# Recurrent Neural Networks

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### Outline

- Problem
- Method
  - Definitions
  - Implementation
- Results
- Conclusion

#### Problem

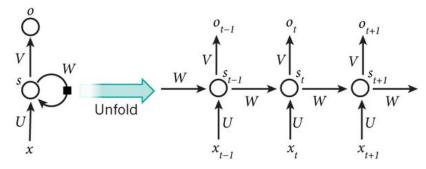
- Recurrent neural networks (RNN) are powerful & popular models that have attracted great interest and shown promise in many NLP tasks.
  - RNN based language models:
    - Common uses include: speech recognition, machine translation, generating image descriptions etc.
    - Allows to score arbitrary sentences based on how likely they are to occur.
    - Also allows us to generate new text... Why? Because it is fun!
  - Aim of this project is to use RNN in order to generate Turkish text based on different training datasets.

### What is recurrent neural network?

- In a traditional neural network, it is assumed that all inputs are independent of each other. But if you want to predict the next word in a sentence, we better know which words **came before it**.
  - RNN are recurrent because they perform the same task for every element of the sequence with the output being depended on the previous computations.
  - One can assume they have a "memory"...
- RNNs have two sources of input: the present and recent past
  - There is information in the input sequence itself, and RNNs use it to perform tasks that other networks can't.

#### How does it work?

- RNN shares the same parameters
   (U,V,W) across all steps
  - Unlike traditional deep neural networks which uses different params at each layer
  - Reduced # of params we need to learn!



 $X_t$  is the input at time step t.

 $s_t$  is the hidden state which is calc. based on the prev. hidden state and the input.  $s_t = f(Ux_t + Ws_{t-1})$ 

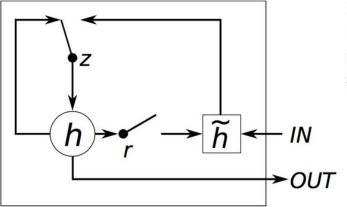
- Training RNNs include backpropagation like others
  - But the gradient at each output depends also on the previous time steps... (Backpropagation through time/BPTT)
  - Vanilla RNNs trained with BPTT have difficulties learning long-term dependencies (that are far apart)
    - Vanishing or exploding gradient problem!

### LSTM and GRU What are they?

- Our goal is to calculate the gradients of the error (w.r.t *U,V* and *W*) and then learn good parameters using Stochastic Gradient Descent.
  - o Gradient values are shrinking exponentially. Eventually vanishing completely...
  - That means those steps doesn't contribute to what we are learning.
  - That's problematic because the meaning of a sentence is often determined by words that aren't very close.
    - Asd
  - Long Short-Term Memory (LSTM) and Gated Recurrent Unit (GRU) are specifically designed to overcome this problem.
    - Same architecture but different update function to compute the hidden state.
    - In this project, GRU have been implemented.

### **Gated Recurrent Units**

- GRU has two gates, a reset gate *r*, and an update gate *z*.
  - Reset gate determines how to combine the new input with the previous memory
  - Update gate defines how much of the previous memory to keep around.
  - (If we set the r to all 1's and z to all 0's = vanilla RNN)



$$z = \sigma(x_t U^z + s_{t-1} W^z)$$

$$r = \sigma(x_t U^r + s_{t-1} W^r)$$

$$h = \tanh(x_t U^h + (s_{t-1} \circ r) W^h)$$

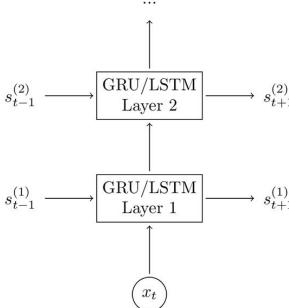
$$s_t = (1 - z) \circ h + z \circ s_{t-1}$$

They are called gates because sigmoid function  $\sigma$  squashes the values of vectors between 0 and 1.

### **Gated Recurrent Units**

Adding a second layer to the network allows our model to capture higher-level interactions

- But they may lead to overfitting
- Performance considerations...



### Implementation

- Download lots of Turkish text (Approx. 800MB of Turkish Wikipedia dump and various e-books)
- 2. Clean text from markup tags, metadata headers etc. (2.195.043 sentences)
- 3. Tokenize text into words (1.041.577 unique word tokens)
- 4. Remove infrequent words and build vocabulary with different sizes (2.000, 8.000, 10.000 etc.)
- 5. Prepend/append <start> and <end> tokens
- 6. Build training data matrices
  - a. The input to RNNs are vectors, not strings. So we map words to indices. (later, we will use them as one-hot vectors)

### Implementation

- 7. Initialize the parameters U, V and W to small random values...
- 8. For each word in the input sentence, forward propagation (GRU, LSTM) make predictions representing the probabilities of the next word.
  - a. Predictions are random at first. Need to measure the errors it made! **Cross-entropy loss function** ( $o_n$  output predictions,  $y_n$  is the correct words)
  - b. Remember, the goal is to find *U*, *V* and *W* that minimize the loss.
    - i. Then calc. SGD to nudge the parameters into the right direction  $L(y,o) = -\frac{1}{N} \sum_{n} y_n \log o_n$
    - ii. SGD also needs learning rate (how big of a step we should make with the gradient values calculated by BPTT)
    - iii. Decrease learning rate if the loss continues to increase!
- 9. For each *N* training examples
  - a. Generate sentences using model parameters at this step.
  - b. Also, save model parameters into a file so that we can use them once we know the model is trained enough.

### What did I use?

- Python 2.7
- NLTK library
  - Work tokenization etc.
- Theano
  - To perform operations on GPU
  - Provides efficient implementations for most of the mathematical functions

## Simulation environment & parameters

Operating system	Debian GNU/Linux 6 (64bit)
CPU	8 GHz
Memory	16GB RAM memory
Vocabulary size	6.000, 8.000, 10.000
Number of epochs	20, 40
Hidden dimension size	100, 120
Number of GRU layers	2
SGD learning rate	0.001

#### Results

- Most of the generated text is nonsense...
  - √ olarak bir gerekir .
  - √ 1967 yılında milli
  - √ bir mecmua).
  - √ köyün yeşil adlı sebebiyle tipi takım göç
  - √ dolayısıyla ilçesinin bilinir .
  - √ milletvekili yataktan dergisinde mektepten
  - √ hiçbir oğlundan kendi söylemeyi :
  - √ yaptığı doğru herhalde evladım ve sevgili arkasını bende .
  - √ anlıyordum söylemişti yeni ?
  - √ hayretle bıraktım olsa diye haydi ...
  - √ gece buyurun ... dedi . ...
  - √ ankara'ya gelirim ... ... tekrar

#### Results

- ...but some of them are getting meaningful and grammatically correct:
  - √ içme suyu şebekesi
  - √ uzun 2010 yılında .
  - √ köyün ekonomisi tarım ve hayvancılığa dayalıdır .
  - √ ptt şubesi ve ptt acentesi yoktur .
  - √ 15 mart 2010 tarihinde yapılan bir nüfus sayımına göre şehrin nüfusunun ve
  - √ köye ulaşımı sağlayan yol asfalt olup
  - √ köyde ilköğretim okulu vardır.
  - ✓ mahallenin adının nereden geldiği ve geçmişi hakkında bilgi yoktur .
- Basically, it just needs parameter fine-tuning & more time to train...

### Conclusion

- RNN (specifically *GRU*) **implementation in Python**.
- A pre-trained language model which can be used to generate TR text.
  - Still in progress... Training is a long process with this amount of data.
- Turkish datasets or the scripts to generate them
  - Along with the source code, they can be accessed at https://github.com/emrekgn/tr-rnn
- Future work
  - Word2vec or Glove can be used to further optimize results

Thank you for listening!