# TensorFlow Lite

The professional course

# What is our goal?

To enable our students to **build mobile products** based on **machine learning at the edge**.



# Why?

- High demand for mobile applications.
- Hardware constraints.
- Privacy concernings.
- Highly profitable opportunities.

### For who is this course?

- Machine learning engineers.
- Mobile software engineers.
- Enthusiasts of **AI** and **mobile applications** in general.

# Who are your instructors?



Michel Meneses
Software Engineer, Machine Learning
M.S. Computer Science, B.S. Computer Engineering (Federal University of Sergipe/Brazil)



Luiz Vitor Reis Software Engineer, Embedded Systems B.S. Mechatronics Engineering (University of Brasília/Brazil)

## What will you learn?

- How to embed machine learning models on mobile devices (FREE).
- How to evaluate the performance of machine learning models on mobile apps.
- How to optimize machine learning models for mobile devices.
- How to develop an optimized machine learning-based mobile app from scratch.

# TensorFlow Lite

Week 1

# Agenda

- 1. Machine learning at the edge
- 2. Introduction to TF Lite
  - a. Goal
  - b. Advantages
  - c. Architecture
- 3. Hands-on project
- 4. Wrap-up

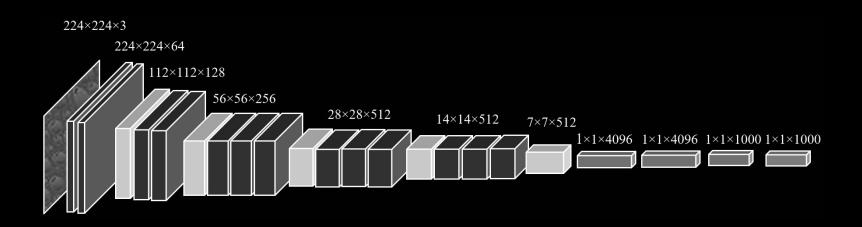
Initially, only simple models were practical (e.g., Viola-Jones face detector).



Since 2012, deep learning has become the state-of-the-art for many ML problems.



However, deep learning models are deep (i.e. large and computationally expensive).



Designing deep learning mobile applications based on the cloud.



Issues of cloud-based architectures: internet need, battery draining and time delay.







Solution: frameworks optimized for running deep learning models at the edge.





TensorFlow Lite was released in 2019 by Google Brain.









#### Advantages:

- **Latency**: there is no round-trip to a server.
- Privacy: no data needs to leave the device.
- **Connectivity**: an Internet connection **is not** required.
- Power consumption: network connections are power-hungry.

TensorFlow Lite is cross-platform.











TensorFlow Lite is cross-platform.











TensorFlow Lite is cross-platform.











#### TF Lite Architecture:



#### Pick a model

Pick a new model or retrain an existing one.



#### Convert

Convert a TensorFlow model into a compressed flat buffer with the TensorFlow Lite Converter.



#### Deploy

Take the compressed .tflite file and load it into a mobile or embedded device.



#### Optimize

Quantize by converting 32-bit floats to more efficient 8-bit integers or run on GPU.

TF Lite Converter (Python):

```
import tensorflow as tf
converter = tf.lite.TFLiteConverter.from_keras_model(model)
tflite_model = converter.convert()
```

TF Lite Interpreter (Android/Java):

```
MappedByteBuffer tfLiteModel = FileUtil.loadMappedFile( context: this, MODEL_FILENAME);
this.interpreter = new Interpreter(tfLiteModel, new Interpreter.Options());
interpreter.run(inputImage.getBuffer(), output.getBuffer());
```

# Hands-on Project

# **Hands-on Project**

Classifying images of dogs (dataset "Stanford Dogs")













# Hands-on Project

#### Steps:

- 1. Build and train a model using TensorFlow
- 2. Convert it to TF Lite
- 3. Embed it in Android
- 4. Run the application

During this week we have learned:

- 1. The importance of machine learning at the edge frameworks
- 2. TensorFlow Lite's architecture
- 3. How to embed a deep learning model in a mobile application using TF Lite

However, there are several open questions:

- 1. How can we assess the performance of our embedded model?
- 2. How can we optimize our embedded model?
- 3. How can we build a commercial mobile product using TensorFlow Lite?

Stay tuned for **Week 2** of *TensorFlow Lite - The professional course!*