

REINVENT THE WHEEL - GRU

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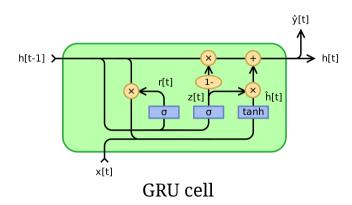


PROBLEM DESCRIPTION AND INTRODUCTION

A Gated recurrent unit (GRU) is a variant of RNN architecture for recurrent networks. It uses gating mechanisms to control the flow of information between its nodes.

This helps retaining long-term dependencies and combat the "short-term memory" issue of vanilla RNNs.

GRU networks outperform LSTM networks thanks to smaller number of parameters



2

Implementation

This project implements a simple character-predicting GRU network. It is implemented in C++ with no external dependencies.

Most hyperparameters as well as training data can be changed using command-line options.

class Matrix

- mathematical operations on matrices
- class GRU
 - the main network class
 - uses Xavier initialization for its parameters
 - implements forward and backward passes

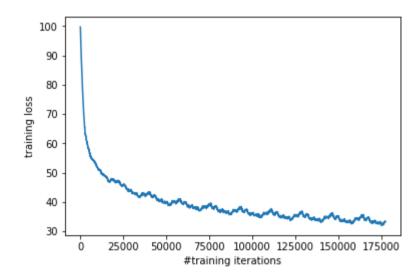
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Training and evaluation

The project's functionality can be demonstrated by running the training on a sample dataset. For this purpose the Dinosaur list dataset was chosen, which contains **1143** unique names.

Training details:

- sequence of 20 characters
- 100 epochs
- learning rate 0.001



4

Results

The network is fully capable of learning and generating valid results:

austrosaurus utarapateks rapator venasugngowx sinornitholog aladromeus toriosaurus zenimaceitisaurus varaikenoplota colopteryx velocirapteros omaisaurus steronthodesceus zhenyurocoese

As we can see, it learned to properly alternate vowels and consonants, and use suffixes such as -saurus, -us, or -raptor.

Resources

- https://colah.github.io/posts/2015-08-Understanding-LSTMs
- https://en.wikipedia.org/wiki/Gated_recurrent_unit
- https://d2l.ai/chapter_recurrent-modern/gru.html
- https://www.kaggle.com/kumazaki98/dinosaur-list