Language Modeling with Torch

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Agenda

- Meeting Torch
 - What is Torch?
 - When should I use Torch?
 - How does Torch work?
- Language Modeling with Torch
 - Understanding Language Models
 - Recurrent Neural Networks
 - Generating sentences



Meeting Torch

Lua, tensors and neural networks



What is Torch? Lua, not Python

Examples of Lua types :

```
String

a = 'hello'

string.sub(a, 'he[l]+)', 'jell')

print(a)

jello

Number

b = 3

print(b + 4)

7
```

- Lua is lightweight, elegant and easy to learn
- Easy to interface with C/CUDA
- Fast for-loops
- Functions as first class citizens, closures

```
Table
c = {1,2,3,key='value'}
c.key = 4
c[2]= 7
print(c)
{1,7,3,key=4}
```



What is Torch? Tensors for linear algebra

```
\begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ \vdots \\ v_n \end{bmatrix}
\begin{bmatrix} v_{11}, v_{12}, v_{13}, \dots, v_{1m} \\ v_{21}, v_{22}, v_{23}, \dots, v_{2m} \\ v_{31}, v_{32}, v_{33}, \dots, v_{3m} \\ \vdots & \vdots & \ddots & \vdots \\ v_{n1}, v_{n2}, v_{n3}, \dots, v_{nm} \end{bmatrix}
```

- Tensors are N-dimension arrays
 - Initialize and matrix-matrix multiply two tensors:

```
a = torch.Tensor(2,3) b = torch.Tensor{\{1\},\{2\},\{3\}\}} a:uniform(0,1) print(b)

print(a) 1

0.6187 0.8752 0.1496 2

0.6979 0.8230 0.1439 [torch.DoubleTensor of size 2x3] [torch.DoubleTensor of size 3x1]
```

c = torch.mm(a,b) c:mm(a,b) c = a * b

- Basic linear algebra sub-routines (BLAS)
- Sub-tensor extraction, etc.
- Can be run on GPU/CPU
- Different types: DoubleTensor, FloatTensor, LongTensor, CudaTensor, etc.



What is Torch? Neural Networks for deep learning

Train neural networks using back-propagation (gradient descent + chain rule)

```
-- model
                              -- data sample
                                                           -- forward
mlp = nn.Sequential()
                              input = torch.randn(3)
                                                           output = mlp:forward(input)
                              target = torch.randn(2)
 :add(nn.Linear(3,4))
                                                           loss = mse:forward(output, target)
 :add(nn.Tanh())
                                                           -- backward
 :add(nn.Linear(4,2))
                                                           gradOutput = mse:backward(output, target)
 :add(nn.Sigmoid())
                                                           mlp:zeroGradParameters()
-- loss function
                                                           mlp:backward(input, gradOutput)
mse = nn.MSECriterion()
                                                           -- update
                                                           mlp:updateParameters(0.1)
```

- Automatic gradient differentiation
 - No compilation (TensorFlow, Theano)
 - Imperative programming (easier to debug than symbolic graphs)
- Extend by implementing nn.Modules or nn.Criterions



When should Torch be used? Research and production

- Deep learning research
 - Assemble existing Modules and Criterions
 - Or implement your own
 - Easy to debug (no compilation or symbolic graph)
 - Lots of packages to pick from (torch, nn, optim, image, dataload, rnn, etc.)
- Production environment
 - Lua was designed for embedded systems
 - Servers (GPU/CPU, Ubuntu/Linux, Max OS X)
 - Smart phones (Android, iOS)
- Fast execution
 - glample/rnn-benchmarks
 - soumith/convnet-benchmarks

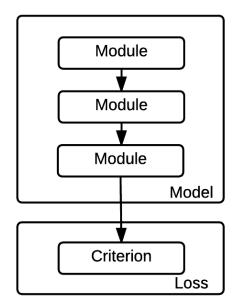


How does Torch work? Mental model

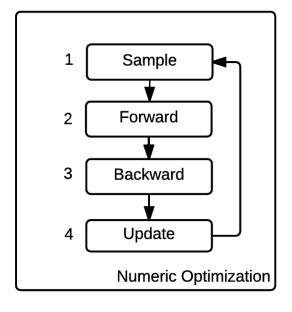
Prepare Dataset

Loader Files Dataset

Define Model and Loss



Train Model



Language Models

Generating sentences using recurrent neural networks



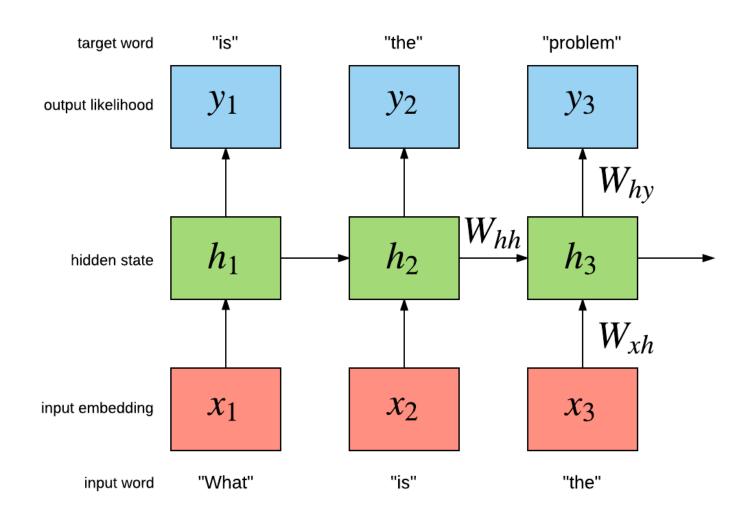
Understanding language models

- Language model dataset is a corpus of text samples
 - "Why does it always rain on Saturdays?"
- Maximize the likelihood of a word given previous words: $P(w_t|w_1,...,w_{t-1})$
- Applications:
 - ► Generate sentences (sample next word given previous words)
 - Auto-complete
 - Measure the probability of a sentence (e.g. does this translation make sense?)

$$P(w_1, w_2, ..., w_T) = \prod_{t=1}^{T} P(w_t | w_1, ..., w_{t-1})$$



Recurrent Neural Networks



Recurrent Neural Networks

- Model architectures
 - Simple Recurrent Neural Networks (Simple RNN)
 - Long Short Term Memory (LSTM)
 - Gated Recurrent Units (GRU)
- Used to model sequential data
 - Video frames, text, time-series
- Different applications
 - Language modeling, sentiment analysis, machine translation
- ► Element-Research/rnn package provides RNNs for Torch
 - Tutorials, documentation, models, criterions and examples



Generating sentences

- Recursively sample sentences (one word at a time)
- <S> The company said its net profit rose to \$ 289 million, or 96 cents per share, in the three months ended on March 31 compared with \$ 173 million, or \$ 0.68 a share, a year ago . </\$>
- <S> But I 've been a bit disappointed with our performance, " said Wenger .
- <S> The first is an even bigger problem .
- <S> The next big thing for him is he will be able to tell the world he is thinking about his future .
- <S> The new rules have been added to the legislation so that they don't have to be approved for public use . </\$>
- <\$> The Pentagon 's top counter-terrorism official, who has been in charge of a new system of intelligence collection and inspection, wrote in an e-mail message that while the new system could be easily implemented, it remains an option.
- <S> "I was trying to get a glass of water . <S> Later he was driven to a nearby house where he was later found to be severely ill .
- 40GB model learns to generate coherent sentences
- Trained on Google 1-billion words dataset



Multi-GPU Multi-Layer LSTM

```
nn.Serial @ nn.Sequential {
  [input -> (1) -> (2) -> (3) -> output]
  (1): nn.ParallelTable {
    input
| `-> (1): nn.Sequential {
                   finput -> (1) -> (2) -> (3) -> (4) -> (5) -> (6) -> (7) -> (8) -> (9) -> (10) -> (11) -> (12) -> output]
                   (1): nn.Convert
                  (2): nn.GPU(2) @ nn.Concat {
                          `-> (1): nn.GPU(1) @ nn.LookupTableMaskZero
`-> (2): nn.GPU(2) @ nn.LookupTableMaskZero
                         ... -> output
                 (3): nn.GPU(2) @ nn.Dropout(0.2, busy)
(4): nn.GPU(1) @ nn.SeqLSTM
(5): nn.GPU(1) @ nn.Dropout(0.2, busy)
(6): nn.GPU(1) @ nn.SeqLSTM
(7): nn.GPU(1) @ nn.Dropout(0.2, busy)
(8): nn.GPU(2) @ nn.SeqLSTM
(9): nn.GPU(2) @ nn.Dropout(0.2, busy)
(10): nn.GPU(2) @ nn.SeqLSTM
(11): nn.GPU(2) @ nn.SeqLSTM
(12): nn.GPU(3) @ nn.SplitTable
            ·-> (2): nn.ldentity
          ... -> output
    (3): nn.GPU(3) @ nn.Sequencer @ nn.Recursor @ nn.MaskZero @ nn.NCEModule(2048 -> 793471)
```



For more info:

http://torch.ch

https://github.com/Element-Research/rnn

