

# Introduction to Theano

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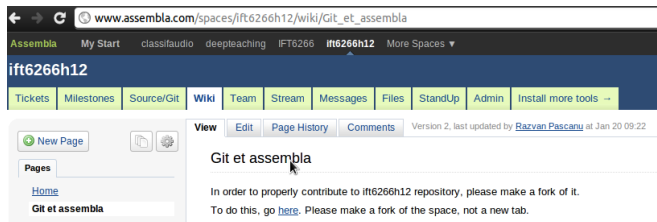
January 22, 2012

- The Tools
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- Theano
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Laboratoire d'Informatique



des Systèmes Adaptatifs  
<http://www.iro.umontreal.ca/~lisa>



- Repo: <http://www.assembla.com/spaces/ift6266h12/wiki>

## TODO:

- Make an assembla account
- Join the team (you can send me the username at `pascanur@iro.umontreal.ca`)
- Create a fork of the repository
- Learn to properly use **Git**

# *RTFM*

<http://deeplearning.net/software/theano/install.html>

- Mandatory

- Assembla [[www.assembla.com](http://www.assembla.com)]
- Git [[www.git-scm.com](http://www.git-scm.com)]
- jobman, jobdispatch [[www.deeplearning.net/software/jobman](http://www.deeplearning.net/software/jobman)]
- NumPy [[www.numpy.org](http://www.numpy.org)]
- python [[www.python.org](http://www.python.org)]
- psql [<http://www.postgresql.org>]
- SciPy [[www.scipy.org](http://www.scipy.org)]
- Theano [[www.deeplearning.net/software/theano](http://www.deeplearning.net/software/theano)]

- Optional

- ipdb [[pypi.python.org/pypi/ipdb](http://pypi.python.org/pypi/ipdb)]
- IPython [[www.ipython.org](http://www.ipython.org)]
- matplotlib [<http://matplotlib.sourceforge.net/>]

```
1      # Compute  $\sum_{i=0}^N x_i^2$ 
2      import theano, theano.tensor as TT
3      x = TT.vector()
4      rval = (x ** 2).sum()
5      fn = theano.function([x], rval,
6                           allow_input_downcast = True)
7
8      assert 14 == fn([1, 2, 3])
```

- execution speed optimization
- symbolic differentiation
- stability optimization

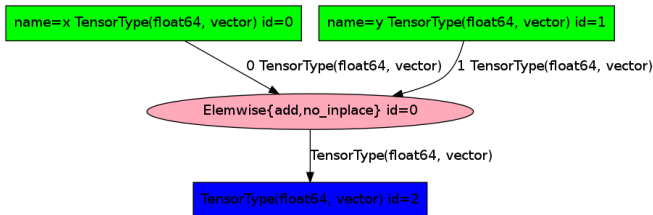
- The fundamental package for scientific computing in Python
- Linear algebra (and more) on multidimensional tensors
- Random number generators
- and much, much more
- all in the friendly python syntax

```
1      import numpy
2      W = numpy.random.uniform(size=(3,3))
3      X = numpy.asarray([1,0,1])
4      print numpy.dot(W,X)
```

See ipython notebook example.

Is as simple as following these simple steps:

- 1 Declare *variables*, specifying the types
- 2 Write down the expression in terms of the variables
- 3 *Compile the function* that will compute the expression
- 4 Call the function on the intended data



See [ipynb notebook example](#).

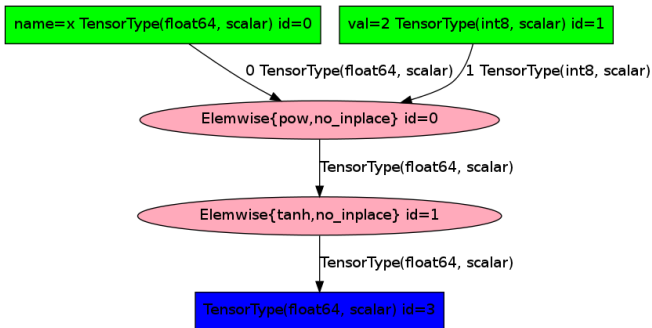


- Theano provides symbolic differentiation
- The computational graph representing the gradients is automatically optimized
- And all this at a single call away: `TT.grad(cost, wrt)`

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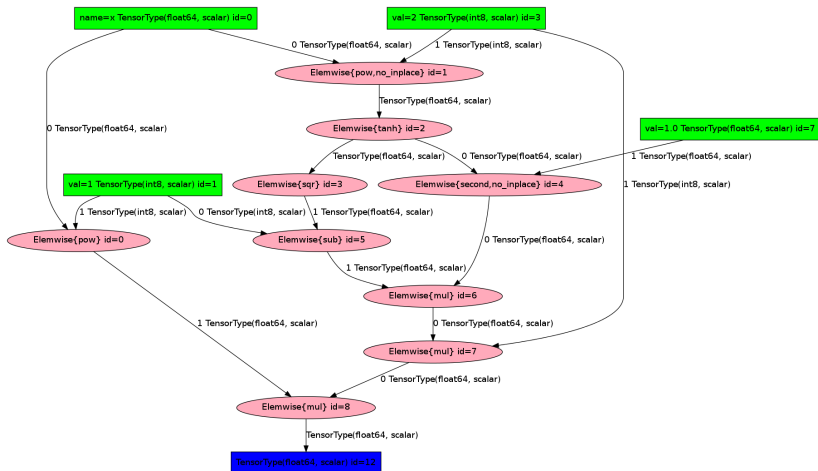
```
1 import theano, theano.tensor as TT
2 x = TT.vector('x')
3 rval = TT.tanh(x ** 2)
4 fn_forward = theano.function([x], rval)
5 gx = TT.grad(rval, x)
6 fn_grad = theano.function([x], gx)
```

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$$\frac{\partial \tanh(x^2)}{\partial x} = \frac{\partial \tanh(x^2)}{\partial x^2} \frac{\partial x^2}{\partial x} = \frac{\partial \tanh(x^2)}{\partial \tanh(x^2)} \frac{\partial \tanh(x^2)}{\partial x^2} \frac{\partial x^2}{\partial x} \quad (1)$$

## Computing derivatives



- Think of them as global variables
- They have a state that stays consistent in between calls
- Especially useful to keep memory on the same device
- Can be updated via `set_value` and `updates` provided to `theano.function`

See ipython notebook example.

```
1  def function(inputs,  
2      outputs=None,  
3      mode=None,  
4      updates=[],  
5      givens=[],  
6      no_default_updates=False,  
7      accept_inplace=False,  
8      name=None,  
9      rebuild_strict=True,  
10     allow_input_downcast=None,  
11     profile=None):
```

- Use **test values**
- Use `theano.printing.Print` to print intermediate results
- Use `theano.printing.pydotprint` to visualize the computational graph
- Run in `DEBUG_MODE`

See ipython notebook example.

See ipython notebook example.

`http://deeplearning.net/tutorial/logreg.html#logreg`



# Questions ?