



Charles B.

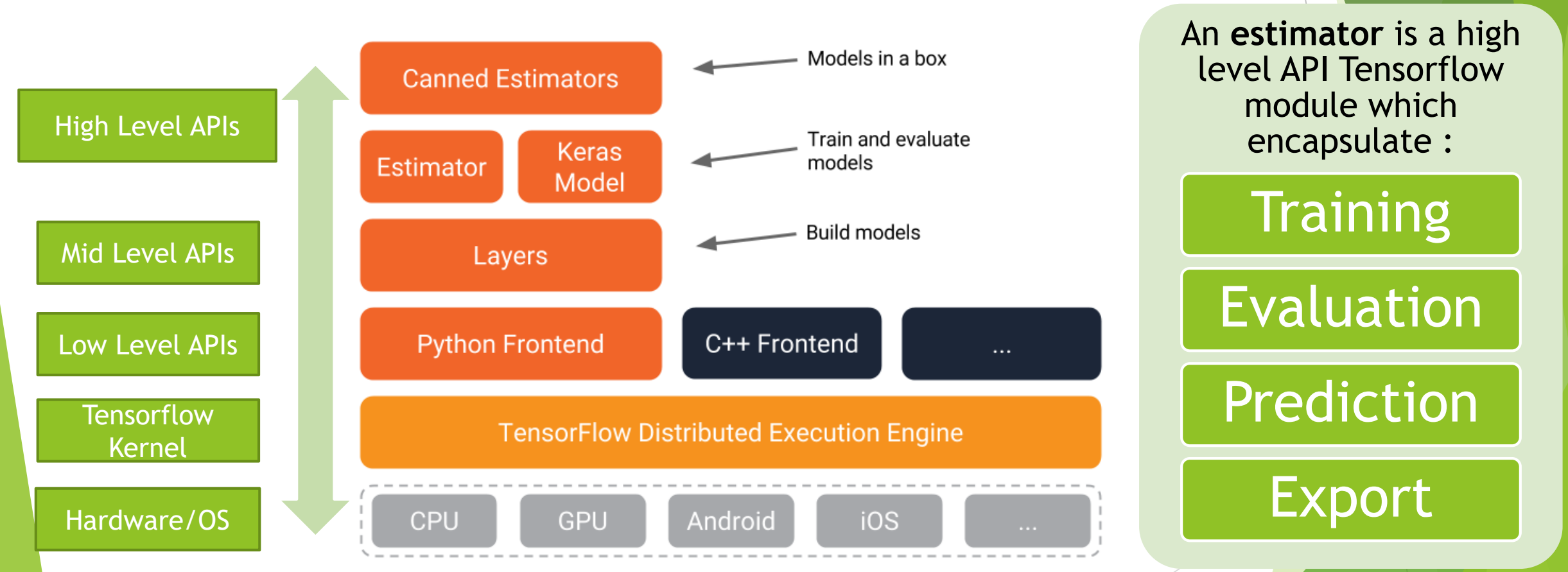
# What is TensorFlow first ?



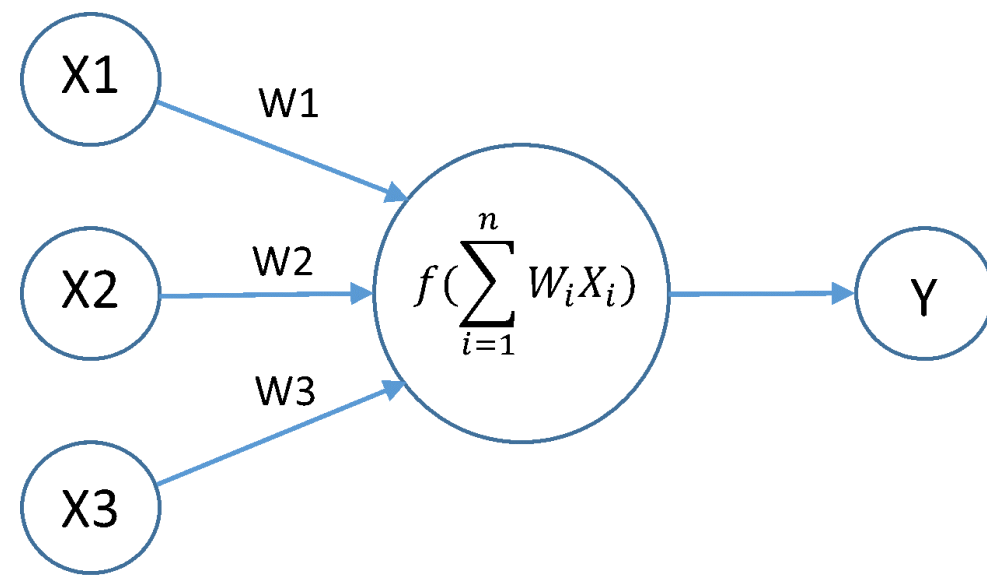
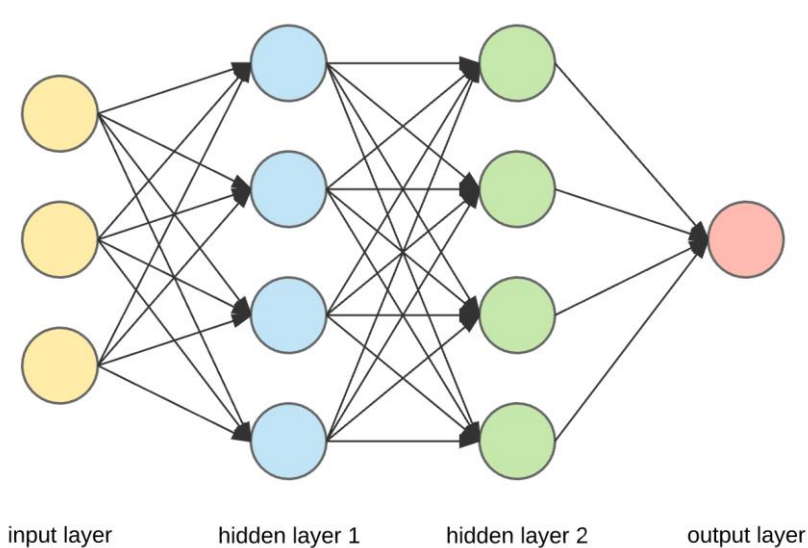
- ▶ Fast, flexible and scalable Google open source library for **Machine Learning** and **Deep Neural Networks** research
- ▶ Performs numerical computations in the form of a **Dataflow graph**. Tensorflow separates definition of computations from their execution.
  - The graph here is a data structure that describes the computation you want to perform
- ▶ **Cross-Platform** : it runs on nearly everything: **GPUs and CPUs**—including mobile and embedded platforms

*Here is a demo if you are intereted in building a DNN fast through an application sandbox : <http://playground.tensorflow.org>*

# Tensorflow overview



# Basic Neural Network Architecture

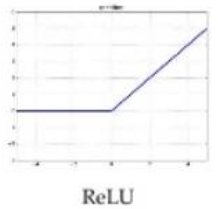
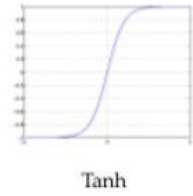
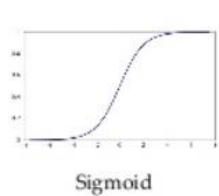


## Non-Linear Activation Function

**f**



- Sigmoid:  $S(t) = \frac{1}{1 + e^{-t}}$
- Tanh:  $\tanh x = \frac{\sinh x}{\cosh x} = \frac{e^x - e^{-x}}{e^x + e^{-x}}$
- Rectified Linear Unit (ReLU):  
 $f(x) = \max(0, x)$



Most popular activation function for DNN as of 2015, avoids saturation issues, makes learning faster

# Data flow

**Phase 1:**  
Create  
Tensors  
(variables)

**Phase 2:**  
Write  
operations  
between  
those Tensors

**Phase 3:**  
Initialize your  
Tensors

**Phase 4:**  
Create a  
Session

**Phase 5:** Run  
the Session on  
the Graph you  
created

# Tensorflow Tensors

- ▶ Computations in Tensorflow are made inside a graph with **Tensors**, the main data structure.
- ▶ A **Tensor** is a generalization of vectors and matrices to potentially higher dimensions. Internally, TensorFlow represents tensors as **n-dimensional arrays** of base datatypes
- ▶ A Tensor has a **data type** (float32, int32 or string) and a **shape** (dimension).

Rank	Math entity
0	Scalar (magnitude only)
1	Vector (magnitude and direction)
2	Matrix (table of numbers)
3	3-Tensor (cube of numbers)
n	n-Tensor (you get the idea)

```
import tensorflow as tf
```

```
floating = tf.Variable(3.14159265359, tf.float64)
```

```
first_primes = tf.Variable([2, 3, 5, 7, 11], tf.int32)
```

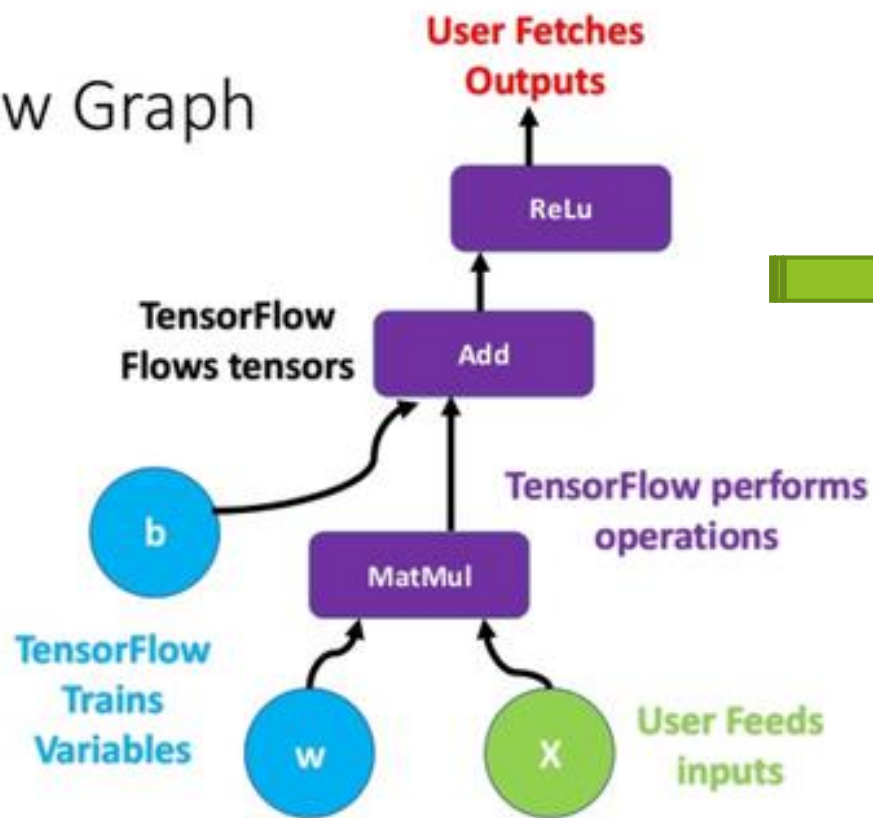
```
squarish_squares = tf.Variable([ [4, 9], [16, 25] ], tf.int32)
```

```
my_image = tf.zeros([10, 299, 299, 3]) # batch x height x width x color
```

# Tensorflow Graph

A **graph** is a series of TensorFlow operations arranged into a graph of nodes. In another words, it means a graph is just an arrangement of nodes that represent the operations in your model.

## TensorFlow Graph



`Z1 = tf.add(tf.matmul(W1,X),b1)`  
`A1 = tf.nn.relu(Z1)`



# Running a Session in the current Graph

```
init = tf.global_variables_initializer()
```

Initializes the variables  
(when run later in the  
session)

```
with tf.Session() as session:
    session.run(init)
    print(session.run(A1))
```

Session() object takes the  
current graph by default,  
tf.get\_default\_graph. But  
we could have precised  
another one with *graph*  
argument

## Working with several graphs

```
g1 = tf.Graph()
g2 = tf.Graph()
with g1.as_default():
    #...add nodes to graph g1
with g2.as_default():
    #...add nodes to graph g2

sess1 = tf.Session(graph = g1)
sess2 = tf.Session(graph = g2)
```



# What is TensorFlow-Hub ?

- ▶ TF-Hub is a library for the **publication**, discovery and **consumption** of state-of-art Machine and Deep Learning Models.
- ▶ It is similar to some Deep Learning frameworks such as Caffe, Keras, etc. but is actually easier for everyone to **publish** and **host models**.
- ▶ It is part of the Tensorflow framework

# Why TensorFlow-Hub ?

We need a **library** for reusable machine learning modules.

A **module** is a self-contained piece of a TensorFlow graph, along with its weights and assets, that can be reused across different tasks in a process known as **Transfer Learning**.

*Weights are the parameters of the pretrained model that we want to use for our own model. They can be either retrained or just fixed.*

It is a very useful technique for people who doesn't have time or skills to **collect a huge amount of data** or doesn't have enough **hardware power** to train their model.

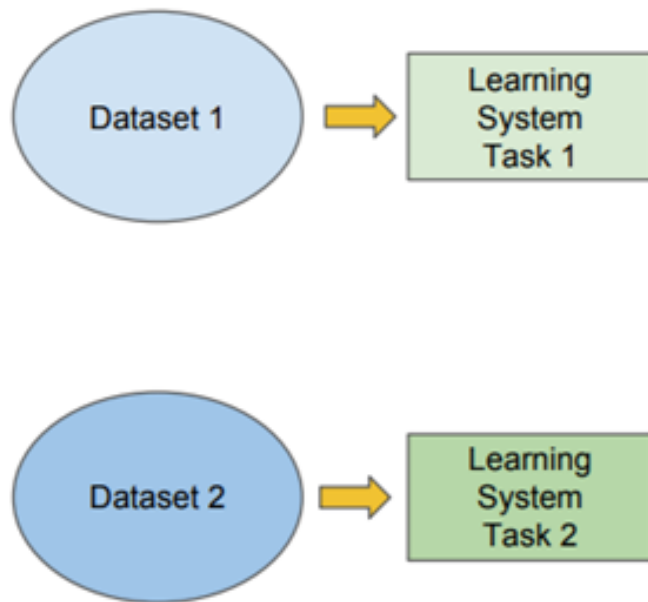
Sharing a **pretrained model** makes it possible for a developer to customize it for their domain.

**« Transfer Learning will be the  
next driver of ML success ! »»**

*Andrew Ng*

# Traditional ML

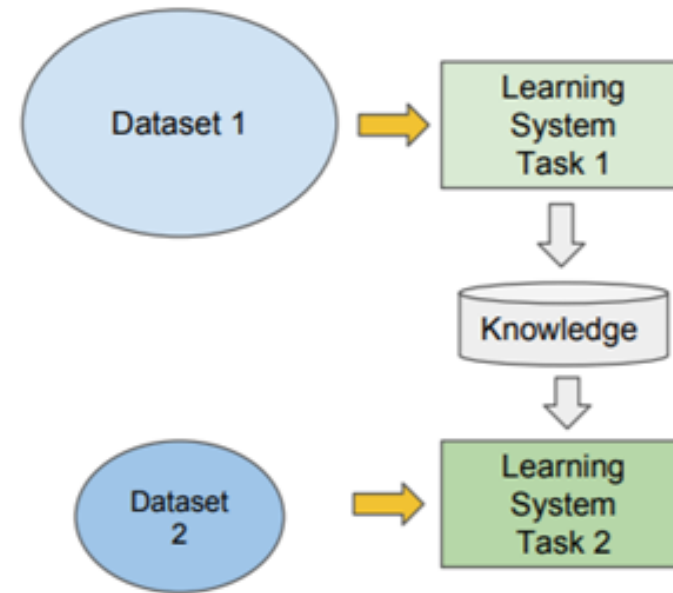
- Isolated, single task learning:
  - Knowledge is not retained or accumulated. Learning is performed w.o. considering past learned knowledge in other tasks



vs

# Transfer Learning

- Learning of a new tasks relies on the previous learned tasks:
  - Learning process can be faster, more accurate and/or need less training data



# Transfer Learning in a nutshell



Train a model with a smaller dataset

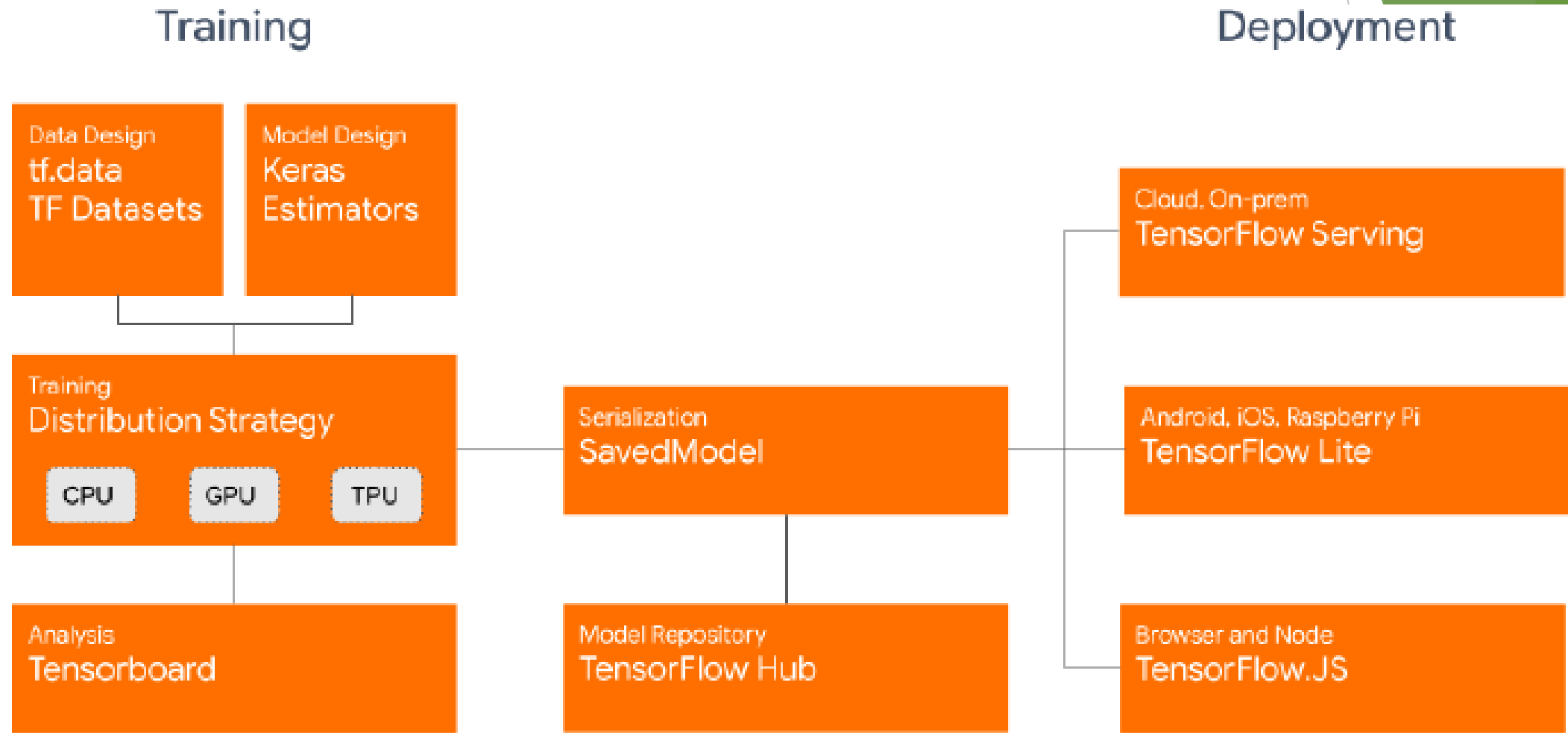


Improve generalization



Speed up model training

# TF pipeline



# Tensorflow-hub Modules



## Image

Here you'll find models that enable:

- Image augmentation
- Image classification
- Feature Vector Extraction

And more.



## Text

Here you'll find models that enable:

- Text embedding



## Video

Here you'll find models that enable:

- Video Classification

# Quick demonstration

Use of a pre-trained word embedding on random sentences.

*Word Embedding is a technique which transforms a word into a vector of real numbers, making possible to cluster vectors which are close to each other (words having a similar meaning/sentiment). It is also helpful to reduce the dimensions of words representation, making training easier through a Machine/Deep Learning model.*

```
!pip install "tensorflow_hub>=0.5.0"
!pip install "tensorflow>=2.0.0"

import tensorflow as tf
import tensorflow_hub as hub

module_url = "https://tfhub.dev/google/tf2-preview/nnlm-en-dim128/1"
embed = hub.KerasLayer(module_url)
embeddings = embed(["A long sentence.", "single-word",
                    "http://example.com"])
print(embeddings.shape)  #(3,128)
```



# Let's inspect this module in details

← tf2-preview/nnlm-en-dim128

Problem domain

Embedding

Publisher

Google

Architecture

Nnln

Data set

News

## TF2 SavedModel

This is a [SavedModel in TensorFlow 2 format](#). Using it requires TensorFlow 2 (or 1.15) and TensorFlow Hub 0.5.0 or newer.

### Overview

This module is in the **SavedModel 2.0** format and was created to help preview TF2.0 functionalities. It is based on <https://tfhub.dev/google/nnlm-en-dim128/1>.

Text embedding based on feed-forward Neural-Net Language Models[1] with pre-built OOV. Maps from text to 128-dimensional embedding vectors.

### Details

Based on NNLM with three hidden layers.

#### Input

The module takes a **batch of sentences in a 1-D tensor of strings** as input.

#### Preprocessing

The module preprocesses its input by **splitting on spaces**.

#### Out of vocabulary tokens

Small fraction of the least frequent tokens and embeddings (~2.5%) are **replaced by hash buckets**. Each hash bucket is initialized using the remaining embedding vectors that hash to the same bucket.

#### Sentence embeddings

Word embeddings are combined into sentence embedding using the `sqrtn` combiner (see [tf.nn.embedding\\_lookup\\_sparse](#)).

# How to use modules in Tensorflow-Hub ?

- ▶ Go to Tensorflow-Hub website (<https://tfhub.dev>) and explore all available modules according to your needs.
- ▶ When you have chosen a module, use the corresponding url in your code or directly use the assets (model's weights). Look at the documentation and also examples to understand how to use it.

**Saved model**

**Want to  
use this  
model?**

Choose a button  
on the right to get  
started.

**Copy URL to clipboard**

**Download Assets**

**Open Colab Notebook**

# Example: Build a sentiment classifier

- ▶ We will use a **TF-Hub text embedding module** to train a simple **sentiment classifier** with a good accuracy. We will then analyze the predictions to make sure our model is reasonable.
- ▶ We will try to solve the *Large Movie Review Dataset v1.0* task from Mass et al. The dataset consists of IMDB (Internet Movie Database) movie reviews labeled by positivity from 1 to 10. The task is to label the reviews as **negative** or **positive**.

# Data preprocessing

```
import tensorflow as tf
import tensorflow_hub as hub
import matplotlib.pyplot as plt
import numpy as np
import os
import pandas as pd
import re
import seaborn as sns
```

```
train_df, test_df = download_and_load_datasets()
train_df.head()
```



	sentence	sentiment	polarity
0	After what was considered to be the official D...	10	1
1	This movie was like "The Disney Channel after ...	1	0
2	Although time has revealed how some of the eff...	8	1
3	A lot has been said about Shinjuku Triad Socie...	10	1
4	In September 2003 36-year-old Jonny Kennedy di...	8	1

0	After what was considered to be the official D...	10	1
1	This movie was like "The Disney Channel after ...	1	0
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# Load all files from a directory in a DataFrame.

```
def load_directory_data(directory):
    data = {}
    data["sentence"] = []
    data["sentiment"] = []
    for file_path in os.listdir(directory):
        with tf.gfile.GFile(os.path.join(directory, file_path), "r") as f:
            data["sentence"].append(f.read())
            data["sentiment"].append(re.match("\d+_(\d+)\.txt", file_path).group(1))
    return pd.DataFrame.from_dict(data)
```

# Merge positive and negative examples, add a polarity column and shuffle.

```
def load_dataset(directory):
    pos_df = load_directory_data(os.path.join(directory, "pos"))
    neg_df = load_directory_data(os.path.join(directory, "neg"))
    pos_df["polarity"] = 1
    neg_df["polarity"] = 0
    return pd.concat([pos_df, neg_df]).sample(frac=1).reset_index(drop=True)
```

# Download and process the dataset files.

```
def download_and_load_datasets(force_download=False):
    dataset = tf.keras.utils.get_file(
        fname="aclImdb.tar.gz",
        origin="http://ai.stanford.edu/~amaas/data/sentiment/aclImdb_v1.tar.gz",
        extract=True)
```

```
train_df = load_dataset(os.path.join(os.path.dirname(dataset),
                                     "aclImdb", "train"))
```

```
test_df = load_dataset(os.path.join(os.path.dirname(dataset),
                                    "aclImdb", "test"))
```

```
return train_df, test_df
```

```
# Training input on the whole training set with no limit on training epochs.
train_input_fn = tf.estimator.inputs.pandas_input_fn(
    train_df, train_df["polarity"], num_epochs=None, shuffle=True)

# Prediction on the whole training set.
predict_train_input_fn = tf.estimator.inputs.pandas_input_fn(
    train_df, train_df["polarity"], shuffle=False)
# Prediction on the test set.
predict_test_input_fn = tf.estimator.inputs.pandas_input_fn(
    test_df, test_df["polarity"], shuffle=False)
```

Input functions

```
embedded_text_feature_column = hub.text_embedding_column(
    key="sentence",
    module_spec="https://tfhub.dev/google/nnlm-en-dim128/1")
```

Text embedding  
module

Build our model

```
estimator = tf.estimator.DNNClassifier(
    hidden_units=[500, 100],
    feature_columns=[embedded_text_feature_column],
    n_classes=2,
    optimizer=tf.train.AdagradOptimizer(learning_rate=0.003))
```

Estimator (first layer: 500  
units, second layer: 100  
units)

```
estimator.train(input_fn=train_input_fn, steps=1000)
```

Model's training

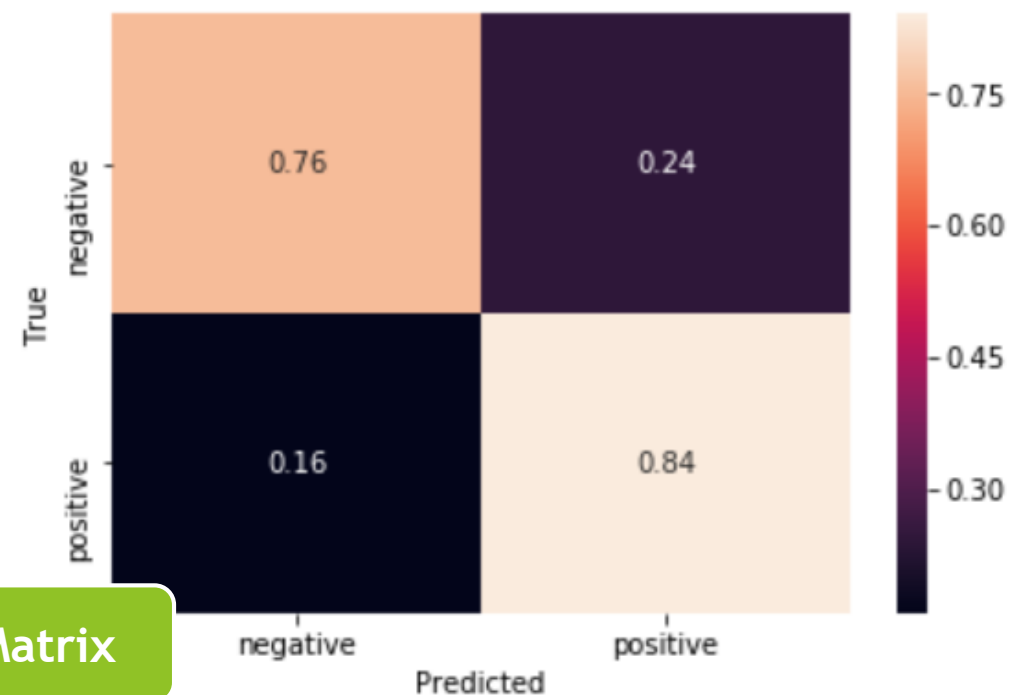
```
train_eval_result = estimator.evaluate(input_fn=predict_train_input_fn)
test_eval_result = estimator.evaluate(input_fn=predict_test_input_fn)

print("Training set accuracy: {accuracy}".format(**train_eval_result))
print("Test set accuracy: {accuracy}".format(**test_eval_result))
```

Predictions on  
train/test sets

```
Training set accuracy: 0.7990000247955322
Test set accuracy: 0.7924399971961975
```

Accuracies



Confusion Matrix

# Additional Ressources

<https://opensource.com/article/17/11/intro-tensorflow>

<https://medium.com/analytics-vidhya/reusing-a-pre-trained-deep-learning-model-on-a-new-task-transfer-learning-1c0a25a92dfb>

<https://medium.com/tensorflow/introducing-tensorflow-hub-a-library-for-reusable-machine-learning-modules-in-tensorflow-cdee41fa18f9>

[https://www.tensorflow.org/hub/tutorials/text\\_classification\\_with\\_tf\\_hub](https://www.tensorflow.org/hub/tutorials/text_classification_with_tf_hub)