山东大学 计算机科学与技术 学院

信息检索与数据挖掘 课程实验报告

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实验题目: 朴素贝叶斯分类器(多分类)

实验内容:

基于向量空间模型的部分方法实现二十分类的朴素贝叶斯分类器 朴素贝叶斯基于的公式为贝叶斯公式,即:

$$p(c_i|x) = \frac{p(x|c_i)p(c_i)}{p(x,c)}$$

其中 ci 为文本类别, x 为文本, 朴素贝叶斯分类器基于假设为各单词之间的顺序对结果无影响, 即若文本 x 中有词 x1, x2, x3, x4.....,则 p(x|ci)=p(x1|ci) p(x2|ci)p(x3|ci)p(x4|ci)...... 虽然假设较为简单, 但分类效果较好。

本次实验使用的是二十分类数据集,取 1/9 为测试集,8/9 为训练集,最终二十分类正确率 80%,且大多数数据的 precision 都达到了 90%以上,代码如下:

```
import os
import chardet
import re
import nltk
import numpy
import math
# 夫除停用词
def cutstopwords(str):
   stopwords = {}.fromkeys([line.rstrip() for line in
open('estopwords.txt')])
   segs = str.replace('\n','').lower().split(' ')
   new str = ''
   for seg in segs:
      if seq not in stopwords:
          new_str = new_str + " " +seg
   return new str
# 去除标点
def cutsyms(str):
   new_str = re.sub('[1234567890,.\'\"\t\n*_+=?/|!@#$%^&*()`~<>:;\-
\[\]]'," ",str)
   return new_str
```

```
# 词干提取
def stemming(str):
   s = nltk.stem.SnowballStemmer('english')
   segs = str.replace('\n', '').lower().split(' ')
   new_str = ''
   for seg in segs:
      new_str = new_str + " " + s.stem(seg)
   return new_str
# 读取文本
def readtxt(path):
   global class_dict
   global class_num
   global class_list
   global num_txt
   global num_dict
   num dict = {}
   num_txt = 0
   all_context = ""
   for dirName, subdirList, fileList in os.walk(path):
      fileList.remove(fileList[0])
      for fname in fileList:
          class_name = dirName.split('/')[6]
          num = class_list[class_name]
          class_num[num] = class_num[num] + 1
          fname = os.path.join(dirName, fname)
          f = open(fname, 'rb')
          data = f.read()
          f.close()
          print(chardet.detect(data))
          print(fname)
open(fname, 'r+', encoding=chardet.detect(data)['encoding'])
          str = fname.read()
          str = cutsyms(str)
          str = cutstopwords(str)
          str = stemming(str)
          str_list = str.replace('\n', '').lower().split(' ')
          for seg in str_list:
```

```
if seg in num_dict.keys():
                num_dict[seg] = num_dict[seg] + 1
             else:
                num_dict[seg] = 1
         for seg in str_list:
             if seg in class dict[class name].keys():
class_dict[class_name].update({seg:class_dict[class_name][seg] + 1})
             else:
                class dict[class name].update({seg:1})
         all_context = all_context + "\n" + str
         num_txt = num_txt + 1
          fname.close()
   return all context
# 统计词出现次数
def wordcount(str):
   strl_ist = str.replace('\n','').lower().split(' ')
   count_dict = {}
   for str in strl_ist:
      if str in count dict.keys():
         count_dict[str] = count_dict[str] + 1
      else:
         count_dict[str] = 1
   # count_list=sorted(count_dict.items(), key=lambda x:x[1], reverse=True)
   count dict.pop('')
   return count_dict
#全部文本读取
num_txt = 0
num dict = {}
class_list = {'alt.atheism':0,'comp.graphics':1,'comp.os.ms-
windows.misc':2,'comp.sys.ibm.pc.hardware':3,'comp.sys.mac.hardware':4,'co
mp.windows.x':5,'misc.forsale':6,'rec.autos':7,'rec.motorcycles':8,'rec.sp
ort.baseball':9, 'rec.sport.hockey':10, 'sci.crypt':11, 'sci.electronics':12,
'sci.med':13,'sci.space':14,'soc.religion.christian':15,'talk.politics.gun
s':16, 'talk.politics.mideast':17, 'talk.politics.misc':18, 'talk.religion.mi
sc':19}
class_num = [0]*20
class_dict = {'alt.atheism':{},'comp.graphics':{},'comp.os.ms-
windows.misc':{},'comp.sys.ibm.pc.hardware':{},'comp.sys.mac.hardware':{},
'comp.windows.x':{},'misc.forsale':{},'rec.autos':{},'rec.motorcycles':{},
```

```
'rec.sport.baseball':{},'rec.sport.hockey':{},'sci.crypt':{},'sci.electron
ics':{},'sci.med':{},'sci.space':{},'soc.religion.christian':{},'talk.poli
tics.guns':{},'talk.politics.mideast':{},'talk.politics.misc':{},'talk.rel
igion.misc':{}}
context = readtxt("/Users/apple/Desktop/ir/train")
#全部文档记数,去除高频低频词
str_dict = wordcount(context)
new_dict = str_dict
str_dict = {}
for seg in new dict:
   if (new_dict[seg] > 10) & (new_dict[seg] < 10000):</pre>
       str_dict[seg] = new_dict[seg]
if '' in str_dict.keys():
   str_dict.pop('')
length = len(str dict)
print(length)
for i in class_list.keys():
   z_num = class_list[i]
   newc_dict = {}
   for seg in str_dict:
       if seg in class dict[i].keys():
          newc_dict[seg] = class_dict[i][seg]
   class_dict[i] = newc_dict
class sum = [0]*20
for i in class_list.keys():
   x_num = class_list[i]
   for j in class_dict[i].keys():
       class_sum[x_num] = class_sum[x_num] + class_dict[i][j]
#classification
classification = [[0 \text{ for } x \text{ in } range(20)] \text{ for } y \text{ in } range(20)]
for dirName, subdirList, fileList in
os.walk('/Users/apple/Desktop/ir/test'):
   fileList.remove(fileList[0])
   for fname in fileList:
       name = dirName.split('/')[6]
       num = class_list[name]
       print(name)
       fname = os.path.join(dirName, fname)
       f = open(fname, 'rb')
```

```
data = f.read()
      f.close()
      fname = open(fname, 'r+',
encoding=chardet.detect(data)['encoding'])
      ins_str = fname.read()
      fname.close()
      ins_dict = {}
      ins_str = cutsyms(ins_str)
      ins_str = cutstopwords(ins_str)
      ins_str = stemming(ins_str)
      ins_dict = wordcount(ins_str)
      class_pro = [0]*20
      for i in class_list.keys():
         y_num = class_list[i]
         for seg in ins_dict.keys():
            if seg in class_dict[i].keys():
                class_pro[y_num] = class_pro[y_num] +
math.log((class_dict[i][seg]+1)/(class_sum[y_num]+20))
            else:
                class_pro[y_num] = class_pro[y_num] +
math.log((1)/(class_sum[y_num]+20))
         class_pro[y_num] = class_pro[y_num] +
math.log((class_num[y_num])/(2000))
      index = numpy.argmax(class_pro)
      classification[num][index] = classification[num][index] + 1
sum = 0
for i in range(20):
   sum = sum + classification[i][i]
print(classification)
print(sum/2000)
评价标准: 召回率与准确率,可以进行综合考虑,取调和均值,本实验可以
得到 20*20 的矩阵, 便于后续处理
```

实验过程中遇到和解决的问题:	
(记录实验过程中遇到的问题, 以及解决过程和实验结果。	,可以适当配以关键
代码辅助说明, 但不要大段贴代码。)	
1. 第三类文本的分类效果不好:分析由于文本较短,	故所得词汇较少,
可以在预处理时对去除的文本进行改进	

结论分析与体会: 朴素贝叶斯虽然简单,但效果较好,很多情况下基于的假设虽然完全正确,但可以解决实际问题