## CSE258 Homework 4

Chenyu Huang, University of California San Diego

03/05/2017

## 1 Homework 4

**Task 1** The 5 most common bigrams in the first 5000 reviews are:

Figure 1: 5 most common bigrams

- **Task 2** Using the most 1000 bigrams from the corpus, the MSE of the linear model is 0.343313.
- **Task 3** With a mixture of the most common bigrams and unigrams, the MSE of the model is 0.289394.
- **Task 4** The 5 words with the most common positive weights and 5 words with most negative weights are as follows:

Top 5 words with the most positive weights:

Top 5 words with the most negative weights:

$$('straw', -0.19930067914769131), \qquad (6)$$

$$('thebackground', -0.22131852632951593), \qquad (7)$$

$$('corn', -0.23815182761392634), \qquad (8)$$

$$('water', -0.27289904891323319), \qquad (9)$$

$$('sortof', -0.63021419493511976) \qquad (10)$$

Homework № 4 Page 1 / 16

## **Task 5** The IDF of the words are:

$$('foam': 1.1139433523068367),$$
 (11)

$$('smell': 0.47712125471966244),$$
 (12)

$$(banana': 1.6720978579357175),$$
 (13)

$$('lactic': 2.9206450014067875),$$
 (14)

$$('tart': 1.806179973983887)$$
 (15)

The TF-IDF of the words in the first review are:

$$('foam': 2.2278867046136734),$$
 (16)

$$('smell': 0.47712125471966244),$$
 (17)

$$(banana': 3.344195715871435),$$
 (18)

$$('lactic': 5.841290002813575),$$
 (19)

$$('tart': 1.806179973983887)$$
 (20)

**Task 6** Using just the 1000 most common unigram feature representation. The cosine similarity between the first and second reviews is: 0.106130241679.

**Task 7** Using only the 1000 most common unigram feature representation. The review with the highest cosine similarity with the first review has the following attributes:

$$BeerId: 52211 \tag{21}$$

$$Profile\ Name: Heatwave 33$$
 (22)

**Task 8** Using tf-idf as their feature representation, the MSE of the unigram model now is: 0.278742

Homework № 4 Page 2 / 16

## 2 Code

Listing 1: task1

```
import numpy
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import operator
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
bigramCount = {}
### Count bigrams
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 r = r.split()
 for i in range(0, len(r) - 1):
   pair = r[i] + " " + r[i+1]
   if pair in bigramCount:
     bigramCount[pair] = bigramCount[pair] + 1
      bigramCount[pair] = 1
print len(bigramCount)
sorted_bigram = sorted(bigramCount.items(), key=operator.itemgetter(1),
   reverse = True)
print sorted_bigram[:5]
```

Homework № 4 Page 3 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import operator
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
bigramCount = {}
### Count bigrams
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 r = r.split()
 for i in range(0, len(r) - 1):
   pair = r[i] + " " + r[i+1]
   if pair in bigramCount:
      bigramCount[pair] = bigramCount[pair] + 1
    else:
      bigramCount[pair] = 1
sorted_bigram = sorted(bigramCount.items(), key = operator.itemgetter(1),
   reverse = True)
words = []
for s in sorted_bigram[:1000]:
 words.extend([s[0]])
### Sentiment analysis
wordId = dict(zip(words, range(len(words))))
wordSet = set(words)
def feature(datum):
  feat = [0] * len(words)
```

Homework № 4 Page 4 / 16

```
r = ''.join([c for c in datum['review/text'].lower() if not c in
    punctuation])
 r = r.split()
 for i in range(0, len(r) - 1):
   w = r[i] + " " + r[i+1]
   if w in words:
     feat[wordId[w]] += 1
 feat.append(1) #offset
 return feat
X = [feature(d) for d in data]
y = [d['review/overall'] for d in data]
#No regularization
\#theta, residuals, rank, s = numpy.linalg.lstsq(X, y)
#With regularization
clf = linear_model.Ridge(1.0, fit_intercept=False)
clf.fit(X, y)
theta = clf.coef_
# predictions = clf.predict(X)
# The mean squared error
print("Mean squared error: %f"
     % np.mean((clf.predict(X) - y) ** 2))
```

Homework № 4 Page 5 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import operator
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
bigramCount = {}
### Count bigrams
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 r = r.split()
 for i in range(0, len(r) - 1):
   pair = r[i] + " " + r[i+1]
    if pair in bigramCount:
      bigramCount[pair] = bigramCount[pair] + 1
    else:
      bigramCount[pair] = 1
unigramCount = {}
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 for w in r.split():
   if w in unigramCount:
     unigramCount[w] += 1
    else:
      unigramCount[w] = 1
combine = dict(bigramCount.items() + unigramCount.items())
combine_sort = sorted(combine.items(), key = operator.itemgetter(1), reverse
   = True)
words = []
```

Homework № 4 Page 6 / 16

```
for s in combine_sort[:1000]:
 words.extend([s[0]])
### Sentiment analysis
wordId = dict(zip(words, range(len(words))))
wordSet = set(words)
def feature(datum):
 feat = [0]*len(words)
 r = ''.join([c for c in datum['review/text'].lower() if not c in
     punctuation])
 r = r.split()
  for w in r:
   if w in words:
     feat[wordId[w]] += 1
 for i in range(0, len(r) - 1):
   w = r[i] + " " + r[i+1]
   if w in words:
      feat[wordId[w]] += 1
 feat.append(1) #offset
  return feat
X = [feature(d) for d in data]
y = [d['review/overall'] for d in data]
#No regularization
 \verb|#theta, residuals, rank, s = numpy.linalg.lstsq(X, y) \\
#With regularization
clf = linear_model.Ridge(1.0, fit_intercept=False)
clf.fit(X, y)
theta = clf.coef_
predictions = clf.predict(X)
print("Mean squared error: %f"
      % np.mean((clf.predict(X) - y) ** 2))
weights = {}
for i in range(len(words)):
 for key, value in wordId.iteritems():
   if value == i:
      weights[key] = theta[i]
      break
weight_sort = sorted(weights.items(), key = operator.itemgetter(1), reverse =
    True)
print weight_sort[:5]
print weight_sort[-5:]
```

Homework № 4 Page 7 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import math
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
idf = {}
for word in ['foam', 'smell', 'banana', 'lactic', 'tart']:
 idf[word] = 0
punctuation = set(string.punctuation)
for word in ['foam', 'smell', 'banana', 'lactic', 'tart']:
 for d in data:
   r = ''.join([c for c in d['review/text'].lower() if not c in punctuation
       ])
   r = r.split()
    if word in r:
     idf[word] += 1
for key in idf:
 idf[key] = math.log10(len(data) / idf[key])
print idf
tf = {'foam':0, 'smell':0, 'banana':0, 'lactic':0, 'tart':0}
tf_idf = {}
for key in idf:
 d = data[0]
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
  r = r.split()
 for w in r:
   if key == w:
      tf[key] += 1
  tf_idf[key] = tf[key] * idf[key]
```

Homework № 4 Page 8 / 16

```
print tf_idf
```

Homework Noman = 4 Page 9 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import math
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
### 1000 most common unigrams
wordCount = defaultdict(int)
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 for w in r.split():
   wordCount[w] += 1
counts = [(wordCount[w], w) for w in wordCount]
counts.sort()
counts.reverse()
words = set([x[1] for x in counts[:1000]])
idf = {}
tf1 = {}
for word in ''.join([c for c in data[0]['review/text'].lower() if not c in
   punctuation]).split():
  # only using 1000 features
 if word in words:
   if word in tf1:
     tf1[word] += 1
    else:
     tf1[word] = 1
    idf[word] = 0
tf2 = {}
for word in ''.join([c for c in data[1]['review/text'].lower() if not c in
   punctuation]).split():
```

Homework № 4 Page 10 / 16

```
# only using 1000 features
  if word in words:
   if word in tf2:
     tf2[word] += 1
   else:
     tf2[word] = 1
    idf[word] = 0
for word in idf:
 for d in data:
   r = ''.join([c for c in d['review/text'].lower() if not c in punctuation
   r = r.split()
   if word in r:
     idf[word] += 1
for key in idf:
 idf[key] = math.log10(5000 / float(idf[key]))
tf_idf1 = {}
for key in tf1:
 tf_idf1[key] = tf1[key] * idf[key]
tf_idf2 = {}
for key in tf2:
 tf_idf2[key] = tf2[key] * idf[key]
def find_cosine_sim(tf_idf1, tf_idf2):
 nominator = 0.0
 deno_part1 = 0.0
  deno_part2 = 0.0
 for key in tf_idf1:
   if key in tf_idf2:
     nominator += tf_idf1[key] * tf_idf2[key]
  for key in tf_idf1:
      deno_part1 += tf_idf1[key] ** 2
  for key in tf_idf2:
      deno_part2 += tf_idf2[key] ** 2
  denominator = math.sqrt(deno_part1) * math.sqrt(deno_part2)
  return nominator / denominator
print find_cosine_sim(tf_idf1, tf_idf2)
```

Homework № 4 Page 11 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import math
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
### 1000 most common unigrams
wordCount = defaultdict(int)
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 for w in r.split():
   wordCount[w] += 1
counts = [(wordCount[w], w) for w in wordCount]
counts.sort()
counts.reverse()
words = set([x[1] for x in counts[:1000]])
idf = {}
for w in words:
 idf[w] = 0
tf1 = {}
for word in ''.join([c for c in data[0]['review/text'].lower() if not c in
  punctuation]).split():
  # only using 1000 features
 if word in words:
   if word in tf1:
     tf1[word] += 1
   else:
      tf1[word] = 1
```

Homework № 4 Page 12 / 16

```
for word in idf:
 for d in data:
   r = ''.join([c for c in d['review/text'].lower() if not c in punctuation
       ])
   r = r.split()
   if word in r:
     idf[word] += 1
for key in idf:
 idf[key] = math.log10(5000 / float(idf[key]))
tf_idf1 = {}
for key in tf1:
 tf_idf1[key] = tf1[key] * idf[key]
def find_cosine_sim(tf_idf1, tf_idf2):
 nominator = 0.0
 deno_part1 = 0.0
  deno_part2 = 0.0
  for key in tf_idf1:
   if key in tf_idf2:
     nominator += tf_idf1[key] * tf_idf2[key]
  for key in tf_idf1:
      deno_part1 += tf_idf1[key] ** 2
  for key in tf_idf2:
     deno_part2 += tf_idf2[key] ** 2
  denominator = math.sqrt(deno_part1) * math.sqrt(deno_part2)
  if denominator == 0:
   return 0
 return nominator / denominator
counter = 1
sim = 0
text = ""
bid = ""
pname = ""
index = 1
for d in data[1:]:
 tf2 = {}
 tf_idf2 = {}
 for word in ''.join([c for c in d['review/text'].lower() if not c in
     punctuation]).split():
   if word in words:
     if word in tf2:
       tf2[word] += 1
      else:
       tf2[word] = 1
  for key in tf2:
    tf_idf2[key] = tf2[key] * idf[key]
```

Homework № 4 Page 13 / 16

```
if find_cosine_sim(tf_idf1, tf_idf2) > sim:
    sim = find_cosine_sim(tf_idf1, tf_idf2)
    text = d['review/text']
    index = counter
    bid = d['beer/beerId']
    pname = d['user/profileName']
    counter += 1

print text
print bid
print pname
```

Homework № 4 Page 14 / 16

```
import numpy as np
import urllib
import scipy.optimize
import random
from collections import defaultdict
import nltk
import string
from nltk.stem.porter import *
from sklearn import linear_model
import operator
def parseData(fname):
 for l in urllib.urlopen(fname):
   yield eval(1)
### Just the first 5000 reviews
print "Reading data..."
data = list(parseData("beer_50000.json"))[:5000]
print "done"
### 1000 most common unigrams
wordCount = defaultdict(int)
punctuation = set(string.punctuation)
for d in data:
 r = ''.join([c for c in d['review/text'].lower() if not c in punctuation])
 for w in r.split():
   wordCount[w] += 1
counts = [(wordCount[w], w) for w in wordCount]
counts.sort()
counts.reverse()
words = set([x[1] for x in counts[:1000]])
idf = {}
for w in words:
 idf[w] = 0
for word in idf:
 for d in data:
   r = ''.join([c for c in d['review/text'].lower() if not c in punctuation
       ])
   r = r.split()
   if word in r:
      idf[word] += 1
## Sentiment analysis
wordId = dict(zip(words, range(len(words))))
def feature(datum):
```

Homework № 4 Page 15 / 16

```
feat = [0] * len(words)
  tf = \{\}
 for word in ''.join([c for c in datum['review/text'].lower() if not c in
     punctuation]).split():
  # only using 1000 features
   if word in words:
     if word in tf:
       tf[word] += 1
      else:
       tf[word] = 1
  tf_idf = {}
 for key in tf:
   tf_idf[key] = tf[key] * idf[key]
   feat[wordId[key]] = tf_idf[key]
  feat.append(1) #offset
 return feat
X = [feature(d) for d in data]
y = [d['review/overall'] for d in data]
#With regularization
clf = linear_model.Ridge(1.0, fit_intercept=False)
clf.fit(X, y)
theta = clf.coef_
# predictions = clf.predict(X)
# The mean squared error
print("Mean squared error: %f"
     % np.mean((clf.predict(X) - y) ** 2))
```

Homework № 4 Page 16 / 16