A Tour of TensorFlow

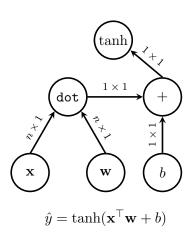


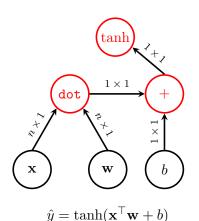
Peter Goldsborough

July 11, 2016

Contents

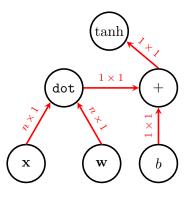
- 1. Computational Paradigms
- 2. Execution Model
- 3. Back Propagation in TensorFlow
- 4. Back Propagation in TensorFlow
- 5. Visualization Tools
- 6. Use Cases
- 7. Walkthrough





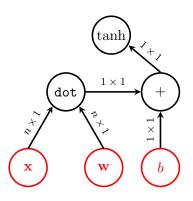
Computational Graphs

1. Operations



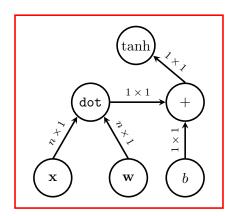
 $\hat{y} = \tanh(\mathbf{x}^{\top}\mathbf{w} + b)$

- 1. Operations
- 2. Tensors



$$\hat{y} = \tanh(\mathbf{x}^{\top}\mathbf{w} + b)$$

- 1. Operations
- 2. Tensors
- 3. Variables



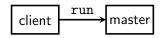
 $\hat{y} = \text{session.run}(\tanh(\mathbf{x}^{\top}\mathbf{w} + b))$

- 1. Operations
- 2. Tensors
- 3. Variables
- 4. Sessions

client

Actors

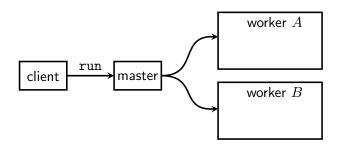
1. Client



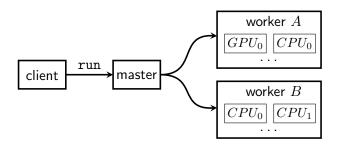
Actors

1. Client

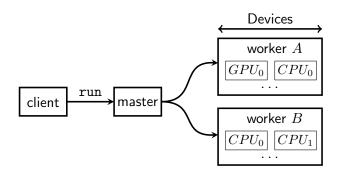
Master



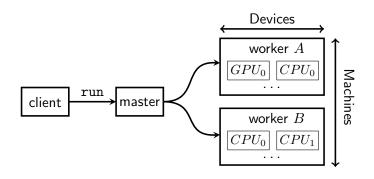
- 1. Client
- MasterWorkers



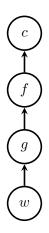
- 1. Client
- 2. Master
- 3. Workers 4. Devices

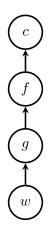


- 1. Client
- 2. Master
- 3. Workers 4. Devices

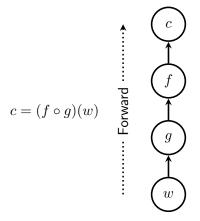


- 1. Client
- 2. Master
- 3. Workers
- 4. Devices

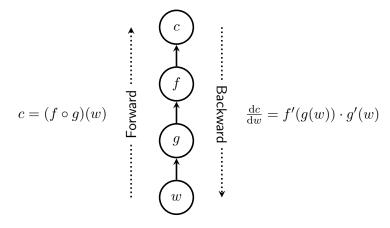




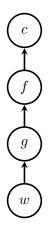
Symbol to Number Differentiation



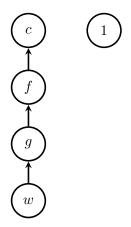
Symbol to Number Differentiation



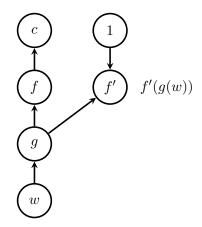
Symbol to Number Differentiation



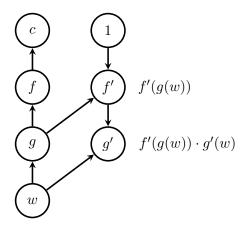
Symbol to Symbol Differentiation



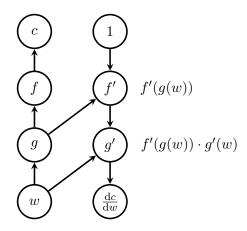
Symbol to Symbol Differentiation



Symbol to Symbol Differentiation



Symbol to Symbol Differentiation



Symbol to Symbol Differentiation

▶ Deep Neural Networks have the tendency of being . . . deep

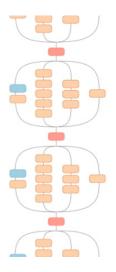
- ▶ Deep Neural Networks have the tendency of being . . . deep
- ► Easy to drown in the complexity of an architecture

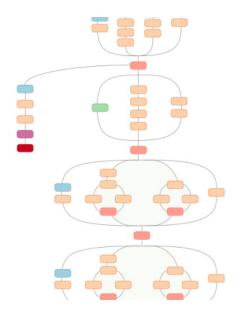
- Deep Neural Networks have the tendency of being . . . deep
- Easy to drown in the complexity of an architecture
- > 36,000 nodes for Google's *Inception* model

- Deep Neural Networks have the tendency of being . . . deep
- Easy to drown in the complexity of an architecture
- > 36,000 nodes for Google's *Inception* model



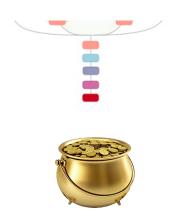








 $Source: \ http://googleresearch.blogspot.de/2016/03/train-your-own-image-classifier-with.html \\$



 $Source: \ http://googleresearch.blogspot.de/2016/03/train-your-own-image-classifier-with.html \\$

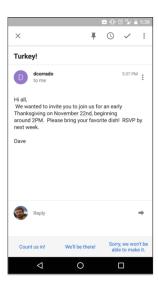
TensorBoard to the Rescue

► Smart email replies in Google *Inbox*



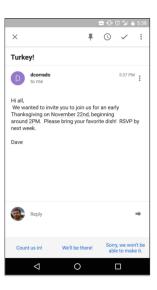
 $Source: \ http://googleresearch.blogspot.de/2015/11/computer-respond-to-this-email.html (computer-respond-to-this-email) (computer$

- ► Smart email replies in Google *Inbox*
- Emails mapped to "thought vectors"



 $Source: \ http://googleresearch.blogspot.de/2015/11/computer-respond-to-this-email.html \\$

- ► Smart email replies in Google *Inbox*
- Emails mapped to "thought vectors"
- LSTMs synthesize valid replies



 $Source: \ http://googleresearch.blogspot.de/2015/11/computer-respond-to-this-email.html (computer-respond-to-this-email) (computer$

► Google DeepMind now using TensorFlow

- Google DeepMind now using TensorFlow
- ► Already for *AlphaGo*



- Google DeepMind now using TensorFlow
- Already for AlphaGo
- According to a DeepMind SWE reasons are:



- Google DeepMind now using TensorFlow
- Already for AlphaGo
- According to a DeepMind SWE reasons are:
 - Python,



- Google DeepMind now using TensorFlow
- Already for AlphaGo
- According to a DeepMind SWE reasons are:
 - Python,
 - Integration with Google Cloud Platform,



- Google DeepMind now using TensorFlow
- Already for AlphaGo
- According to a DeepMind SWE reasons are:
 - Python,
 - Integration with Google Cloud Platform,
 - Support for TPUs,



- Google DeepMind now using TensorFlow
- Already for AlphaGo
- According to a DeepMind SWE reasons are:
 - Python,
 - Integration with Google Cloud Platform,
 - Support for TPUs,
 - Ability to run on many GPUs.



Walkthrough

Thank You