


# Implementation of Auto-Complete with MapReduce

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## Overview

## Visualization

The following pictures show screenshots of the visulization of autocomplte.



### Autocomplete using PHP/Ajax and jQuery

Type a word :

- with the
- with a
- with his
- will i
- will not
- with that
- will be
- with that word
- wise and
- wife and

See [www.qianqianshan.com](http://www.qianqianshan.com) for more details.



### Autocomplete using PHP/Ajax and jQuery

Type a word :

- with the
- with a
- with his
- with that
- with that word
- within the
- within a

See [www.qianqianshan.com](http://www.qianqianshan.com) for more details.

## Preliminaries

- Operating system: The operating system is Ubuntu 18.04 LTS (Bionic Beaver).
- Docker container (<https://www.docker.com/why-docker>) version 18.09.5:

Docker enables users to bundle an application together with its preferred execution environment to be executed on a target machine. (<https://hadoop.apache.org/>)

See <https://docs.docker.com/install/> for more details on docker installation. Frequently used commands of docker can be found at <https://docs.docker.com/get-started/>.

- Hadoop cluster is launched within docker containers by following instructions at <https://devhub.io/repos/joway-hadoop-cluster-docker>, 3 containers with 1 master and 2 slaves will be started.

Output by running script `start-container.sh`:

```
start hadoop-master container...
start hadoop-slave1 container...
start hadoop-slave2 container...
```

- Visualize the auto-complete process with PHP, MySQL and jQuery as shown at <http://www.bewebdeveloper.com/tutorial-about-autocomplete-using-php-mysql-and-jquery>. LAMP (Linux, Apache HTTP server, MySQL, PHP) software bundle is used for the visulization (Wiki).

## N-Gram Model

N-gram model is a language model which predicts the next word from the previous  $(N - 1)$  words, where N-gram is an N-token sequence of words such as "I like apple"(trigram), "New York"(bigram) and so on. The intuition of N-gram model is that we can *approximate* the probability of a word given the entire history by the probability given the last  $(N - 1)$  words.

### Notation

Denote a sequence of  $n$  words as

$$w_1^n = w_1, w_2, \dots, w_n$$

and we have a computer-readable collection of text called *corpus*,  $C$ , that can be used to count frequencies of words or sequences. Denote the count of N-gram of words  $w_1, \dots, w_N$  as  $C(w_1, \dots, w_N)$ .

## Maximum Likelihood estimation (MLE) of Probability of Next Word

We obtain the MLE estimate of the probabilities of N-gram model by normalizing the counts from corpus  $C$ , that is, the ratio of counts of sequence  $w_{n-N+1}^{n-1}w_n$  over that of  $w_{n-N+1}^{n-1}$ ,

$$P(w_n|w_{n-N+1}^{n-1}) = \frac{C(w_{n-N+1}^{n-1}w_n)}{C(w_{n-N+1}^{n-1})}$$

## N-Gram Model Implementation

There are three main steps to implement N-gram model:

- Calculate the counts of each n-gram for  $n = 1, 2, \dots, N$  based on raw data stored on hadoop distributed file systems (hdfs)
- Calculate the counts of each following word given the previous word sequence and save the top  $k$  frequent following words for each possible combination of previous word sequence into MySQL database.

## Optimization of N-Gram Model in MapReduce

In the N-gram model, we predict the next word based on MLE of probabilities of words, this can be simplified by justing checking which word has the highest frequency to show up next given the previous  $(N - 1)$  words. For example, users input *I like*, and we have

$$C(apple|I \quad like) = 500$$

and

$$C(banana|I \quad like) = 200$$

as the denominator  $C(Ilike)$  is the same for the above two cases, the frequencies reflect their corresponding probabilities. Then we may want to recommend the trigram sequence as *I like apple* as first.

## Discussion

- N-gram model is sensitive to the corpus we are using. The probabilities are often an implication of specific facts of the used corpus.
- N-gram model may assign some N-grams zero probability due to the limited corpus we are using. One can use smoothing methods to assign these kinds of N-grams non-zero probability, for example, Laplace smoothing, Good-Turing discounting and so on.

## Run Auto-Complete in Hadoop

### Set up LAMP

- `sudo apt-get update`
- `sudo apt-get -y install wget screen unzip`, install wget, screen and unzip if not installed.

- `screen -S lamp`, run the installation process in background to avoid unexpected interruption.
- `git clone https://github.com/teddysun/lamp.git`, download lamp. See more details at <https://github.com/teddysun/lamp>.
- `cd lamp`
- `chmod +x *.sh`, change script mode to executable.
- `./lamp.sh`, install lamp.

Other installation options are shown at <https://www.linode.com/docs/web-servers/lamp/install-lamp-stack-on-ubuntu-18-04/>.

## Set up MySQL

### Create a database

- CTL + ALT + T to open a new terminal
- `cd /usr/local/mysql/bin/`
- `sudo ./mysql -uroot -p`, run as root password is also required to enter MySQL monitor, use the password you set up when installing LAMP.
- `create database test;`, create a database called `test`. Database can be dropped with `drop database test;`
- `use test;`, switch to test database
- `create table output(starting_phrase VARCHAR(250), following_word VARCHAR(250), count INT);`, create a table called `output` with three columns: `starting_phrase`, `following_word`, `count`. Table can be deleted with `drop table output;`.
- `GRANT ALL PRIVILEGES ON *.* TO 'root'@ '%' IDENTIFIED BY 'password' WITH GRANT OPTION; FLUSH PRIVILEGES;`, grant MySQL remote login with the `password` replaced by the password of your MySQL server.
- `SHOW VARIABLES WHERE Variable_name = 'port';` check the port number to be used later.

### Set up Hadoop in docker

- `./start-container.sh`, start docker
- `./start-hadoop.sh`, start hadoop
- `cd src`
- `wget http://central.maven.org/maven2/mysql/mysql-connector-java/5.1.39/mysql-connector-java-5.1.39-bin.jar` download `mysql-connector-java-5.1.39-bin.jar` from <https://mvnrepository.com/artifact/mysql/mysql-connector-java/5.1.39> so we can write to local mysql after running MapReduce jobs.
- `hdfs dfs -mkdir /mysql`, create a folder `mysql` in hdfs within hadoop.
- `hdfs dfs -put mysql-connector-java-*.jar /mysql/`, copy the previous downloaded `mysql-connector*.jar` to `mysql` folder of hdfs.
- `wget --no-check-certificate https://github.com/QianqianShan/Auto_Complete/archive/v1.0.tar.gz`, download source code and corpus.
- `tar -xzf v1.0.tar.gz`, unzip the file, there will be a folder `Auto_Complete-1.0`.
- `cd Auto_Complete-1.0`.

- `hdfs dfs -mkdir -p input`
- `hdfs dfs -rm -r /output`, if there exists an output folder.
- `hdfs dfs -put bookList/* input/`

**Set up your IP, password and path to mysql connector in Driver.java:**

1 - local\_ip\_address: can be found by

- Method 1: `hostname -I`
- Method 2: `ifconfig | grep inet | grep broadcast`

2 - MySQL\_port:

- `SHOW VARIABLES WHERE Variable_name = 'port'`; check the port number to be used later.

3 - your\_password: the password you set when you set up your MySQL database.

4 - hdfs\_path\_to\_mysql-connector: for example, `/root/src/mysql-connector-java-5.1.39.jar`.

### Run Auto-Complete

- `hadoop com.sun.tools.javac.Main *.java`
- `jar cf ngram.jar *.class`
- `hadoop jar ngram.jar Driver input /output 2 3 4`, 2 3 4 are arguments passed to code, 2 is gram size, 3 is threshold size, 4 is following word size.
- Once the MapReduce job is done, use `select * from output limit 10`; to check if your local database has been written with data.

## References

1. <https://www.docker.com>
2. <http://hadoop.apache.org/>
3. <https://devhub.io/repos/joway-hadoop-cluster-docker>
4. <https://github.com/teddysun/lamp>
5. Speech and Language Processing, Daniel Jurafsky & James H. Martin (Chapter 3)
6. MapReduce Tutorial on <https://hadoop.apache.org>