## Deep Learning in Data Science, DD2424

# Short report on lab assignment 4

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#### 1 Introduction

In this assignment, I will train an RNN to synthesize English text character by character. I will train a vanilla RNN with outputs, as described in lecture 9, using the text from the book The Goblet of Fire by J.K. Rowling. The variation of SGD I will use for the optimization will be AdaGrad.

#### 2 Read in the data

The first step is to read in the training data from the text file.

In addition to the raw text, a unique set of characters present in the text is identified. This set forms the basis of the character-level encoding scheme used to transform the text data into a numerical format, a prerequisite for machine learning algorithms. To facilitate this transformation, two conversion functions are provided: char\_to\_idx and idx\_to\_char. The former function converts a character into a one-hot encoded numerical representation, while the latter function serves as the inverse first one, converting the numerical representation back into the corresponding character.

#### 3 Build a vanilla RNN

After I defined the functions to load and prepare the data, I wrote the code to implement a vanilla RNN. I defined the following functions:

- initialization() to randomly initialize the parameters of the network
- synthesize\_text() to synthesize text character by character
- forward() which implements the forward pass of the network
- computeCost() which computes the loss the network
- computeGrads() which computes the gradients through analytical calculations
- computeGradsNum() which computes the numerical value of the gradients
- fit() which allows me to train the network and implement the backward pass

Before starting to train my network I checked the correctness of my functions by computing the differences between the analytical and numerical gradients.

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Figure 1: Gradients check for two epochs first (left), second (right)

## 4 Train your RNN using AdaGrad

After I verified the correctness of my implementation, I proceeded to train my network. I trained my RNN for 7 epochs and I reported below the smooth loss and the synthesized every 10000 iterations.

Iter	Smooth Loss	Synthesized Text
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		KOha_rHX OyA)PURO•?HMAtnCbOC'LMhOwU-HRSPg
		zb 7E: DDB_EGeMnEa $\tilde{A}_{4}^{1}$ 0GE_7dHGYfR/-YNQniH)LR1(si
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Table 1: Iterations, Smooth Loss, and Synthesized Text for 7 epoch of training

I included also the plot representing the evolution of the smooth loss during the training:

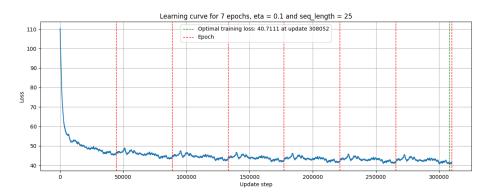


Figure 2: Loss plot of the vanilla RNN

Finally, I report a 1000-character-long text synthesized by my best network:

#### Final synthesized text

e mloling fotil, Gellaw the grow wot his been weont cokents in cullun wazt, commomse-had Hermoslien uf Hagd gotwrefione unt, the mow the now mast,". . . Herding sew his hand." The tave aronk ear ene he pog romenon hive, reaget of at them.. . tof the Doubl want his is!" "Prack. "He said you and was said Hery waig and coidy. "Anthiming shipped nat weveation a the pariting the you wopply in how him." "Oo and dirnust. "De wass think agrear. Mug, gemelf. Harry. Fometerdhy singly wearts is ell, porking out Murse said it hims to wither as. I't eyall Vome?" "Ther, Ron. He ass the Snappen. Ited fronged folf-perd in. The ubuot! I you dayestione linkinl on the so'll soom tork faile antsen everar, she sowe dizard." "Arted the Hed weed Dumm, a sumbreemen, morst'rb!" "I lookevedowed notilf Harry nets and luverreccend so other at sweet to Neghed to betss it, you donce alt. "I go. . . . TTHough a dont engraltlisc. "Iting he't to and thicking mald. Harry cisir. he whough ere, moun the gree f

## 5 Bonus Points

#### 5.1 Adam optimizer

In this section, I tried to modify my code in order to use the Adam optimizer instead of AdaGrad. I trained my network for 2 epochs using AdaGrad and then Adam optimizer. I did not notice a great difference in performance, but the execution time was lower.

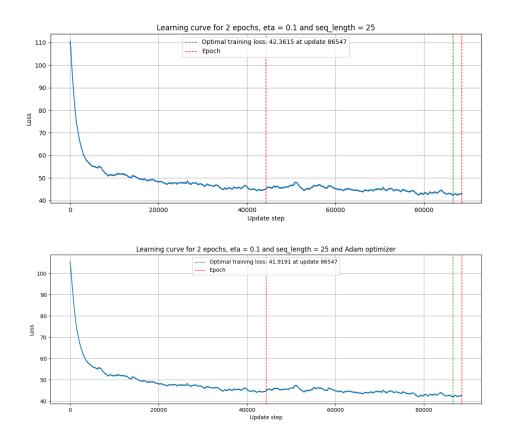


Figure 3: Loss plot of the vanilla RNN using AdaGrad (top) and Adam (bottom)