3.6 Featurizing text data with tfidf weighted word-vectors

In [16]:

```
import pandas as pd
import matplotlib.pyplot as plt
import re
import time
import warnings
import numpy as np
from nltk.corpus import stopwords
from sklearn.preprocessing import normalize
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
warnings.filterwarnings("ignore")
import sys
import os
import pandas as pd
import numpy as np
from tqdm import tqdm
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.preprocessing import StandardScaler
from sklearn.model selection import train test split
from scipy.sparse import coo matrix, vstack
import pandas as pd
import matplotlib.pyplot as plt
import re
import time
import warnings
import sqlite3
from sqlalchemy import create engine # database connection
import csv
import os
warnings.filterwarnings("ignore")
import datetime as dt
import numpy as np
from nltk.corpus import stopwords
from sklearn.decomposition import TruncatedSVD
from sklearn.preprocessing import normalize
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.manifold import TSNE
import seaborn as sns
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion matrix
from sklearn.metrics.classification import accuracy score, log loss
from sklearn.feature_extraction.text import TfidfVectorizer
from collections import Counter
from scipy.sparse import hstack
from sklearn.multiclass import OneVsRestClassifier
from sklearn.svm import SVC
from sklearn.model_selection import StratifiedKFold
from collections import Counter, defaultdict
from sklearn.calibration import CalibratedClassifierCV
from sklearn.naive_bayes import MultinomialNB
from sklearn.naive bayes import GaussianNB
from sklearn.model_selection import train test split
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import normalized mutual info score
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import cross val score
from sklearn.linear_model import SGDClassifier
from mlxtend.classifier import StackingClassifier
from sklearn import model selection
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import precision_recall_curve, auc, roc_curve
```

```
In [17]:
```

In [18]:

```
print (df.shape)
df.head()
```

(70000, 6)

Out[18]:

	id	qid1	qid2	question1	question2	is_duplicate
0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0
1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0
2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0
3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24}[/math] i	0
4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0

In [19]:

```
df_without_qid = df.drop(columns=['qid1', 'qid2','is_duplicate'])
df_without_qid.head()
```

Out[19]:

	id	question1	question2
0	0	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh
1	1	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto
2	2	How can I increase the speed of my internet co	How can Internet speed be increased by hacking
3	3	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24}[/math] i
4	4	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?

In [20]:

```
if os.path.isfile(r'D:\AppliedAI\Homework-n-Assignments\# 20 Quora\nlp_features_train.csv'):
    dfnlp = pd.read_csv("nlp_features_train.csv",encoding='latin-1')
    dfnlp = dfnlp.head(70000)
else:
    print("download nlp_features_train.csv from drive or run previous notebook")

if os.path.isfile(r'D:\AppliedAI\Homework-n-Assignments\# 20
Quora\df_fe_without_preprocessing_train.csv'):
    dfppro = pd.read_csv("df_fe_without_preprocessing_train.csv",encoding='latin-1')
    dfppro = dfppro.head(70000)
else:
```

```
print("download df_fe_without_preprocessing_train.csv from drive or run previous notebook")

df1 = dfnlp.drop(['qid1','qid2','question1','question2'],axis=1)

df2 = dfppro.drop(['qid1','qid2','question1','question2','is_duplicate'],axis=1)

#df3 = df.drop(['qid1','qid2','question1','question2','is_duplicate'],axis=1)
```

In [21]:

```
print (df1.shape)
df1.head()
```

(70000, 17)

Out[21]:

	id	is_duplicate	cwc_min	cwc_max	csc_min	csc_max	ctc_min	ctc_max	last_word_eq	first_word_eq	abs_len_diff
0	0	0	0.999980	0.833319	0.999983	0.999983	0.916659	0.785709	0.0	1.0	2.0
1	1	0	0.799984	0.399996	0.749981	0.599988	0.699993	0.466664	0.0	1.0	5.0
2	2	0	0.399992	0.333328	0.399992	0.249997	0.399996	0.285712	0.0	1.0	4.0
3	3	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0	0.0	2.0
4	4	0	0.399992	0.199998	0.999950	0.666644	0.571420	0.307690	0.0	1.0	6.0
4	4										

In [22]:

```
print (df2.shape)
df2.head()
```

(70000, 12)

Out[22]:

	id	freq_qid1	freq_qid2	q1len	q2len	q1_n_words	q2_n_words	word_Common	word_Total	word_share	freq_q1+q2	f
0	0	1	1	66	57	14	12	10.0	23.0	0.434783	2	С
1	1	4	1	51	88	8	13	4.0	20.0	0.200000	5	3
2	2	1	1	73	59	14	10	4.0	24.0	0.166667	2	С
3	3	1	1	50	65	11	9	0.0	19.0	0.000000	2	С
4	4	3	1	76	39	13	7	2.0	20.0	0.100000	4	2
4		•	·		•	•			•	•		▶ I

In [23]:

```
df4 =pd.merge(df1, df2, on='id')
df4.head()
```

Out[23]:

	id	is_duplicate	cwc_min	cwc_max	csc_min	csc_max	ctc_min	ctc_max	last_word_eq	first_word_eq	 freq_qid2
0	0	0	0.999980	0.833319	0.999983	0.999983	0.916659	0.785709	0.0	1.0	 1
1	1	0	0.799984	0.399996	0.749981	0.599988	0.699993	0.466664	0.0	1.0	 1
2	2	0	0.399992	0.333328	0.399992	0.249997	0.399996	0.285712	0.0	1.0	 1
3	3	0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0	0.0	 1
4	4	0	0.399992	0.199998	0.999950	0.666644	0.571420	0.307690	0.0	1.0	 1

5 rows × 28 columns

```
•
                                                                                                    P
In [24]:
df final =pd.merge(df without qid, df4, on='id')
#print (df_final.shape)
In [25]:
df final train, df final test = train test split(df final, test size=0.33, shuffle=False)
#print (df final train.shape)
In [26]:
q1 train = df final train['question1'].values
q1_test = df_final_test['question1'].values
q2 train = df final train['question2'].values
q2 test = df final test['question2'].values
y train = df final train['is duplicate'].values
y test = df final test['is duplicate'].values
In [27]:
q_train = np.concatenate([q1_train,q2_train])
q_test = np.concatenate([q1_test,q2_test])
print (q_train.shape)
print (q test.shape)
(93800,)
(46200,)
In [28]:
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
lst train q1=[]
lst test q1=[]
lst_of_lst_train_q1 = []
lst_of_lst_test_q1 = []
for sentance in tqdm(q1 train):
   lst train q1.append(sentance.strip())
for sentance in tqdm(lst_train_q1):
    lst_of_lst_train_q1.append(sentance.split())
for sent in tqdm(q1_test):
   lst test q1.append(sent.strip())
for sent in tqdm(lst test q1):
    lst_of_lst_test_q1.append(sent.split())
w2v model self taught train q1=Word2Vec(lst of lst train q1,min count=1,size=50, workers=4)
w2v words train q1 = list(w2v model self taught train q1.wv.vocab)
lst train q2=[]
lst_test_q2=[]
lst_of_lst_train_q2 = []
lst_of_lst_test_q2 = []
for sentance in tqdm (q2 train):
    lst_train_q2.append(sentance.strip())
for sentance in tqdm(lst_train_q2):
    lst_of_lst_train_q2.append(sentance.split())
for sent in tqdm(q2_test):
    lst test q2.append(sent.strip())
for sent in tqdm(lst test q2):
```

```
lst_of_lst_test_q2.append(sent.split())
w2v model self taught train q2=Word2Vec(lst of lst train q1,min count=1,size=50, workers=4)
w2v words train q2 = list(w2v model self taught train q2.wv.vocab)
                                                                 | 46900/46900
100%1
[00:00<00:00, 1959525.62it/s]
100%|
                                                               1 46900/46900
[00:00<00:00, 783407.57it/s]
                                                                  23100/23100
100%|
[00:00<00:00, 2105674.97it/s]
100%|
                                                                        | 23100/23100
[00:00<00:00, 661765.48it/s]
100%|
                                                                   1 46900/46900
[00:00<00:00, 1751891.22it/s]
100%|
                                                                  46900/46900
[00:00<00:00, 644195.60it/s]
100%|
[00:00<00:00, 1930125.15it/s]
100%|
                                                                        | 23100/23100
[00:00<00:00, 643396.41it/s]
```

In [29]:

```
tf_idf_vect_q1 = TfidfVectorizer(min_df=4,ngram_range=(1,3),max_features=2500)
tfidf_q1_train = tf_idf_vect_q1.fit_transform(q1_train)
dictionary_q1 = dict(zip(tf_idf_vect_q1.get_feature_names(), list(tf_idf_vect_q1.idf_)))
tfidfw2v_q1 = tf_idf_vect_q1.get_feature_names()
tfidf_q1_test = tf_idf_vect_q1.transform(q1_test)

#############################

tf_idf_vect_q2 = TfidfVectorizer(min_df=4,ngram_range=(1,3),max_features=2500)
tfidf_q2_train = tf_idf_vect_q2.fit(q2_train)
dictionary_q2 = dict(zip(tf_idf_vect_q2.get_feature_names(), list(tf_idf_vect_q2.idf_)))
tfidfw2v_q2 = tf_idf_vect_q2.get_feature_names()
tfidf_q2_test = tf_idf_vect_q2.transform(q2_test)
```

In [30]:

```
#tf_idf_vect = TfidfVectorizer(min_df=4,ngram_range=(1,3),max_features=2500)
#tfidf_q1_train = tf_idf_vect.fit_transform(q1_train)
#tfidf_q1_test = tf_idf_vect.transform(q1_test)

#tfidf_q2_train = tf_idf_vect.fit_transform(q2_train)
#tfidf_q2_test = tf_idf_vect.transform(q2_test)
```

In [31]:

```
tfidf w2v q1 train = []; # the tfidf-w2v for each sentence/review is stored in this list
row=0:
for sent4 in tqdm(lst of lst_train_q1): # for each review/sentence
   sent vec4 = np.zeros(50) # as word vectors are of zero length
    weight sum4 =0; # num of words with a valid vector in the sentence/review
    for word4 in sent4: # for each word in a review/sentence
        if word4 in w2v_words_train_q1 and word4 in tfidfw2v_q1:
            vec4 = w2v_model_self_taught_train_q1.wv[word4]
            tf_idf_train_q1 = dictionary_q1[word4]*(sent4.count(word4)/len(sent4))
            sent_vec4 += (vec4 * tf_idf_train_q1)
            weight_sum4 += tf_idf_train_q1
    if weight sum4 != 0:
       sent vec4 /= weight sum4
    tfidf_w2v_q1_train.append(sent_vec4)
    row += 1
100%|
                                                                                | 46900/46900 [01:
07<00:00, 699.64it/s]
```

In [32]:

```
row=0;
for sent5 in tqdm(lst_of_lst_test_q1): # for each review/sentence
   sent vec5 = np.zeros(50) # as word vectors are of zero length
    weight sum5 =0; # num of words with a valid vector in the sentence/review
    for word5 in sent5: # for each word in a review/sentence
        if word5 in w2v words train q1 and word5 in tfidfw2v q1:
            vec5 = w2v model self taught train q1.wv[word5]
            tf idf test q1 = dictionary q1[word5]*(sent5.count(word5)/len(sent5))
            sent vec5 += (vec5 * tf idf test q1)
            weight_sum5 += tf_idf_test_q1
    if weight sum5 != 0:
       sent vec5 /= weight sum5
    tfidf_w2v_q1_test.append(sent_vec5)
    row += 1
                                                                          23100/23100 [00:
100%|
45<00:00, 509.46it/s]
In [33]:
tfidf w2v q2 train = []; # the tfidf-w2v for each sentence/review is stored in this list
for sent2 in tqdm(lst of lst train q2): # for each review/sentence
    sent vec2 = np.zeros(50) # as word vectors are of zero length
    weight sum2 =0; # num of words with a valid vector in the sentence/review
    for word2 in sent2: # for each word in a review/sentence
        if word2 in w2v words train q2 and word4 in tfidfw2v q2:
            vec2 = w2v model self taught train q2.wv[word2]
            tf idf train q2 = dictionary q2[word2]*(sent2.count(word2)/len(sent2))
            sent_vec2 += (vec2 * tf_idf_train_q2)
            weight_sum2 += tf_idf_train_q2
    if weight sum2 != 0:
        sent vec2 /= weight sum2
    tfidf_w2v_q2_train.append(sent_vec2)
100%|
                                                                            | 46900/46900 [01:
24<00:00, 551.88it/s]
In [34]:
tfidf w2v q2 test = []; # the tfidf-w2v for each sentence/review is stored in this list
for sent3 in tqdm(lst of lst test q2): # for each review/sentence
   sent vec3 = np.zeros(50) # as word vectors are of zero length
    weight sum3 =0; # num of words with a valid vector in the sentence/review
    for word3 in sent3: # for each word in a review/sentence
        if word3 in w2v words train q2 and word5 in tfidfw2v q2:
            vec3 = w2v model self taught train q2.wv[word3]
            tf idf test q2 = dictionary q2[word3]*(sent3.count(word3)/len(sent3))
            sent vec3 += (vec3 * tf idf test q2)
            weight_sum3 += tf idf test q2
    if weight sum3 != 0:
       sent vec3 /= weight sum3
    tfidf_w2v_q2_test.append(sent_vec3)
    row += 1
100%|
                                                                          23100/23100 [00:
44<00:00, 524.40it/s]
q1 train std =StandardScaler(with mean=False, with std=False).fit transform(tfidf w2v q1 train)
q1 test std =StandardScaler(with mean=False, with std=False).fit transform(tfidf w2v q1 test)
q2 train std =StandardScaler(with mean=False, with std=False).fit transform(tfidf w2v q2 train)
\verb|q2_test_std| = StandardScaler(with_mean=\textbf{False}, with_std=\textbf{False}).fit_transform(tfidf_w2v_q2_test)|
In [38]:
```

#https://stackoverflow.com/questions/45961747/append-tfidf-to-pandas-dataframe

```
#nttps://www.researcngate.net/post/How to append TF-1DF vector into pandas datairame
#q1 train std df = pd.DataFrame(q1 train std.toarray())
#q1 test std df = pd.DataFrame(q1 test std.toarray())
#q2_train_std_df = pd.DataFrame(q2_train_std.toarray())
#q2 test std df = pd.DataFrame(q2 test std.toarray())
q1 train std df = pd.DataFrame(q1 train std)
q1 test std df = pd.DataFrame(q1 test std)
q2 train std df = pd.DataFrame(q2 train std)
q2 test std df = pd.DataFrame(q2 test std)
In [39]:
from scipy.sparse import coo_matrix, hstack
import scipy.sparse as ss
train_after_dropping = df_final_train.drop(['id','question1','question2','is_duplicate'],axis=1)
print (train after dropping.shape)
X_train = pd.concat([train_after_dropping, q2_train_std_df,q1_train_std_df], axis=1)
test after dropping = df final test.drop(['id','question1','question2','is duplicate'],axis=1)
X_test1 = pd.concat([q2_test_std_df,q1_test_std_df],axis=1)
X test = ss.hstack([test after dropping, X test1])
print("Number of data points in train data :", X_train.shape)
print("Number of data points in test data :", X test.shape)
(46900, 26)
Number of data points in train data: (46900, 126)
Number of data points in test data: (23100, 126)
In [40]:
print("-"*10, "Distribution of output variable in train data", "-"*10)
train distr = Counter(y_train)
train len = len(y train)
print("Class 0: ",int(train_distr[0])/train_len,"Class 1: ", int(train_distr[1])/train_len)
print("-"*10, "Distribution of output variable in train data", "-"*10)
test distr = Counter(y_test)
test_len = len(y_test)
print("Class 0: ",int(test distr[1])/test len, "Class 1: ",int(test distr[1])/test len)
----- Distribution of output variable in train data ------
Class 0: 0.626226012793177 Class 1: 0.37377398720682303
----- Distribution of output variable in train data -----
Class 0: 0.36978354978354977 Class 1: 0.36978354978354977
In [41]:
\# This function plots the confusion matrices given y_i, y_ihat.
def plot confusion matrix(test y, predict y):
    C = confusion_matrix(test_y, predict_y)
    \# C = 9,9 matrix, each cell (i,j) represents number of points of class i are predicted class j
    A = (((C.T)/(C.sum(axis=1))).T)
    #divid each element of the confusion matrix with the sum of elements in that column
    \# C = [[1, 2],
         [3, 4]]
    \# C.T = [[1, 3],
            [2, 4]]
    # C.sum(axis = 1) axis=0 corresonds to columns and axis=1 corresponds to rows in two
diamensional array
    \# C.sum(axix = 1) = [[3, 7]]
    \# ((C.T)/(C.sum(axis=1))) = [[1/3, 3/7]
                                [2/3, 4/7]]
    \# ((C.T)/(C.sum(axis=1))).T = [[1/3, 2/3]
                                [3/7, 4/7]]
    \# sum of row elements = 1
```

```
B = (C/C.sum(axis=0))
    #divid each element of the confusion matrix with the sum of elements in that row
          [3, 4]]
   # C.sum(axis = 0) axis=0 corresonds to columns and axis=1 corresponds to rows in two
diamensional array
   \# C.sum(axix = 0) = [[4, 6]]
   \# (C/C.sum(axis=0)) = [[1/4, 2/6],
                            [3/4, 4/6]]
   plt.figure(figsize=(20,4))
   labels = [1,2]
   # representing A in heatmap format
   cmap=sns.light_palette("blue")
   plt.subplot(1, 3, 1)
   sns.heatmap(C, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=labels)
   plt.xlabel('Predicted Class')
   plt.ylabel('Original Class')
   plt.title("Confusion matrix")
   plt.subplot(1, 3, 2)
   sns.heatmap(B, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=labels)
   plt.xlabel('Predicted Class')
   plt.ylabel('Original Class')
   plt.title("Precision matrix")
   plt.subplot(1, 3, 3)
    # representing B in heatmap format
   sns.heatmap(A, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=labels)
   plt.xlabel('Predicted Class')
   plt.ylabel('Original Class')
   plt.title("Recall matrix")
   plt.show()
```

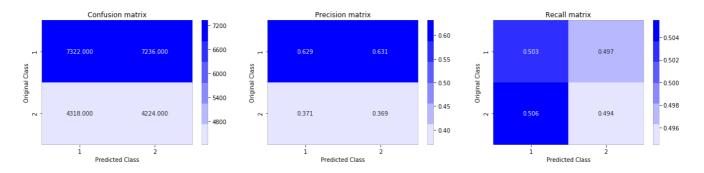
4.4 Building a random model (Finding worst-case log-loss)

In [42]:

```
# we need to generate 9 numbers and the sum of numbers should be 1
# one solution is to genarate 9 numbers and divide each of the numbers by their sum
# ref: https://stackoverflow.com/a/18662466/4084039
# we create a output array that has exactly same size as the CV data
predicted_y = np.zeros((test_len,2))
for i in range(test_len):
    rand_probs = np.random.rand(1,2)
    predicted_y[i] = ((rand_probs/sum(sum(rand_probs)))[0])
print("Log loss on Test Data using Random Model",log_loss(y_test, predicted_y, eps=1e-15))

predicted_y =np.argmax(predicted_y, axis=1)
plot_confusion_matrix(y_test, predicted_y)
```

 ${\tt Log~loss~on~Test~Data~using~Random~Model~0.8888108603784698}$



4.5 XGBoost

In [44]:

```
|print (type(X train))
print (type(X test))
X_test = pd.DataFrame(X_test.toarray())
print (type(X test))
X_train = X_train.as_matrix()
X test = X test.as matrix()
<class 'pandas.core.frame.DataFrame'>
<class 'scipy.sparse.coo.coo matrix'>
<class 'pandas.core.frame.DataFrame'>
In [45]:
print (type(X train))
print (type(X_test))
<class 'numpy.ndarray'>
<class 'numpy.ndarray'>
In [46]:
from xgboost import XGBClassifier
from scipy.stats import uniform, randint
from sklearn.model_selection import cross val score, GridSearchCV, KFold, RandomizedSearchCV,
train test split
import warnings
warnings.filterwarnings("ignore")
XGB = XGBClassifier(booster='gbtree',objective = 'binary:logistic',eval metric= 'logloss')
param_grid = {'max_depth':[3,9,11],'n_estimators':[50,100,300,500]}
xgboost GBDT = RandomizedSearchCV(XGB,param grid,cv=3, verbose=2)
xgboost GBDT.fit(X train,y train)
xgboost_GBDT.best_params_
Fitting 3 folds for each of 10 candidates, totalling 30 fits
[Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
[CV] n estimators=100, max depth=3 ......
[CV] ...... n_estimators=100, max_depth=3, total= 20.3s
[Parallel(n jobs=1)]: Done 1 out of 1 | elapsed: 20.2s remaining:
[CV] n estimators=100, max depth=3 ......
[CV] ...... n estimators=100, max depth=3, total= 19.7s
[CV] ...... n estimators=100, max depth=3, total= 19.6s
[CV] n estimators=50, max depth=3 ......
[CV] ...... n estimators=50, max depth=3, total= 10.2s
[CV] n estimators=50, max depth=3 .....
[CV] ...... n estimators=50, max depth=3, total= 10.2s
[CV] n estimators=50, max depth=3 ......
[CV] ...... n estimators=50, max depth=3, total= 10.1s
[CV] n_estimators=50, max_depth=11 .....
[CV] n_estimators=50, max_depth=11 ......
[CV] ...... n_estimators=50, max_depth=11, total= 34.2s
[CV] n_estimators=50, max_depth=11 .....
[CV] n estimators=300, max depth=9 ......
[CV] ..... n estimators=300, max depth=9, total= 2.8min
[CV] n estimators=300, max depth=9 .....
[CV] ...... n_estimators=300, max_depth=9, total= 2.7min
[CV] n estimators=300, max depth=9 .....
[CV] ...... n estimators=300, max depth=9, total= 2.7min
[CV] n estimators=100, max depth=9 ......
[CV] ...... n estimators=100, max depth=9, total= 54.2s
```

```
[CV] n estimators=100, max depth=9 ......
[CV] n estimators=300, max depth=11 .....
[CV] ...... n estimators=300, max depth=11, total= 3.3min
[CV] n_estimators=300, max_depth=11 ......
[CV] ...... n_estimators=300, max_depth=11, total= 3.5min
[CV] n estimators=300, max depth=11 .....
[CV] ...... n_estimators=300, max_depth=11, total= 3.4min
[CV] n estimators=100, max depth=11 ......
[CV] ...... n_estimators=100, max_depth=11, total= 1.1min
[CV] n_estimators=100, max_depth=11 ......
[CV] ...... n estimators=100, max depth=11, total= 1.1min
[CV] ..... n estimators=100, max depth=11, total= 1.1min
[CV] n estimators=500, max depth=3 ......
[CV] ...... n estimators=500, max depth=3, total= 1.6min
[CV] n estimators=500, max_depth=3 .....
[CV] ...... n_estimators=500, max_depth=3, total= 1.6min
[CV] n_estimators=500, max_depth=3 ......
[CV] ...... n_estimators=500, max_depth=3, total= 1.6min
[CV] n_estimators=300, max_depth=3 ......
[CV] ...... n_estimators=300, max_depth=3, total= 1.0min
[CV] n estimators=300, max depth=3 .....
[CV] ...... n_estimators=300, max_depth=3, total= 1.0min
[CV] n estimators=300, max depth=3 ......
[CV] ...... n estimators=300, max depth=3, total= 59.8s
[CV] n estimators=500, max depth=9 .....
[CV] ..... n estimators=500, max depth=9, total= 4.5min
[CV] n estimators=500, max depth=9 ......
[CV] ...... n_estimators=500, max_depth=9, total= 4.6min
[CV] n estimators=500, max depth=9 ......
[CV] ...... n_estimators=500, max_depth=9, total= 4.5min
[Parallel(n jobs=1)]: Done 30 out of 30 | elapsed: 49.3min finished
Out[46]:
{'n estimators': 300, 'max depth': 9}
In [47]:
from xgboost import XGBClassifier
from scipy.stats import uniform, randint
from sklearn.model_selection import cross_val_score, GridSearchCV, KFold, RandomizedSearchCV,
train test split
import warnings
besthyperpara xgboost = XGBClassifier(booster='gbtree',objective = 'binary:logistic',eval metric= '
logloss', max depth=9, n estimators=300)
besthyperpara xgboost.fit(X train,y train)
sig clf = CalibratedClassifierCV(besthyperpara xgboost, method="sigmoid")
sig clf.fit(X train, y train)
predict_y_xgboo = sig_clf.predict_proba(X_test)
In [48]:
print (log loss(y test, predict y xgboo, labels=besthyperpara xgboost.classes , eps=1e-15))
0.35881632092238686
In [49]:
predicted y xgboo =np.argmax(predict y xgboo,axis=1)
plot_confusion_matrix(y_test, predicted_y_xgboo)
       Confusion matrix
                                  Precision matrix
                                                            Recall matrix
                      12000
```

0.75

10000

[UV] n_estimators=100, max_depth=3, total= 34.28

