

## DEEP LEARNING

sample



#### **ABOUT THE SPEAKER**



Dr Assan Sanogo
Data Scientist x Pharmacist
Independent Data scientist
Develops data sciences program
Coaches > 100 junior data scientists

aws certified

Solutions
Architect

Associate

aWS certified
Cloud
Practitioner

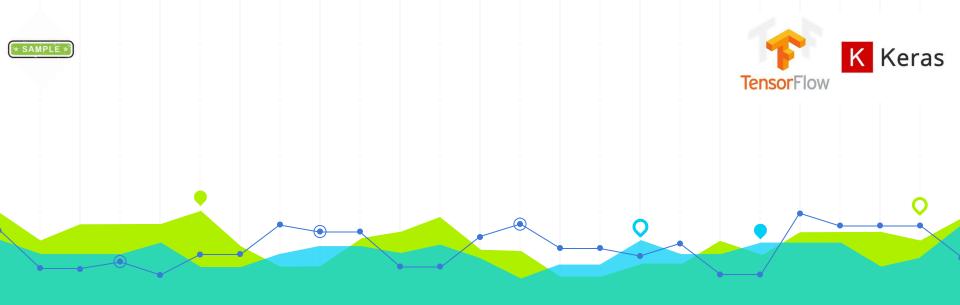
aws certified
SysOps
Administrator
Associate



#### **COURSE STRUCTURE**



- 1. What is Deep Learning?
- 2. What is Tensorflow & Keras
- 3. Image Data
- 4. The structure of a Neural network
- 5. Convolutional layers
- 6. Pooling Layers
- 7. Activation
- 8. Loss function
- 9. Optimizer
- 10. Callbacks
- 11. Training
- 12. Evaluation



## What is Deep Learning?

Let's start with the basics



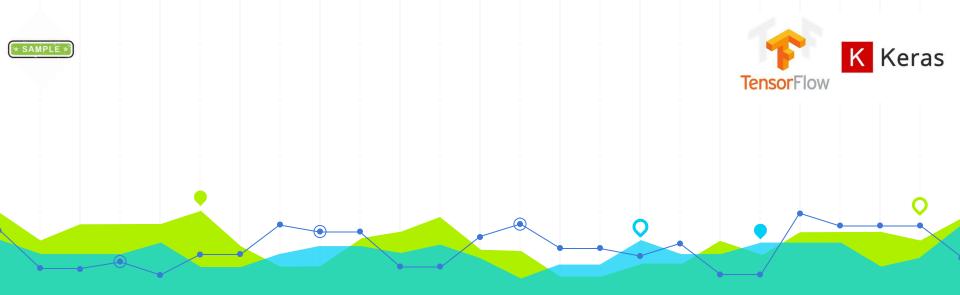


#### 1. WHAT IS DEEP LEARNING?

- Deep Learning is a subfield of machine learning concerned with algorithms inspired by the structure and function of the brain called artificial neural networks.
- Deep Learning is the key technology behind autonomous cars, enabling them to recognize a
  cat or dog, or to distinguish digits
- Deep learning requires large amounts of labeled data. For example, autonomous car development requires millions of images and thousands of hours of recording.
- **Deep learning** requires a significant **computing power**. High-performance GPUs offer a parallel architecture that is efficient for this type of algorithm.

Deep Learning has set the standard in recognition and detection accuracy





## What is Keras & Tensorflow?

Simplified modular deep learning

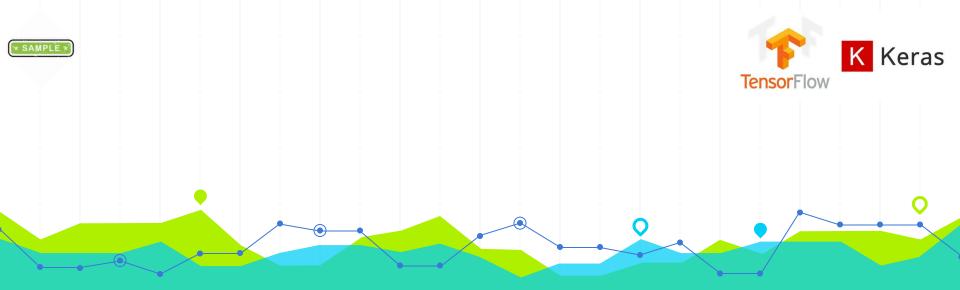




#### 2. WHAT IS TENSORFLOW.KERAS?

- Keras is the high-level API of TensorFlow
- highly-productive interface for solving deep learning problems.
- provides **essential abstractions** and building blocks
- allows to develop deep learning solutions very fast

keras and tf.keras are two separate projects. (although they share many commonalities)



## Image data Images are tensors







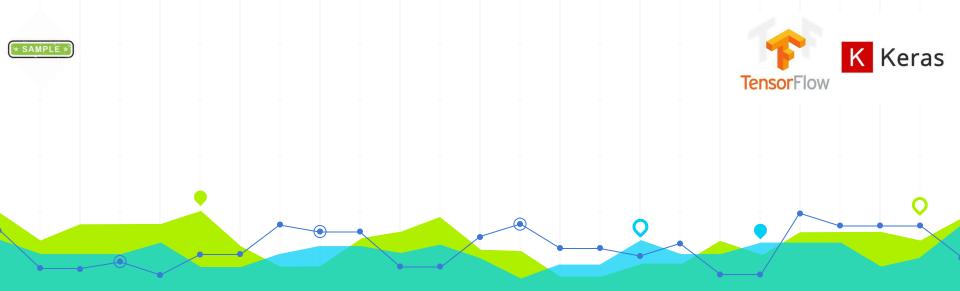
#### 3. THE DATA

- Dataset : Pictures of Dogs and Cats
- 25 000 Images (.jpg)
- Various Image size



The images vary in quality in simplicity (occlusion, multiple animals, human presence etc.)





# The Structure of a DL model

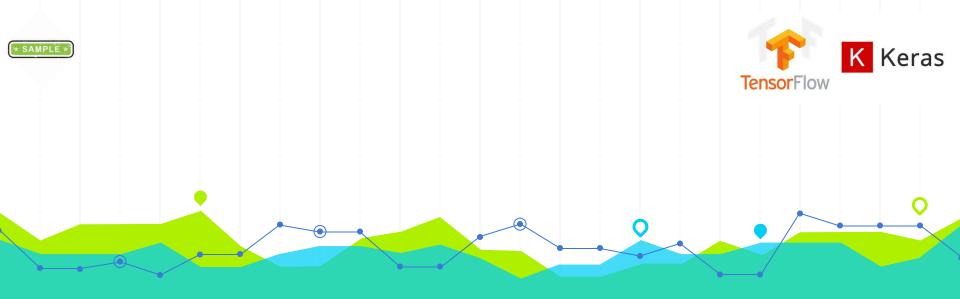
A matter of building blocks





#### A. STRUCTURE OF A CONVNET MODEL

- The main purpose of the **convolutional layers** is to extract meaningful features from the input images.
- To identify these features, CNNs analyze images thanks to "groups of pixels" called **filters.**
- Filters act as feature detectors from the original image.
- During the training process, the CNN slides the filter over all locations of the image and calculates the dot product for each feature at a time.
- The results of these calculations are stored in a so-called feature map (sometimes also named activation map).
- A feature map displays where a certain feature was identified in the original image.
- Then, the values from the **feature map** are transformed with an **activation function**
- These feature maps are used as **input** of the next layer



# Layers Let's dig in the building blocks

5



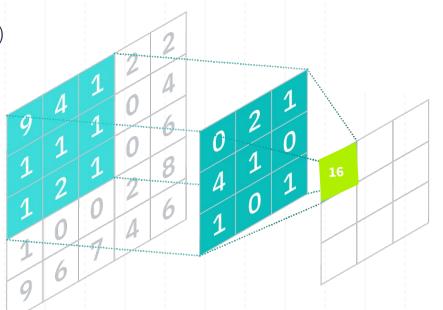


Output array

#### A. WHAT IS A CONVOLUTION?

Input image

- The learned filter (dark turquoise)
- The image (light turquoise)
- Output (green)
- Image size reduction
- Learnable filters



Filter





#### **B. WHAT IS A MAX POOLING LAYER?**

- The image (light turquoise)
- Image size reduction (padding)
- No Learnable filters

1	3	5	1
76	32	12	6
26	07	8	26
16	17	68	32



76	12
26	68





#### **C. WHAT IS AN AVERAGE POOLING LAYER?**

- The image (light turquoise)
- Image size reduction (padding)
- No Learnable filters

1	3	5	1
76	32	12	6
26	37	68	26
16	37	8	32

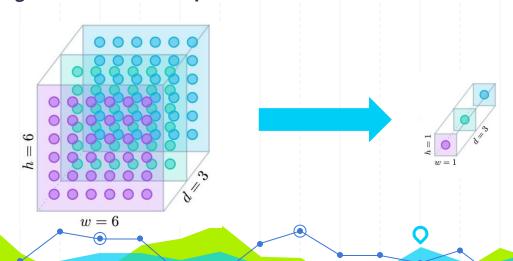






#### D. WHAT IS A GLOBAL AVERAGE POOLING LAYER?

- Transfert learning
- Mapping of model feature maps to a vector

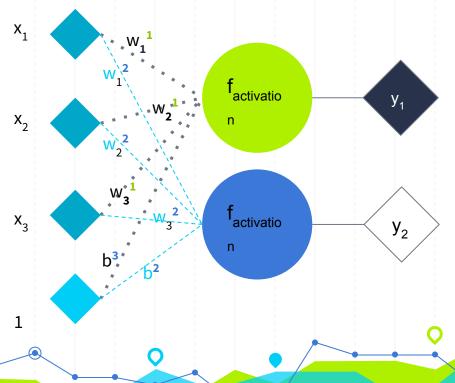


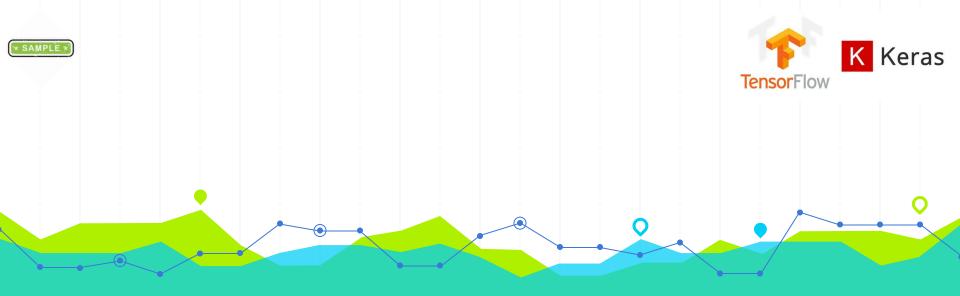




#### D. WHAT IS A DENSE LAYER?

- Dense layers are fully connected to previous layers
- Concentrates the number of learnable parameters
- Output after Activation
- Numerous notations





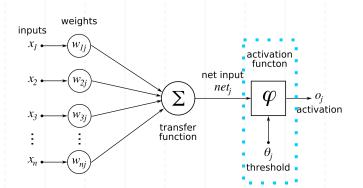
## **Activation functions**

Set activate and Fire!





- Function added into an artificial neural network
   in order to make the network learn complex patterns in the data.
- Activation function are either **linear** or **nonlinear** (sigmoid, tanH, reLu)
- In analogy to **biological neurons**, the activation function represents the final step when "activated" neurons fire.



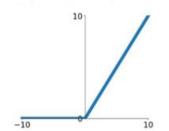




- reLU
- Sigmoid
- TanH

#### ReLU

 $\max(0, x)$ 



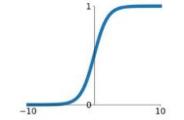
**ReLU, Sigmoid & tanH are NON LINEAR activation functions** 





- reLU
- **Sigmoid**
- TanH

Sigmoid 
$$\sigma(x) = \frac{1}{1+e^{-x}}$$



**ReLU, Sigmoid & tanH are NON LINEAR activation functions** 

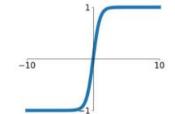




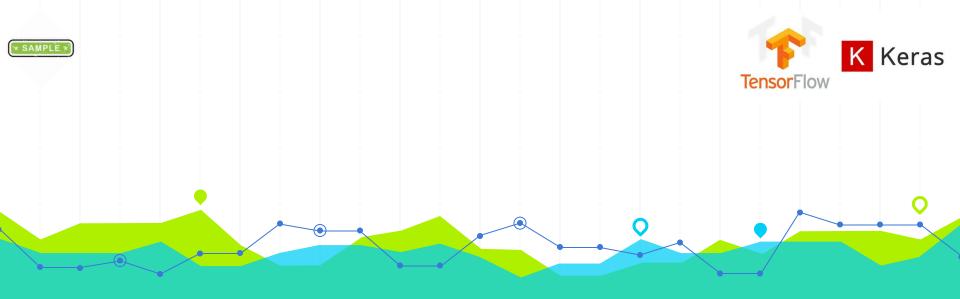
- reLU
- Sigmoid
- TanH

### tanh

tanh(x)



**ReLU, Sigmoid & tanH are NON LINEAR activation functions** 



## **Loss function**

The objective function has to be minimized



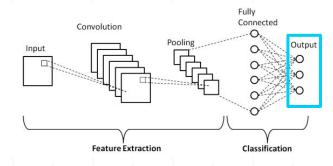


#### **WHICH LOSS FUNCTION?**

- Binary classification (n=2 classes) Last activation function: Sigmoid
  - Email spam detection (spam or not)
  - Churn prediction (churn or not)
- Categorical classification (n>2 classes)
  Last activation function: Softmax
  - Digits classification (1 to 10)
  - Vehicles classification (cars, motorcycles, trucks)

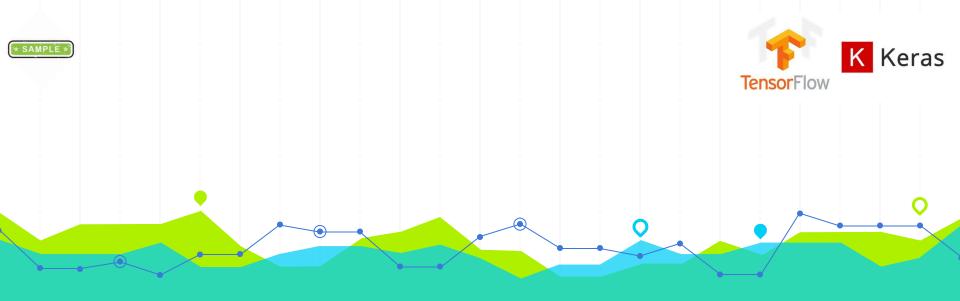


#### **WHICH LOSS FUNCTION?**



- Binary classification (n=2 classes)
- **Last activation** function: **Sigmoid**

- Email spam detection (spam or not)
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## **Optimizers**

minimize the objective function





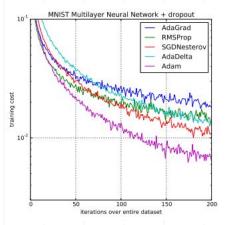
#### **WHICH OPTIMIZER?**

SGD: stochastic gradient descent

Efficient way to fit linear classifiers and regressors under convex loss functions

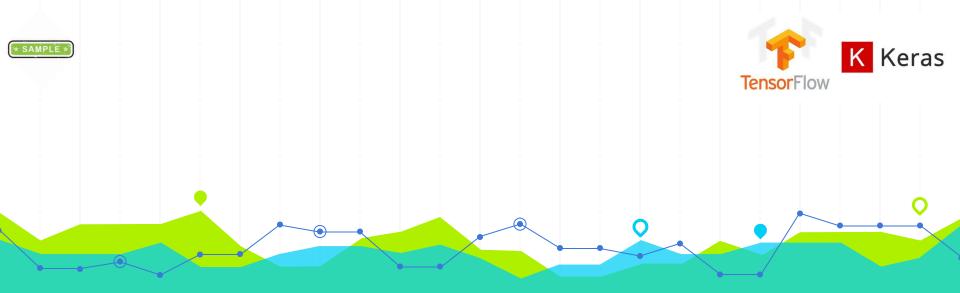
Adam : adaptive moment estimation

The Adam optimization algorithm is an extension to stochastic gradient descent



Comparison of Adam to Other Optimization Algorithms Training a Multilayer Perceptron Taken from Adam: A Method for Stochastic Optimization, 2015.





## **Callbacks**

Save the model progress

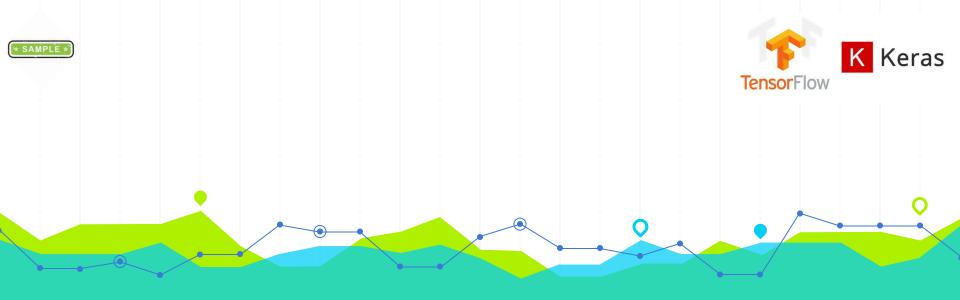






#### WHAT IS A CALLBACK?

- A callback is an object that can perform actions at various stages of training
  - @ the start or end of an epoch
  - before or after a single batch.
- Callbacks for monitoring metrics: tensorboard
- Callbacks for saving model: modelcheckpoint
- Callbacks to avoid overfitting: earlystopping.



## **Training**

Learning will happen step by step & 1 epoch at a time





#### A. TRAIN A NEURAL NETWORK - VOCAB

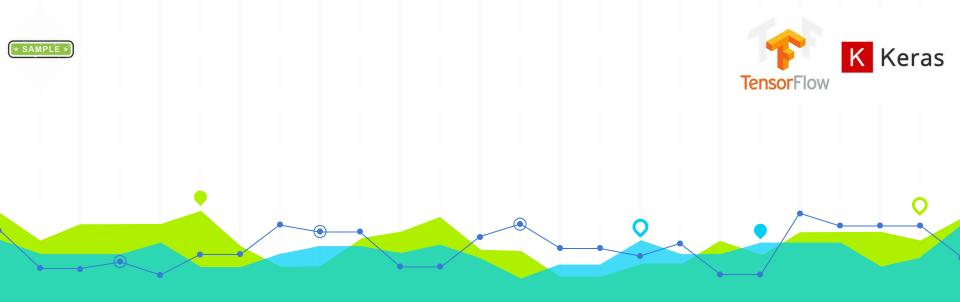
- Epochs number: an hyperparameter which describes the number of times the neural network will process the whole dataset.
- Steps: an hyperparameter which is calculated by using the total number of observations and the batch size
- Batch size: an hyperparameter that sets the number of samples to feed the neural network before updating the learnable model parameters.
- Fine-tuning: Fine-tuning is one approach to transfer learning where you change the model output to fit the new task and train only the final layers of the model.





#### **B.** TRAIN A NEURAL NETWORK STEP BY STEP?

- Prepare the data (preprocessing)
- Choose a pretrained model
- Define optimizers
- Choose the objective function
- Compile the model
- Define the metrics
- Launch the training
- Save the model



## **Evaluation**

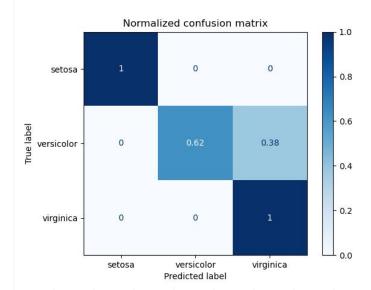
Time to check the model performance





#### **TRAINED MODEL: EVALUATION**

- Confusion matrix
- Precision-recall

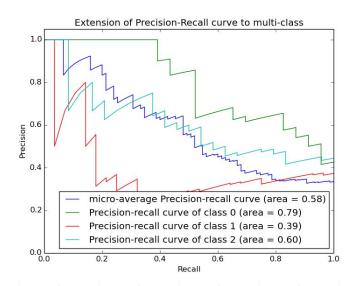






#### **TRAINED MODEL: EVALUATION**

- Confusion matrix
- Precision-recall



# THANKS!

### Any questions?

You can ask for the **full** version

@ predicteev.consulting@gmail.com