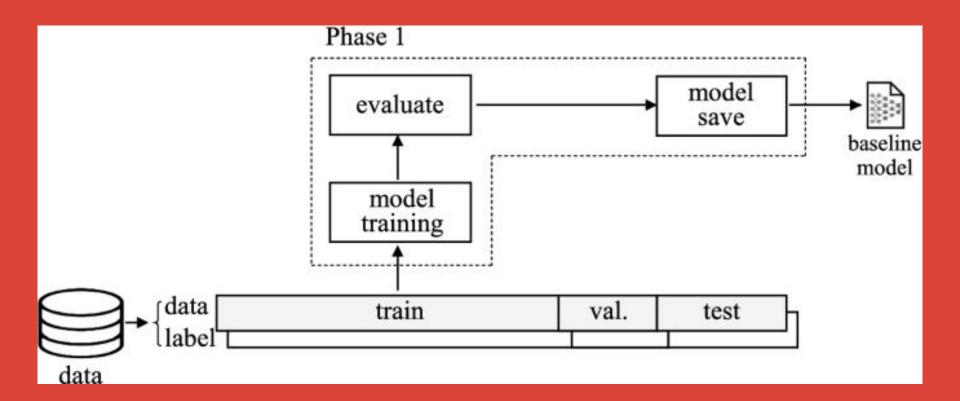
## TensorFlow and Code Review

Deep Learning in Remote Sensing

Episode-2

İrem KÖMÜRCÜ iremkomurcu.com iremkomurcubm@gmail.com

#### Create General Baseline Model



## Data Visualization of the Splits

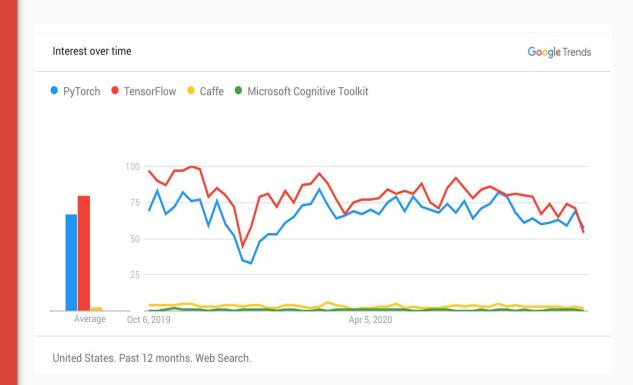




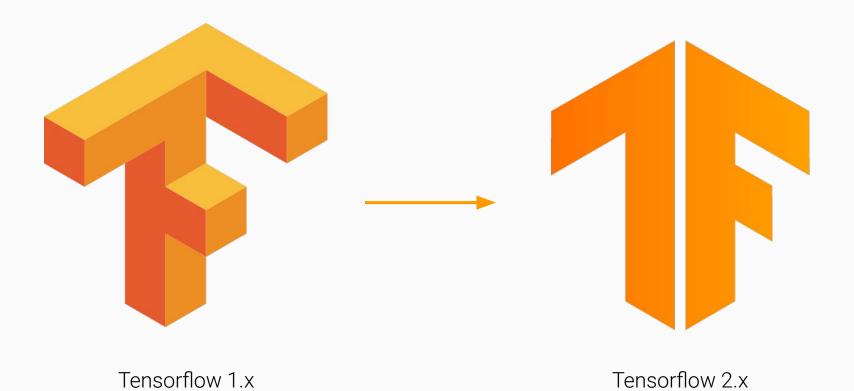
- Create, Google Brain Team
- Open Source
- Comprehensive, flexible ecosystem of tools, libraries
- Community Resources
- Easy model building
- Supporting multi language
- Google Cloud Platform, Google Cloud Speech etc

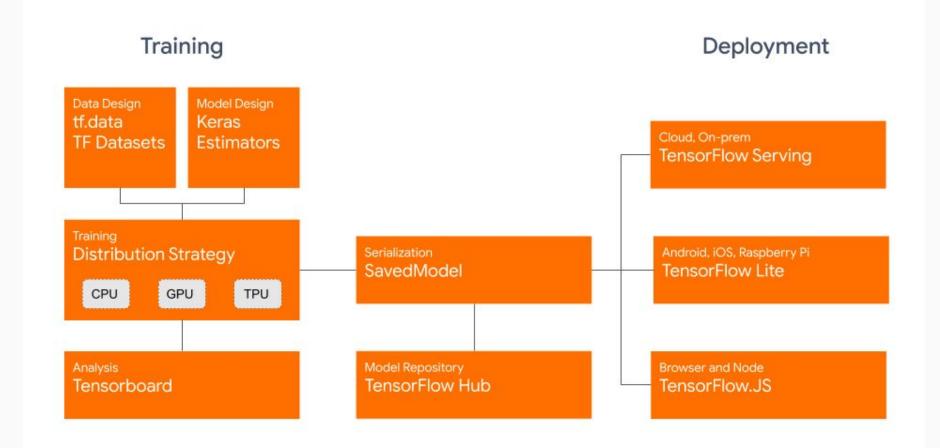
## Some **DL Frameworks**

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













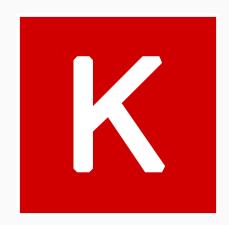
import tensorflow as tf

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



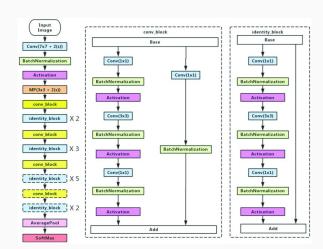


### **Data Preprocessing**

```
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

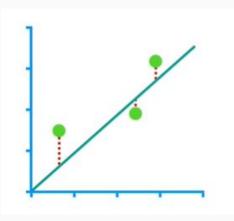
v = beta_2 * v + (1 - beta_2) * np.power(g, 2)

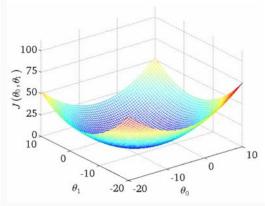
m_hat = m / (1 - np.power(beta_1, t))

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 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
- ...

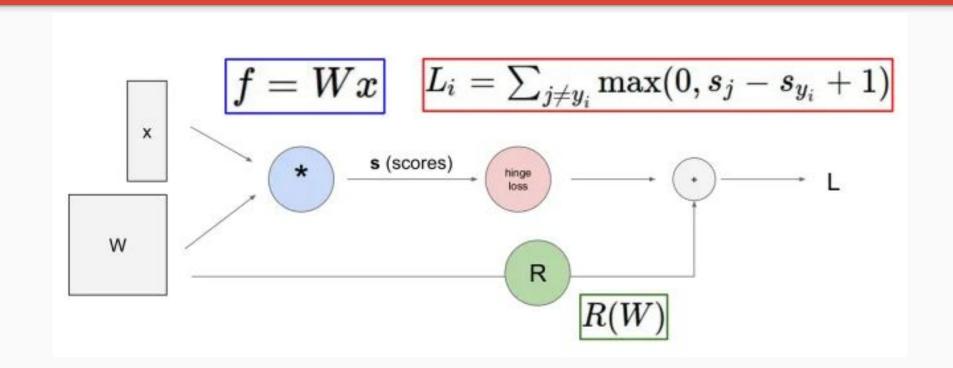
```
from tensorflow import keras
from tensorflow.keras import layers

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

```
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')

### 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),
)
```

#### Regression

- o MSAE
- o R Square
- Adjusted R Square

#### Classification

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

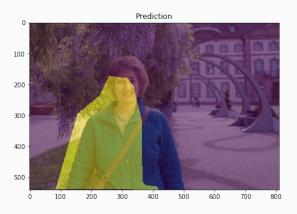
## Unsupervised Models

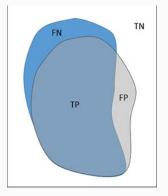
- Rand Index
- Mutual Information

#### Others

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





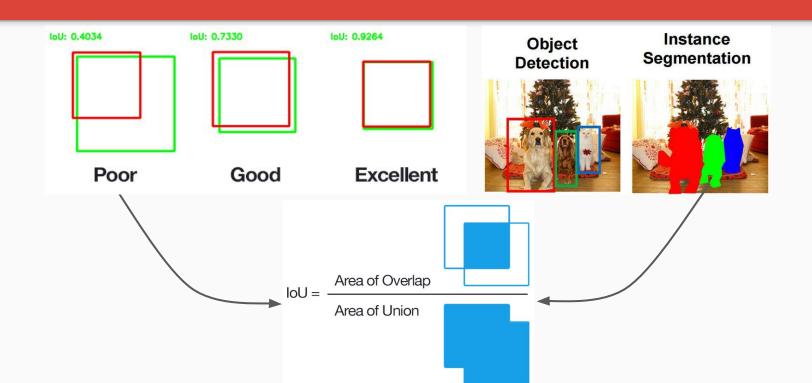


	Actual = Yes	Actual = No
Predicted = Yes	TP	FP
Predicted = No	FN	TN

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

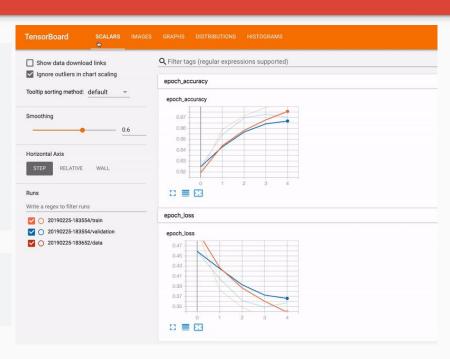
Source 1: deepai.org Source 2: kdnuggets.com



#### TensorBoard

### for visual debugging

%tensorboard --logdir logs/fit

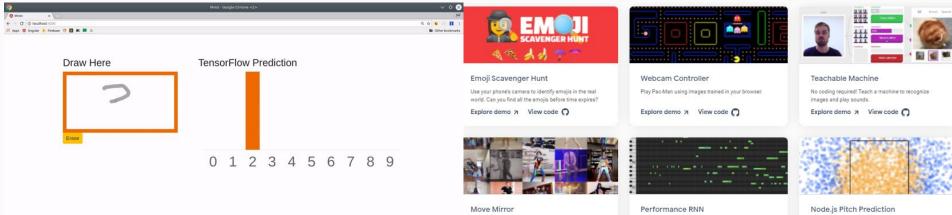


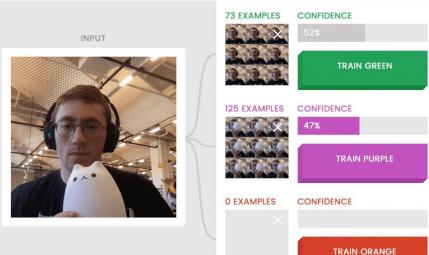
## TensorFlow,js for machine learning on the web



## TensorFlow.js

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 camera.

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 ()
Try it on iOS ()



#### Speech recognition

Explore an app that uses a microphone to spot keywords



#### Gesture recognition

Train a neural network to recognize destures caught on



#### Smart reply

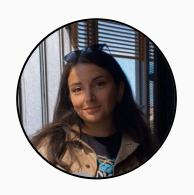
Generate reply suggestions to input conversational chat

#### **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?

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