

further analysis

May 16, 2016

```
In [1]: %matplotlib inline
```

```
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np

# Make the graphs a bit prettier, and bigger
pd.set_option('display.mpl_style', 'default')
plt.rcParams['figure.figsize'] = (15, 5)
plt.rcParams['font.family'] = 'sans-serif'

# This is necessary to show lots of columns in pandas 0.12.
# Not necessary in pandas 0.13.
pd.set_option('display.width', 5000)
pd.set_option('display.max_columns', 60)
```

```
In [2]: cd md
```

```
/home/raisa/md
```

```
In [54]: all_df=[]
         nfiles=15
         for i in range(nfiles):
             filename = 'msample%d.csv' % i
             print i
             all_df.append(pd.read_csv(filename, header=None))
```

```
0
1
2
3
4
5
6
7
8
9
10
11
12
13
14
```

```
In [55]: all_df[0]
```

Out [55] :	0	1	2	3	4
0	2	U26@DOM1	U26@DOM1	C616	U26
1	9	U101@DOM1	U101@DOM1	C1862	C1862
2	33	C2025\$@DOM1	C2025\$@DOM1	C467	C467
3	47	C2653\$@DOM1	C2653\$@DOM1	C2653	C2653
4	54	C2653\$@DOM1	C2653\$@DOM1	C2653	C586
5	55	C2653\$@DOM1	C2653\$@DOM1	C2653	C2653
6	95	U66@DOM1	U66@DOM1	C832	C832
7	128	C1114\$@DOM1	C1114\$@DOM1	C1115	C1114
8	164	C1114\$@DOM1	C1114\$@DOM1	C1115	C1114
9	174	C2692\$@DOM1	C2692\$@DOM1	C528	C528
10	205	U252@DOM1	U252@DOM1	C2627	C1315
11	213	C599\$@DOM1	C599\$@DOM1	C553	C553
12	239	C3390\$@DOM1	C3390\$@DOM1	C3392	C3392
13	243	U22@DOM1	U22@DOM1	C477	U22
14	286	C1607\$@DOM1	C1607\$@DOM1	C457	C457
15	308	C2096\$@?	C2096\$@?	C25240	C25240
16	335	U4@DOM1	U4@DOM1	C229	C229
17	355	C1714\$@DOM1	C1714\$@DOM1	C612	C612
18	363	C1527\$@DOM1	C1527\$@DOM1	C1527	C612
19	454	C2096\$@?	C2096\$@?	C457	C457
20	489	C2902\$@DOM1	C2902\$@DOM1	C2902	C1065
21	523	C4334\$@DOM1	C4334\$@DOM1	C4334	C2106
22	554	C2653\$@DOM1	C2653\$@DOM1	C2653	C2653
23	571	U1825@?	U1825@?	C612	C612
24	623	U22@DOM1	U22@DOM1	C506	U22
25	641	C860\$@DOM1	C860\$@DOM1	C860	C457
26	673	C2043\$@DOM1	C2043\$@DOM1	C529	C529
27	677	C2759\$@DOM1	C2759\$@DOM1	C2759	C2759
28	773	LOCAL SERVICE@C3049	LOCAL SERVICE@C3049	C3049	C3049
29	834	C2480\$@DOM1	C2480\$@DOM1	C2479	C2479
...
400682	5010840	U9@DOM1	C586\$@DOM1	C586	C586
400683	5010841	U59@?	U59@?	C1634	C1634
400684	5010861	U8929@?	U8929@?	C19037	C19037
400685	5010873	U59@?	U59@?	C1634	C1634
400686	5010874	U9@?	U9@?	C222	C222
400687	5010879	U22@DOM1	U22@DOM1	C849	U22
400688	5010879	U9@DOM1	U9@DOM1	C222	C222
400689	5010884	NETWORK SERVICE@C25102	NETWORK SERVICE@C25102	C25102	C25102
400690	5010900	C23484\$@DOM1	C23484\$@DOM1	C23484	C586
400691	5010907	C1692\$@DOM1	C1692\$@DOM1	C1692	C1692
400692	5010916	C743\$@DOM1	C743\$@DOM1	C586	C586
400693	5010938	U9@?	U9@?	C222	C222
400694	5010963	C2344\$@DOM1	C2344\$@DOM1	C457	C457
400695	5010970	U22@DOM1	U22@DOM1	C246	U22
400696	5011005	U59@?	U59@?	C1634	C1634
400697	5011008	C3188\$@DOM1	C3188\$@DOM1	C3188	C3188
400698	5011014	U101@?	U101@?	C3415	C3415
400699	5011015	U59@?	U59@?	C589	C589
400700	5011043	U10107@DOM1	U10107@DOM1	C419	C419
400701	5011067	C27118\$@DOM1	C27118\$@DOM1	C1369	C1369
400702	5011071	C21596\$@DOM1	C21596\$@DOM1	C21596	C612
400703	5011083	U9@?	U9@?	C222	C222

MICROSOFT_AUTH

400704	5011087	U9@DOM1	U9@DOM1	C222	C222
400705	5011110	C398\$@DOM1	C398\$@DOM1	C1767	C1767
400706	5011116	C1791\$@DOM1	C1791\$@DOM1	C1065	C1065
400707	5011120	C1617\$@DOM1	C1617\$@DOM1	C1618	C457
400708	5011157	C7780\$@DOM1	C7780\$@DOM1	C7780	C528
400709	5011161	U22@DOM1	U22@DOM1	C965	U22
400710	5011167	U6715@DOM1	U6715@DOM1	C10781	C10781
400711	5011195	U199@DOM1	U1825@DOM1	C1929	C1929

[400712 rows x 9 columns]

```
In [56]: Y=[]
         for i in range(nfiles):
             Y.append(all_df[i][8]=='Success')
```

```
In [13]: Y[1]
```

```
Out[13]: 0      False
         1      False
         2      False
         3      True
         4      True
         5      False
         6      True
         7      True
         8      True
         9      False
        10      False
        11      False
        12      True
        13      False
        14      True
        15      True
        16      False
        17      True
        18      True
        19      True
        20      False
        21      False
        22      True
        23      True
        24      True
        25      True
        26      False
        27      False
        28      True
        29      False
        ...
    400276      True
    400277      False
    400278      True
    400279      False
    400280      False
    400281      True
    400282      True
```

```

400283    False
400284     True
400285    False
400286     True
400287     True
400288    False
400289     True
400290    False
400291     True
400292     True
400293    False
400294    False
400295    False
400296     True
400297    False
400298    False
400299     True
400300    False
400301     True
400302    False
400303    False
400304     True
400305    False
Name: 8, dtype: bool

```

```

In [57]: def map_user(x):
        if x.startswith('C'):
            return 'C'
        elif x.startswith('U'):
            return 'U'
        else:
            return x

```

```

In [68]: X=[]
        for i in range(nfiles):
            df=all_df[i]
            df["source_user"], df["source_domain"] = zip(*df[1].str.split('@').tolist())
            df["source_user"]=df["source_user"].str.rstrip('$')
            df["destination_user"], df["destination_domain"] = zip(*df[2].str.split('@').tolist())
            df["destination_user"]=df["destination_user"].str.rstrip('$')
            df['source_class']=df['source_user'].map(map_user)
            df['destination_class']=df['destination_user'].map(map_user)
            x=pd.DataFrame.from_items([
                ('time', (df[0]%(24*60*60)).astype(int))])
            x['same_user']=(df['destination_user']==df['source_user'])
            x['same_domain']=(df['destination_domain']==df['source_domain'])
            x['source_user_comp_same']=(df[3]==df['source_user'])
            x['destination_user_comp_same']=(df['destination_user']==df[4])
            x['same_comp']=(df[3]==df[4])
            x['source_domain_comp_same']=(df[3]==df['source_domain'])
            x['destination_domain_comp_same']=(df['destination_domain']==df[4])

            for j in [5,6, 7]:
                for label in sorted(df[j].unique()):
                    if label=='?':

```

```

        if j==5:
            x['?_authentication type']=(df[j]==label)
        elif j==6:
            x['?_logon type']=(df[j]==label)
        else:
            x[label]=(df[j]==label)
    for cl in ['source_class', 'destination_class']:
        for label in sorted(df[cl].unique()):
            if cl=='source_class':
                x['source_'+label]=(df[cl]==label)
            else:
                x['destination_'+label]=(df[cl]==label)
X.append(x)

```

In [62]: X[1]

```

Out[62]:
      time same_user same_domain source_user_comp_same destination_user_comp_same same_comp so
0         2         True         True                False                True         False
1         3         True         True                False                False        True
2        11         True         True                False                False        False
3       140         True         True                 True                False        False
4       176         True         True                False                False         True
5       185         True         True                False                False         True
6       224         True         True                 True                False        False
7       250         True         True                False                False         True
8       252         True         True                False                False        False
9       333         True         True                 True                True         True
10      348         True         True                False                True         False
11      416         True         True                False                True         False
12      459         True         True                 True                True         True
13      470         True         True                 True                True         True
14      485         True         True                False                False        False
15      490         True         True                 True                False        False
16      510         True         True                False                True         False
17      542         True         True                False                False         True
18      551         True         True                 True                False        False
19      570         True         True                False                False         True
20      588         True         True                False                True         False
21      623         True         True                False                True         False
22      679         True         True                False                True         False
23      704         True         True                False                False         True
24      726         True         True                False                False         True
25      745         True         True                 True                False        False
26      750         True         True                False                False         True
27      859         True         True                False                False         True
28      936         True         True                False                False         True
29      975         True         True                False                False         True
...      ...      ...      ...      ...      ...
400276  86055         True         True                False                False        False
400277  86057         True         True                False                False         True
400278  86061         True         True                 True                True         True
400279  86073         True         True                False                False         True
400280  86089         True         True                False                False         True
400281  86093         True         True                False                False        False
400282  86104         True         True                False                False         True

```

400283	86127	True	True	False	False	True
400284	86131	True	True	False	False	True
400285	86151	True	True	False	False	True
400286	86179	True	True	True	False	False
400287	86251	True	True	False	False	True
400288	86259	True	True	False	False	True
400289	86267	True	True	True	True	True
400290	86267	True	True	False	False	True
400291	86275	True	True	False	False	True
400292	86280	True	True	False	False	True
400293	86281	True	True	False	False	True
400294	86281	True	True	False	True	False
400295	86287	True	True	False	False	True
400296	86298	True	True	False	False	True
400297	86341	True	True	False	False	True
400298	86347	True	True	False	False	True
400299	86354	True	True	False	False	True
400300	86356	True	True	False	False	True
400301	86372	True	True	False	False	True
400302	86373	True	True	False	False	True
400303	86374	True	True	False	False	True
400304	86391	True	True	False	False	True
400305	86393	True	True	False	False	False

[400306 rows x 56 columns]

In [63]: X[0].columns

Out[63]: Index([u'time', u'same_user', u'same_domain', u'source_user_comp.same', u'destination_user_comp.s

In [64]: [len(entry.columns) for entry in X]

Out[64]: [53, 56, 53, 54, 54, 52, 56, 56, 57, 55, 55, 54, 55, 54, 54]

In [65]: all_col = set(sum([list(entry.columns) for entry in X], []))
[all_col.difference(list(entry.columns)) for entry in X]

Out[65]: [{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
'CygwinLsa',
'MICROSOFT_AUTHENTICA',
'MICROSOFT_AUTHENTICATION_P',
'MICROSOFT_AUTHENTICATION_PA',
'MICROSOFT_AUTHENTICATION_PACK'},
{'CygwinLsa', 'MICROSOFT_AUTHENTICA', 'MICROSOFT_AUTHENTICATION_P'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
'CygwinLsa',
'MICROSOFT_AUTHENTICA',
'MICROSOFT_AUTHENTICATION_P',
'MICROSOFT_AUTHENTICATION_PA',
'MICROSOFT_AUTHENTICATION_PAC'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
'CygwinLsa',
'MICROSOFT_AUTHENTICA',
'MICROSOFT_AUTHENTICATION_P',
'MICROSOFT_AUTHENTICATION_PAC'}],

```

{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
 'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PAC'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
 'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PAC',
 'MICROSOFT_AUTHENTICATION_PACK',
 'Setuid'},
{'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_PAC'},
{'CygwinLsa', 'MICROSOFT_AUTHENTICATION_P'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
 'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_PACKAGE_V1'},
{'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA',
 'MICROSOFT_AUTHENTICATION_PAC'},
{'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA',
 'MICROSOFT_AUTHENTICATION_PAC'},
{'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA'},
{'CygwinLsa',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA',
 'Setuid'},
{'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE',
 'MICROSOFT_AUTHENTICA',
 'MICROSOFT_AUTHENTICATION_P',
 'MICROSOFT_AUTHENTICATION_PA',
 'MICROSOFT_AUTHENTICATION_PAC'}]

```

```

In [69]: col_set = [set(entry.columns) for entry in X]
common_subset = set.intersection(*col_set)
drop_cols = [e.difference(common_subset) for e in col_set]
for entry, to_drop in zip(X, drop_cols):
    print 'dropping', to_drop
    for item in to_drop:
        del entry[item]

```

```

dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACKAGE_V1', 'MICROSOFT_AUTHENTICATION_PAC'])
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_
dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACK', 'MICROSOFT_AUTHENTICATION_PACKAGE_V1'])
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'MICROSOFT_AUTHENTICATION_PACKAGE_V1', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'MICROSOFT_AUTHENTICATION_PACKAGE_V1', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'MICROSOFT_AUTHENTICATION_PACKAGE_V1'])
dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACK', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'CygwinLsa', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'MICROSOFT_AUTHENTICATION_P', 'CygwinLsa', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PA', 'Setuid', 'MICROSOFT_AUTHENTICATION_P', 'MICROSOFT_AUTHENTIC
dropping set(['Setuid', 'CygwinLsa', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_AUTHENTICATION_
dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACK', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_
dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACK', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_
dropping set(['MICROSOFT_AUTHENTICATION_PACK', 'ACRONIS_RELOGON_AUTHENTICATION_PACKAGE', 'MICROSOFT_
dropping set(['Setuid', 'MICROSOFT_AUTHENTICATION_PACK', 'CygwinLsa', 'MICROSOFT_AUTHENTICATION_PACKAGE_V

```

```

In [70]: col0 = list(X[0].columns)
         for i in range(1,nfiles):
             col_i = list(X[i].columns)
             assert col0 == col_i, 'mismatch in %r:\n%s\n%s' % (i, col0, col_i)

```

```

In [71]: from sklearn import linear_model
         clf_l1_LR = linear_model.LogisticRegression(C=1000, penalty='l1', tol=0.001).fit(X[0], Y[0])
         scores=[]
         scores.append(clf_l1_LR.score(X[0], Y[0]))
         print 'score for training set', scores[0]
         for i in range(1,nfiles):
             scores.append(clf_l1_LR.score(X[i], Y[i]))
             print 'score for test set', i, scores[i]

```

```

score for training set 0.944072051748
score for test set 1 0.94448247091
score for test set 2 0.943976919929
score for test set 3 0.944386639788
score for test set 4 0.944560448937
score for test set 5 0.943735713999
score for test set 6 0.944166904201
score for test set 7 0.943538001825
score for test set 8 0.944438192553
score for test set 9 0.943566597067
score for test set 10 0.944126539894
score for test set 11 0.944858468573
score for test set 12 0.944788959785
score for test set 13 0.944127431039
score for test set 14 0.943777194073

```

```

In [72]: print 'mean', np.mean(scores), 'std', np.std(scores)

```

```

mean 0.944173502288 std 0.00039856965612

```

Logistic regression with Lasso (L1 penalty) computed over 15 non-overlapping subsets of auth.txt.gz gave me a score with mean 0.9442 and std 0.0004. I believe I am sampling from a normal distribution, which means I have a very narrow gaussian. This in turn means that further sampling will not change my results significantly.

```

In [ ]:

```