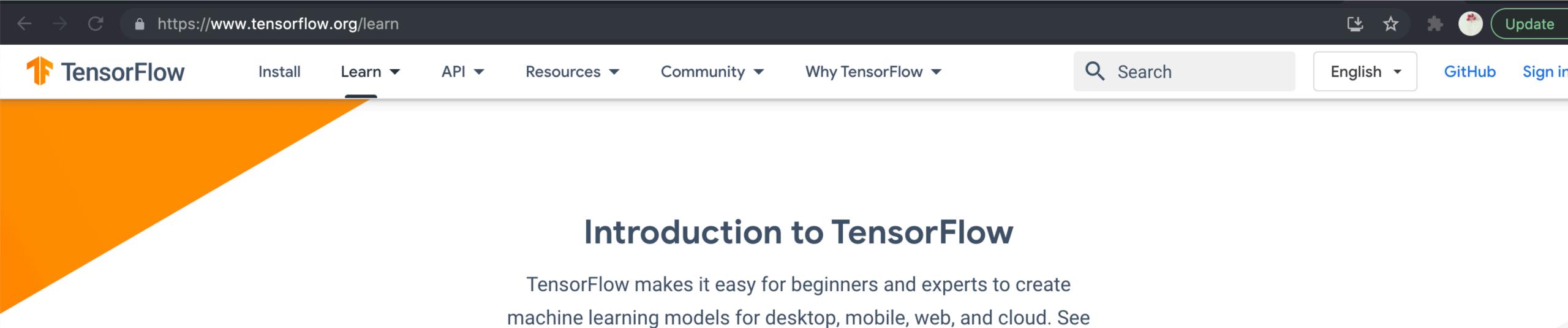
### **TensorFlow Object Detection API**

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This primer will shortly review some notions and resources that you will need to get started with the TensorFlow Object Detection (TFOD) API.



the sections below to get started.



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Deploy a production-ready ML
pipeline for training and inference
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(TFX).

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**README.md** 



### Welcome to the Model Garden for TensorFlow

The TensorFlow Model Garden is a repository with a number of different implementations of state-of-the-art (SOTA) models and modeling solutions for TensorFlow users. We aim to demonstrate the best practices for modeling so that TensorFlow users can take full advantage of TensorFlow for their research and product development.

To improve the transparency and reproducibility of our models, training logs on TensorBoard.dev are also provided for models to the extent possible though not all models are suitable.

Directory	Description		
official	<ul> <li>A collection of example implementations for SOTA models using the latest TensorFlow 2's high-level APIs</li> <li>Officially maintained, supported, and kept up to date with the latest TensorFlow 2 APIs by TensorFlow</li> <li>Reasonably optimized for fast performance while still being easy to read</li> </ul>		
research	<ul> <li>A collection of research model implementations in TensorFlow 1 or 2 by researchers</li> <li>Maintained and supported by researchers</li> </ul>		
community	• A curated list of the GitHub repositories with machine learning models and implementations powered by TensorFlow 2		
orbit	• A flexible and lightweight library that users can easily use or fork when writing customized training loop code in TensorFlow 2.x. It seamlessly integrates with tf.distribute and supports running on different device types (CPU, GPU, and TPU).		

#### Contributors 737

















+ 726 contributors

#### Languages

- Python 90.2%
- Jupyter Notebook 7.0%
- C++ 1.8%

Other 1.0%

### **TensorFlow 2 Detection Model Zoo**



We provide a collection of detection models pre-trained on the COCO 2017 dataset. These models can be useful for out-of-the-box inference if you are interested in categories already in those datasets. You can try it in our inference colab

They are also useful for initializing your models when training on novel datasets. You can try this out on our few-shot training colab.

Please look at this guide for mobile inference.

Finally, if you would like to train these models from scratch, you can find the model configs in this directory (also in the linked tar.gz s).

Model name	Speed (ms)	COCO mAP	Outputs
CenterNet HourGlass104 512x512	70	41.9	Boxes
CenterNet HourGlass104 Keypoints 512x512	76	40.0/61.4	Boxes/Keypoints
CenterNet HourGlass104 1024x1024	197	44.5	Boxes
CenterNet HourGlass104 Keypoints 1024x1024	211	42.8/64.5	Boxes/Keypoints
CenterNet Resnet50 V1 FPN 512x512	27	31.2	Boxes
CenterNet Resnet50 V1 FPN Keypoints 512x512	30	29.3/50.7	Boxes/Keypoints
CenterNet Resnet101 V1 FPN 512x512	34	34.2	Boxes
CenterNet Resnet50 V2 512x512	27	29.5	Boxes
CenterNet Resnet50 V2 Keypoints 512x512	30	27.6/48.2	Boxes/Keypoints
CenterNet MobileNetV2 FPN 512x512	6	23.4	Boxes
CenterNet MobileNetV2 FPN Keypoints 512x512	6	41.7	Keypoints
EfficientDet D0 512x512	39	33.6	Boxes













latest

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**Training Custom Object Detector** 

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**Common issues** 

Docs » TensorFlow 2 Object Detection API tutorial



### **TensorFlow 2 Object Detection API tutorial**

#### Important

This tutorial is intended for TensorFlow 2.5, which (at the time of writing this tutorial) is the latest stable version of TensorFlow 2.x.

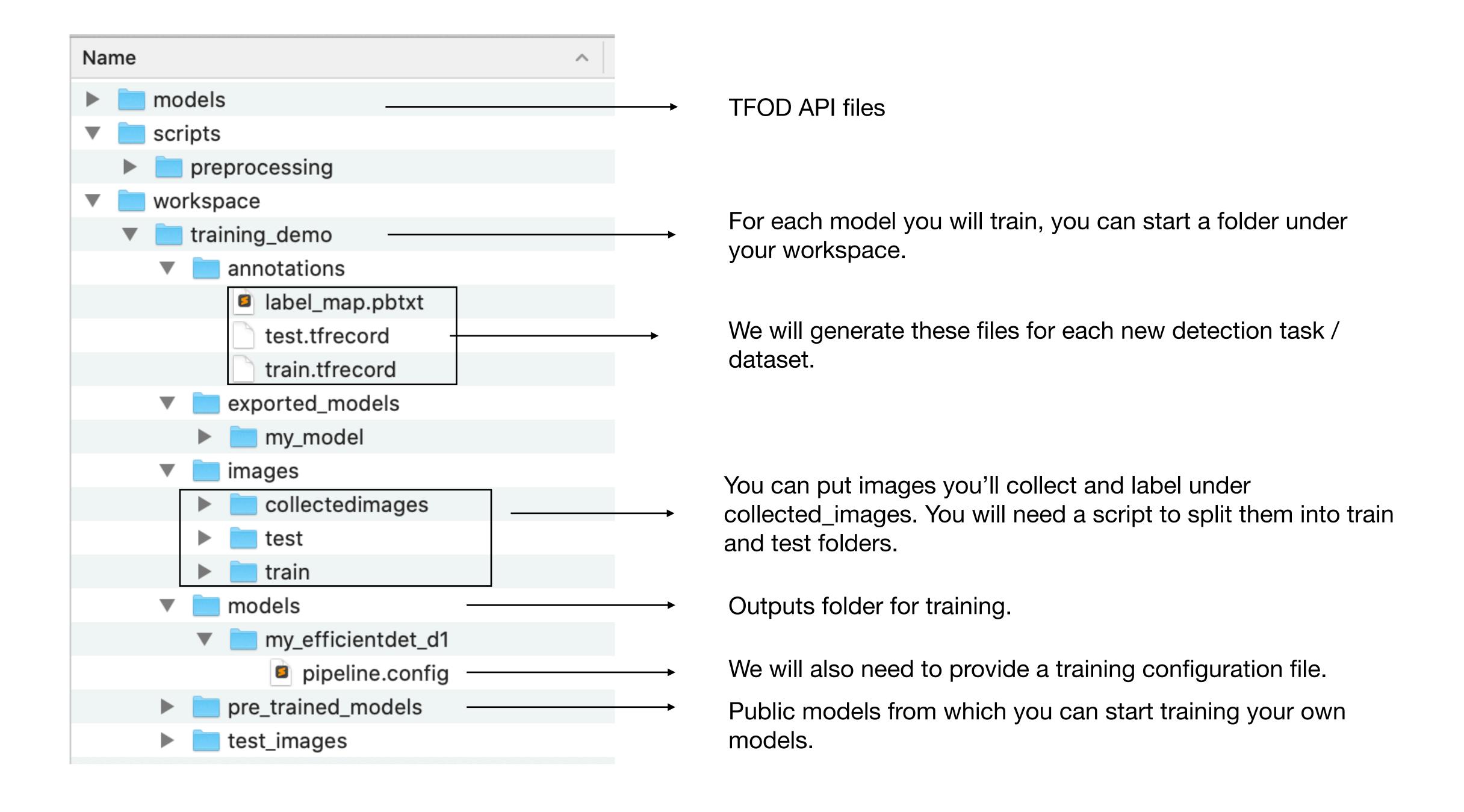
A version for TensorFlow 2.2 can be found here.

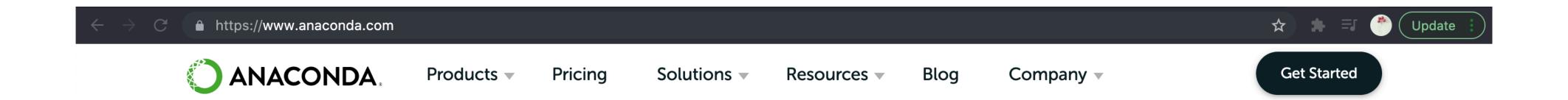
A version for TensorFlow 1.14 can be found here.

This is a step-by-step tutorial/guide to setting up and using TensorFlow's Object Detection API to perform, namely, object detection in images/video.

The software tools which we shall use throughout this tutorial are listed in the table below:

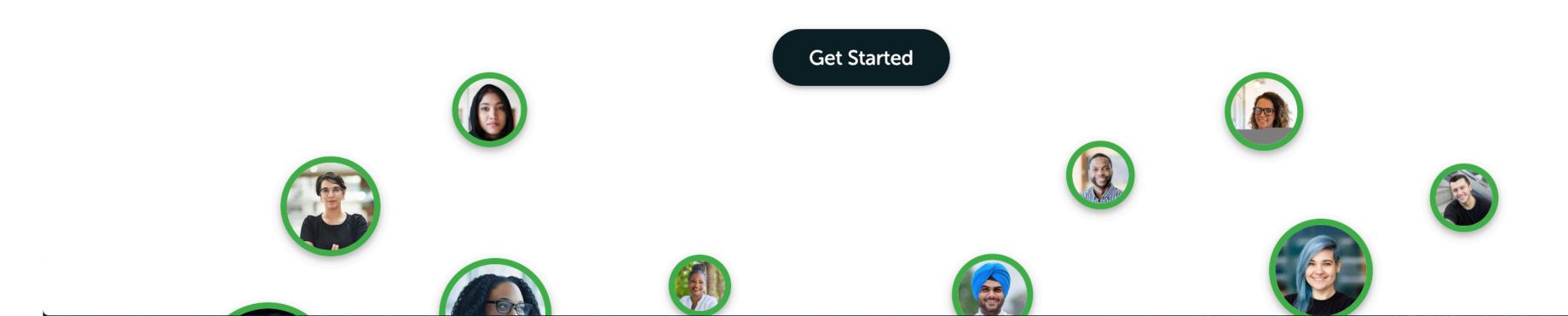
Target Software versions					
OS	Windows, Linux				
Python	3.9 <sup>1</sup>				
TensorFlow	2.5.0				
CUDA Toolkit	11.2				
CuDNN	8.1.0				
Anaconda	Python 3.8 (Optional)				

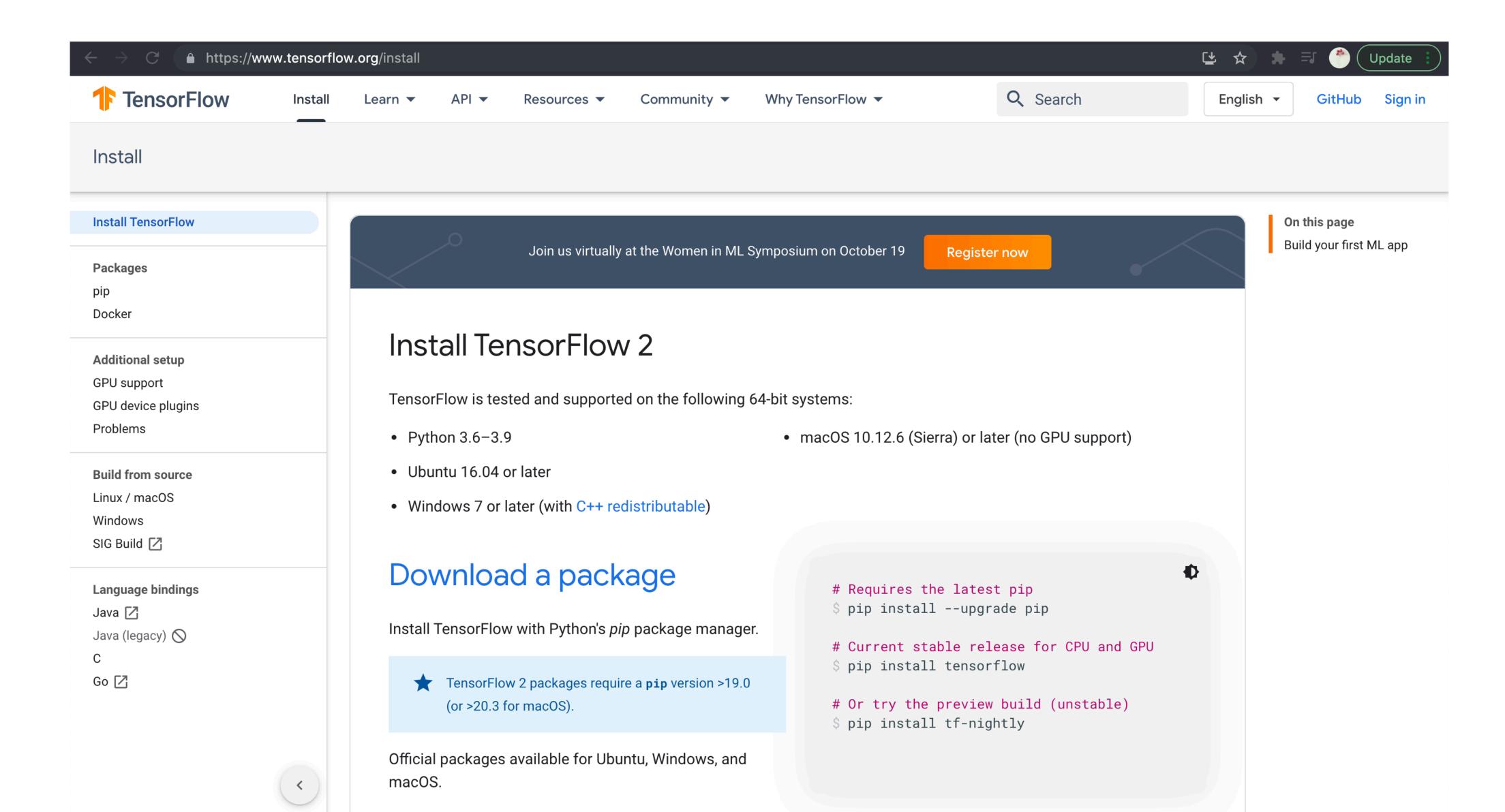


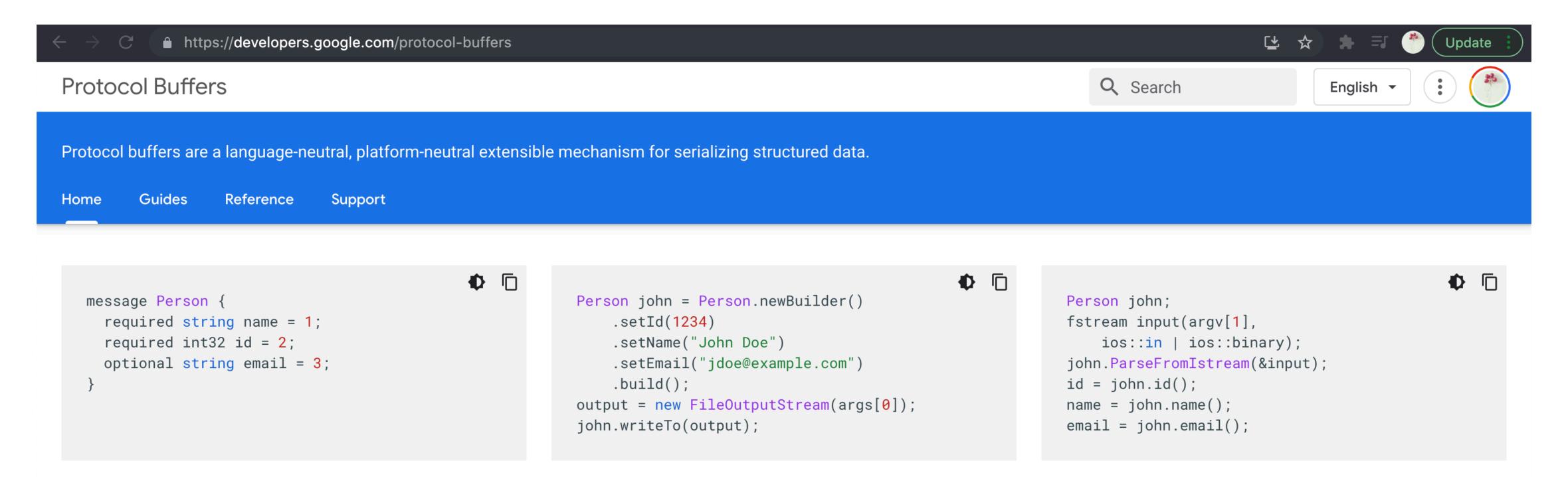


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#### What are protocol buffers?

Learn more

Protocol buffers are Google's language-neutral, platform-neutral, extensible mechanism for serializing structured data – think XML, but smaller, faster, and simpler. You define how you want your data to be structured once, then you can use special generated source code to easily write and read your structured data to and from a variety of data streams and using a variety of languages.

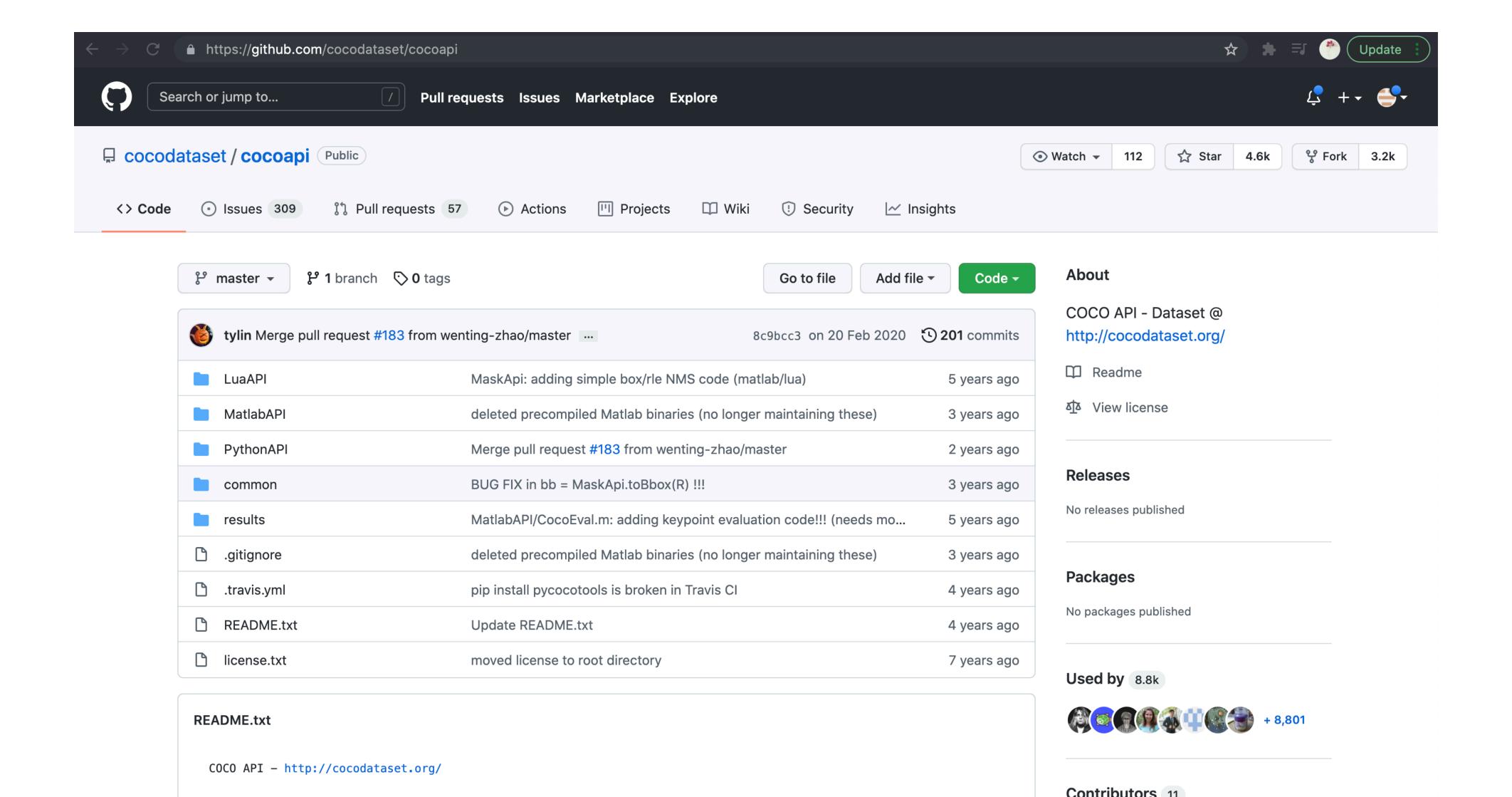
#### Pick your favorite language

Protocol buffers currently support generated code in Java, Python, Objective-C, and C++. With our new proto3 language version, you can also work with Dart, Go, Ruby, and C#, with more languages to come.

#### How do I start?

- 1. Download and install the protocol buffer compiler.
- 2. Read the overview.
- 3. Try the tutorial for your chosen language.

C++ C# Dart Go Java Kotlin Python



# Things you will need to do

- 1. Organise your workspace (folder structure)
- 2. (Create a label map and annotate images)
- 3. Generate (Tensorflow record files) tfrecords
- 4. Provide a configuration file for the training pipeline
- 5. Train your model and check its progress
- 6. Export the resulting model and use it to detect objects