

We'll be starting shortly!

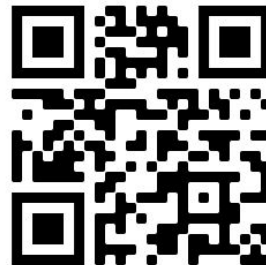
To help us run the workshop smoothly, please kindly:

- Submit all questions using the Q&A function
- If you have an urgent request, please use the "Raise Hand" function

Thank you!

**Exclusive Trainings
by Zenika and Trent!**

<https://bit.ly/CodeMasterClass>





Introduction to TensorFlow

Harry POMMIER
Data Science Consultant @Zenika France, Nantes



Trent Global College



- ▶ Imda Funded Tech Bootcamp

<http://bit.ly/codingdip>

Supported By:



Our Graduates



Sam Tan
Data Analyst
SISTIC Singapore



Lee Tong Lin
Junior Software Developer
Love, Bonito



Jimmy Hsu
IT Engineer
Singtel



Lynette Lim
Front-End Developer
Sparkline Pte Ltd



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- ▶ TECHNOLOGICAL AND ORGANIZATIONAL INNOVATION. BUILDING TOGETHER A SIMPLER, MORE OPEN, FREER AND ACCESSIBLE WORLD
- ▶ Zenika is specialized in software architecture and Agile methods with a threefold-expertise in consulting, realization and training.

www.zenika.sg

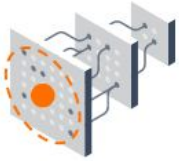
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What is TensorFlow?



- Open source platform for machine learning (ML)
- Flexible ecosystem of tools, libraries and resources
- Build and deployment of ML powered applications
- Develop by Google Brain



Easy model building



Robust ML production anywhere



Powerful experimentation for
research



<https://www.tensorflow.org>

Some TensorFlow Use Cases

- Computer Vision
 - **Image Classification**
 - Object Detection
 - Activity Recognition
 - Image Reconstruction, Colorization
- Natural Language Processing (NLP)
 - Text Classification
 - Machine Translation
 - Sentiment Analysis
 - Speech Recognition

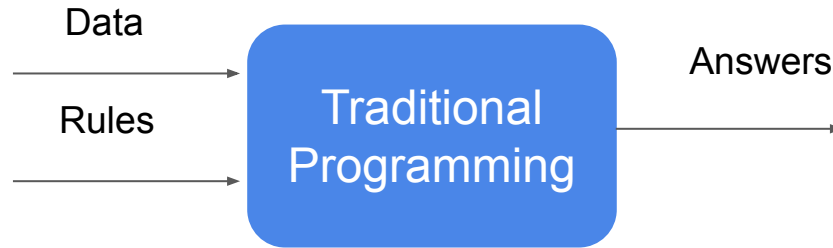


Prerequisite Concepts

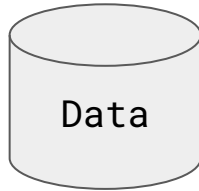
1. Machine Learning: a New Paradigm
2. Tensors
3. Computational Graphs



Prerequisite 1/3: Machine Learning Paradigm



Traditional Programming



Rule

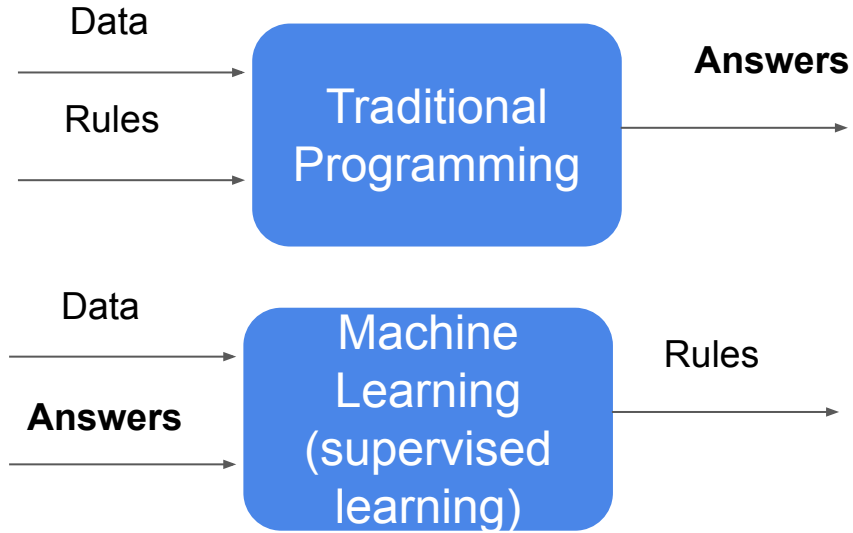
`x = [-5., 10.]`

```
def my_mean(x):  
    return sum(x)/len(x)
```

Answer

`2.5`

A New Paradigm



Why we need ML ?

You want to develop an **activity recognition** app for watersports (e.g. for a connected watch).
With **traditional programming**:



```
if depth == 0:  
    activity =  
    "swimming"  
else:  
    activity =  
    "unknown"
```



```
if depth == 0:  
    if speed > 4:  
        activity =  
        "rowing"  
    else:  
        activity =  
        "swimming"  
else:  
    activity =  
    "unknown"
```



```
if depth == 0:  
    if speed > 4:  
        activity =  
        "rowing"  
    else:  
        activity =  
        "swimming"  
elif depth < 0:  
    activity =  
    "diving"  
else:  
    activity =  
    "unknown"
```



Surfing ?



Water polo ?



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You want to develop an **activity recognition** app for watersports (e.g. for a connected watch).
With **Machine Learning**:



```
[[ 7.1, 1.4],  
 [-7.3, -1.8],  
 [-7. , -6.9],  
 [ 6.7, -3.2],  
 [ 0.3, 6.5]]
```

```
label =  
"swimming"
```



```
[[ 6.8, -2.6],  
 [-8.1, -3.5],  
 [-5.2, -1. ],  
 [-1.6, -5.4],  
 [-8.9, -9. ]]
```

```
label =  
"rowing"
```



```
[[ -1.5, 2.7],  
 [ 7.1, -10. ],  
 [-2.2, 1.4],  
 [ 2.2, 9.8],  
 [-2.6, 5.8]]
```

```
label =  
"diving"
```



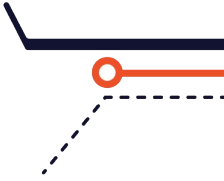
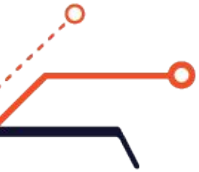
```
[[ -1.2, -4.],  
 [ 9.9, 3.1],  
 [-0.8, 0.1],  
 [-0.8, -6.6],  
 [-6.3, 1.1]]
```

```
label =  
"water polo"
```



```
[[ -1.5, -8.7],  
 [-3.8, -9.9],  
 [-9.9, -5.8],  
 [ 8.6, 7.6],  
 [-7. , -0.3]]
```

```
label =  
"surfing"
```



You want to develop an **activity recognition** app for watersports (e.g. for a connected watch).
With **Machine Learning**:



[[7.1, 1.4],
[-7.3, -1.8],
[-7. , -6.9],
[6.7, -3.2],
[0.3, 6.5]]

label =
"swimming"



[[6.8, -2.6],
[-8.1, -3.5],
[-5.2, -1.],
[-1.6, -5.4],
[-8.9, -9.]]

label =
"rowing"



[[-1.5, 2.7],
[7.1, -10.],
[-2.2, 1.4],
[2.2, 9.8],
[-2.6, 5.8]]

label =
"diving"



[[-1.2, -4.],
[9.9, 3.1],
[-0.8, 0.1],
[-0.8, -6.6],
[-6.3, 1.1]]

label =
"water polo"



[[-1.5, -8.7],
[-3.8, -9.9],
[-9.9, -5.8],
[8.6, 7.6],
[-7. , -0.3]]

label =
"surfing"

Dataset
Features + Labels
(Data + Answers)

a **feature** is an individual
measurable property or
characteristic
e.g. speed, depth, GPS
position, time, pixel value

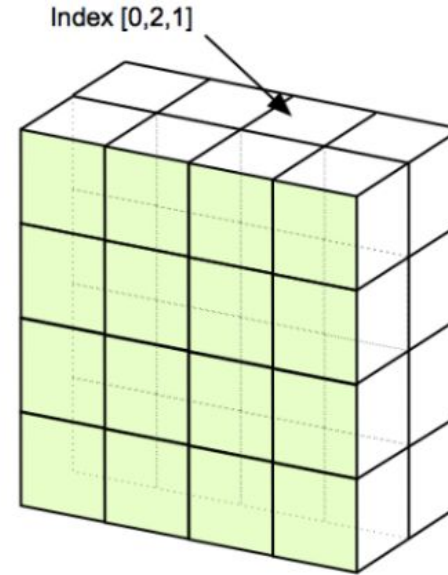
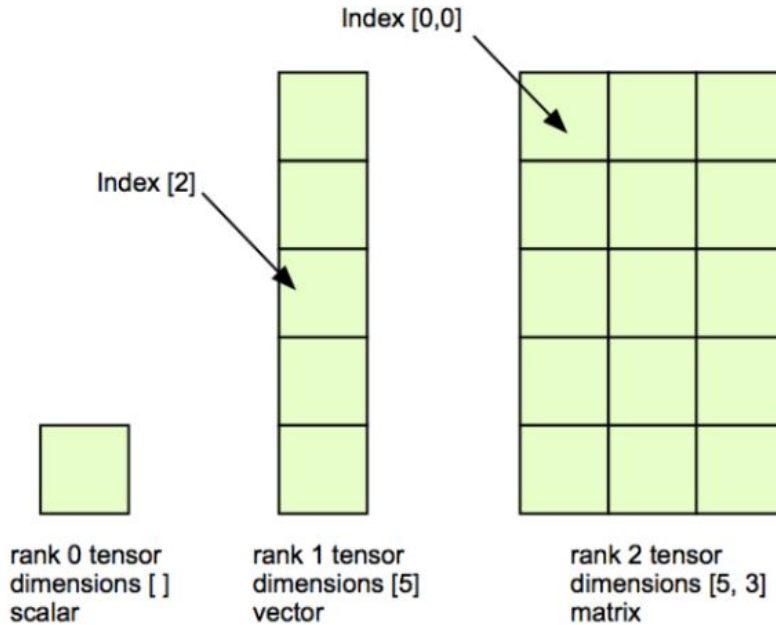
Stored in
Tensors

Prerequisite 2/3: Tensors

- Tensors are multidimensional arrays with a uniform type (float, string, boolean...)
- Can be seen as a generalization of scalars, vectors and matrices in N-dimensions, the dimension of a tensor is called the *rank*
 - scalar: rank = 0
 - vector: rank = 1
 - matrix: rank = 2
 - tensor: rank > 3
- Useful to represent data



Some Tensors



rank 3 tensor
dimensions [4, 4, 2]

rank 4
rank 5
...

shape

Some Tensors

e.g. Grayscale image



48 x 28 pixels

```
array([[255, 255, 128, ..., 128, 255, 255],
       [255, 255, 13, ..., 64, 255, 255],
       [255, 255, 13, ..., 64, 255, 255],
       ...,
       [255, 255, 255, ..., 255, 255, 255],
       [255, 255, 255, ..., 255, 255, 255],
       [255, 255, 255, ..., 255, 255, 255]], dtype=uint8)
```

rank 2 tensor, shape (48, 28)

Some Tensors

e.g. RGB image



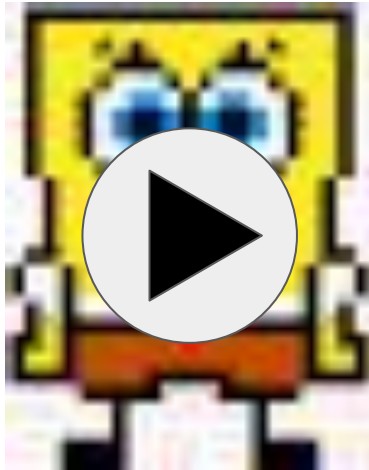
48 x 28 pixels

```
B array([[236, 236, 128, ..., 128, 236, 236],
        [255, 255, 10, ..., 64, 255, 236],
        ...,
        [255, 255, 11, ..., 64, 255, 236],
        ...,
        [255, 255, 128, ..., 128, 255, 255],
        [255, 255, 13, ..., 64, 255, 255],
        [255, 255, 13, ..., 64, 255, 255],
        ...,
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255]], dtype=uint8)
G array([[236, 236, 128, ..., 128, 236, 236],
        [255, 255, 10, ..., 64, 255, 236],
        ...,
        [255, 255, 11, ..., 64, 255, 236],
        ...,
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255]], dtype=uint8)
R array([[255, 255, 128, ..., 128, 255, 255],
        [255, 255, 13, ..., 64, 255, 255],
        [255, 255, 13, ..., 64, 255, 255],
        ...,
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255],
        [255, 255, 255, ..., 255, 255, 255]], dtype=uint8)
```

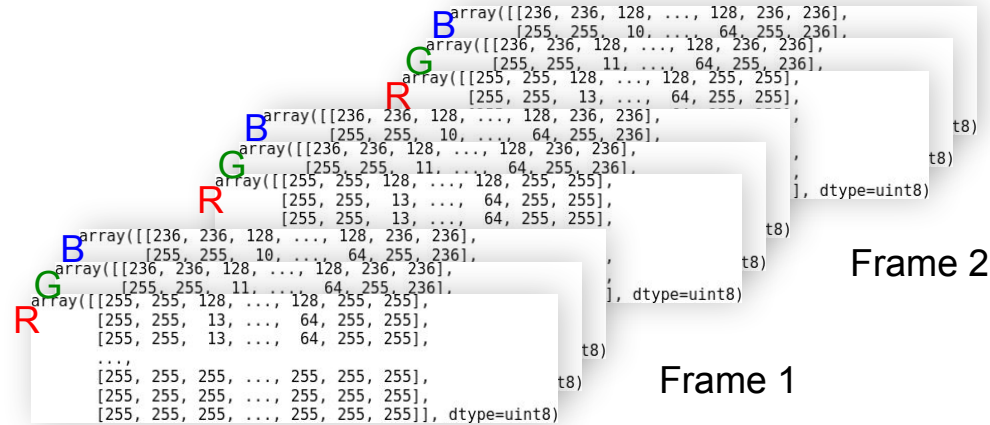
rank 3 tensor, shape (48, 28, 3)

Some Tensors

e.g. RGB video



48 x 28 pixels, 200 frames



rank 4 tensor, shape (200, 48, 28, 3)



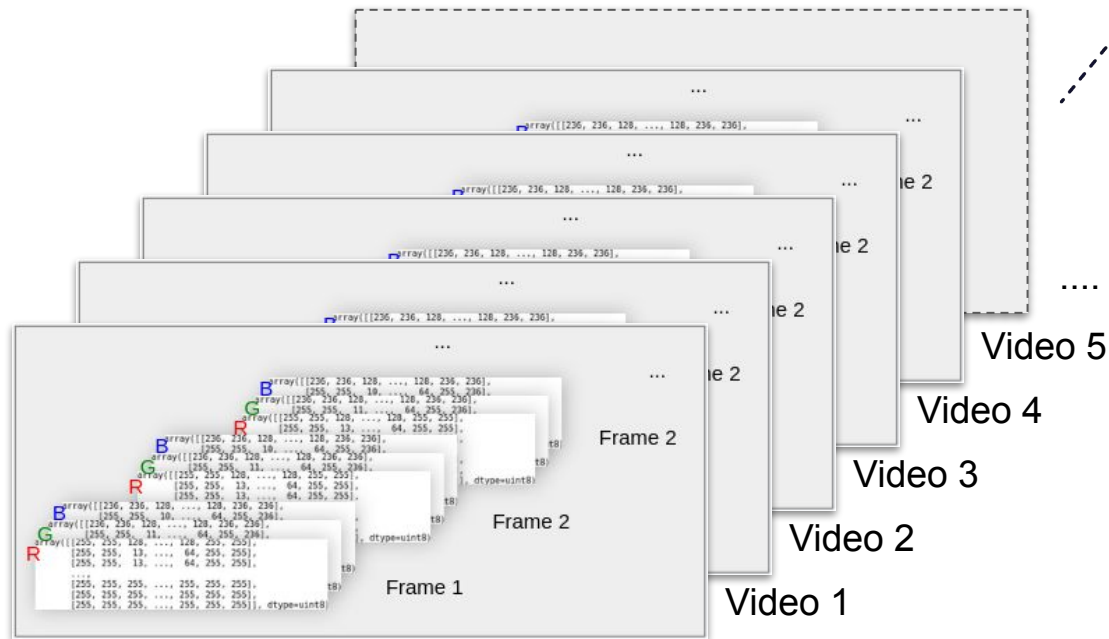
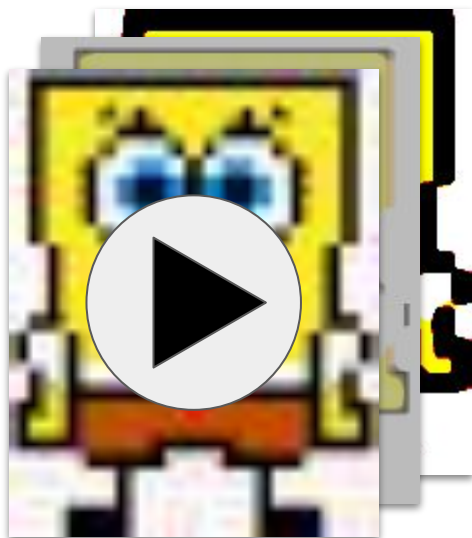
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Some Tensors

e.g. Dataset of 10,000 videos



rank 5 tensor, shape (10000, 200, 48, 28, 3)

48 x 28 pixels, 200 frames, 10,000 examples



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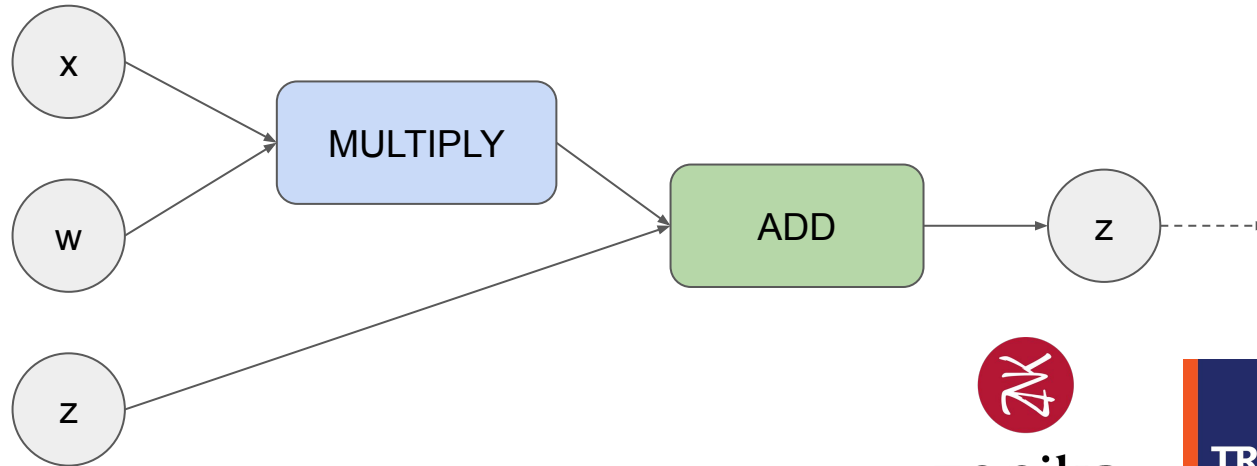
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Prerequisite 3/3: Computational Graphs

- TensorFlow uses a **dataflow graph** to represent your computation in terms of the dependencies between individual operations

Example: $z = w * x + b$

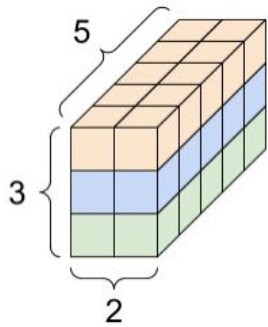


Why Computational Graphs?

- **Parallelism:** easy for the system to identify operations that can execute in parallel.
- **Portability:** graph is defined in high level language, independent representation of the code in your model.
- **Compilation:** graph is compiled with TensorFlow's XLA (Accelerated Linear Algebra) compiler to generate faster code (e.g. fusing together adjacent operations, ...)
- **Distributed execution:** it is possible for TensorFlow to partition your program across multiple devices (CPUs, GPUs, and TPUs) attached to different machines.



TensorFlow Logo

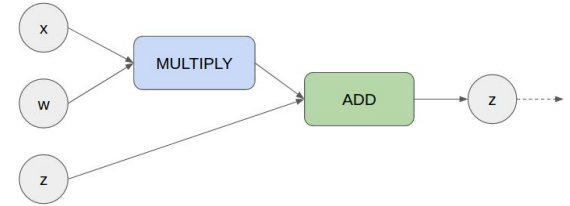


Tensor



TensorFlow

Flow



```
tf.Tensor(  
[[[ 0  1  2  3  4]  
 [ 5  6  7  8  9]]  
  
[[10 11 12 13 14]  
 [15 16 17 18 19]]  
  
[[20 21 22 23 24]  
 [25 26 27 28 29]]], shape=(3, 2, 5), dtype=int32)
```



Simple Computation with TensorFlow 1

TF 1.x
(before 10/2019)

```
import tensorflow as tf
print("TensorFlow version: {}".format(tf.__version__))
```

TensorFlow version: 1.15.0

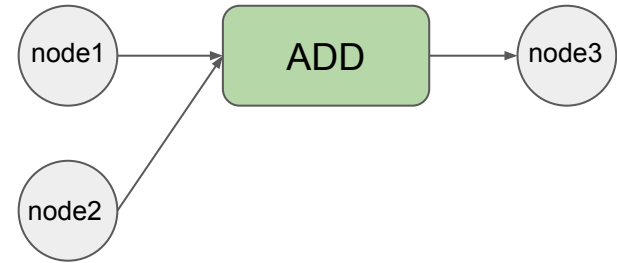
Construction phase

```
node1 = tf.constant([1, 2, 5])
node2 = tf.constant([0, 10, -5])
node3 = tf.add(node1, node2)
init = tf.global_variables_initializer()
```

Execution phase

```
with tf.Session() as sess:
    init.run()
    print(node3.eval())
```

[1 12 0]



Simple Computation with TensorFlow 2

TF 2.x
(since 10/2019)

```
import tensorflow as tf
print("TensorFlow version: {}".format(tf.__version__))
```

TensorFlow version: 2.2.0

```
node1 = tf.constant([1, 2, 5])
node2 = tf.constant([0, 10, -5])
node3 = tf.add(node1, node2)

node3.numpy() #Eager execution by default
array([ 1, 12,  0], dtype=int32)
```

- Makes it easy to get started with TensorFlow and debug models
- Reduces boilerplate code

First Neural Network with TensorFlow

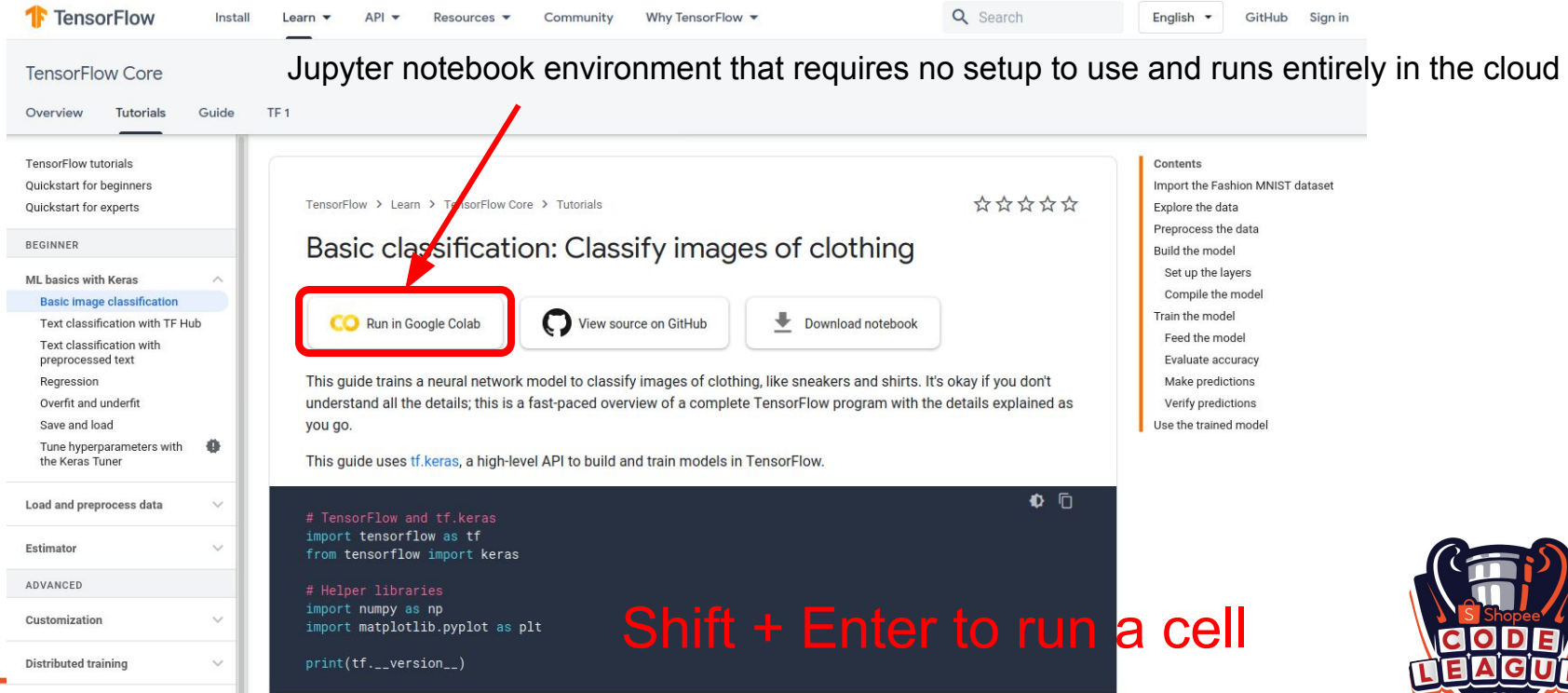


Image Classification

- <https://www.tensorflow.org/tutorials/keras/classification>
- **Keras**: high-level API for building and training deep learning models. It's used for fast prototyping.



Run in Google Colab <https://www.tensorflow.org/tutorials/keras/classification>



TensorFlow Core

Overview Tutorials Guide TF 1

TensorFlow tutorials
Quickstart for beginners
Quickstart for experts

BEGINNER

ML basics with Keras

Basic image classification

Text classification with TF Hub
Text classification with preprocessed text
Regression
Overfit and underfit
Save and load
Tune hyperparameters with the Keras Tuner

Load and preprocess data

Estimator

ADVANCED




Customization

Distributed training

TensorFlow > Learn > TensorFlow Core > Tutorials

Basic classification: Classify images of clothing

☆☆☆☆☆

 Run in Google Colab  View source on GitHub  Download notebook

This guide trains a neural network model to classify images of clothing, like sneakers and shirts. It's okay if you don't understand all the details; this is a fast-paced overview of a complete TensorFlow program with the details explained as you go.

This guide uses `tf.keras`, a high-level API to build and train models in TensorFlow.

```
# TensorFlow and tf.keras
import tensorflow as tf
from tensorflow import keras

# Helper libraries
import numpy as np
import matplotlib.pyplot as plt

print(tf.__version__)
```

Contents

- Import the Fashion MNIST dataset
- Explore the data
- Preprocess the data
- Build the model
 - Set up the layers
 - Compile the model
- Train the model
 - Feed the model
 - Evaluate accuracy
 - Make predictions
 - Verify predictions
- Use the trained model

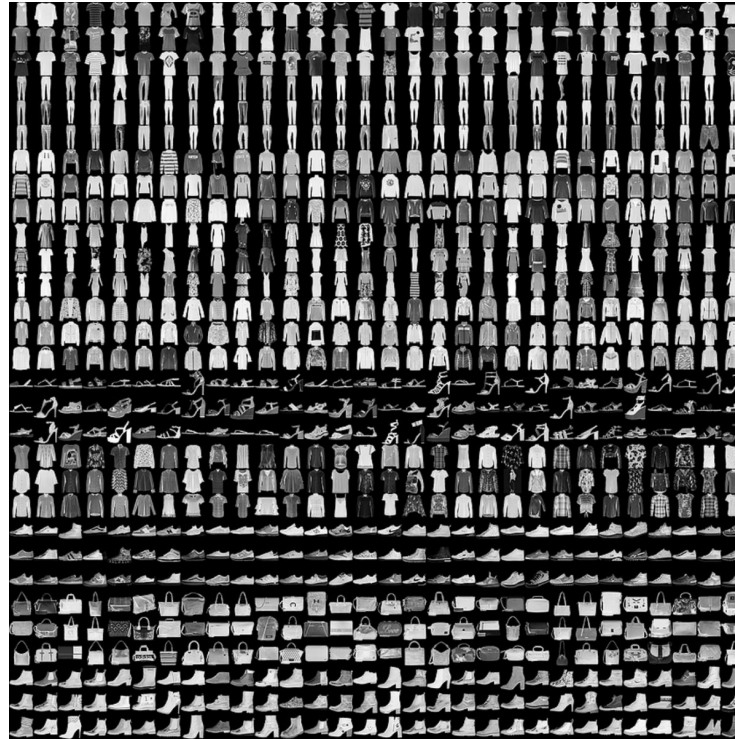
Shift + Enter to run a cell



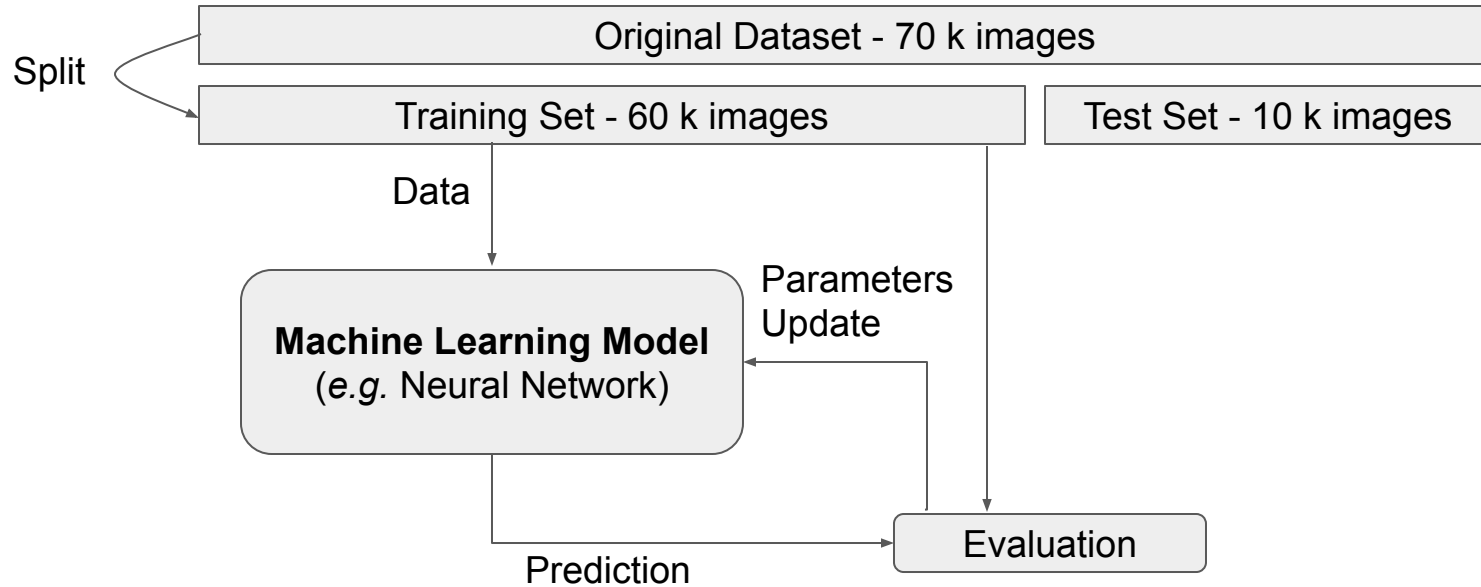
Fashion MNIST

- 70 k grayscale images
- 28 x 28 pixels
- 10 categories/classes

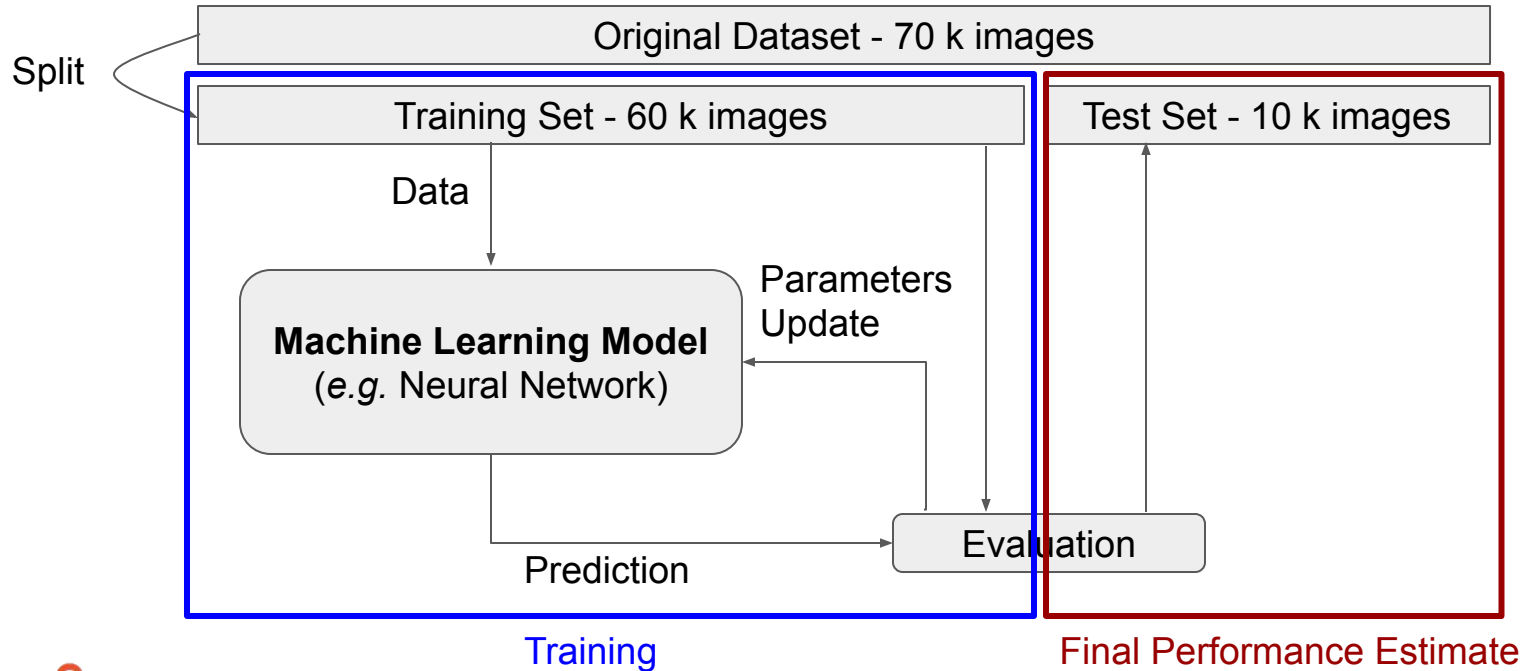
Label	Class
0	T-shirt/top
1	Trouser
2	Pullover
3	Dress
4	Coat
5	Sandal
6	Shirt
7	Sneaker
8	Bag
9	Ankle boot



Basic Model Training



Basic Model Training



Libraries Importation

```
# TensorFlow and tf.keras
import tensorflow as tf
from tensorflow import keras
```

Keras: a high-level API to build and train models in TensorFlow

```
# Helper libraries
import numpy as np
import matplotlib.pyplot as plt
```

NumPy: Scientific Computing with Python

Pyplot: a plotting framework
(e.g. visualize images, curves, histograms...)

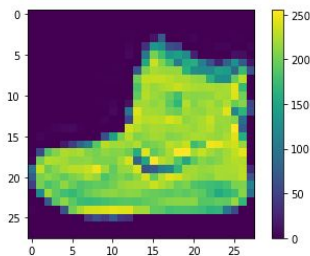
```
print(tf.__version__)
```

```
2.2.0
```

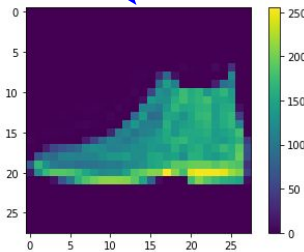


Dataset Importation

```
fashion_mnist = keras.datasets.fashion_mnist  
(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()
```



09



09

Split

Original Dataset - 70 k images

Training Set - 60 k images

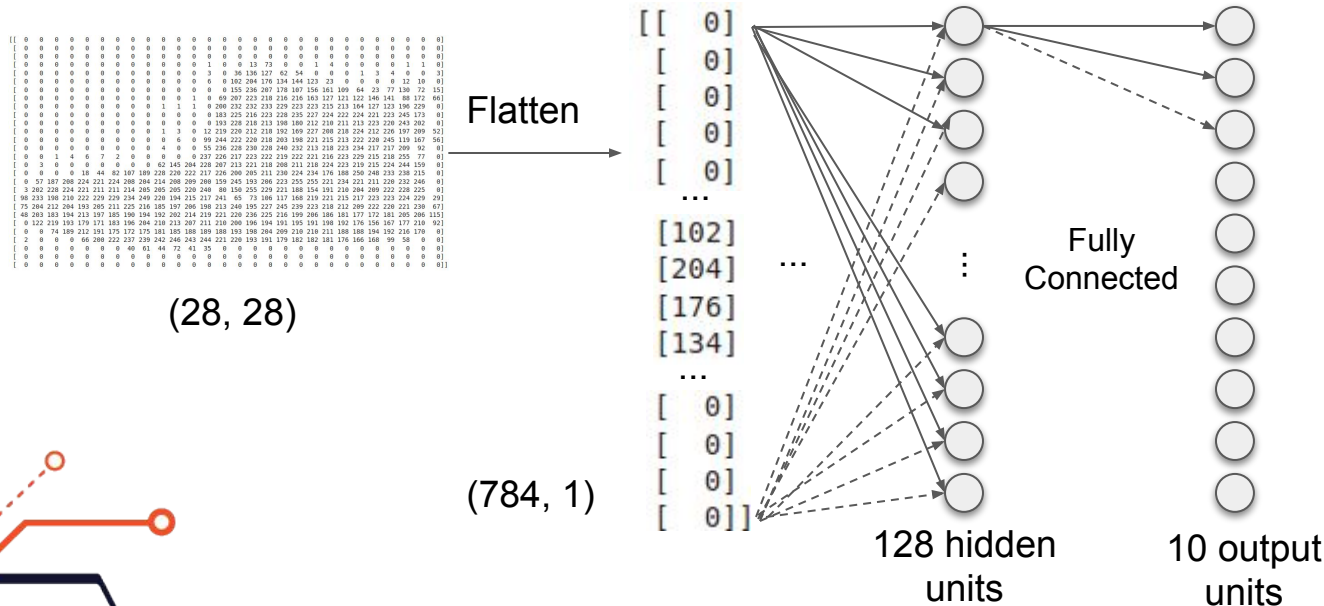
Test Set - 10 k images



ML Model Construction

```
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(128, activation='relu'),
    keras.layers.Dense(10)
])
```

A Sequential model is appropriate for a **plain stack of layers** where each layer has **exactly one input tensor and one output tensor**.



Label	Class
0	T-shirt/top
1	Trouser
2	Pullover
3	Dress
4	Coat
5	Sandal
6	Shirt
7	Sneaker
8	Bag
9	Ankle boot



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Model Compilation

Optimization algorithm (update parameters according to the **loss**)

A **loss** function measures how well the output of a model for a given input matches the target output. The goal is to minimize this difference during training.

```
model.compile(optimizer='adam',  
              loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),  
              metrics=['accuracy'])
```

Accuracy is the number of correctly predicted data points out of all the data points



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Model Training

An **epoch** is a complete learning pass through the entire dataset

```
model.fit(train_images, train_labels, epochs=10)
```

```
1875/1875 [=====] - 3s 2ms/step - loss: 0.3138 - accuracy: 0.8854
Epoch 5/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2948 - accuracy: 0.8914
Epoch 6/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2803 - accuracy: 0.8958
Epoch 7/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2678 - accuracy: 0.9006
Epoch 8/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2590 - accuracy: 0.9043
Epoch 9/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2464 - accuracy: 0.9078
Epoch 10/10
1875/1875 [=====] - 3s 2ms/step - loss: 0.2402 - accuracy: 0.9099

<tensorflow.python.keras.callbacks.History at 0x7fe472eff4a8>
```



Evaluate Accuracy

```
test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)

print('\nTest accuracy:', test_acc)
```

```
313/313 - 1s - loss: 0.3382 - accuracy: 0.8840
```

```
Test accuracy: 0.8840000033378601
```

88% of images are correctly classified



Make predictions

```
probability_model = tf.keras.Sequential([model,  
                                         tf.keras.layers.Softmax()]])
```

The **Softmax** layer normalizes the output into a probability distribution.

Model output **without** Softmax:

```
array([-11.4, -15.4, -15.1, -14.5, -18.8, -1.4, -13.2, 3.39, -16.1, 7.92], dtype=float32)
```

Model output **with** Softmax:

```
array([4.05e-09, 7.34e-11, 9.56e-11, 1.89e-10, 2.36e-12, 8.91e-05, 6.72e-10, 0.0107, 3.56e-11, 0.989], dtype=float32)
```



Make predictions

```
probability_model = tf.keras.Sequential([model,  
                                         tf.keras.layers.Softmax()]])
```

```
predictions = probability_model.predict(test_images)
```

```
array([1.2275562e-07, 8.9534730e-08, 7.6224147e-09, 1.9726743e-08,  
       6.1515345e-07, 4.6329503e-03, 4.1164008e-06, 1.5043605e-02,  
       2.8817641e-07, 9.8031825e-01], dtype=float32)
```

The **Softmax** layer normalizes the output into a probability distribution.



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Make predictions

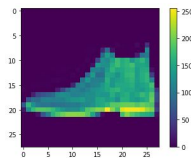
```
probability_model = tf.keras.Sequential([model,  
                                         tf.keras.layers.Softmax()]])
```

```
predictions = probability_model.predict(test_images)
```

```
array([1.2275562e-07, 8.9534730e-08, 7.6224147e-09, 1.9726743e-08,  
       6.1515345e-07, 4.6329503e-03, 4.1164008e-06, 1.5043605e-02,  
       2.8817641e-07, 9.8031825e-01], dtype=float32)
```

```
np.argmax(predictions[0])
```

9



Ankel boot !

The **Softmax** layer normalizes the output into a probability distribution.



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Make predictions

```
probability_model = tf.keras.Sequential([model,  
                                         tf.keras.layers.Softmax()]])
```

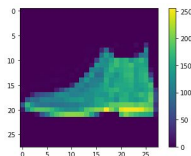
The **Softmax** layer normalizes the output into a probability distribution.

```
predictions = probability_model.predict(test_images)
```

```
array([1.2275562e-07, 8.9534730e-08, 7.6224147e-09, 1.9726743e-08,  
       6.1515345e-07, 4.6329503e-03, 4.1164008e-06, 1.5043605e-02,  
       2.8817641e-07, 9.8031825e-01], dtype=float32)
```

```
np.argmax(predictions[0])
```

9



→ Ankel boot !

Make prediction with
your own (28, 28) image:

```
probability_model.predict(my_image)
```



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Key Take Aways

- Tensors are multidimensional arrays, useful to represent data
- With Machine Learning, we generally try to **deduce rules from data and answers**
- Data (features) and answers (labels) constitute a **dataset**
- Model training: we split the dataset into a **training set** and a **testing set**
- Model performance during training is measured with the **loss**
- Model training is **iterative**



Next ?

- TensorFlow Documentation, <https://www.tensorflow.org/learn>
- Courses/Training
- MOOC
- Books, <https://www.deeplearningbook.org>
- Blogs, <https://towardsdatascience.com/>
- Kaggle Competitions, <https://www.kaggle.com>
- Code your own ML project !





COURSE BREAKDOWN

01

Introduction and
Reminders about
Machine Learning

02

The Basics of
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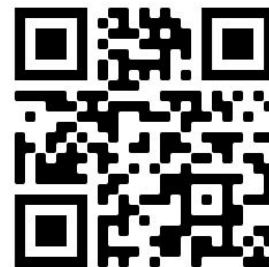
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Thanks for attending !



Neural Style Transfer with TensorFlow, Harry POMMIER



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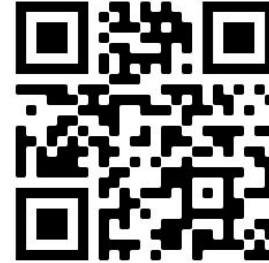


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