Software Packages for Deep Learning

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Outline

Introduction

Python

Torch

Caffe

TensorFlow

MxNET

Comparison

Machine Learning



ML gives computers the ability to learn without being explicitly programmed [Samuel 1959]

ML explores the study and construction of algorithms that can learn from and make predictions on data

Data mining, computational statistics, optimization, ...

Fourth paradigm, big data, deep learning, artificial intelligence

General Tasks of ML



Classification: Inputs are divided into two or more classes, and the learner must produce a model that assigns unseen inputs to one or more (multi-label classification) of these classes

Clustering: Inputs are divided into groups. Unlike in classification, the groups are not known beforehand, making this typically an unsupervised task

Regression: Similar to classification, but the outputs are continuous rather than discrete

Other tasks: density estimation, dimensionality reduction, ...

Packages for General Machine Learning



What is the purpose?

Solving problems from practical applications (user interface)

Developing algorithms and optimizing implementation (development)

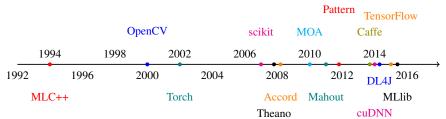
Theoretical analysis for machine learning

What do we want for a ML package?

Easy for new tasks and new network structures (less steep learning curve)

Easy for debugging (with good support and large community)

Performance and scalability



Deep Learning



Comparison



Table: Framework Comparison: Basic information

| Viewpoint | Torch | Caffe | TensorFlow | MXNet |
|--------------------|----------------------------------|--------------------|------------|---------------|
| Started | 2002 | 2013 | 2015 | 2015 |
| Main Developers | Facebook, Twitter, Google, | BVLC (Berkeley) | Google | DMLC |
| License | BSD | BSD | Apache | Apache |
| Core | C/Lua | C++ | C++ | C++ |
| Languages | C/Lua | C++ | Python | Python |
| Supported | Lua | C++/Python | C++/Python | C++/Python |
| Interface | Lua | Matlab | R/Java/Go | R/Julia/Scala |

Comparison



Table: Framework Comparision: Performance

| Viewpoint | Torch | Caffe | TensorFlow | MXNet | |
|------------|------------|-----------|-------------|-------------|--|
| Pretrained | Yes | Yes | No | Yes | |
| Models | 168 | 168 | NO | | |
| Low-level | Good | Good | Fairly good | Very few | |
| Operators | Good | Good | ranny good | very lew | |
| High-level | Good | Good | Good | Good | |
| Support | Good | Good | Good | Good | |
| Speed | Great | Great | Not so good | Excellent | |
| One-GPU | Great | Great | Not so good | Excellent | |
| Memory | Great | Great | Not so good | Excellent | |
| Management | Great | Great | Not so good | Executent | |
| Parallel | Multi-GPU | Multi-GPU | Multi-GPU | Distributed | |
| Support | Multi-OF U | With-OF C | With-OF C | Distributed | |

Python: A general-purpose programming language



Created by Guido van Rossum in 1989 and first released in 1991

Named after "the Monty Python" (British comedy group)

An interpreted language—simple, clear, and readable

Python has many excellent packages for machine learning

The language of choice in introductory programming courses

| Data from Indeed.com 2016 | SQL |
|---------------------------------|------------|
| | JAVA |
| | JAVASCRIPT |
| | C# |
| | PYTHON |
| | C++ |
| | PHP |
| | IOS |
| | RUBY/RAILS |

| Feb | Change \$ | Programming language | \$ | Share \$ | Trends ♦ |
|-----|------------|-------------------------|-----------|----------|-----------------|
| 1 | | Java | | 22.6 % | -1.3 % |
| 2 | | Python | | 14.7 % | +2.8 % |
| 3 | | PHP | | 9.4 % | -1.2 % |
| 4 | | C# | | 8.3 % | -0.3 % |
| 5 | ↑ ↑ | Javascript | | 7.7 % | +0.4 % |
| 6 | | C | | 7.0 % | -0.2 % |
| 7 | ↓↓ | C++ | | 6.9 % | -0.6 % |
| 8 | | Objective-C | | 4.2 % | -0.6 % |
| 9 | 1 | R | | 3.4 % | +0.4 % |
| 10 | ↓ | Swift | | 2.9 % | +0.1 % |

Python for Scientific Computing



Why Python for scientific computing?

Strong introspection¹ capabilities (???What does even mean???)

Full modularity, supporting hierarchical packages

Exception-based error handling

Dynamic data types and automatic memory management

Why consider such a slow language for simulation?

Good for proof-of-concept

Implementation time versus execution time

Code readability and maintenance — short code, fewer bugs

Well-written Python code is "fast enough" for most computational tasks

Time critical parts executed through compiled language or available packages

Built-in Data Structures



```
Numeric types—int, float, complex, ex: a=1, b=1.0, c=1L, d=0xf, e=010, f=1+2j Sequence types—list, tuple, str, dict, ex: g=[3.14, True, 'Yes', [1], (1L,)] + [False] + [None]*3, h=(3.14, True, 'Yes', [1], ()), i='Hello' + "," + "'world!"", j={1: 'int', 'pi': 3.14}
```

Control Flow



If-then-else

```
a = 1
if a > 0:
print "a_is_positive"
elif a=0:
print "a_is_zero"
else:
print "a_is_negative"
```

For loop

• While loop

Functions and Modules



Defining functions

Using modules

There are 3 different ways to use modules. Examples are below. 1. import math This will only introduce the name math into the name space in which the import command was issued. The names within the math module will not appear in the enclosing namespace: they must be accessed through the name math. For example: math.sin(3.14). 2. from math import * This does not introduce the name math into the current namespace. It does however introduce all public names of the math module into the current namespace, directly using: sin(3.14) 3. from math import sin This will only import the sin function from math module and introduce the name sin into the current namespace, but it will not introduce the name math into the current namespace, directly using: sin(3.14)

Programming interface



Example 1



Programming interface



Example 1



Computational graph



Programming interface



Visualization



Example 1



Programming interface



Example 1



Numerical tests



