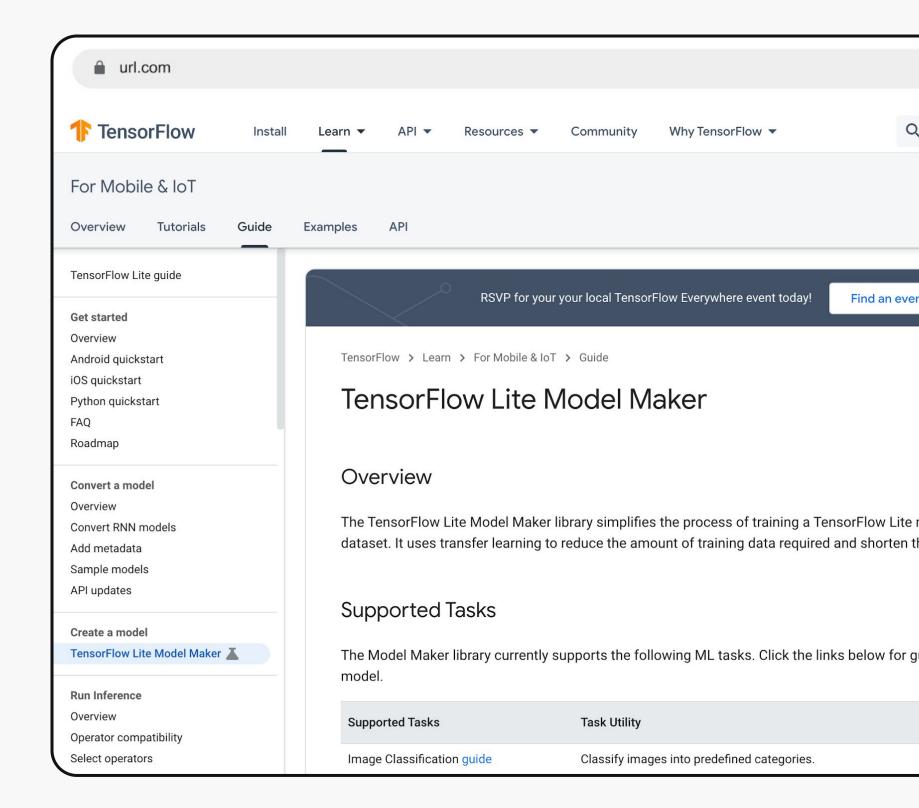
Text-based

Object Detection

Audio

...and more

https://goo.gle/3wPwcNP



```
from tflite_model_maker import image_classifier
from tflite_model_maker.image_classifier import DataLoader
# Load input data specific to an on-device ML app.
data = DataLoader.from_folder('toys_photos/')
train_data, test_data = data.split(0.9)
# Customize the TensorFlow model.
model = image_classifier.create(train_data)
# Evaluate the model.
loss, accuracy = model.evaluate(test_data)
# Export to Tensorflow Lite model and label file in `export_dir`.
model.export(export_dir='/tmp/')
```

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```

Google Cloud AutoML

Train high-quality custom machine learning models with minimal effort and machine learning expertise.













AutoML Video Intelligence



AutoML Natural Language

cloud.google.com/automl

Summary

01

Use Turnkey APIs

Easy-to-use APIs that provide complete ML-powered use cases for your apps

02

Use Existing models

Need a model? Tensorflow Hub is here to help! 03

Build Custom models

If no model solves your problem, you can customize them to your data with Google Cloud AutoML or TensorFlow Lite Model Maker

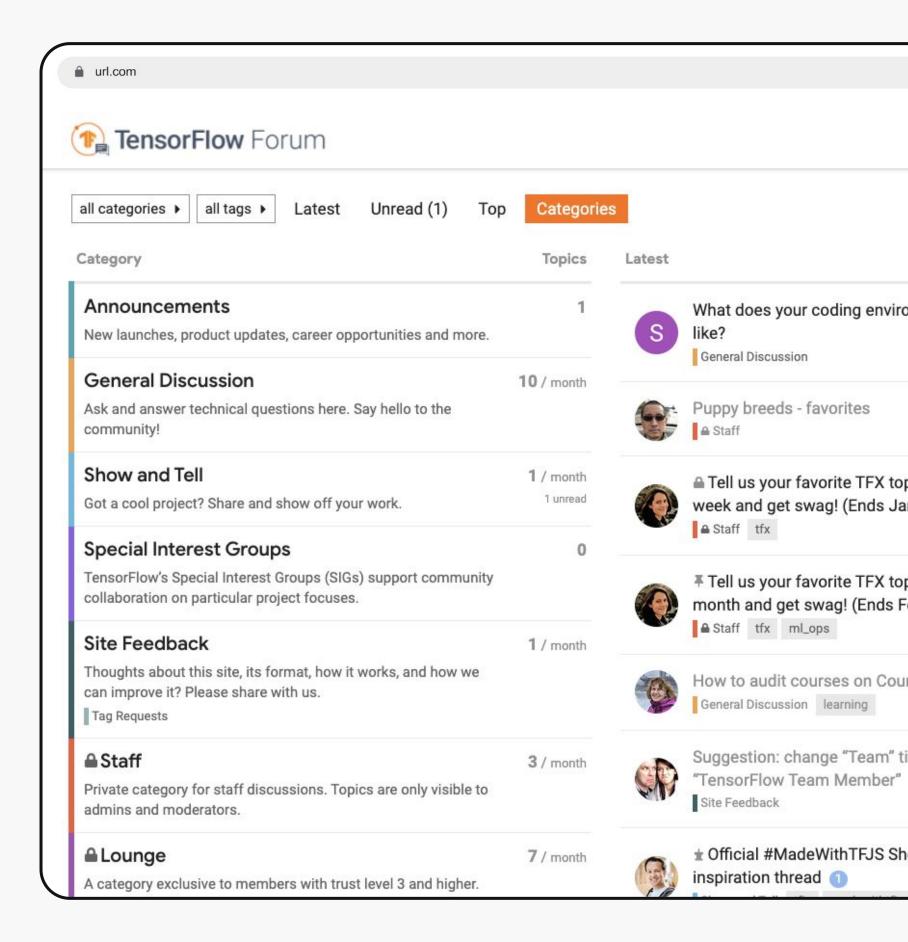
Training Models

Let's continue the conversation in the TensorFlow Forum!

A place for constructive conversation, support, inspiration, and sharing of best practices between the TensorFlow community.

Create your account and join the conversation!

discuss.tensorflow.org



Thank you!



Luiz Gustavo Martins

Google Al Developer Advocate He / him



gusthema 🎔