

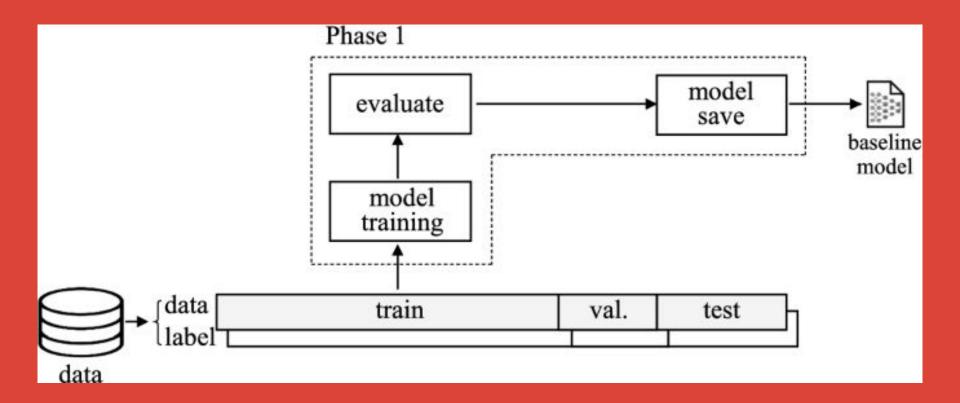
TensorFlow and Code Review

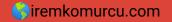
Deep Learning in Remote Sensing

Episode-2

İrem KÖMÜRCÜ iremkomurcu.com iremkomurcubm@gmail.com

Create General Baseline Model





Data Visualization of the Splits

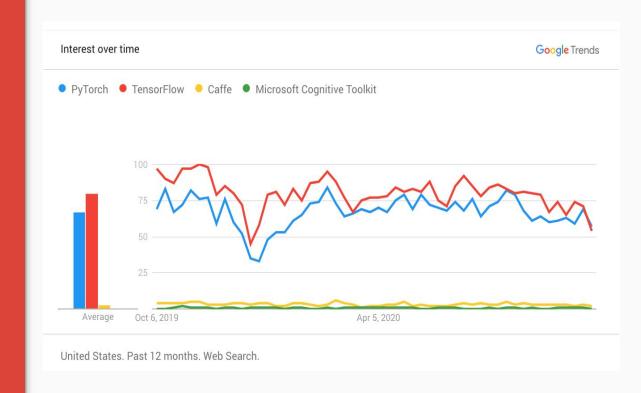




- Create, Google Brain Team
- Open Source
- Comprehensive, flexible ecosystem of tools, libraries
- Community Resources
- Easy model building
- Supporting multi language
- Google Cloud Platform, Google Cloud Speech etc

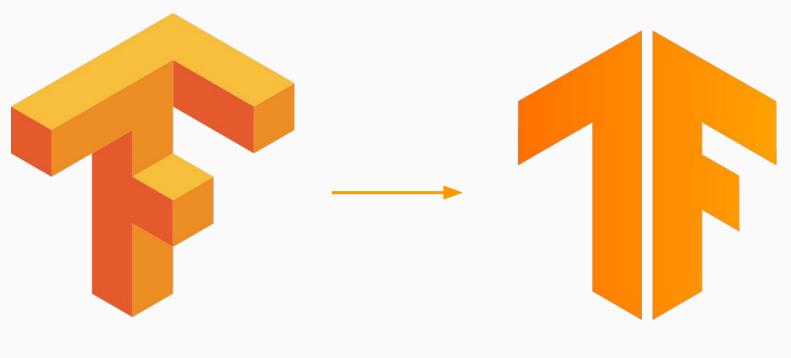
Some **DL Frameworks**

- Tensorflow
- Keras
- Pytorch
- MXNet
- Theano
- Caffe

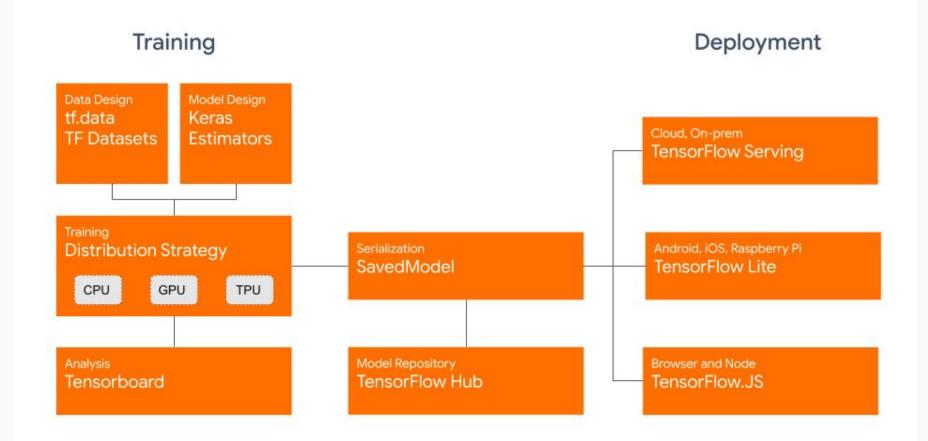


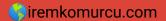
Source: tooploox.com





Tensorflow 1.x Tensorflow 2.x









import tensorflow as tf

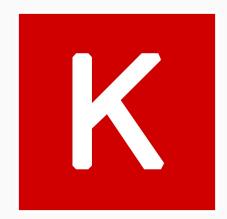
from tensorflow.keras.layers import Dense, Flatten, Conv2D
from tensorflow.keras import Model



Tensorflow & Keras

- Data preprocessing
- Model
- Applications
- Layer
- Optimizer
- Activation
- Loss Function
- Callbacks
- Metrics



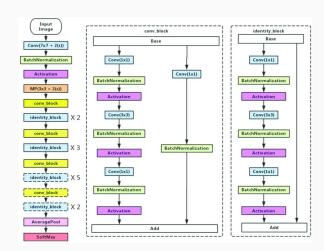


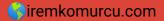
Data Preprocessing

```
tf.keras.preprocessing.image dataset from directory(
    directory,
    labels="inferred",
    label mode="int",
                                                     tf.keras.preprocessing.text.Tokenizer(
    class_names=None,
                                                         num words=None,
    color mode="rgb",
                                                         filters='!"#$%&()*+,-./:;<=>?@[\\]^_`{|}~\t\n',
    batch size=32,
                                                         lower=True,
    image size=(256, 256),
                                                         split=" ",
    shuffle=True,
                                                         char level=False,
    seed=None,
                                                         oov token=None,
    validation split=None,
                                                         document count=0,
                                                         **kwargs
    subset=None,
    interpolation="bilinear",
    follow links=False,
```

Tensorflow & Keras Applications

- Resnet50
- VGG16
- Xception
- EfficientNet
- Inception_v3
- Densenet
- Mobilenet





Optimizer

Adam Optimizer Formula

```
for t in range(num_iterations):

g = compute_gradient(x, y)

m = beta_1 * m + (1 - beta_1) * g

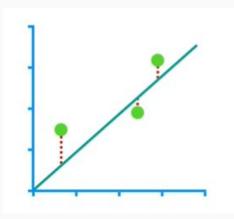
v = beta_2 * v + (1 - beta_2) * np.power(g, 2)

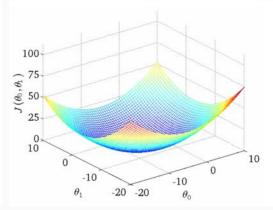
m_hat = m / (1 - np.power(beta_1, t))

v_hat = v / (1 - np.power(beta_2, t))

w = w - step_size * m_hat / (np.sqrt(v_hat) + epsilon)

Adam.pv hosted with by by GitHub
view raw
```





```
tf.keras.optimizers.Adam(
    learning_rate=0.001,
    beta_1=0.9,
    beta_2=0.999,
    epsilon=1e-07,
    amsgrad=False,
    name="Adam",
    **kwargs
)
```

Optimizer

- Adam
- SGD
- RMSprop
- Adadelta
- Adagrad
- Adamax
- ...

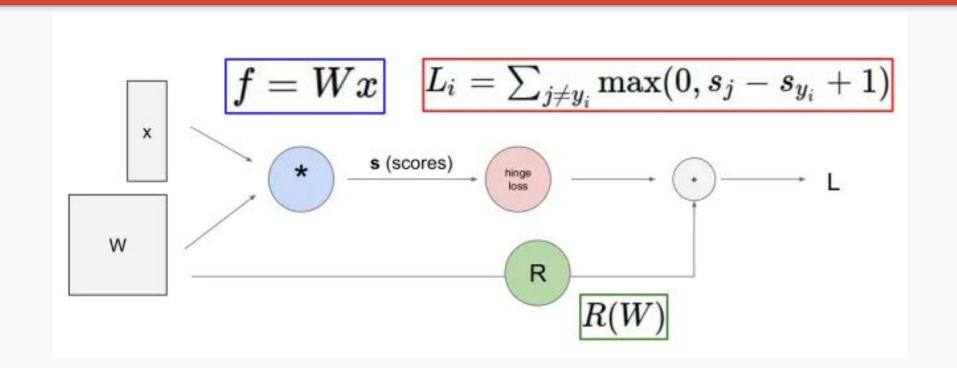
```
from tensorflow import keras
from tensorflow.keras import layers

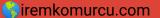
model = keras.Sequential()
model.add(layers.Dense(64, kernel_initializer='uniform', input_shape=(10,)))
model.add(layers.Activation('softmax'))

opt = keras.optimizers.Adam(learning_rate=0.01)
model.compile(loss='categorical_crossentropy', optimizer=opt)
```

```
# pass optimizer by name: default parameters will be used
model.compile(loss='categorical_crossentropy', optimizer='adam')
```

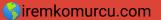
Loss Function





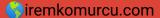
Loss Function

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opt = keras optimizers Adam(learning rate=0.01)
model.compile(loss='categorical crossentropy', optimizer=opt)
# pass optimizer by name: default parameters will be used
model.compile(loss='categorical crossentropy', optimizer='adam')
```



Callbacks

```
my_callbacks = [
    tf.keras.callbacks.EarlyStopping(patience=2),
    tf.keras.callbacks.ModelCheckpoint(filepath='model.{epoch:02d}-{val_loss:.2f}.h5'),
    tf.keras.callbacks.TensorBoard(log_dir='./logs'),
]
model.fit(dataset, epochs=10, callbacks=my_callbacks)
```



Training

```
print("Fit model on training data")
history = model.fit(
    x_train,
    y_train,
    batch_size=64,
    epochs=2,
    # We pass some validation for
    # monitoring validation loss and metrics
    # at the end of each epoch
    validation_data=(x_val, y_val),
)
```

Metrics

Regression

- o MSAE
- o R Square
- Adjusted R Square

Classification

- o Precision-Recall
- o ROC-AUC
- Accuracy
- o Log-Loss

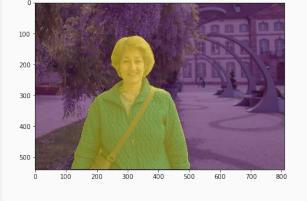
Unsupervised Models

- Rand Index
- Mutual Information

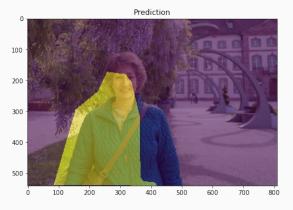
Others

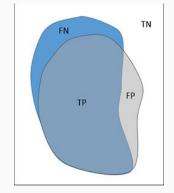
- CV Error
- Heuristic methods to find K
- BLEU Score (NLP)

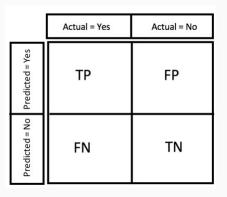




Ground Truth







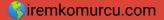
Metrics

Metrics

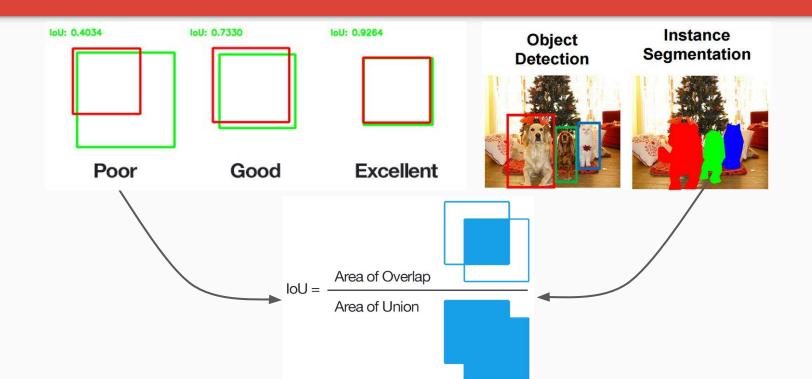
| Metric | Formula |
|----------------------------|---|
| True positive rate, recall | $\frac{\mathrm{TP}}{\mathrm{TP} + \mathrm{FN}}$ |
| False positive rate | $\frac{\text{FP}}{\text{FP+TN}}$ |
| Precision | $\frac{\mathrm{TP}}{\mathrm{TP} + \mathrm{FP}}$ |
| Accuracy | $\frac{\text{TP+TN}}{\text{TP+TN+FP+FN}}$ |
| F-measure | $\frac{2 \cdot \text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}}$ |

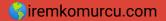
| | Actual = Yes | Actual = No |
|-----------------|--------------|-------------|
| Predicted = Yes | TP | FP |
| Predicted = No | FN | TN |

Source 1: deepai.org Source 2: kdnuggets.com



Metrics

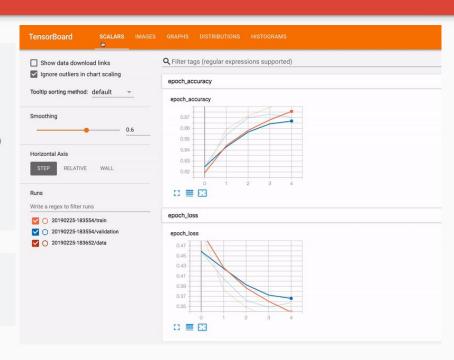




TensorBoard

for visual debugging

%tensorboard --logdir logs/fit





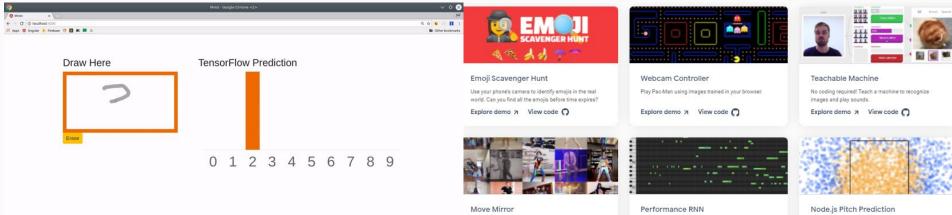
TensorFlow.js for machine learning on the web

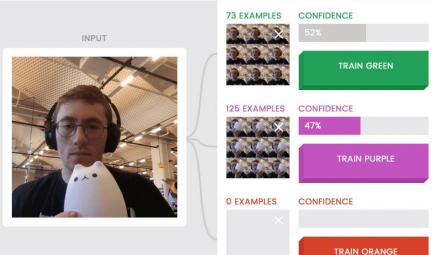




iremkomurcu.com

https://www.tensorflow.org/js/demos





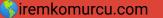




TensorFlow.js

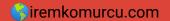
```
<script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs@1.0.0/dist/tf.min.js"></script>
```

- Using Script Tags
- Installation from NPM and using a build tool like Parcel, WebPack or Rollup



TensorFlow Lite for mobile and embedded ML





TensorFlow Lite

https://www.tensorflow.org/lite/examples/

TensorFlow Lite example apps

A collection of TensorFlow Lite apps.



Image classification

Test an image classification solution with a pre-trained model that can recognize 1000 different types of items from input frames on a mobile cemera.

Try it on Android ()
Try it on iOS ()

Try it on Raspberry Pi



Object detection

Explore an app using a pre-trained model that draws and labels bounding boxes around 1000 different recognizable objects from input frames on a mobile camera.

Try it on Android \bigcap Try it on iOS \bigcap

Try it on Raspberry Pi 🕠



Pose estimation

Explore an app that estimates poses of people in an image.

Try it on Android Ω Try it on iOS Ω



Speech recognition

Explore an app that uses a microphone to spot keywords



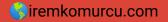
Gesture recognition

Train a neural network to recognize gestures caught on



Smart reply

Generate reply suggestions to input conversational chat



Practice Session

Use of GitHub Github Code Review Segmentation Example

Please visit on YouTube video to talk about this presentation and practice session. You can find the video link in the my GitHub repo.

THANKS



Does anyone have any questions?

iremkomurcubm@gmail.com iremkomurcu.com







