

In [1]:

```
import pandas as pd
import numpy as np
# import scipy.stats as sps
# import tensorflow
import matplotlib.pyplot as plt
import seaborn as sns
# import sklearn as skl
# from sklearn import preprocessing
from sklearn.decomposition import PCA
# from sklearn.model_selection import train_test_split
# from keras.layers import Input, Dense
# from keras.models import Model
%matplotlib inline
```

In [2]:

```
%%time
fa_dir = '/Users/stevecoggeshall/Documents/Teaching/Fraud Analytics/2018 USC fra
ud class'
myvars = pd.read_csv(fa_dir + '/data/NY property/NY property vars 1 million zsca
le.csv', index_col=0)
```

CPU times: user 35.1 s, sys: 3.11 s, total: 38.2 s
Wall time: 40.4 s

In [3]:

```
def mem_usage(pandas_obj):
    if isinstance(pandas_obj, pd.DataFrame):
        usage_b = pandas_obj.memory_usage(deep=True).sum()
    else: # we assume if not a df it's a series
        usage_b = pandas_obj.memory_usage(deep=True)
    usage_mb = usage_b / 1024 ** 2 # convert bytes to megabytes
    return "{:03.2f} MB".format(usage_mb)
```

In [4]:

```
print(mem_usage(myvars))
```

488.00 MB

In [5]:

```
numrecords = len(myvars)
print(numrecords)
```

1048575

In [6]:

```
myvars.shape
```

Out[6]:

(1048575, 60)

In [7]:

```
%%time
mydata = (myvars - myvars.mean()) / myvars.std()
```

CPU times: user 1.42 s, sys: 1.86 s, total: 3.28 s
Wall time: 2.22 s

In [8]:

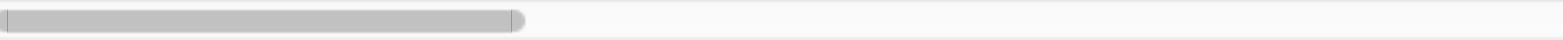
```
mydata.head(25)
```

Out[8]:

	fv_la_z3	vl_la_z3	vt_la_z3	fv_la_z5	vl_la_z5	vt_la_z5	fv_la_tc
RECORD							
1	-0.017442	-0.006894	-0.035573	0.039096	-0.001195	-0.017468	-0.003404
2	0.041219	-0.007313	-0.023622	0.031649	-0.017204	-0.035889	-0.042937
3	0.173077	-0.057336	-0.080118	0.173978	-0.059177	-0.083077	0.252584
4	-0.126544	-0.052837	-0.077530	-0.127203	-0.054533	-0.080394	-0.083387
5	-0.185336	-0.053291	-0.077321	-0.186301	-0.055001	-0.080177	-0.147934
6	0.092040	0.010679	-0.005715	0.048598	-0.001939	-0.017496	0.039271
7	-0.145741	-0.039019	0.018256	-0.134313	-0.034903	0.028601	-0.127953
8	-0.057213	-0.014697	-0.042009	-0.017558	-0.022163	-0.028557	-0.038114
9	-0.022529	-0.029363	-0.030958	-0.012399	-0.025098	-0.017080	-0.039074
10	0.032587	-0.001736	-0.029278	0.014435	-0.006640	-0.023259	0.040257

11	-0.185120	-0.057079	-0.074450	-0.186346	-0.059049	-0.078506	-0.146887
12	0.070893	-0.010213	-0.032185	0.049757	-0.014666	-0.026673	0.073688
13	0.021683	-0.009986	-0.026622	0.070190	0.003577	-0.012810	-0.065035
14	-0.036051	-0.028151	-0.042927	-0.024571	-0.019479	-0.031873	-0.079766
15	0.191803	0.002140	0.000678	0.085621	-0.018391	-0.019235	0.035982
16	-0.186832	-0.056908	-0.078343	-0.186813	-0.058392	-0.081706	-0.149652
17	-0.057750	-0.003240	-0.038228	-0.014818	0.002996	-0.021360	-0.038583
18	-0.127829	0.028800	0.060426	-0.140528	0.001892	0.011178	-0.117673
19	-0.127546	0.212890	0.069285	-0.105719	0.354861	0.162641	-0.107952
20	0.044302	-0.005226	-0.025818	0.025804	-0.003795	-0.012105	0.006627
21	-0.122235	0.054226	0.023287	-0.133842	0.045778	0.037200	-0.125405
22	0.071487	0.001660	-0.010291	0.032116	0.000741	0.003441	0.025216
23	-0.089114	-0.027748	-0.053597	-0.064053	-0.028246	-0.040597	-0.065956
24	-0.032166	-0.007253	-0.035733	0.007916	-0.006832	-0.012249	-0.016255
25	-0.181604	-0.057720	-0.076454	-0.182239	-0.059737	-0.080377	-0.129295

25 rows × 60 columns



In [9]:

```
mydata_transpose = mydata.transpose()  
mydata_transpose.head()
```

Out[9]:

RECORD	1	2	3	4	5	6	7
fv_la_z3	-0.017442	0.041219	0.173077	-0.126544	-0.185336	0.092040	-0.145741
vl_la_z3	-0.006894	-0.007313	-0.057336	-0.052837	-0.053291	0.010679	-0.039019
vt_la_z3	-0.035573	-0.023622	-0.080118	-0.077530	-0.077321	-0.005715	0.018256
fv_la_z5	0.039096	0.031649	0.173978	-0.127203	-0.186301	0.048598	-0.134313
vl_la_z5	-0.001195	-0.017204	-0.059177	-0.054533	-0.055001	-0.001939	-0.034903

5 rows × 1048575 columns

In [10]:

```
mydata.describe()
```

Out[10]:

	fv_la_z3	vl_la_z3	vt_la_z3	fv_la_z5	vl_la_z5	vt_la_z5
count	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06
mean	3.060097e-17	-2.210921e-17	1.850339e-17	-6.476865e-17	1.034353e-17	-8.01919e-17
std	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00
min	-1.878308e-01	-5.857795e-02	-8.083236e-02	-1.888085e-01	-6.045835e-02	-8.3102e-02
25%	-1.147560e-01	-4.308624e-02	-5.781444e-02	-1.101547e-01	-4.328888e-02	-5.9102e-02
50%	-2.695101e-02	-1.952813e-02	-3.772837e-02	-1.869088e-02	-1.809100e-02	-3.3102e-02
75%	4.210179e-02	-7.795125e-04	-1.631675e-02	3.964738e-02	-2.969563e-05	-9.3103e-03
max	4.698105e+02	7.180752e+02	6.741482e+02	4.722559e+02	7.411263e+02	6.98105e+02

8 rows × 60 columns

In [12]:

```
mydata_transpose.shape
```

Out[12]:

(60, 1048575)

In [13]:

```
%%time
covariance = mydata_transpose.dot(mydata)/(len(mydata)-1)
```

CPU times: user 264 ms, sys: 19.3 ms, total: 283 ms
Wall time: 321 ms

In [14]:

```
print(covariance)
```

	fv_la_z3	vl_la_z3	vt_la_z3	fv_la_z5	vl_la_z5	vt_l
a_z5 \						
fv_la_z3	1.000000	0.661558	0.703109	0.984641	0.655264	0.69
7466						
vl_la_z3	0.661558	1.000000	0.896382	0.658099	0.988189	0.89
8941						
vt_la_z3	0.703109	0.896382	1.000000	0.690665	0.872938	0.97
3475						
fv_la_z5	0.984641	0.658099	0.690665	1.000000	0.664654	0.70
6027						
vl_la_z5	0.655264	0.988189	0.872938	0.664654	1.000000	0.90
3967						
vt_la_z5	0.697466	0.898941	0.973475	0.706027	0.903967	1.00
0000						
fv_la_tc	0.929804	0.468063	0.488307	0.928143	0.471315	0.49
8382						
vl_la_tc	0.898276	0.603545	0.525884	0.902470	0.609779	0.54
1689						
vt_la_tc	0.919974	0.449050	0.464041	0.920059	0.452272	0.47
2782						
fv_la_bo	0.980145	0.578312	0.631494	0.975604	0.577150	0.63
4613						
vl_la_bo	0.659617	0.965445	0.853572	0.660410	0.971275	0.87
1254						
vt_la_bo	0.679287	0.840405	0.932289	0.676686	0.841361	0.93
9203						
fv_la_none	0.981100	0.640188	0.673661	0.974729	0.640611	0.67
9224						
vl_la_none	0.665680	0.963388	0.856543	0.667762	0.968534	0.87
1974						
vt_la_none	0.698902	0.842603	0.927386	0.696030	0.839643	0.92
7041						
fv_ba_z3	0.064839	0.101390	0.065028	0.059791	0.100733	0.06
4093						

vl_ba_z3 8137	0.058533	0.143621	0.092611	0.057661	0.146131	0.09
vt_ba_z3 8367	0.048991	0.118709	0.091145	0.048145	0.122055	0.09
fv_ba_z5 5693	0.076409	0.126589	0.081066	0.086353	0.130773	0.08
vl_ba_z5 4191	0.084968	0.207339	0.127010	0.087405	0.212945	0.13
vt_ba_z5 4476	0.080276	0.188089	0.126097	0.084135	0.193656	0.13
fv_ba_tc 7986	0.026074	0.041777	0.027894	0.025560	0.041855	0.02
vl_ba_tc 8827	0.013590	0.028128	0.017955	0.013579	0.028465	0.01
vt_ba_tc 4835	0.024286	0.043439	0.032982	0.024147	0.044149	0.03
fv_ba_bo 7393	0.055685	0.090210	0.057429	0.053227	0.090081	0.05
vl_ba_bo 8429	0.082890	0.202461	0.122705	0.082550	0.207192	0.12
vt_ba_bo 9397	0.056771	0.136790	0.093745	0.056717	0.140632	0.09
fv_ba_none 4131	0.051533	0.085357	0.053972	0.049337	0.085388	0.05
vl_ba_none 6756	0.049897	0.116467	0.073521	0.048807	0.117714	0.07
vt_ba_none 7677	0.046305	0.099666	0.073532	0.045801	0.101195	0.07
fv_bv_z3 7805	0.033118	0.045648	0.028469	0.028533	0.044024	0.02
vl_bv_z3 2630	0.022681	0.060382	0.039915	0.022040	0.059909	0.04
vt_bv_z3 9173	0.021965	0.057712	0.045098	0.021011	0.058099	0.04
fv_bv_z5 8294	0.038927	0.057070	0.035434	0.045084	0.058817	0.03
vl_bv_z5 9628	0.039655	0.102997	0.065177	0.041020	0.103964	0.06
vt_bv_z5 4146	0.039152	0.098321	0.068477	0.041300	0.099781	0.07
fv_bv_tc 4596	0.003745	0.007164	0.004595	0.003521	0.007042	0.00
vl_bv_tc 4203	0.002593	0.006306	0.004031	0.002491	0.006258	0.00
vt_bv_tc 6452	0.003478	0.007818	0.006014	0.003343	0.007855	0.00
fv_bv_bo 4348	0.025995	0.039945	0.024205	0.023713	0.038996	0.02
vl_bv_bo 9057	0.028627	0.074642	0.046359	0.028341	0.075214	0.04
vt_bv_bo 1003	0.025068	0.065778	0.047232	0.024704	0.066644	0.05
fv_bv_none	0.021448	0.038052	0.022911	0.019959	0.037478	0.02

3352						
vl_bv_none 0714	0.017007	0.044621	0.029103	0.016460	0.044447	0.03
vt_bv_none 9503	0.018051	0.047336	0.036393	0.017509	0.047713	0.03
fv_none_z3 8436	0.177369	0.359698	0.237394	0.173296	0.364614	0.23
vl_none_z3 8764	0.129240	0.332888	0.228778	0.129624	0.343199	0.24
vt_none_z3 1174	0.058500	0.153041	0.135912	0.058131	0.160115	0.15
fv_none_z5 7086	0.177101	0.382789	0.245188	0.186077	0.393398	0.25
vl_none_z5 9769	0.220313	0.546208	0.330333	0.224899	0.560332	0.34
vt_none_z5 5764	0.147064	0.351955	0.239068	0.152544	0.360511	0.25
fv_none_tc 9436	0.171417	0.239643	0.169878	0.171277	0.240631	0.16
vl_none_tc 3794	0.164677	0.324901	0.211442	0.170984	0.331049	0.22
vt_none_tc 2478	0.124498	0.193677	0.154744	0.126082	0.197333	0.16
fv_none_bo 6271	0.172195	0.363638	0.233462	0.170773	0.370171	0.23
vl_none_bo 7546	0.204808	0.509954	0.318523	0.205671	0.525670	0.33
vt_none_bo 0264	0.113062	0.281339	0.204539	0.113341	0.291843	0.22
fv_none_none 3325	0.176130	0.304244	0.206157	0.172319	0.303796	0.20
vl_none_none 5891	0.171358	0.383553	0.245093	0.169557	0.388324	0.25
vt_none_none 9821	0.129092	0.248747	0.191687	0.127179	0.252119	0.19
	fv_la_tc	vl_la_tc	vt_la_tc	fv_la_bo	...	
\						
fv_la_z3	0.929804	0.898276	0.919974	0.980145	...	
vl_la_z3	0.468063	0.603545	0.449050	0.578312	...	
vt_la_z3	0.488307	0.525884	0.464041	0.631494	...	
fv_la_z5	0.928143	0.902470	0.920059	0.975604	...	
vl_la_z5	0.471315	0.609779	0.452272	0.577150	...	
vt_la_z5	0.498382	0.541689	0.472782	0.634613	...	
fv_la_tc	1.000000	0.936945	0.987323	0.936251	...	
vl_la_tc	0.936945	1.000000	0.948862	0.894860	...	
vt_la_tc	0.987323	0.948862	1.000000	0.937089	...	
fv_la_bo	0.936251	0.894860	0.937089	1.000000	...	
vl_la_bo	0.476802	0.610570	0.459600	0.596221	...	
vt_la_bo	0.476782	0.510678	0.454797	0.639534	...	
fv_la_none	0.942226	0.908246	0.927906	0.969087	...	
vl_la_none	0.507356	0.644955	0.485268	0.586057	...	
vt_la_none	0.543733	0.572528	0.515408	0.633022	...	

fv_ba_z3	0.052233	0.076318	0.045395	0.045743	...
vl_ba_z3	0.040643	0.083448	0.039059	0.041583	...
vt_ba_z3	0.035494	0.067250	0.036056	0.034754	...
fv_ba_z5	0.064155	0.094082	0.056220	0.055799	...
vl_ba_z5	0.057868	0.117482	0.054340	0.060265	...
vt_ba_z5	0.058284	0.108910	0.055732	0.057474	...
fv_ba_tc	0.029095	0.040504	0.025956	0.019423	...
vl_ba_tc	0.013284	0.024728	0.012729	0.010000	...
vt_ba_tc	0.027139	0.040423	0.026460	0.017990	...
fv_ba_bo	0.047218	0.069076	0.040952	0.041490	...
vl_ba_bo	0.056227	0.115945	0.052748	0.059306	...
vt_ba_bo	0.040198	0.077828	0.039092	0.040913	...
fv_ba_none	0.043315	0.064168	0.037581	0.038348	...
vl_ba_none	0.037056	0.074024	0.035130	0.036094	...
vt_ba_none	0.038489	0.066048	0.037535	0.033846	...
fv_bv_z3	0.026459	0.036462	0.021625	0.022914	...
vl_bv_z3	0.015340	0.033091	0.014970	0.015767	...
vt_bv_z3	0.015364	0.031138	0.015984	0.015184	...
fv_bv_z5	0.033211	0.046559	0.027324	0.028352	...
vl_bv_z5	0.026689	0.056119	0.025414	0.027596	...
vt_bv_z5	0.027833	0.054500	0.027145	0.027515	...
fv_bv_tc	0.003814	0.006250	0.003361	0.002780	...
vl_bv_tc	0.002216	0.005047	0.002233	0.001887	...
vt_bv_tc	0.003513	0.006460	0.003639	0.002557	...
fv_bv_bo	0.021910	0.031332	0.017919	0.019243	...
vl_bv_bo	0.019171	0.041134	0.018214	0.020322	...
vt_bv_bo	0.017496	0.035923	0.017541	0.017936	...
fv_bv_none	0.015970	0.025391	0.013304	0.015763	...
vl_bv_none	0.011866	0.025682	0.011496	0.012025	...
vt_bv_none	0.013276	0.026792	0.013651	0.012919	...
fv_none_z3	0.130290	0.220862	0.119799	0.125799	...
vl_none_z3	0.087051	0.180291	0.086395	0.092120	...
vt_none_z3	0.041259	0.079055	0.045113	0.041662	...
fv_none_z5	0.131102	0.230262	0.120748	0.127454	...
vl_none_z5	0.146547	0.298867	0.136588	0.156814	...
vt_none_z5	0.103921	0.193878	0.098457	0.105560	...
fv_none_tc	0.198375	0.246716	0.177645	0.125694	...
vl_none_tc	0.163173	0.279493	0.155155	0.120661	...
vt_none_tc	0.144650	0.190953	0.139715	0.090776	...
fv_none_bo	0.125246	0.219473	0.115117	0.124240	...
vl_none_bo	0.135400	0.280384	0.128730	0.146260	...
vt_none_bo	0.076872	0.151366	0.076770	0.081220	...
fv_none_none	0.153305	0.226163	0.140085	0.127359	...
vl_none_none	0.130704	0.249337	0.123653	0.124241	...
vt_none_none	0.111922	0.172787	0.108203	0.092851	...

	vt_none_z5	fv_none_tc	vl_none_tc	vt_none_tc	fv_non
e_bo \					
fv_la_z3	0.147064	0.171417	0.164677	0.124498	0.17
2195					
vl_la_z3	0.351955	0.239643	0.324901	0.193677	0.36
3638					
vt_la_z3	0.239068	0.169878	0.211442	0.154744	0.23

3462					
fv_la_z5	0.152544	0.171277	0.170984	0.126082	0.17
0773					
vl_la_z5	0.360511	0.240631	0.331049	0.197333	0.37
0171					
vt_la_z5	0.255764	0.169436	0.223794	0.162478	0.23
6271					
fv_la_tc	0.103921	0.198375	0.163173	0.144650	0.12
5246					
vl_la_tc	0.193878	0.246716	0.279493	0.190953	0.21
9473					
vt_la_tc	0.098457	0.177645	0.155155	0.139715	0.11
5117					
fv_la_bo	0.105560	0.125694	0.120661	0.090776	0.12
4240					
vl_la_bo	0.347217	0.236449	0.322078	0.184392	0.37
2947					
vt_la_bo	0.203134	0.144753	0.182989	0.120254	0.21
2548					
fv_la_none	0.148545	0.212992	0.191548	0.152392	0.18
0246					
vl_la_none	0.349898	0.309520	0.389226	0.238359	0.39
0907					
vt_la_none	0.225827	0.260304	0.262614	0.205255	0.25
6394					
fv_ba_z3	0.144124	0.197882	0.213340	0.158151	0.17
9203					
vl_ba_z3	0.238215	0.155190	0.269535	0.217059	0.18
9976					
vt_ba_z3	0.291853	0.143589	0.297789	0.311022	0.18
7771					
fv_ba_z5	0.182741	0.215023	0.224653	0.175128	0.20
3812					
vl_ba_z5	0.288181	0.174473	0.266425	0.192352	0.24
8031					
vt_ba_z5	0.338106	0.193214	0.294584	0.255227	0.26
3227					
fv_ba_tc	0.061399	0.109101	0.112994	0.096109	0.07
6923					
vl_ba_tc	0.044006	0.059692	0.084362	0.072777	0.04
4232					
vt_ba_tc	0.094764	0.123568	0.172215	0.179730	0.08
3849					
fv_ba_bo	0.131524	0.178874	0.193723	0.142483	0.16
3019					
vl_ba_bo	0.246973	0.171774	0.267966	0.188909	0.23
0272					
vt_ba_bo	0.241793	0.136349	0.242186	0.226566	0.18
7100					
fv_ba_none	0.121326	0.161376	0.178061	0.129997	0.14
7475					
vl_ba_none	0.180218	0.158429	0.247699	0.180319	0.16
5744					

vt_ba_none 4979	0.212809	0.159505	0.265803	0.244817	0.16
fv_bv_z3 7175	0.086636	0.111210	0.124424	0.086996	0.09
vl_bv_z3 5043	0.124060	0.071579	0.129677	0.105616	0.08
vt_bv_z3 1247	0.170730	0.078595	0.166228	0.170161	0.10
fv_bv_z5 0849	0.108351	0.122879	0.131735	0.096566	0.11
vl_bv_z5 2503	0.184328	0.100594	0.165155	0.125466	0.13
vt_bv_z5 0498	0.231871	0.118934	0.191853	0.174334	0.16
fv_bv_tc 4613	0.013786	0.019088	0.021300	0.018481	0.01
vl_bv_tc 0834	0.012564	0.015210	0.023157	0.021908	0.01
vt_bv_tc 6982	0.022401	0.024057	0.037831	0.041379	0.01
fv_bv_bo 1000	0.075374	0.091176	0.104347	0.071244	0.08
vl_bv_bo 2947	0.119650	0.070202	0.117278	0.086804	0.09
vt_bv_bo 3954	0.158270	0.076727	0.149160	0.143291	0.10
fv_bv_none 5091	0.074086	0.069407	0.078850	0.057029	0.07
vl_bv_none 6253	0.091459	0.057641	0.100011	0.077605	0.06
vt_bv_none 5778	0.138276	0.069348	0.139726	0.137758	0.08
fv_none_z3 7465	0.871997	0.760034	0.609315	0.630714	0.98
vl_none_z3 9836	0.688792	0.336202	0.652159	0.643435	0.47
vt_none_z3 8771	0.636395	0.263236	0.502289	0.640302	0.39
fv_none_z5 3880	0.923738	0.710192	0.574710	0.588053	0.96
vl_none_z5 6316	0.795634	0.474670	0.611989	0.456116	0.72
vt_none_z5 7003	1.000000	0.566948	0.539684	0.584984	0.86
fv_none_tc 7778	0.566948	1.000000	0.802419	0.827169	0.72
vl_none_tc 8595	0.539684	0.802419	1.000000	0.877501	0.56
vt_none_tc 6632	0.584984	0.827169	0.877501	1.000000	0.58
fv_none_bo 0000	0.867003	0.727778	0.568595	0.586632	1.00
vl_none_bo	0.740297	0.425269	0.682415	0.574441	0.64

9270					
vt_none_bo	0.800650	0.406473	0.572758	0.676196	0.65
0577					
fv_none_none	0.692869	0.859811	0.714075	0.672871	0.86
3318					
vl_none_none	0.613231	0.615334	0.855667	0.667400	0.61
2909					
vt_none_none	0.727611	0.643948	0.745097	0.797583	0.67
5812					

	vl_none_bo	vt_none_bo	fv_none_none	vl_none_none	vt
_none_none					
fv_la_z3	0.204808	0.113062	0.176130	0.171358	
0.129092					
vl_la_z3	0.509954	0.281339	0.304244	0.383553	
0.248747					
vt_la_z3	0.318523	0.204539	0.206157	0.245093	
0.191687					
fv_la_z5	0.205671	0.113341	0.172319	0.169557	
0.127179					
vl_la_z5	0.525670	0.291843	0.303796	0.388324	
0.252119					
vt_la_z5	0.337546	0.220264	0.203325	0.255891	
0.199821					
fv_la_tc	0.135400	0.076872	0.153305	0.130704	
0.111922					
vl_la_tc	0.280384	0.151366	0.226163	0.249337	
0.172787					
vt_la_tc	0.128730	0.076770	0.140085	0.123653	
0.108203					
fv_la_bo	0.146260	0.081220	0.127359	0.124241	
0.092851					
vl_la_bo	0.523810	0.279906	0.303220	0.383251	
0.239193					
vt_la_bo	0.289066	0.168485	0.179623	0.214809	
0.151295					
fv_la_none	0.204348	0.113486	0.206639	0.189286	
0.149812					
vl_la_none	0.517011	0.277704	0.377937	0.444452	
0.294164					
vt_la_none	0.305101	0.179987	0.289978	0.282159	
0.231727					
fv_ba_z3	0.165770	0.104816	0.220427	0.227081	
0.168080					
vl_ba_z3	0.320880	0.274317	0.197155	0.318228	
0.279695					
vt_ba_z3	0.390634	0.431346	0.179142	0.351810	
0.402510					
fv_ba_z5	0.200096	0.127471	0.225359	0.224477	
0.173946					
vl_ba_z5	0.365697	0.257226	0.217996	0.309654	
0.241651					
vt_ba_z5	0.393817	0.337406	0.235587	0.339720	

0.319239				
fv_ba_tc	0.067803	0.043023	0.093167	0.092010
0.069156				
vl_ba_tc	0.057934	0.045497	0.052136	0.069412
0.055521				
vt_ba_tc	0.122014	0.123694	0.096455	0.134359
0.136265				
fv_ba_bo	0.149345	0.094771	0.199892	0.206840
0.152225				
vl_ba_bo	0.346686	0.244831	0.220361	0.318323
0.244672				
vt_ba_bo	0.326642	0.312636	0.171899	0.286788
0.293265				
fv_ba_none	0.140535	0.087786	0.179299	0.189064
0.137004				
vl_ba_none	0.245181	0.193207	0.201568	0.292433
0.232269				
vt_ba_none	0.284429	0.289621	0.197274	0.311569
0.313640				
fv_bv_z3	0.081493	0.054456	0.126413	0.135580
0.095401				
vl_bv_z3	0.145653	0.132238	0.092577	0.154017
0.137805				
vt_bv_z3	0.205400	0.234529	0.101083	0.198014
0.222879				
fv_bv_z5	0.096836	0.066743	0.130453	0.134355
0.099804				
vl_bv_z5	0.202138	0.161603	0.128305	0.194198
0.161570				
vt_bv_z5	0.237880	0.232369	0.150221	0.225280
0.224694				
fv_bv_tc	0.012040	0.008296	0.014962	0.014871
0.010857				
vl_bv_tc	0.014228	0.012195	0.011209	0.015630
0.013185				
vt_bv_tc	0.025886	0.028519	0.016483	0.025924
0.027885				
fv_bv_bo	0.068695	0.045174	0.104677	0.114342
0.078832				
vl_bv_bo	0.144789	0.112912	0.091091	0.139370
0.113448				
vt_bv_bo	0.189592	0.198491	0.099119	0.177672
0.187828				
fv_bv_none	0.067226	0.046169	0.078786	0.083807
0.060118				
vl_bv_none	0.106544	0.093003	0.074539	0.118407
0.101177				
vt_bv_none	0.165079	0.184726	0.089237	0.166099
0.180333				
fv_none_z3	0.649121	0.652012	0.884135	0.641566
0.705884				
vl_none_z3	0.903358	0.921200	0.421323	0.770132
0.831260				

vt_none_z3	0.711115	0.941938	0.325008	0.593286
0.830450				
fv_none_z5	0.673143	0.660666	0.818964	0.605231
0.661253				
vl_none_z5	0.900224	0.638287	0.590669	0.708766
0.569478				
vt_none_z5	0.740297	0.800650	0.692869	0.613231
0.727611				
fv_none_tc	0.425269	0.406473	0.859811	0.615334
0.643948				
vl_none_tc	0.682415	0.572758	0.714075	0.855667
0.745097				
vt_none_tc	0.574441	0.676196	0.672871	0.667400
0.797583				
fv_none_bo	0.649270	0.650577	0.863318	0.612909
0.675812				
vl_none_bo	1.000000	0.835656	0.541784	0.810951
0.744611				
vt_none_bo	0.835656	1.000000	0.511617	0.678138
0.877525				
fv_none_none	0.541784	0.511617	1.000000	0.767331
0.772338				
vl_none_none	0.810951	0.678138	0.767331	1.000000
0.851030				
vt_none_none	0.744611	0.877525	0.772338	0.851030
1.000000				

[60 rows x 60 columns]

In [15]:

```
# %%time
# eig_vals, eig_vecs = np.linalg.eig(covariance)
```

In [16]:

```
# print('Eigenvectors \n%s' %eig_vecs)
```

In [17]:

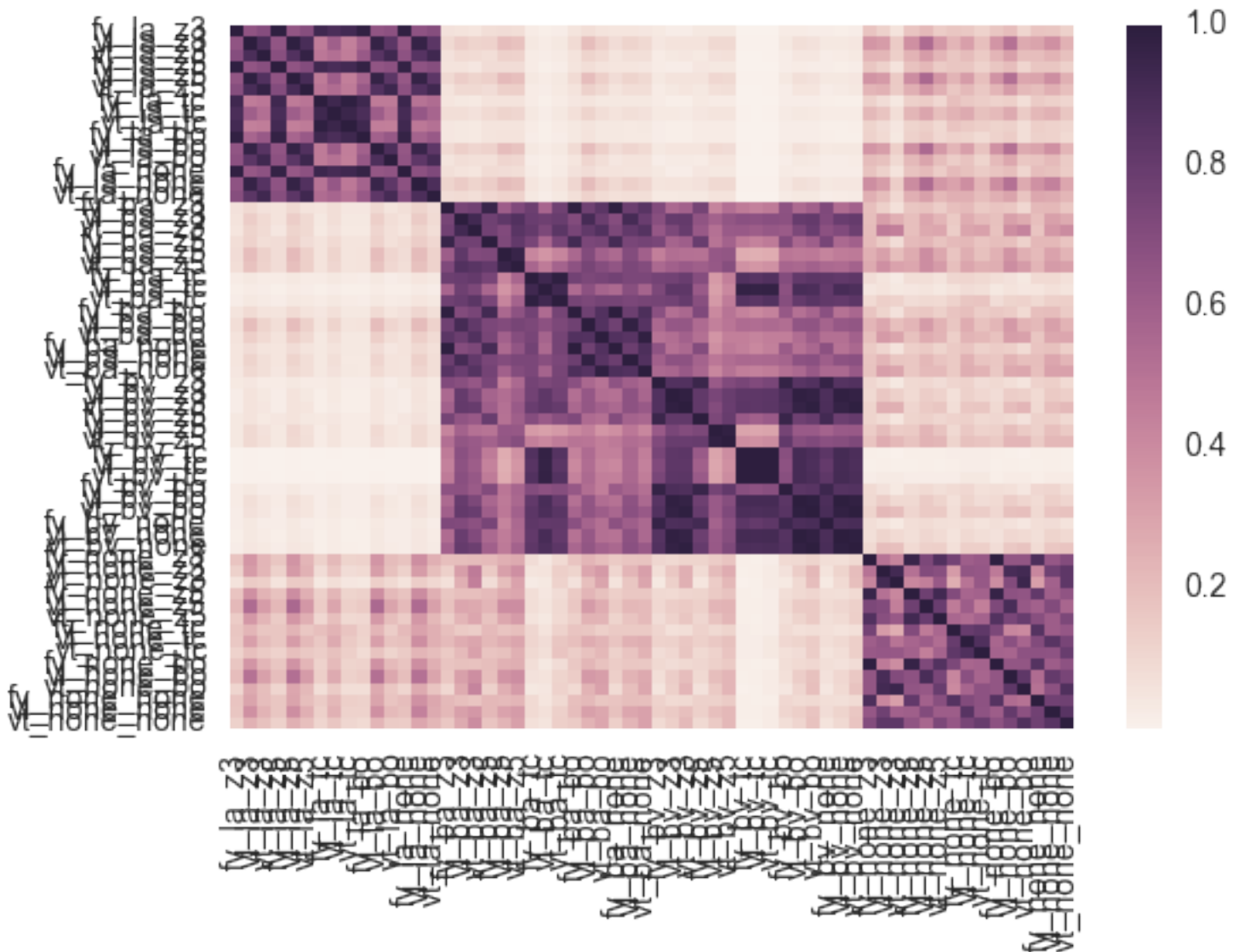
```
# print('Eigenvalues \n%s' %eig_vals)
```

In [18]:

```
sns.heatmap(covariance)
```

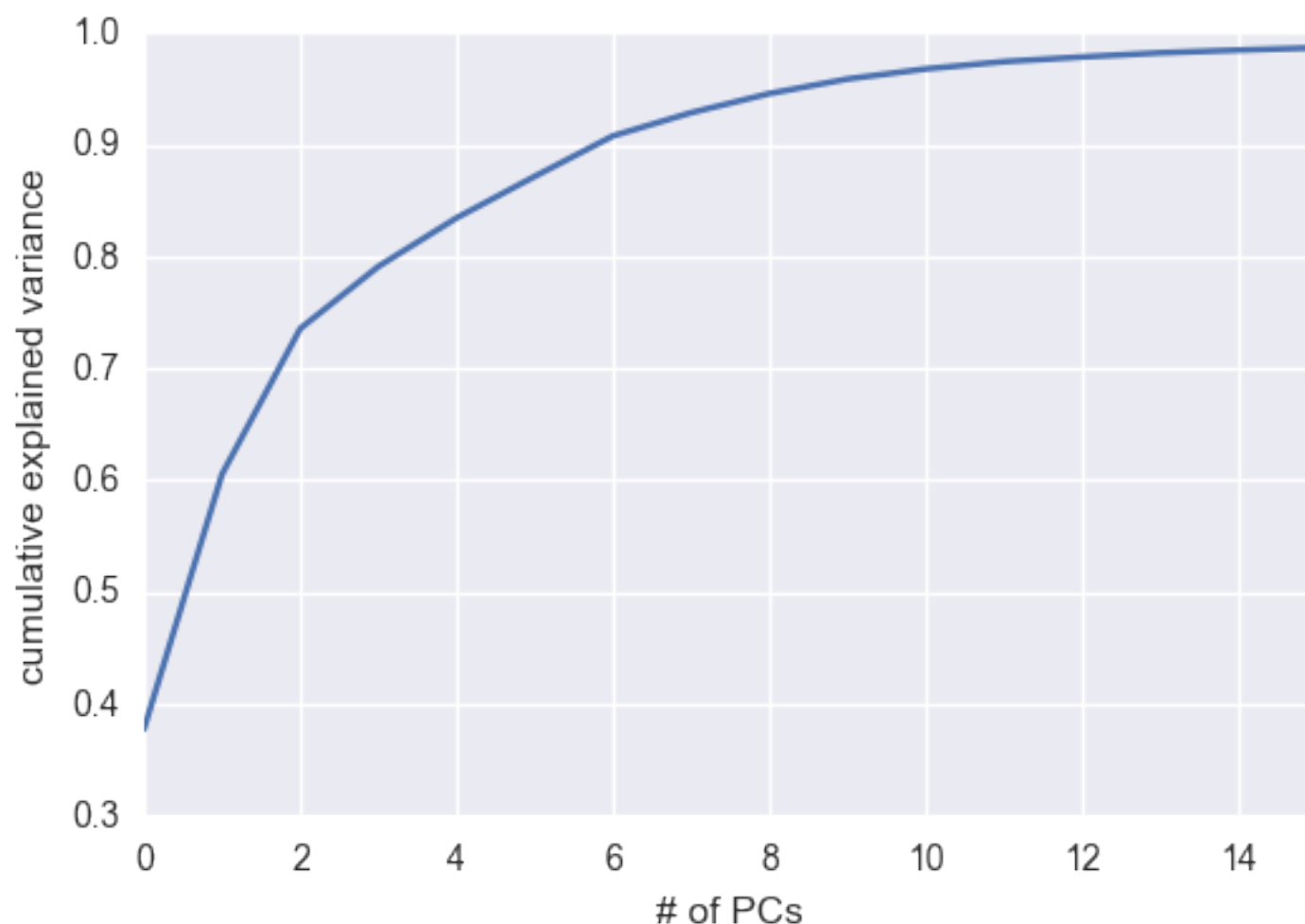
Out[18]:

<matplotlib.axes._subplots.AxesSubplot at 0x131c809b0>



In [19]:

```
%%time
pca = PCA(n_components=50).fit(mydata)
plt.plot(np.cumsum(pca.explained_variance_ratio_))
plt.xlabel('# of PCs')
plt.ylabel('cumulative explained variance')
plt.xlim(xmax=15)
plt.ylim(ymax=1)
plt.show()
```



CPU times: user 8.39 s, sys: 2.78 s, total: 11.2 s
Wall time: 8.78 s

In [20]:

```
%%time
NPCs = 10
PCs = pd.DataFrame(PCA(n_components=NPCs).fit_transform(mydata))
```

CPU times: user 14.3 s, sys: 5.65 s, total: 19.9 s
Wall time: 13.3 s

In [21]:

```
PCs.shape
```

Out[21]:

```
(1048575, 10)
```

In [22]:

```
PCs.head()
```

Out[22]:

	0	1	2	3	4	5	6	7
0	-0.034264	-0.085701	0.071586	0.029461	0.019315	0.033291	0.010059	-0.012814
1	-0.077732	-0.081204	0.019688	0.036318	0.013258	-0.001528	0.035214	-0.032494
2	-0.266122	-0.057719	0.223981	-0.026517	0.132294	0.140012	0.171081	-0.169957
3	-0.314207	-0.293961	-0.059221	-0.030026	-0.164627	0.034447	0.135383	-0.101394
4	-0.125387	-0.358425	-0.204000	-0.045647	-0.201698	0.046472	0.039459	0.040546

In [23]:

```
%%time
PCs_zscale = (PCs - PCs.mean())/ PCs.std()
```

CPU times: user 529 ms, sys: 202 ms, total: 732 ms
Wall time: 480 ms

In [24]:

```
PCs_zscale.head()
```

Out[24]:

	0	1	2	3	4	5	6	7
0	-0.007197	-0.023181	0.025625	0.016139	0.011971	0.022319	0.006828	-0.01144
1	-0.016327	-0.021964	0.007048	0.019895	0.008217	-0.001024	0.023903	-0.0297
2	-0.055897	-0.015612	0.080176	-0.014526	0.081994	0.093864	0.116126	-0.1522
3	-0.065997	-0.079512	-0.021199	-0.016449	-0.102034	0.023094	0.091895	-0.0908
4	-0.026337	-0.096948	-0.073024	-0.025006	-0.125009	0.031155	0.026784	0.0363

In [25]:

```
PCs.describe()
```

Out[25]:

	0	1	2	3	4	5
count	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06
mean	-3.875824e-16	5.998621e-17	4.646066e-17	2.754557e-17	4.269339e-17	-1.048575e-16
std	4.760963e+00	3.697088e+00	2.793606e+00	1.825435e+00	1.613460e+00	1.404858e+00
min	-3.512390e+01	-1.369115e+03	-1.105806e+03	-8.284555e+02	-5.254451e+02	-4.098575e+02
25%	-1.681646e+01	-2.223510e-01	-1.112228e-01	-2.895072e-02	-1.733281e-01	-1.048575e-01
50%	-7.233993e-02	-1.148778e-01	-1.323046e-02	-2.949506e-03	-3.031621e-02	1.104858e-02
75%	-9.371534e-03	-9.002384e-03	7.212620e-02	2.789066e-02	6.037596e-02	2.409858e-02
max	3.362582e+03	2.091879e+03	1.223548e+03	9.599008e+02	7.627368e+02	8.204858e+02



In [26]:

```
PCs_zscale.describe()
```

Out[26]:

	0	1	2	3	4	5
count	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06	1.048575e+06
mean	6.911923e-17	2.452717e-18	-3.321742e-17	-1.569507e-17	-1.031590e-17	-5.031590e-18
std	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00
min	-7.377478e-02	-3.703225e+02	-3.958347e+02	-4.538400e+02	-3.256635e+02	-2.566635e+02
25%	-3.532156e-02	-6.014220e-02	-3.981333e-02	-1.585963e-02	-1.074263e-01	-7.4263e-01
50%	-1.519439e-02	-3.107250e-02	-4.735977e-03	-1.615783e-03	-1.878956e-02	7.6e-02
75%	-1.968412e-03	-2.434993e-03	2.581831e-02	1.527891e-02	3.742017e-02	1.6e-01
max	7.062819e+02	5.658181e+02	4.379815e+02	5.258477e+02	4.727335e+02	5.5e+02

In [27]:

```
Scores = pd.DataFrame(np.ones(numrecords), columns = ['s1'])
Scores['s2'] = np.ones(numrecords)
```

In [28]:

```
%%time
Scores['s1'] = PCs_zscale.abs().sum(axis=1)
PCs_zscale_sq = PCs_zscale **2
Scores['s2'] = PCs_zscale_sq.abs().sum(axis=1)
```

CPU times: user 1.25 s, sys: 91.1 ms, total: 1.34 s
Wall time: 557 ms

In [29]:

```
Scores.head(10)
```

Out[29]:

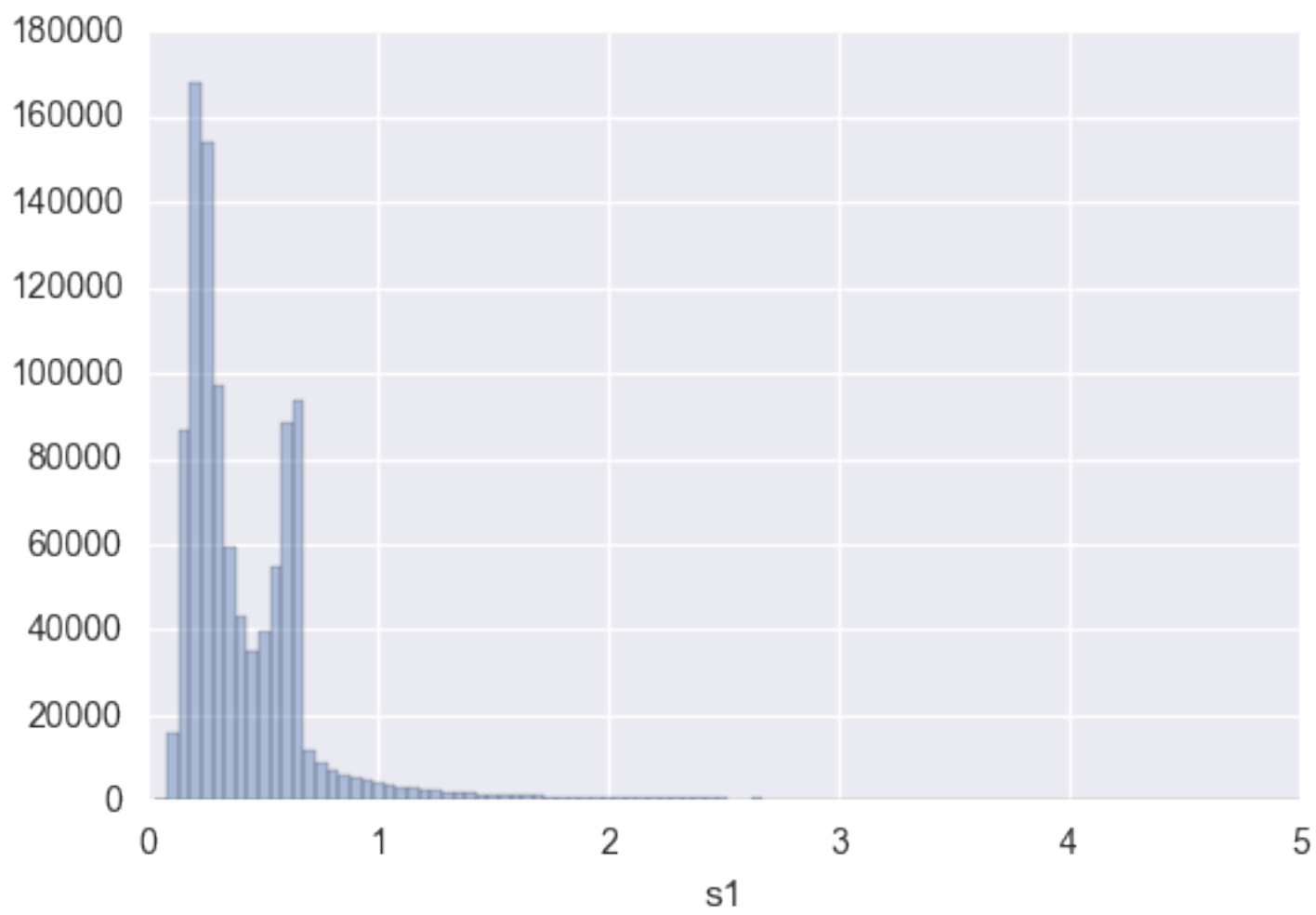
	s1	s2
0	0.214021	0.007214
1	0.166653	0.003586
2	0.623062	0.062306
3	0.555506	0.041120
4	0.586902	0.055246
5	0.256063	0.012991
6	0.672738	0.068247
7	0.173454	0.004553
8	0.157571	0.003098
9	0.171562	0.006851

In [30]:

```
xhigh = 5
sns.plt.xlim(0,xhigh)
temp = Scores[ Scores['s1'] <= xhigh]
sns.distplot(temp['s1'], bins = 100, kde = False)
```

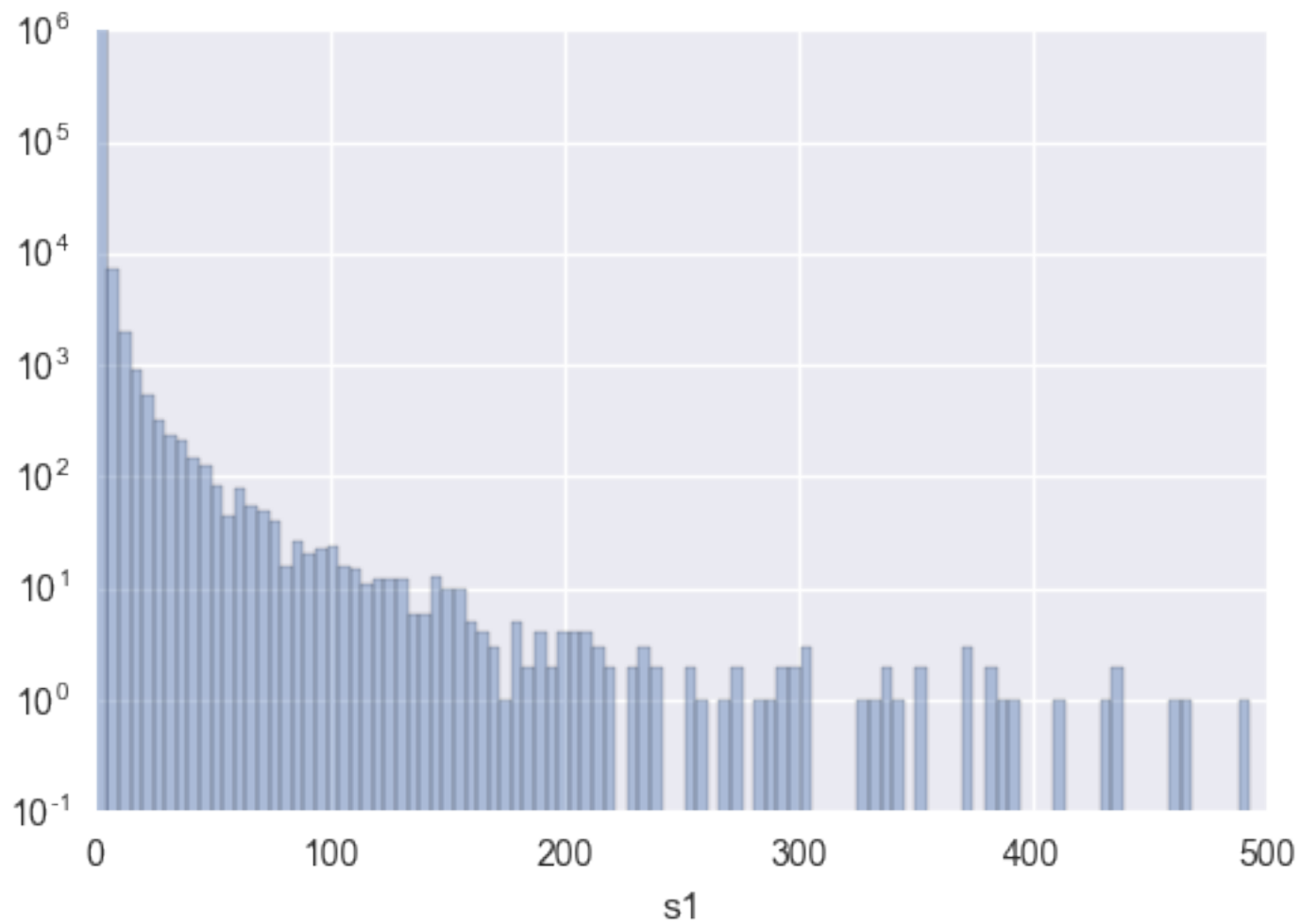
Out[30]:

<matplotlib.axes._subplots.AxesSubplot at 0x11b2114a8>



In [31]:

```
xhigh = 500
temp = Scores[ Scores['s1'] <= xhigh ]
sns.plt.xlim(0, xhigh)
sns.plt.ylim(.1, 10**6)
ax = sns.distplot(temp['s1'], bins = 100, kde=False)
ax.set_yscale('log')
plt.savefig('log.png')
```

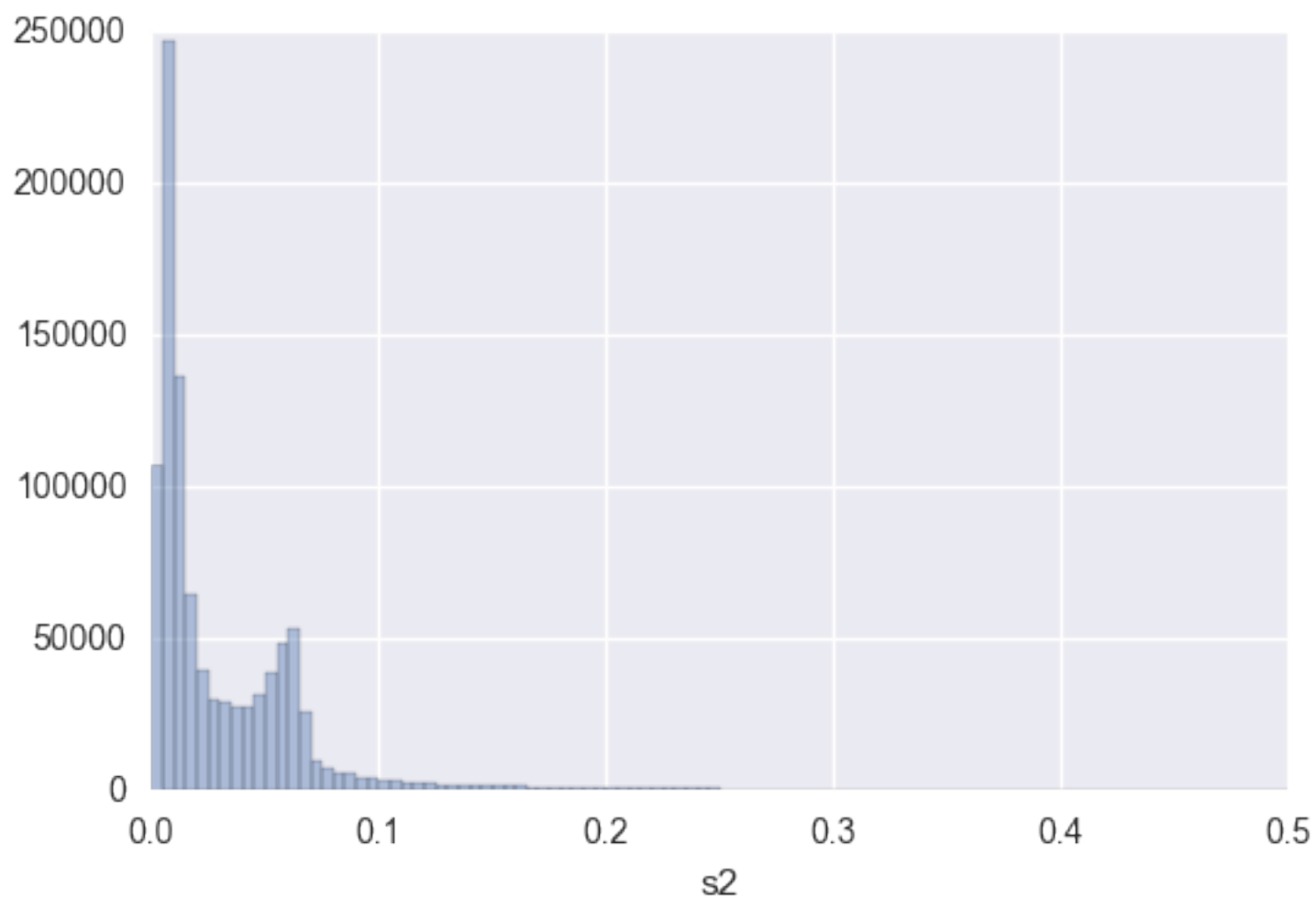


In [32]:

```
xhigh = .5  
sns.plt.xlim(0,xhigh)  
temp = Scores[Scores['s2'] <= xhigh]  
sns.distplot(temp['s2'], bins = 100, kde = False)
```

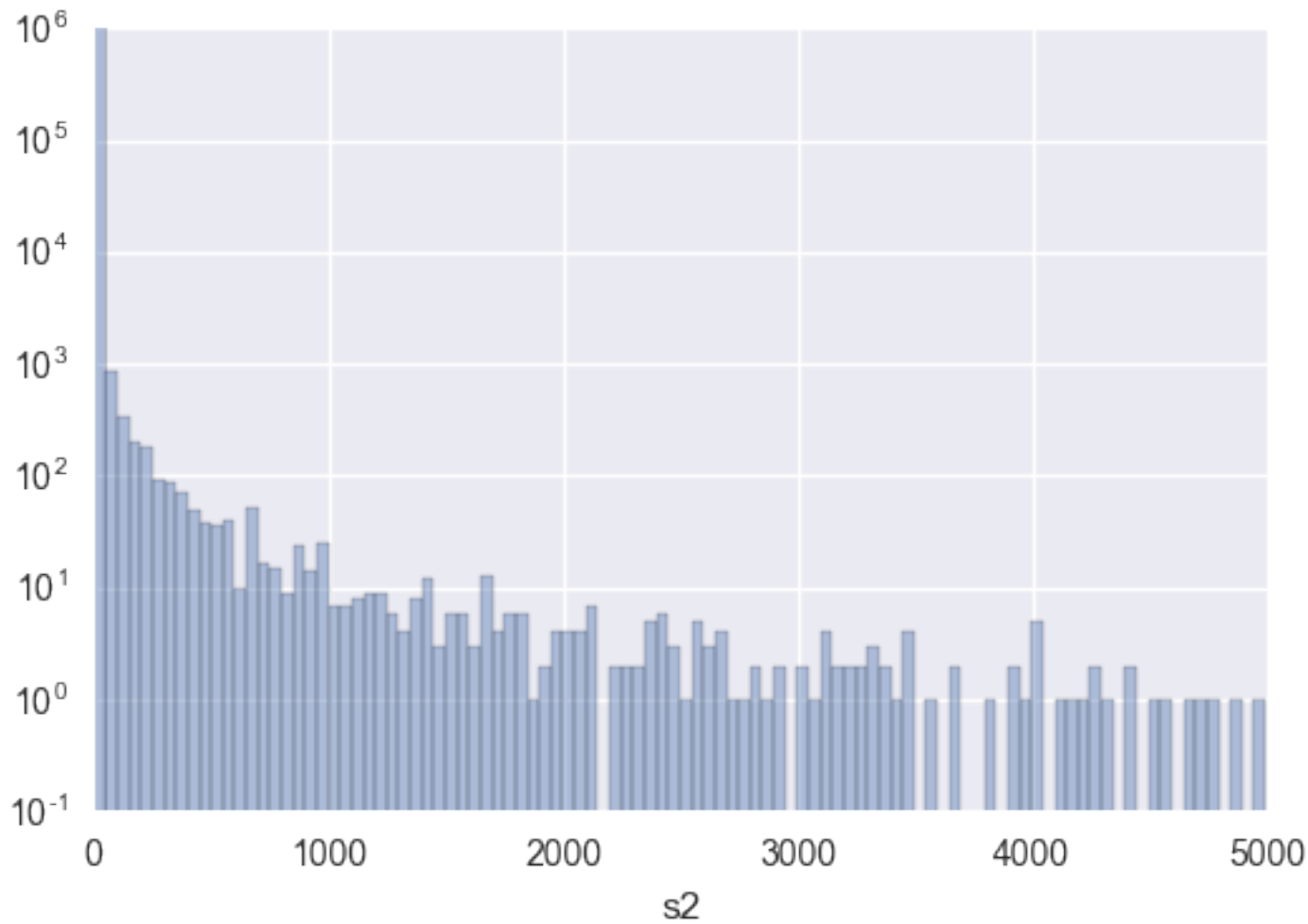
Out[32]:

<matplotlib.axes._subplots.AxesSubplot at 0x11b10f278>



In [33]:

```
xhigh = 5000
temp = Scores[Scores['s2'] <= xhigh]
sns.plt.xlim(0, xhigh)
sns.plt.ylim(.1, 10**6)
ax = sns.distplot(temp['s2'], bins = 100, kde=False)
ax.set_yscale('log')
plt.savefig('log.png')
```



In [34]:

```
Scores['record'] = Scores.index + 1
```

In [35]:

```
Scores.head()
```

Out[35]:

	s1	s2	record
0	0.214021	0.007214	1
1	0.166653	0.003586	2
2	0.623062	0.062306	3
3	0.555506	0.041120	4
4	0.586902	0.055246	5

In [36]:

```
Scores.sort_values('s1').tail(12)
```

Out[36]:

	s1	s2	record
977470	1316.559485	3.603928e+05	977471
24585	1424.827559	3.637766e+05	24586
648674	1484.598645	3.929890e+05	648675
902255	1528.395442	4.163797e+05	902256
787891	1829.780582	5.960451e+05	787892
970080	1835.168673	6.397387e+05	970081
294060	1873.017687	4.426446e+05	294061
1046263	1910.958069	6.109203e+05	1046264
78803	2181.573331	7.841722e+05	78804
315452	2325.012043	6.381404e+05	315453
5392	2430.121225	1.028361e+06	5393
376242	2598.794418	1.020363e+06	376243

In [37]:

```
Scores.sort_values('s2').tail(12)
```

Out[37]:

	s1	s2	record
977470	1316.559485	3.603928e+05	977471
24585	1424.827559	3.637766e+05	24586
648674	1484.598645	3.929890e+05	648675
902255	1528.395442	4.163797e+05	902256
294060	1873.017687	4.426446e+05	294061
787891	1829.780582	5.960451e+05	787892
1046263	1910.958069	6.109203e+05	1046264
315452	2325.012043	6.381404e+05	315453
970080	1835.168673	6.397387e+05	970081
78803	2181.573331	7.841722e+05	78804
376242	2598.794418	1.020363e+06	376243
5392	2430.121225	1.028361e+06	5393

In []: