R Notebook

Code ▼

Step 1 - Load library and source code

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```
if (!require("devtools")){install.packages("devtools")}
if (!require("pacman")) {
   ## devtools is required
  library(devtools)
  install github("trinker/pacman")
if (!require("RWeka")){install.packages("RWeka")}
if (!require("tm")) {install.packages("tm")}
if (!require("tidytext")) {install.packages("tidytext")}
if (!require("tidyverse")) {install.packages("tidyverse")}
if (!require("DT")) {install.packages("DT")}
library(devtools)
library(pacman)
library(tm)
library(tidytext)
library(tidyverse)
library(DT)
library(RWeka)
pacman::p_load(knitr, readr, stringr, tesseract, vecsets)
#source('../lib/ifCleanToken.R')
file name vec <- list.files("../data/ground truth") #100 files in total
```

Step 2 - read the files and conduct Tesseract OCR

Although we have processed the Tesseract OCR and save the output txt files in the data folder, we include this chunk of code in order to make clear the whole pipeline to you.

```
# for(i in c(1:length(file_name_vec))){
    current_file_name <- sub(".txt","",file_name_vec[i])</pre>
    ## png folder is not provided on github (the code is only on demonstration purpose)
    current_tesseract_txt <- tesseract::ocr(paste("../data/png/",current_file_name,".png",sep</pre>
#
=""))
#
#
    ### clean the tessetact text (separate line by "\n", delete null string, transter to lower c
ase)
#
    clean_tesseract_txt <- strsplit(current_tesseract_txt,"\n")[[1]]</pre>
#
    clean_tesseract_txt <- clean_tesseract_txt[clean_tesseract_txt!=""]</pre>
    ### save tesseract text file
#
    writeLines(clean_tesseract_txt, paste("../data/tesseract/",current_file_name,".txt",sep=""))
# }
```

Step 3 - Error detection

```
#input: t is a txt file path; output is a cleaned txt
clean txt<-function(t){</pre>
  current_txt <- readLines(t,warn=FALSE,encoding = "UTF-8")</pre>
  current_txt =gsub("[^A-Za-z ]","",current_txt)
  current_txt =trimws(current_txt)
  current_txt =gsub("\\s+"," ",current_txt)
  return(tolower(current_txt))
}
#get all the cleaned ground truth txt and cleaned orc txt
ground truth<-list()</pre>
orc_txt<-list()</pre>
for(i in 1:length(file name vec)){
  ground_truth[[i]] <- clean_txt(paste("../data/ground_truth/",</pre>
                                               file_name_vec[i],sep=""))
  orc_txt[[i]] <- clean_txt(paste("../data/tesseract/",</pre>
                                            file_name_vec[i],sep=""))
}
#check the number of lines
not_equal_txt<-matrix(ncol=4)</pre>
for(i in 1:length(file_name_vec)){
  if(length(ground_truth[[i]])!=length(orc_txt[[i]])){
    not_equal_txt<-rbind(not_equal_txt,c(length(ground_truth[[i]]),</pre>
                                            length(orc txt[[i]]),
                                            file_name_vec[i],i))
  }
}
#check the txt manually and make correction
ground_truth[[3]]<-ground_truth[[3]][1:291]</pre>
orc_txt[[3]]<-orc_txt[[3]][1:291]
#delete No.10 txt
#delete No.22 txt
ground_truth[[23]]<-ground_truth[[23]][1:222]</pre>
orc txt[[23]]<-orc txt[[23]][1:222]
ground_truth[[34]]<-ground_truth[[34]][1:466]</pre>
orc txt[[34]]<-orc txt[[34]][1:466]
ground_truth[[41]]<-ground_truth[[41]][1:740]</pre>
orc_txt[[41]]<-orc_txt[[41]][1:740]
ground_truth[[61]]<-ground_truth[[61]][-c(498,499)]</pre>
ground_truth[[63]]<-ground_truth[[63]][1:674]</pre>
orc txt[[63]]<-orc txt[[63]][1:674]
ground_truth[[68]]<-ground_truth[[68]][1:891]</pre>
orc txt[[68]]<-orc txt[[68]][1:891]
#delete No.70
ground_truth[[72]]<-ground_truth[[72]][-499]</pre>
#delete No.80
ground truth[[100]]<-ground truth[[100]][1:803]</pre>
```

```
#the txt pairs we can use
txt file num<-c(1:9,11:21,23:69,71:79,81:100)# total 96 txt pairs
#remove the lines with different length in pairs
for(i in txt_file_num){
 gt<-str_count(ground_truth[[i]],'\\w+')</pre>
 ot<-str_count(orc_txt[[i]],'\\w+')</pre>
 jud<-gt==ot
 ground_truth[[i]]<-ground_truth[[i]][jud]</pre>
 orc_txt[[i]]<-orc_txt[[i]][jud]</pre>
}
#the txt we can use now are ground_truth[[txt_file_num]] and orc_txt[[txt_file_num]]
#for text error detection and correction
source("./lib/ErrorDetection/2gram_error_detector.R")
source("./lib/ErrorDetection/orc txt error detector.R")
ground_truth_use = ground_truth[txt_file_num]
orc_txt_use = orc_txt[txt_file_num]
truth line = list()
truth word = list()
for(i in 1:length(ground_truth_use)){
truth_line[i] = paste(ground_truth_use[[i]], collapse = " ")
truth_word[i] = list(str_split(truth_line[[i]]," ")[[1]])
}
orc_line = list()
orc_word = list()
for(i in 1:length(orc txt use)){
 orc_line[i] = paste(orc_txt_use[[i]], collapse = " ")
orc_word[i] = list(str_split(orc_line[[i]]," ")[[1]])
}
```

```
# while(orc_word[[5]][nchar(orc_word[[5]])>1]){
# error = neighbor_words(orc_word[[5]],truth_word[[5]]) # test for 1 document
# }
#
#
# orc5 = matrix(0,ncol=11,nrow=length(orc_word[[5]]))
                                                                                             #
# colnames(orc5) = c('word','correct_word',"neighbor1","neighbor2","neighbor3",
#
                                                      "neighbor4", "neighbor5", "neighbor6",
#
                                                     "neighbor7", "neighbor8", "truth")
#
# for(i in 1:nrow(orc5)){
   orc5[i,] = error[[i]]
# }
# orc5 = write.csv(orc5,"./output/orc5.csv")
# # error = sapply(orc_dict,neighbor_words,truth_dict) # for all documents
# orc all = matrix(0,ncol=10,nrow=length(unlist(orc dict)))
# colnames(orc_all) = c('word','correct_word',"neighbor1","neighbor2","neighbor3",
#
                                                      "neighbor4", "neighbor5", "neighbor6",
#
                                                     "neighbor7", "neighbor8")
#
# file_name = file_name_vec[txt_file_num]
# for ( i in 1:length(file name)){
#
#
#
    error = neighbor_words(orc_dict[[i]],truth_dict[[i]])
#
#
    orc = matrix(0,ncol=10,nrow=length(orc_dict[[i]]))
#
#
    colnames(orc) = c('word','correct_word',"neighbor1","neighbor2","neighbor3",
#
                                                     "neighbor4", "neighbor5", "neighbor6",
                                                     "neighbor7", "neighbor8")
#
#
#
#
    for(j in 1:nrow(orc)){
#
    orc[j,] = error[[j]]
# }
#
#
    current_file_name <- sub(".txt","",file_name[i])</pre>
#
#
    write.csv(orc, paste("./output/forcorrection/",current_file_name,".csv",sep=""))
#
#
    print(i)
#
# }
```

Step 4 - Error correction

Generate n-gram candidate database

Create a corpus based on the ground_truth text

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```

```
corpus <- VCorpus(VectorSource(ground_truth))%>%
  tm_map(removeWords, character(0))%>%
  tm_map(stripWhitespace)
dtm.docs <- DocumentTermMatrix(corpus)</pre>
```

Create n-grams candidate set

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```
# Functions
OnegramTokenizer <- function(x) {RWeka::NGramTokenizer(x, RWeka::Weka control(min=1, max=1))}</pre>
# BigramTokenizer <- function(x) {RWeka::NGramTokenizer(x, RWeka::Weka_control(min=2, max=2))}
ThreegramTokenizer <- function(x) {RWeka::NGramTokenizer(x, RWeka::Weka_control(min=3, max=3))}</pre>
# FourgramTokenizer <- function(x) {RWeka::NGramTokenizer(x, RWeka::Weka_control(min=4, max=4))}
FivegramTokenizer <- function(x) {RWeka::NGramTokenizer(x, RWeka::Weka control(min=5, max=5))}
# #Onegram
options(mc.cores=1)
dtm.docs.1g <- DocumentTermMatrix(corpus, control=list(tokenize=OnegramTokenizer))</pre>
# # Bigrams
# options(mc.cores=1)
# dtm.docs.2g <- DocumentTermMatrix(corpus, control=list(tokenize=BigramTokenizer))</pre>
# #Threegrams
options(mc.cores=1)
dtm.docs.3g <- DocumentTermMatrix(corpus, control=list(tokenize=ThreegramTokenizer))</pre>
# #Fourgrams
# options(mc.cores=1)
# dtm.docs.4g <- DocumentTermMatrix(corpus, control=list(tokenize=FourgramTokenizer))</pre>
#Fivegrams
options(mc.cores=1)
dtm.docs.5g <- DocumentTermMatrix(corpus, control=list(tokenize=FivegramTokenizer))</pre>
```

```
# To get the onegram dist, we use the slam package for ops with simple triplet mat
sums.1g <- colapply_simple_triplet_matrix(dtm.docs.1g,FUN=sum)
sums.1g <- sort(sums.1g, decreasing=T)
write.csv(sums.1g, file = "onegram.csv")

# To get the 3gram dist, we use the slam package for ops with simple triplet mat
sums.3g <- colapply_simple_triplet_matrix(dtm.docs.3g,FUN=sum)
sums.3g <- sort(sums.3g, decreasing=T)
write.csv(sums.3g, file = "3-gram.csv")

# To get the fivegram dist, we use the slam package for ops with simple triplet mat
sums.5g <- colapply_simple_triplet_matrix(dtm.docs.5g,FUN=sum)
sums.5g <- sort(sums.5g, decreasing=T)
data.frame(sums.5g)[21,]

write.csv(sums.5g, file = "5-gram.csv")</pre>
```

Create relaxed context n-grams candidate set Create 5-gram relaxed context candidates set

```
# read in 5-gram candidates frequency data
five.g <- read.csv("5-gram.csv")</pre>
colnames(five.g) <- c("5-gram", "frequency")</pre>
# split each 5-grams to 5 individual words
1 <-lapply(as.character(five.g$`5-gram`), strsplit, " ")</pre>
relaxedL <- 1
relaxed1 <- c()
relaxed2 <- c()
relaxed3 <- c()
relaxed4 <- c()
relaxed5 <- c()
for(i in 1:length(1)){
  #replace first word with *
  relaxedL[[i]][[1]][1] <- "*"
  # paste five words back together to 5-gram
  relaxed1 <- rbind(relaxed1, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  #replace second word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][2] <- "*"
  # paste five words back together to 5-gram
  relaxed2 <- rbind(relaxed2, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  #replace third word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][3] <- "*"
  # paste five words back together to 5-gram
  relaxed3 <- rbind(relaxed3, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  #replace fourth word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][4] <- "*"
  # paste five words back together to 5-gram
  relaxed4 <- rbind(relaxed4, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  #replace fifth word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][5] <- "*"
  relaxed5 <- rbind(relaxed5, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
}
relaxed1.df <- data.frame(relaxed1, five.g$frequency)</pre>
relaxed2.df <- data.frame(relaxed2, five.g$frequency)</pre>
relaxed3.df <- data.frame(relaxed3, five.g$frequency)</pre>
relaxed4.df <- data.frame(relaxed4, five.g$frequency)</pre>
relaxed5.df <- data.frame(relaxed5, five.g$frequency)</pre>
# merge frequency information for same relaxed context 5-gram candidates
relaxed1.df <- relaxed1.df %>%
  group_by(relaxed1) %>%
```

```
summarise(frequency = sum(five.g.frequency)) %>%
  arrange(frequency)
relaxed2.df <- relaxed2.df %>%
  group by(relaxed2) %>%
  summarise(frequency = sum(five.g.frequency)) %>%
  arrange(frequency)
relaxed3.df <- relaxed3.df %>%
  group_by(relaxed3) %>%
  summarise(frequency = sum(five.g.frequency)) %>%
  arrange(frequency)
relaxed4.df <- relaxed4.df %>%
  group by(relaxed4) %>%
  summarise(frequency = sum(five.g.frequency)) %>%
  arrange(frequency)
relaxed5.df <- relaxed5.df %>%
  group by(relaxed5) %>%
  summarise(frequency = sum(five.g.frequency)) %>%
  arrange(frequency)
write.csv(relaxed1.df, file = "relaxed1.df.csv")
write.csv(relaxed2.df, file = "relaxed2.df.csv")
write.csv(relaxed3.df, file = "relaxed3.df.csv")
write.csv(relaxed4.df, file = "relaxed4.df.csv")
write.csv(relaxed5.df, file = "relaxed5.df.csv")
```

Create 3-gram relaxed context candidates set

```
# read in 3-gram data
three.g <- read.csv("3-gram.csv")</pre>
colnames(three.g) <- c("3-gram", "frequency")</pre>
# split 3 grams lines to individual words
1 <-lapply(as.character(three.g$`3-gram`), strsplit, " ")</pre>
relaxedL <- 1
relaxed1.3gram <- c()</pre>
relaxed2.3gram <- c()
relaxed3.3gram <- c()
for(i in 1:length(1)){
  # replace first word with *
  relaxedL[[i]][[1]][1] <- "*"
  # combine three words back as * word1 word2 and append to dataframe
  relaxed1.3gram <- rbind(relaxed1.3gram, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  # replace second word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][2] <- "*"
  # combine three words back as 'word1 * word3' and append to dataframe
  relaxed2.3gram <- rbind(relaxed2.3gram, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
  # replace third word with *
  relaxedL[[i]][[1]] <- l[[i]][[1]]
  relaxedL[[i]][[1]][3] <- "*"
  # combine three words back as 'word1 word2 *' and append to dataframe
  relaxed3.3gram <- rbind(relaxed3.3gram, paste(relaxedL[[i]][[1]], collapse = ' '))</pre>
}
relaxed1.3g.df <- data.frame(relaxed1.3gram, three.g$frequency)</pre>
relaxed2.3g.df <- data.frame(relaxed2.3gram, three.g$frequency)</pre>
relaxed3.3g.df <- data.frame(relaxed3.3gram, three.g$frequency)</pre>
# merge frequency information for same relaxed context 3-gram candidates
relaxed1.3g.df <- relaxed1.3g.df %>%
  group_by(relaxed1.3gram) %>%
  summarise(frequency = sum(three.g.frequency)) %>%
  arrange(frequency)
relaxed2.3g.df <- relaxed2.3g.df %>%
  group by(relaxed2.3gram) %>%
  summarise(frequency = sum(three.g.frequency)) %>%
  arrange(frequency)
relaxed3.3g.df <- relaxed3.3g.df %>%
  group_by(relaxed3.3gram) %>%
```

```
summarise(frequency = sum(three.g.frequency)) %>%
arrange(frequency)

write.csv(relaxed1.3g.df, file = "relaxed1.3g.df.csv")
write.csv(relaxed2.3g.df, file = "relaxed2.3g.df.csv")
write.csv(relaxed3.3g.df, file = "relaxed3.3g.df.csv")
```

Step 5 - Performance measure

```
# word level
count intersect<-function(truth,orc){</pre>
  ground_truth_vec <- str_split(paste(truth, collapse = " ")," ")[[1]]</pre>
  orc_vec<-str_split(paste(orc, collapse = " ")," ")[[1]]</pre>
  return(length(vecsets::vintersect(ground_truth_vec,orc_vec)))
}
word recall before<-rep(NA,length(txt file num))
word precis before<-rep(NA,length(txt file num))</pre>
# word_recall_after<-rep(NA,length(txt_file_num))</pre>
# word precis after<-rep(NA,length(txt file num))</pre>
for(i in txt file num){
  truth_length<-length(str_split(paste(ground_truth[[i]], collapse = " ")," ")[[1]])</pre>
  orc_length<-length(str_split(paste(orc_txt[[i]], collapse = " ")," ")[[1]])</pre>
  old_vec_len<-count_intersect(ground_truth[[i]],orc_txt[[i]])</pre>
  word recall before[i]<-old vec len/truth length
  word_precis_before[i]<-old_vec_len/orc_length
}
#postprocessing for one txt
rlt<-read.csv("../output/rltX2.csv")</pre>
rlt<-as.data.frame(rlt)</pre>
word wise after<-sum(as.character(rlt$Correction)==as.character(rlt$GroundTruth))/nrow(rlt)#0.75
5
#character level
intersect number<-function(a,b){</pre>
  ta<-data.frame(table(a))</pre>
  tb<-data.frame(table(b))</pre>
  mt<-merge(ta,tb,by.x = "a",by.y = "b")</pre>
  mt$min_freq<-apply(mt[,c(2,3)],1,min)</pre>
  return(sum(mt$min_freq))
}
compare word<-function(bi words){</pre>
  if(nchar(bi words[1])==0 |nchar(bi words[2])==0 ){
    return(0)
  }else{
    a<-strsplit(bi_words[1],split="")[[1]]</pre>
    b<-strsplit(bi_words[2],split="")[[1]]</pre>
    return(intersect_number(a,b))
  }
}
chara recall before<-rep(NA,length(txt file num))</pre>
chara_precis_before<-rep(NA,length(txt_file_num))</pre>
#chara_recall_after<-rep(NA,length(txt_file_num))</pre>
#chara_precis_after<-rep(NA,length(txt_file_num))</pre>
for(i in txt_file_num){
```

```
gt<-str split(paste(ground truth[[i]], collapse = " ")," ")[[1]]</pre>
  ot<-str_split(paste(orc_txt[[i]], collapse = " ")," ")[[1]]</pre>
  exa<-cbind(gt,ot)</pre>
  rt<-apply(exa,1,compare word)</pre>
  chara recall before[i]<-sum(rt)/sum(nchar(gt))#ratio of correct with ground truth
  chara precis before[i]<-sum(rt)/sum(nchar(ot))#ratio of correct with ORC output
}
#postprocessing for one txt
rlt_chara<-apply(rlt[,1:2],1,compare_word)</pre>
length intersect chara<-sum(rlt chara)</pre>
chara wise after recall<-length intersect chara/sum(nchar(as.character(rlt$GroundTruth)))#0.928
chara wise after precis<-length intersect chara/sum(nchar(as.character(rlt$ORC)))</pre>
#0.951554
OCR_performance_table <- data.frame("Tesseract" = rep(NA,4),</pre>
                                      "Tesseract with postprocessing" = rep(NA,4))
row.names(OCR performance table) <- c("word wise recall", "word wise precision",
                                                   "character_wise_recall", "character_wise_precisi
on")
OCR performance table["word wise recall", "Tesseract"] <-
  mean(word recall before,na.rm = T)
OCR_performance_table["word_wise_precision","Tesseract"] <-</pre>
  mean(word precis before,na.rm = T)
OCR_performance_table["word_wise_recall","Tesseract_with_postprocessing"]<-
  word wise after
OCR_performance_table["word_wise_precision","Tesseract_with_postprocessing"]<-
  word wise after
OCR performance table["character wise recall", "Tesseract"] <-
  mean(chara recall before, na.rm = T)
OCR performance table["character wise precision", "Tesseract"] <-
  mean(chara precis before, na.rm = T)
OCR_performance_table["character_wise_recall", "Tesseract_with_postprocessing"] <-
  chara_wise_after_recall
OCR performance table["character wise precision", "Tesseract with postprocessing"] <-
  chara wise after precis
kable(OCR performance table, caption="Summary of OCR performance")
```

```
In [ ]:
```

```
# -*- coding: utf-8 -*-
Created on Sun Nov 18 15:18:26 2018
@author: Chenghao
11 11 11
from pyxdameraulevenshtein import damerau levenshtein distance
from nltk.metrics.distance import edit distance
import py_common_subseq.py_common_subseq as CS #Need to change xrange() as range
()
import math
import pandas as pd
# Candidate search
class project4():
   def candidate search(Dictionary, We, threshold):
        candidate = {}
        for Wc in Dictionary:
            dist = damerau levenshtein distance(Wc, We)
            if dist <= threshold:</pre>
                candidate[Wc] = Dictionary[Wc]
        return(candidate)
#-----
# Feature scoring
# Levenshtein edit distance
   def distance_score(candidates, We, threshold):
        Score = {}
        for Wc in candidates:
            score = 1 - edit distance(Wc, We, substitution cost=1, transposition
s=False)/(threshold + 1)
            Score[Wc] = score
        return(Score)
# String similarity
   def similarity_score(candidates, We, a1=0.25, a2=0.25, a3=0.25, a4=0.25):
        Score = {}
        for Wc in candidates:
            common subsequences = CS.find common subsequences(Wc, We)
            lcs = sorted(common subsequences, key=lambda x: len(x))[-1]
            IniLetter = We[0]
            EndLetter = We[-1]
            MidLetter = We[math.ceil(len(We)/2)]
            #LCS 1
            common_subseq_IntLetter = set([])
            for W in common subsequences:
                if W.startswith(IniLetter):
                    common_subseq_IntLetter.add(W)
            if len(common_subseq_IntLetter) == 0:
                lcs1 = ''
            else:
```

```
lcs1 = sorted(common subseq IntLetter, key=lambda x: len(x))[-1]
           #LCS z
           common subseq EndLetter = set([])
           for W in common subsequences:
              if W.endswith(EndLetter):
                  common subseq EndLetter.add(W)
           if len(common subseq EndLetter) == 0:
              lcsz = ''
           else:
              lcsz = sorted(common subseq EndLetter, key=lambda x: len(x))[-1]
           #LCS n
           common subseq MidLetter = set([])
           for W in common subsequences:
              if W.startswith(MidLetter):
                  common subseq MidLetter.add(W)
           if len(common subseq MidLetter) == 0:
              lcsn = ''
           else:
              lcsn = sorted(common subseq MidLetter, key=lambda x: len(x))[-1]
           denom = len(Wc) + len(We)
# ------
#
            nlcs = (2*len(lcs)**2)/denom
#
            nmnlcs1 = (2*len(lcs1)**2)/denom
#
            nmnlcsn = (2*len(lcsn)**2)/denom
            nmnlcsz = (2*len(lcsz)**2)/denom
 ______
           # original paper
           nlcs = (2*len(lcs))/denom
           nmnlcs1 = (2*len(lcs1))/denom
           nmnlcsn = (2*len(lcsn))/denom
           nmnlcsz = (2*len(lcsz))/denom
           score = a1*nlcs + a2*nmnlcs1 + a3*nmnlcsn + a4*nmnlcsz
           Score[Wc] = score
       return(Score)
# Language popularity
   def popularity score(candidates):
       Score = {}
       denom = max(candidates.values())
       for Wc in candidates:
           score = candidates[Wc]/denom
           Score[Wc] = score
       return(Score)
# Lexicon existance
   def existance score(candidates, lexicon): # lexicon is a set {candidate1, ca
ndidate2}
       Score = {}
       for Wc in candidates:
           score = int(Wc in lexicon)
           Score[Wc] = score
       return(Score)
```

```
# Exact-context popularity
    def exact popularity score(candidates, We, five gram E, five gram dic): # fi
ve gram E is 5-gram with We, five gram dic is dictionary {5 gram w/ Wc: frequenc
y}
        Score = {}
        Numer = \{\}
        for Wc in candidates:
            five gram C = []
            for five gram in five gram E:
                five gram C.append(five gram.replace(We, Wc, 1))
            numer = 0
            for five_gram_c in give_gram_C:
                numer = numer + five_gram_dic[five_gram_c]
            Numer[Wc] = numer
        Denom = max(Numer.values())
        for Wc in candidates:
            Score[Wc] = Numer[Wc]/Denom
        return(Score)
# Relaxed-context popularity
    def relaxed popularity score(candidates, We, five gram E, five gram dic X):
        Score = {}
        Numer = {}
        for Wc in candidates:
            five gram C = []
            grams C X = []
            for five gram in five gram E:
                five gram C.append(five gram.replace(We, Wc, 1))
            for i in range(5):
                five gram s = list(jieba.cut(five gram C[i]))
                for k in range(5):
                    if -i+4==k:
                        continue
                    else:
                        five_gram_s_copy = copy.deepcopy(five_gram_s)
                        five_gram_s_copy[2*k] = "*"
                        gram_c_x = "".join(five_gram_s_copy)
                        grams_C_X.append(gram_c_x)
            numer = 0
            for five gram c x in grams C X:
                numer = numer + five gram dic X[five gram c x]
            Numer[Wc] = numer
        Denom = max(Numer.values())
        for Wc in candidates:
            Score[Wc] = Numer[Wc]/Denom
        return(Score)
```