

# Deep Learning with Python

Dan Guest   Lots of other people

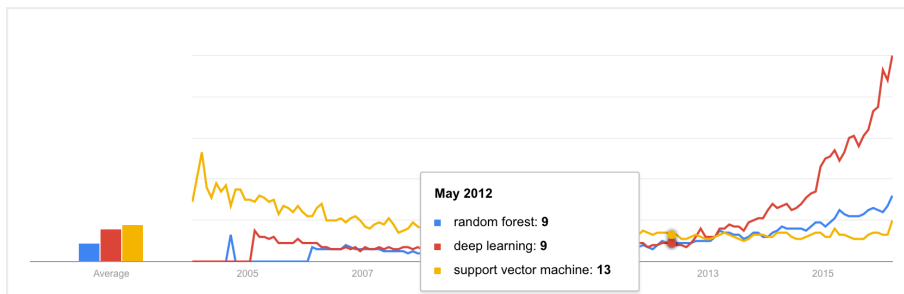
UC Irvine

Lots of other institutes

March 4, 2016

# Setting the stage: What's Cool

## ► Results from Google Trends



- Obviously this is mostly buzz
- But deep learning *has* become very popular

# Deep Learning



What society thinks I do



What my friends think I do



What other computer scientists think I do



What mathematicians think I do



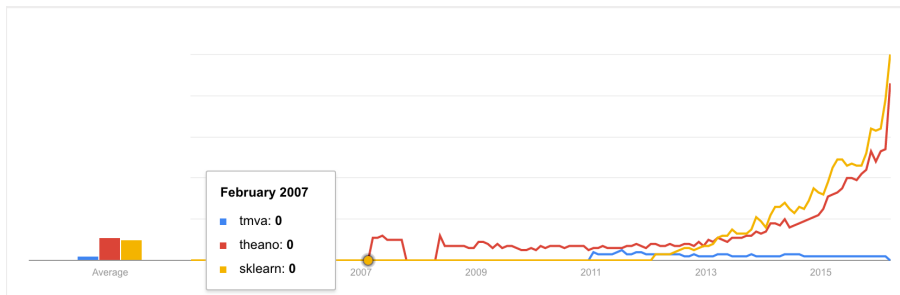
What I think I do

```
from theano import *
```

What I actually do

# Setting the stage: What do people use?

## ► Google Trends



- Also arguably just buzz
- But could predict where the software will go
- We should be careful: we don't want to miss the boat

## What about us (ATLAS in my case)?

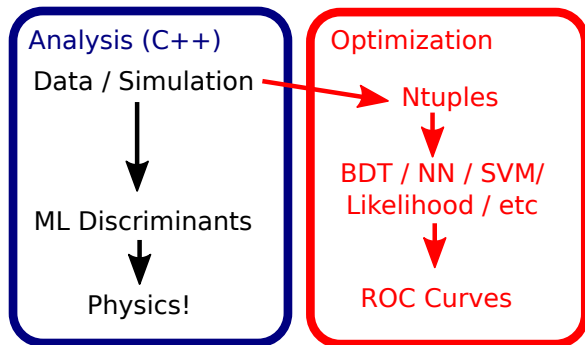
- ▶ Everyone uses TMVA
  - ▶ Lots of talk about adding features
- ▶ But other (python-based) packages have some advantages:
  - ▶ Supported by outside community
  - ▶ We don't have to write yet another framework
  - ▶ Python is easy
  - ▶ **The software is already there**

### So what's holding us back?

- ▶ Inertia: TMVA is already used
- ▶ Lack of “glue packages” to work with our existing code
  - ▶ ROOT doesn't interface well with numpy/hdf5/python

# Our software framework

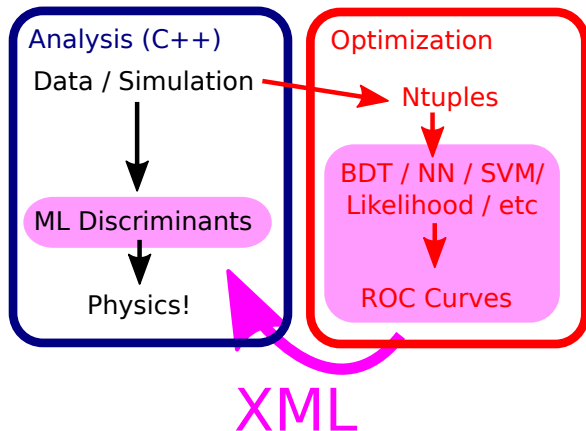
- ▶ Our analysis runs (almost entirely) in C++
- ▶ Optimization is unconstrained



- ▶ Lots of interesting results (Jet Images, etc) that haven't been ported back to C++

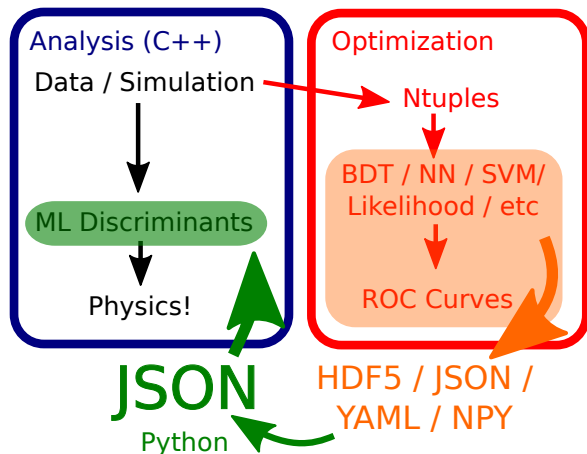
# ROOT Solution

- ▶ TMVA works in C++!
- ▶ We can convert to TMVA xml
- ▶ ...sometimes



- ▶ We can port sklearn BDTs to TMVA

# Our Solution



- ▶ Use whatever we want (Keras)
- ▶ Write the C++ class

- ▶ Use `root_numpy` to convert ROOT to numpy
- ▶ Save NN as JSON, write minimal C++ code for classification



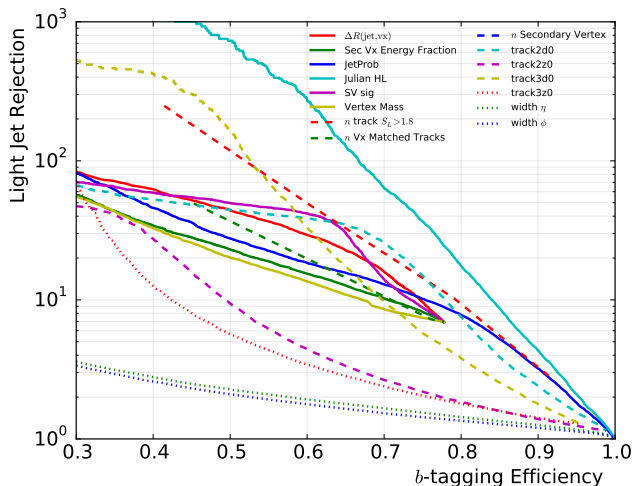
# Lightweight Neural Networks

- ▶ Very powerful (I'm not going to show you a cat)
- ▶ Not very complicated!

$$\mathbf{y} = A_{N-1}f(\cdots A_2f(A_1\mathbf{x} + \mathbf{b}_1) + \mathbf{b}_2\cdots) + \mathbf{b}_{N-1}$$

- ▶ With a few activation functions we cover 90% of use cases
- ▶ Code is on github: <https://github.com/dguest/lw-client>
  - ▶ Thoroughly unimpressive!
  - ▶ Use Eigen, Boost, but **no other dependencies**
  - ▶ Currently supports dense layers, but CovNNs should also be easy
  - ▶ RNNs... maybe a bit more work
    - ▶ First we have to prove they are useful!
  - ▶ Supports AGILEPack output, plan to add Keras, pylearn2, anything else?

# Does it work?



It works ok!

# What Now?

- ▶ Don't write more frameworks, we already have them!
  - ▶ Python-based (numpy, pandas, sklearn, theano) analysis is extremely popular outside HEP
  - ▶ but more “glue package” support would help
- ▶ If you want to use deep learning *right now* we should talk