Interdisciplinary Center for Applied Machine Learning



Applied Machine Learning Academy

### **Programming Languages and Frameworks for Data Science**

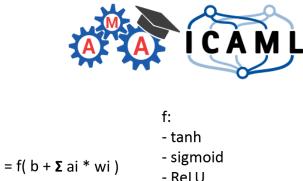
AMA / ICAML - 01.10.2019

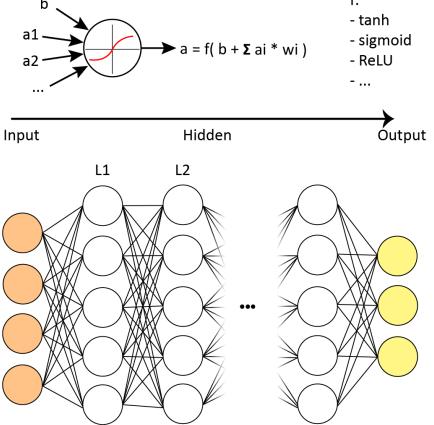


### **Deep Learning**

# **Deep Learning**

- Subfield of machine learning
  - Commonly (wrongly) known as Artificial Intelligence
- Based on Deep Neural Networks
  - Artificial neurons, usually in layers
  - Very complex mappings can be modeled
- Training requires:
  - Specific hardware (GPU / TPU)
  - Many training examples (labeled data)
  - Time (and experience)
- Training based on gradient descend
  - Minimization of a differentiable loss function
- Specific architectures / loss functions for different data-types and applications



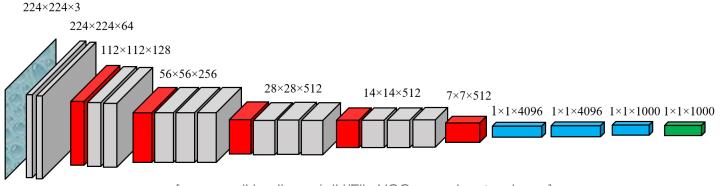


# DL – Applications / Architectures



#### Image classification

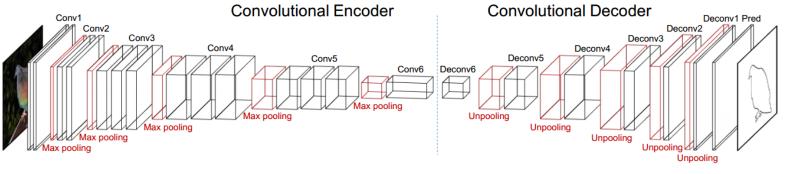
- Convolutional Neural Network
- Convolutions and
- Pooling / downsampling
- Predicts one class per image



[source: wikipedia.org/wiki/File:VGG\_neural\_network.png]

#### Image segmentation

- Fully Convolutional Network
- Encoder similar to CNN
- Decoder uses upsampling via unpooling or transposed convolution
- Predicts one class per pixel



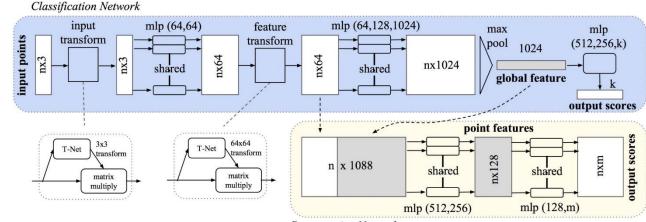
[source: nsarafianos.github.io/icip16]

• Both architectures can also be used for regression

# DL – Applications / Architectures

#### Point cloud classification

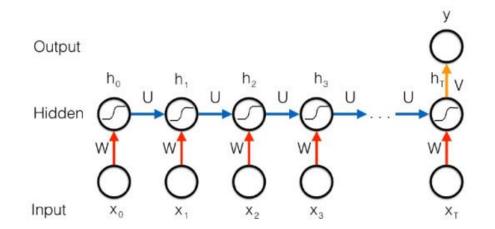
- E.g. PointNet
- Requires special layers for unordered data
- Predicts one class for cloud/ one class per point



Segmentation Network

[source: http://stanford.edu/~rqi/pointnet/images/pointnet.jpg]

- Natural language processing
  - E.g. for language understanding or translation
  - Due to variable input length: Recurrent Neural Networks
  - Process ,word-by-word' with memory



[source: naviglinlp.blogspot.com/2018/04/lecture-10-06042018-recurrent-neural.html]

#### ..... and there are MUCH more!

### Deep Learning – Use cases



#### Researcher

- Training from scratch (or retraining)
- Optimizes network for own needs
- Needs access to all details
- Knows/sets all parameters
- Is interested in fast training

#### **Common user**

- Mostly retrains pre-trained models based on well studied architectures
- Usually doesn't modify the code
- Parameters and details are less important
- Is interested in fast inference

#### $\rightarrow$ Requires low-level API

→ Requires high-level API + optimized deployment

# Frameworks for Deep Learning

- Many frameworks few popular
- Mostly provide similar functionality
  - All can be used for ,usual' cases
  - Choice depends on restrictions / preference
- Most essential features:
  - GPU / TPU support
  - Automated gradient calculation
  - Frequently used functions / layers
- Most supported language: Python
- Powered by *big players*:
  - TensorFlow (Google Brain)
  - PyTorch / Caffe2 (Facebook)
  - Cognitive Toolkit CNTK (Microsoft)

Deep Learning Framework Power Scores 2018

100 96.77 80 60 Score 51.55 40 22.72 17.15 20 12.02 8.37 3.65 2.71 1.181.06 Keras TensorFlow Pytorch MENET Carre Theano Framework

towardsdatascience.com/deep-learning-framework-power-scores-2018-23607ddf297a



### DL – Frameworks and Languages



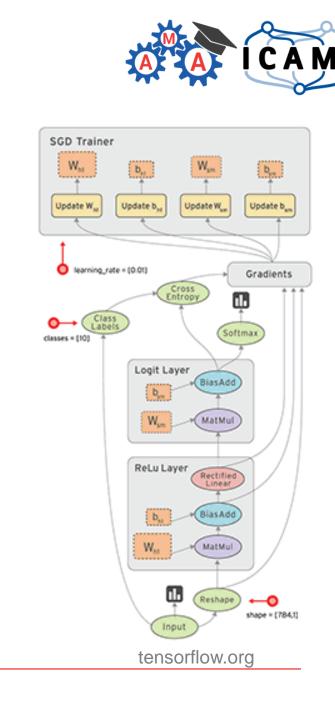
	Python	C++	R	Java	Matlab	Julia
TensorFlow	$\checkmark$	$\checkmark$	$\checkmark$	(√)		(√)
Keras	$\checkmark$		(√)			(√)
PyTorch	$\checkmark$					(√)
Caffe	$\checkmark$	$\checkmark$	(√)		$\checkmark$	(√)
CNTK	$\checkmark$	$\checkmark$	(√)			(√)
MXNet	$\checkmark$	$\checkmark$	(√)			$\checkmark$
DeepLearning4J	(√)		(√)	$\checkmark$		(√)
Chainer	$\checkmark$					
Caffe2	$\checkmark$	$\checkmark$				

 $(\checkmark)$  : access via wrapper or Keras

• Most supported language: **Python** 

# TensorFlow

- Currently most popular Framework
  - Low-level
  - High-level (Keras)
- Based on two-step approach
  - 1. Define computation graph (including rules for updats etc..)
  - 2. Run graph on any hardware (either for training or prediction)
- Requires some time to get started
  - Concepts are not usual in ,regular' programming
  - Huge amount of functions which are updated quite often
- Documentation and online support is very good
- Additionally offers:
  - Easy to use architectures and trained models
  - Complex, but powerful visualization tool: TensorBoard



# PyTorch

- Growing attention
- Low- and high-level functions
- Uses dynamic graphs:
  - Allows to change computation graph between iterations
  - Makes usage simpler and more like ,usual' programming
- Worse documentation and less functionalities than TF
- Easy access to pretrained models

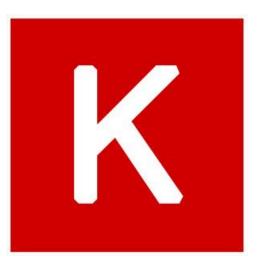




### Keras

- High-level API for:
  - Tensorflow
  - Theano
  - CNTK
  - PlaidML
- Allows training models in very few lines of code
- Currently not so well documented
- Good for beginners, that just want to replicate models
- Often hides relevant information from user
- Will be default API in TensorfFlow 2.0

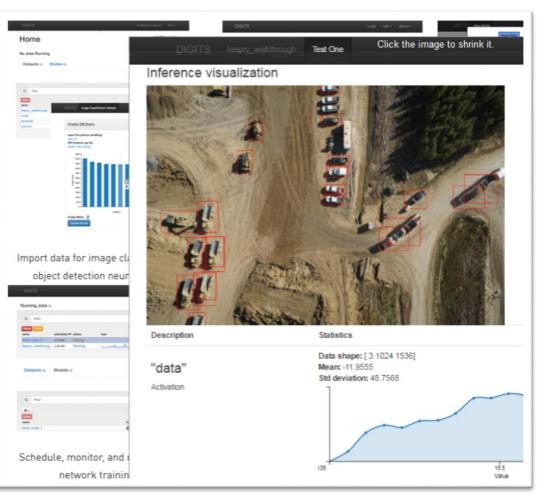




keras.io

# Digits

- Deep Learning GPU Training System (Nvidia)
- Server application / access via browser
- Features:
  - Model choice (including simple modifications)
  - Data overview and processing
  - Training / Evaluation / Prediction
  - Visualization of Graph / Results
- Very high-level!
- Does not require any programming skills



developer.nvidia.com/digits

