KDD2014

March 21, 2016

0.1 KDD2014 Project

import necessary library

```
In [1]: # imports
        import pandas as pd
        import matplotlib.pyplot as plt
        import csv
        import pandas as pd
        import numpy as np
        from sklearn.preprocessing import LabelEncoder, OneHotEncoder
        from sklearn.linear_model import LogisticRegression

        # this allows plots to appear directly in the notebook
        %matplotlib inline
```

load the data

```
In [2]: # load the data
        print('loading the data...')
        projects = pd.read_csv('./data/projects.csv')
        outcomes = pd.read_csv('./data/outcomes.csv')
        sample = pd.read_csv('./data/sampleSubmission.csv')
        print('complete..')
```

```
loading the data...
complete..
```

check data information

```
In [3]: print("projects", projects.shape)
        print("outcomes", outcomes.shape)
```

```
projects (664098, 35)
outcomes (619326, 12)
```

```
In [4]: projects.head(10)
```

```
Out[4]:                          projectid                    teacher_acctid  \
        0  316ed8fb3b81402ff6ac8f721bb31192  42d43fa6f37314365d08692e08680973
        1  90de744e368a7e4883223ca49318ae30  864eb466462bf704bf7a16a585ef296a
        2  32943bb1063267de6ed19fc0ceb4b9a7  37f85135259ece793213aca9d8765542
        3  bb18f409abda2f264d5acda8cab577a9  2133fc46f951f1e7d60645b0f9e48a6c
        4  24761b686e18e5eace634607acbcc19f  867ff478a63f5457eaf41049536c47cd
        5  eac7d156205f1333de3887d656f46611  ff064802c18e68db7ddb7ea0bf7732e8
```

```
6  5a3bfdf2e05781ccd0654dee0d51d1cd  085794a9e315b88cb7aec548831572f5
7  afda16eb54d8992db7bb42923b2a0c69  1d94a31c2dc38d51350eb26a4a2d892c
8  2ab3efb23acc84017cd7896f62c2e889  5a497c425e05bb193564ef3ce2cbf3b6
9  3118962680bb062323c3566197487315  eb0855cc6ea55d173ca8f8e4fd6a9e18

                          schoolid  school_ncesid  school_latitude  \
0  c0e6ce89b244764085691a1b8e28cb81   6.362701e+10        36.576340
1  d711e47810900c96f26a5d0be30c446d   4.837020e+11        32.911179
2  665c3613013ba0a66e3a2a26b89f1b68   4.103270e+11        45.166039
3  4f12c3fa0c1cce823c7ba1df57e90ccb   3.600153e+11        40.641727
4  10179fd362d7b8cf0e89baa1ca3025bb   6.227100e+10        34.043939
5  929336e5a242d2e67f1a591196ba7701   4.018700e+10        33.298792
6  dc34b021fd177cfdb22f22055ca9d04f   1.006000e+10        32.838778
7  86fff7caa07e4f595a716d8a1960f09e   1.301290e+11        33.896013
8  ed170a147cd8ee2391b54b2f8b206160   2.612000e+11        42.422054
9  71b4dee33799c2111773bb4444766bf0   3.402640e+11        39.951767

   school_longitude  school_city school_state  school_zip school_metro  \
0       -119.608713        Selma           CA       93662          NaN
1        -96.723640       Dallas           TX       75243        urban
2       -122.414576       Colton           OR       97017        rural
3        -73.965655     Brooklyn           NY       11226        urban
4       -118.288371  Los Angeles           CA       90006        urban
5       -111.827793     Chandler           AZ       85225     suburban
6        -85.517527    Lafayette           AL       36862        rural
7        -84.554213     Marietta           GA       30060          NaN
8        -83.209656      Detroit           MI       48235        urban
9        -75.117188       Camden           NJ        8102        urban

      ...     resource_type    poverty_level    grade_level  \
0     ...             Books  highest poverty     Grades 6-8
1     ...             Books  highest poverty  Grades PreK-2
2     ...        Technology     high poverty  Grades PreK-2
3     ...             Books  highest poverty     Grades 3-5
4     ...             Other  highest poverty  Grades PreK-2
5     ...        Technology  highest poverty     Grades 3-5
6     ...          Supplies  highest poverty  Grades PreK-2
7     ...             Books  highest poverty  Grades PreK-2
8     ...        Technology  highest poverty  Grades PreK-2
9     ...        Technology  highest poverty  Grades PreK-2

  fulfillment_labor_materials total_price_excluding_optional_support  \
0                          30                                 555.81
1                          30                                 296.47
2                          30                                 430.89
3                          30                                 576.07
4                          30                                 408.40
5                          30                                 750.92
6                          30                                2291.48
7                          30                                 459.36
8                          30                                 325.39
9                          30                                 567.84

  total_price_including_optional_support students_reached  \
```

```
0                                  653.89              32
1                                  348.79              22
2                                  506.93              17
3                                  677.73              12
4                                  480.47              24
5                                  883.44              20
6                                 2695.86             320
7                                  540.42              18
8                                  382.81              25
9                                  668.05              28

  eligible_double_your_impact_match eligible_almost_home_match date_posted
0                                 f                         f  2014-05-12
1                                 f                         f  2014-05-12
2                                 f                         f  2014-05-11
3                                 f                         f  2014-05-11
4                                 f                         f  2014-05-11
5                                 f                         f  2014-05-11
6                                 f                         f  2014-05-11
7                                 f                         f  2014-05-11
8                                 f                         f  2014-05-11
9                                 f                         f  2014-05-11

[10 rows x 35 columns]
```

In [5]:
```
outcomes.head()
```

Out[5]:
```
                          projectid is_exciting  \
0  ffffc4f85b60efc5b52347df489d0238           f
1  ffffac55ee02a49d1abc87ba6fc61135           f
2  ffff97ed93720407d70a2787475932b0           f
3  ffff418bb42fad24347527ad96100f81           f
4  ffff2d9c769c8fb5335e949c615425eb           t

  at_least_1_teacher_referred_donor fully_funded at_least_1_green_donation  \
0                               NaN            f                        NaN
1                                 f            t                          t
2                                 f            t                          t
3                                 f            f                          t
4                                 t            t                          t

  great_chat three_or_more_non_teacher_referred_donors  \
0          f                                       NaN
1          f                                         t
2          t                                         t
3          t                                         f
4          t                                         f

  one_non_teacher_referred_donor_giving_100_plus  \
0                                            NaN
1                                              f
2                                              t
3                                              f
4                                              t
```

```
  donation_from_thoughtful_donor  great_messages_proportion  \
0                            NaN                        NaN
1                              f                         57
2                              f                        100
3                              f                        100
4                              f                         63

   teacher_referred_count  non_teacher_referred_count
0                     NaN                         NaN
1                       0                           7
2                       0                           3
3                       0                           1
4                       6                           2
```

In [6]:
```
# sort the data based on id
projects = projects.sort('projectid')
sample = sample.sort('projectid')
outcomes = outcomes.sort('projectid')
```

```
/home/ubuntu/anaconda3/lib/python3.5/site-packages/ipykernel/__main__.py:2: FutureWarning: sort(columns=
  from ipykernel import kernelapp as app
/home/ubuntu/anaconda3/lib/python3.5/site-packages/ipykernel/__main__.py:3: FutureWarning: sort(columns=
  app.launch_new_instance()
/home/ubuntu/anaconda3/lib/python3.5/site-packages/ipykernel/__main__.py:4: FutureWarning: sort(columns=
```

In [7]:
```
projects.head()
```

Out[7]:

```
                                projectid                    teacher_acctid  \
148979  00001ccc0e81598c4bd86bacb94d7acb  96963218e74e10c3764a5cfb153e6fea
437277  00002bff514104264a6b798356fdd893  3414541eb63108700b188648f866f483
405458  00002d691c05c51a5fdfbb2baef0ba25  7ad6abc974dd8b62773f79f6cbed48d5
91352   0000b38bbc7252972f7984848cf58098  e1aa1ae5301d0cda860c4d9c89c24919
49606   0000ee613c92ddc5298bf63142996a5c  e0c0a0214d3c2cfdc0ab6639bc3c5342

                                schoolid  school_ncesid  school_latitude  \
148979  9f3f9f2c2da7edda5648ccd10554ed8c   1.709930e+11        41.807654
437277  cbaae3265eda78d330cb8ab1a9217071   6.032700e+10        35.203447
405458  56502bae9e97bab5eb54f9001878f469   6.029700e+10        34.137997
91352   30fcfca739b17be54ce3f1ee46980340   2.311400e+11        44.437717
49606   38bb0d62aa613c2f933de56c9df855b7   5.101260e+11        38.851982

        school_longitude   school_city school_state  school_zip school_metro  \
148979        -87.673257       Chicago           IL       60609        urban
437277       -118.840956         Arvin           CA       93203          NaN
405458       -118.062795       Arcadia           CA       91007        urban
91352         -70.201292     Livermore           ME        4253        rural
49606         -77.145287  Falls Church           VA       22041     suburban

              ...     resource_type     poverty_level    grade_level  \
148979        ...           Supplies   highest poverty  Grades PreK-2
437277        ...           Supplies   highest poverty  Grades PreK-2
405458        ...              Books  moderate poverty     Grades 3-5
91352         ...         Technology   highest poverty  Grades PreK-2
49606         ...         Technology      high poverty  Grades PreK-2
```

```
       fulfillment_labor_materials total_price_excluding_optional_support  \
148979                          30                                1273.82
437277                          35                                 477.32
405458                          35                                 892.31
91352                           30                                 547.86
49606                           30                                 384.86

       total_price_including_optional_support students_reached  \
148979                                1498.61               31
437277                                 561.55               20
405458                                1049.78              250
91352                                  644.54               36
49606                                  452.78               19

       eligible_double_your_impact_match eligible_almost_home_match  \
148979                                 f                          f
437277                                 t                          f
405458                                 f                          f
91352                                  t                          f
49606                                  f                          f

       date_posted
148979  2013-04-14
437277  2010-09-08
405458  2010-12-10
91352   2013-09-27
49606   2013-12-11

[5 rows x 35 columns]
```

check the missing data

In [8]:
```
totalCount = projects.shape[0]

for i in range(1,projects.shape[1]):
    nullcount = projects[projects[projects.columns[i]].isnull()].shape[0]
    percentage=float(nullcount)/float(totalCount) *100
    if(percentage>0):
        print(projects.columns[i],percentage,'%')
```

```
school_ncesid 6.4351948055859225 %
school_zip 0.0006023207418182256 %
school_metro 12.333721830211806 %
school_district 0.14275001581091948 %
school_county 0.002559863152727459 %
teacher_prefix 0.0006023207418182256 %
primary_focus_subject 0.0058726272327277 %
primary_focus_area 0.0058726272327277 %
secondary_focus_subject 31.304566494704094 %
secondary_focus_area 31.304566494704094 %
resource_type 0.006776108345455037 %
grade_level 0.0013552216690910076 %
fulfillment_labor_materials 5.282654066116748 %
students_reached 0.021984707076365236 %
```

fill up the missing data

```
In [9]: projects = projects.fillna(method='pad')

In [10]: projects.head(10)

Out[10]:                              projectid                   teacher_acctid  \
148979  00001ccc0e81598c4bd86bacb94d7acb  96963218e74e10c3764a5cfb153e6fea
437277  00002bff514104264a6b798356fdd893  3414541eb63108700b188648f866f483
405458  00002d691c05c51a5fdfbb2baef0ba25  7ad6abc974dd8b62773f79f6cbed48d5
91352   0000b38bbc7252972f7984848cf58098  e1aa1ae5301d0cda860c4d9c89c24919
49606   0000ee613c92ddc5298bf63142996a5c  e0c0a0214d3c2cfdc0ab6639bc3c5342
255442  0000fa3aa8f6649abab23615b546016d  2a578595fe351e7fce057e048c409b18
189646  0000fb6aea57099cc5b051acb7f52a9e  ad51bb5eabffc738775887955421fe75
616019  0001120447a33dd9ffeefa107ed04c43  a799e714a102967d674b258e5ea19231
301504  0001146d343ea9452089d0e302496c06  3f71761d508f95684f2924763175dbe8
511584  0001151477ea5349a0aa64ed1d83f0bc  f30b9edaea56bbade550e2f0da5db4f9

                                schoolid  school_ncesid  school_latitude  \
148979  9f3f9f2c2da7edda5648ccd10554ed8c   1.709930e+11        41.807654
437277  cbaae3265eda78d330cb8ab1a9217071   6.032700e+10        35.203447
405458  56502bae9e97bab5eb54f9001878f469   6.029700e+10        34.137997
91352   30fcfca739b17be54ce3f1ee46980340   2.311400e+11        44.437717
49606   38bb0d62aa613c2f933de56c9df855b7   5.101260e+11        38.851982
255442  3432ed3d4466fac2f2ead83ab354e333   6.409801e+10        34.296596
189646  d4f02777656b5ee806965ae2186e0adb   4.702940e+11        35.037663
616019  c843a6322e90dc34304b60b43f4c2205   4.502580e+11        34.571828
301504  18e8fc522b79044cf70938cbefce41bb   4.503870e+11        34.977550
511584  6cd638cff9af07d02c72bb1cc25612d5   2.612000e+11        42.367172

        school_longitude   school_city school_state  school_zip school_metro  \
148979        -87.673257       Chicago           IL       60609        urban
437277       -118.840956         Arvin           CA       93203        urban
405458       -118.062795       Arcadia           CA       91007        urban
91352         -70.201292     Livermore           ME        4253        rural
49606         -77.145287  Falls Church           VA       22041     suburban
255442       -119.296596       Ventura           CA       93001        urban
189646        -90.092321       Memphis           TN       38109        urban
616019        -80.615642       Kershaw           SC       29067        rural
301504        -81.012395     Rock Hill           SC       29732        urban
511584        -82.985527       Detroit           MI       48214        urban

             ...     resource_type     poverty_level    grade_level  \
148979       ...           Supplies   highest poverty  Grades PreK-2
437277       ...           Supplies   highest poverty  Grades PreK-2
405458       ...              Books  moderate poverty     Grades 3-5
91352        ...         Technology   highest poverty  Grades PreK-2
49606        ...         Technology      high poverty  Grades PreK-2
255442       ...              Books   highest poverty     Grades 3-5
189646       ...              Books   highest poverty     Grades 6-8
616019       ...              Books      high poverty     Grades 6-8
301504       ...         Technology      high poverty     Grades 3-5
511584       ...           Supplies   highest poverty     Grades 3-5

        fulfillment_labor_materials total_price_excluding_optional_support  \
148979                           30                                1273.82
437277                           35                                 477.32
```

```
405458                          35                                  892.31
91352                           30                                  547.86
49606                           30                                  384.86
255442                          35                                  240.10
189646                          30                                  382.71
616019                          17                                  296.00
301504                          35                                  300.97
511584                           9                                  675.24

       total_price_including_optional_support students_reached  \
148979                                 1498.61               31
437277                                  561.55               20
405458                                 1049.78              250
91352                                   644.54               36
49606                                   452.78               19
255442                                  282.47               28
189646                                  450.25               90
616019                                  360.98               35
301504                                  354.08               21
511584                                  823.46              300

       eligible_double_your_impact_match eligible_almost_home_match  \
148979                                 f                           f
437277                                 t                           f
405458                                 f                           f
91352                                  t                           f
49606                                  f                           f
255442                                 t                           f
189646                                 f                           t
616019                                 f                           f
301504                                 f                           t
511584                                 f                           f

       date_posted
148979  2013-04-14
437277  2010-09-08
405458  2010-12-10
91352   2013-09-27
49606   2013-12-11
255442  2012-04-07
189646  2012-11-17
616019  2007-08-12
301504  2011-12-08
511584  2009-08-28

[10 rows x 35 columns]
```

In [11]:
```
# split the training data and testing data
dates = np.array(projects.date_posted)
```

In [12]:
```
print(dates)
```

```
['2013-04-14' '2010-09-08' '2010-12-10' ..., '2010-09-11' '2011-06-11'
 '2009-10-11']
```

```
In [13]: train_idx = np.where(dates < '2014-01-01')[0]
         test_idx = np.where(dates >= '2014-01-01')[0]

In [14]: print(train_idx)

[     0      1      2 ..., 664095 664096 664097]

In [15]: print(test_idx)

[    33     52     53 ..., 664041 664088 664094]

In [16]: # check the data of training data
         print(projects.iloc[train_idx[9999]])

projectid                                    04253b89bf4b42f1e5f29b23418d36c3
teacher_acctid                               8aac3a46b9ac0df676cfc740d272a425
schoolid                                     8827b62394964f26a2764245dac80110
school_ncesid                                                     6.2808e+10
school_latitude                                                      37.9825
school_longitude                                                    -121.717
school_city                                                           Oakley
school_state                                                              CA
school_zip                                                             94561
school_metro                                                        suburban
school_district                                 Oakley Union Elem Sch Dist
school_county                                                  Contra Costa
school_charter                                                            f
school_magnet                                                             f
school_year_round                                                         t
school_nlns                                                               f
school_kipp                                                               f
school_charter_ready_promise                                              f
teacher_prefix                                                          Ms.
teacher_teach_for_america                                                 f
teacher_ny_teaching_fellow                                                f
primary_focus_subject                                       Special Needs
primary_focus_area                                          Special Needs
secondary_focus_subject                                           Literacy
secondary_focus_area                                 Literacy & Language
resource_type                                                        Books
poverty_level                                           moderate poverty
grade_level                                                   Grades 3-5
fulfillment_labor_materials                                           30
total_price_excluding_optional_support                            199.08
total_price_including_optional_support                            234.21
students_reached                                                      10
eligible_double_your_impact_match                                         f
eligible_almost_home_match                                                f
date_posted                                                   2013-10-04
Name: 87071, dtype: object

In [17]: #preprocessing the data based on different types of attr
         projects_numeric_columns = ['school_latitude', 'school_longitude',
                                     'fulfillment_labor_materials',
                                     'total_price_excluding_optional_support',
                                     'total_price_including_optional_support']
```

```
In [18]: projects_id_columns = ['projectid', 'teacher_acctid', 'schoolid', 'school_ncesid']

In [19]: projects_categorial_columns = np.array(list(set(projects.columns).difference(set(projects_nume

In [20]: print(projects_categorial_columns)

['teacher_teach_for_america' 'school_charter_ready_promise' 'grade_level'
 'teacher_prefix' 'school_metro' 'poverty_level' 'primary_focus_area'
 'primary_focus_subject' 'school_kipp' 'resource_type' 'students_reached'
 'school_district' 'eligible_double_your_impact_match' 'school_city'
 'school_year_round' 'school_state' 'school_zip' 'school_nlns'
 'school_charter' 'eligible_almost_home_match' 'secondary_focus_area'
 'school_magnet' 'secondary_focus_subject' 'teacher_ny_teaching_fellow'
 'school_county']

In [21]: projects_categorial_values = np.array(projects[projects_categorial_columns])

In [22]: projects[projects_categorial_columns].head()

Out[22]:
```

	teacher_teach_for_america	school_charter_ready_promise	grade_level
148979	f	f	Grades PreK-2
437277	f	f	Grades PreK-2
405458	f	f	Grades 3-5
91352	f	f	Grades PreK-2
49606	f	f	Grades PreK-2

	teacher_prefix	school_metro	poverty_level	primary_focus_area
148979	Mrs.	urban	highest poverty	Math & Science
437277	Mrs.	urban	highest poverty	Literacy & Language
405458	Mr.	urban	moderate poverty	Literacy & Language
91352	Mrs.	rural	highest poverty	Literacy & Language
49606	Ms.	suburban	high poverty	Literacy & Language

	primary_focus_subject	school_kipp	resource_type	...
148979	Mathematics	f	Supplies	...
437277	Literacy	f	Supplies	...
405458	Literature & Writing	f	Books	...
91352	Literature & Writing	f	Technology	...
49606	ESL	f	Technology	...

	school_state	school_zip	school_nlns	school_charter
148979	IL	60609	f	f
437277	CA	93203	f	f
405458	CA	91007	f	f
91352	ME	4253	f	f
49606	VA	22041	f	f

	eligible_almost_home_match	secondary_focus_area	school_magnet
148979	f	Music & The Arts	f
437277	f	Literacy & Language	f
405458	f	Literacy & Language	f
91352	f	Math & Science	f
49606	f	Special Needs	t

	secondary_focus_subject	teacher_ny_teaching_fellow	school_county

```
148979              Visual Arts                          f          Cook
437277    Literature & Writing                           f          Kern
405458                 Literacy                          f   Los Angeles
91352               Mathematics                          f  Androscoggin
49606             Special Needs                          f       Fairfax

[5 rows x 25 columns]
```

In [23]:
```
print(projects_categorial_values)
```

```
[['f' 'f' 'Grades PreK-2' ..., 'Visual Arts' 'f' 'Cook']
 ['f' 'f' 'Grades PreK-2' ..., 'Literature & Writing' 'f' 'Kern']
 ['f' 'f' 'Grades 3-5' ..., 'Literacy' 'f' 'Los Angeles']
 ...,
 ['f' 'f' 'Grades PreK-2' ..., 'Other' 'f' 'Kings (Brooklyn)']
 ['f' 'f' 'Grades PreK-2' ..., 'Literacy' 'f' 'Clayton']
 ['f' 'f' 'Grades 3-5' ..., 'Mathematics' 'f' 'Florence']]
```

In [24]:
```
# only use the category attr
print(projects_categorial_columns)
print(projects_categorial_columns.shape)
print(projects_categorial_values[:, 0].shape)
```

```
['teacher_teach_for_america' 'school_charter_ready_promise' 'grade_level'
 'teacher_prefix' 'school_metro' 'poverty_level' 'primary_focus_area'
 'primary_focus_subject' 'school_kipp' 'resource_type' 'students_reached'
 'school_district' 'eligible_double_your_impact_match' 'school_city'
 'school_year_round' 'school_state' 'school_zip' 'school_nlns'
 'school_charter' 'eligible_almost_home_match' 'secondary_focus_area'
 'school_magnet' 'secondary_focus_subject' 'teacher_ny_teaching_fellow'
 'school_county']
(25,)
(664098,)
```

In [25]:
```
# encode the category value and reform the original data
label_encoder = LabelEncoder()

# set up the encoding model, using the first row of data
projects_data = label_encoder.fit_transform(projects_categorial_values[:,0])
```

In [26]:
```
# use the model to transform the following data
for i in range(1, projects_categorial_values.shape[1]):
    label_encoder = LabelEncoder()
    projects_data = np.column_stack((projects_data, label_encoder.fit_transform(projects_categ
```

In [27]:
```
projects_data = projects_data.astype(float)
print('The shape of the project data', projects_data.shape)
```

```
The shape of the project data (664098, 25)
```

In [28]:
```
# one hot encoding
enc = OneHotEncoder()
enc.fit(projects_data)
projects_data = enc.transform(projects_data)
```

In [29]:
```
print(projects_data)
```

```
(0, 36457)        1.0
  (0, 36090)        1.0
  (0, 36089)        1.0
  (0, 36061)        1.0
  (0, 36059)        1.0
  (0, 36052)        1.0
  (0, 36050)        1.0
  (0, 36048)        1.0
  (0, 29542)        1.0
  (0, 19391)        1.0
  (0, 19375)        1.0
  (0, 11779)        1.0
  (0, 10366)        1.0
  (0, 7439)        1.0
  (0, 93)        1.0
  (0, 58)        1.0
  (0, 54)        1.0
  (0, 44)        1.0
  (0, 24)        1.0
  (0, 17)        1.0
  (0, 15)        1.0
  (0, 11)        1.0
  (0, 7)        1.0
  (0, 2)        1.0
  (0, 0)        1.0
  :        :
  (664097, 36627)        1.0
  (664097, 36090)        1.0
  (664097, 36080)        1.0
  (664097, 36061)        1.0
  (664097, 36058)        1.0
  (664097, 36052)        1.0
  (664097, 36050)        1.0
  (664097, 36048)        1.0
  (664097, 23853)        1.0
  (664097, 19418)        1.0
  (664097, 19375)        1.0
  (664097, 13051)        1.0
  (664097, 10366)        1.0
  (664097, 3859)        1.0
  (664097, 152)        1.0
  (664097, 59)        1.0
  (664097, 54)        1.0
  (664097, 44)        1.0
  (664097, 24)        1.0
  (664097, 16)        1.0
  (664097, 15)        1.0
  (664097, 11)        1.0
  (664097, 4)        1.0
  (664097, 2)        1.0
  (664097, 0)        1.0
```

In [30]:
```
#Predicting
train = projects_data[train_idx]
```

```
        test = projects_data[test_idx]
        print('shape of test', test.shape)
        clf = LogisticRegression()

shape of test (44772, 37794)

In [31]: print(test)

(0, 0)        1.0
  (0, 2)        1.0
  (0, 6)        1.0
  (0, 9)        1.0
  (0, 13)        1.0
  (0, 16)        1.0
  (0, 20)        1.0
  (0, 30)        1.0
  (0, 54)        1.0
  (0, 59)        1.0
  (0, 262)        1.0
  (0, 8171)        1.0
  (0, 10366)        1.0
  (0, 12981)        1.0
  (0, 19375)        1.0
  (0, 19399)        1.0
  (0, 20111)        1.0
  (0, 36048)        1.0
  (0, 36050)        1.0
  (0, 36053)        1.0
  (0, 36056)        1.0
  (0, 36061)        1.0
  (0, 36086)        1.0
  (0, 36090)        1.0
  (0, 36638)        1.0
  :        :
  (44771, 0)        1.0
  (44771, 2)        1.0
  (44771, 7)        1.0
  (44771, 11)        1.0
  (44771, 13)        1.0
  (44771, 16)        1.0
  (44771, 23)        1.0
  (44771, 42)        1.0
  (44771, 54)        1.0
  (44771, 57)        1.0
  (44771, 86)        1.0
  (44771, 7786)        1.0
  (44771, 10366)        1.0
  (44771, 17894)        1.0
  (44771, 19375)        1.0
  (44771, 19405)        1.0
  (44771, 23207)        1.0
  (44771, 36048)        1.0
  (44771, 36050)        1.0
  (44771, 36052)        1.0
  (44771, 36057)        1.0
```

```
  (44771, 36061)        1.0
  (44771, 36079)        1.0
  (44771, 36090)        1.0
  (44771, 37348)        1.0
```

In [32]:
```
# set the target labels
labels = np.array(outcomes.is_exciting)
clf.fit(train, labels=='t')
```

Out[32]:
```
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
          intercept_scaling=1, max_iter=100, multi_class='ovr', n_jobs=1,
          penalty='l2', random_state=None, solver='liblinear', tol=0.0001,
          verbose=0, warm_start=False)
```

In [33]:
```
# perform the prediction
preds = clf.predict(test)
```

In [34]:
```
print(preds.shape)
```

```
(44772,)
```

In [35]:
```
print(preds)
```

```
[False False False ..., False False False]
```

In [36]:
```
f = lambda x : 1 if x else 0
f = np.vectorize(f)
preds = f(preds)
preds.astype(int)
print(preds)
```

```
[0 0 0 ..., 0 0 0]
```

In [37]:
```
#Save prediction into a file
sample['is_exciting'] = preds
```

In [38]:
```
sample.head()
```

Out[38]:

	projectid	is_exciting
44771	00034e54ed99042609edad55031c8861	0
44770	00057f424b5498c7ece13d13ce3e2178	0
44769	00059e63bb0567708b2b0c9e3d9c43d6	0
44768	0008c67f27dd29ea7be5a7cc5a866df8	0
44767	000b6e707ad50a597ab46eedff6bde05	0

In [39]:
```
sample.tail()
```

Out[39]:

	projectid	is_exciting
4	fff745e9c0b8cc9e73e8c4c9a0ef4292	0
3	fff8beec6de8c9411520d15d1f6979bf	0
2	fff979abefa35a6bdd133b4e4150b737	0
1	fffeb510ee37a0bb01079f06bf141246	0
0	ffff7266778f71242675416e600b94e1	0

In [40]:
```
sample.to_csv('predictions.csv', index = False)
```

In [ ]: