

# Next-word Prediction with Approach Federated Learning

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**Abstract**—When a user is typing text on a mobile device it can be useful to suggest the next word as this will optimize typing time and also avoid possible errors. However, this data has private information, which limits its movement to a centralized environment. In this paper we will demonstrate how to predict the next word while guaranteeing users privacy without moving the data around.

**Index Terms**—Federated Learning, Text Generation, Data Privacy

## I. INTRODUCTION

This work has how goals to predict the next word, ensuring data privacy. Shakespeare pieces will be used as input data. These will be obtained, described, pre-processed and explored for a better understanding. Next, the federation environment will be created where each character in Shakespeare's works will be a participating user, and their speeches will be the input dataset in the prediction model. From this scenario, in which data is only found on mobile devices, Federated Learning will be used to perform the model training in a shared way without moving the data to a centralized environment. For this, steps will be carried out to load a pre-trained global model from a central server, data pre-processing and model training on the user's own device. Then the model parameters will be forwarded to the central server to update the federated average and train the global model.

## II. GETTING AND DESCRIBE DATA

Data were obtained from the Gutenberg project which has all Shakespeare play compiled into a single file. This text file is 12,788 lines, 1,134,708 tokens, 5,465,129 characters and 91 unique characteres. The data has many problems with formatting, in addition, in some pieces there is information about the Gutenberg project and usage license. This information is not valuable to predict the next word from the purpose of this work.

## III. PREPROCESSING TEXT

To begin with, it is necessary to remove pieces unnecessary texts such as information about the Gutenberg project, terms of use license and poems. This data cleansing will help to get text with greater predictive value for the next-word on into Shakespeare's plays. Natural Language Processing (NLP) techniques for data processing will be applied.

### A. Bag-of-Words

A bag-of-words (BoW) is a model maps a document into a vector as  $v = [x_1, x_2, \dots, x_n]$ , where  $x_i$  denotes the occurrence of the  $i$ th word in basic terms [1]. In this case a BoW was generated in resulting a sparse array of data. To better visualize the BoW, only the twenty most common tokens in all parts were keep.

### B. Term Frequency and Inverse Document Frequency

Term Frequency and Inverse Document Frequency (TF-IDF) determine the importance of a feature in a text document [2]. In this approach, instead of filling the document vectors with the raw count, as in the BoW approach, it is filled with a score. This technique comes in very handy for the next word prediction problem because it calculates the rarity of a word in a text.

### C. Latent Dirichlet Allocation

As mentioned in [3] the latent Dirichlet allocation (LDA) is a topic model which discovers latent structures in data as a set of topics from a collection of texts. In this work, the ten most important topics containing twenty words each were calculated.

## IV. DATA EXPLORATION

Using the BoW technique, the first twenty unigrams and bigrams of the text were calculated and how result, the figures 1 and 2 generated.

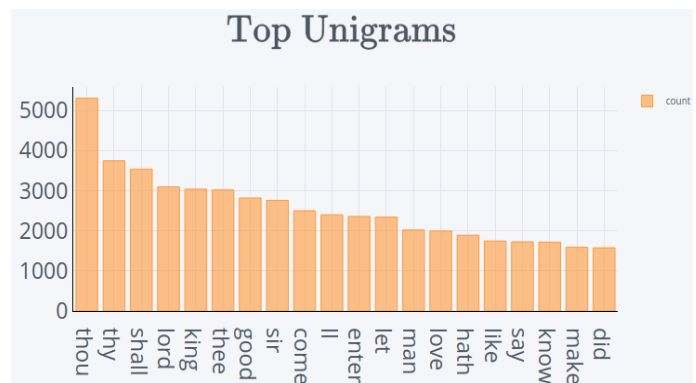


Fig. 1. The distribution of top unigrams after data cleansing.

