

Appendix 1. Jupyter notebook, python code

November 9, 2017

This is converted to PDF for sake of readability, the code itself can be found on the following github repository: <https://github.com/tvanbiemen/EconOfCs>

```
In [16]: #Basic packages
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy import stats
from matplotlib_venn import venn3, venn3_circles
```

```
In [17]: #Extra packages
import random
import socket
import struct
```

```
In [18]: #Dataset 1, this takes some time to load
#The website is sometimes down..... http://downforeveryoneorjustme.com/cybercrime
cybercrime = pd.read_html('http://cybercrime-tracker.net/index.php?s=0&m=4')
cybercrime = cybercrime[0] #We want the first table that is found
cybercrime.head()
```

```
Out[18]:
```

| | -::DATE | -::URL | \ |
|---|------------|---|---|
| 0 | 08-11-2017 | mfmqatar.com/deux/windsor/admin.php | |
| 1 | 08-11-2017 | 191.101.245.31/panel/ | |
| 2 | 08-11-2017 | tapkinoski.gdn.cp-tr-1.webhostbox.net/index.php | |
| 3 | 08-11-2017 | finexp.us/hk/hs/HSB/config.php?account=taliban | |
| 4 | 08-11-2017 | brainboom.000webhostapp.com/ | |

| | -::IP | -::TYPE | Unnamed: 4 |
|---|----------------|----------|------------|
| 0 | 198.38.82.246 | Pony | NaN |
| 1 | NaN | AZORult | NaN |
| 2 | 5.100.156.10 | TVRAT | NaN |
| 3 | 5.206.224.14 | Formbook | NaN |
| 4 | 145.14.145.219 | Betabot | NaN |

```
In [19]: #Dataset 2
Zeus = pd.read_html('https://zeustracker.abuse.ch/monitor.php?filter=all',)
Zeus = Zeus[1] # we need the second table that is found
Zeus.head()
```

```
Out [19]:
```

| | Dateadded | Malware | Host | IP address | Level |
|---|------------|---------|-----------------------------|----------------|-------|
| 0 | 2017-11-07 | Citadel | slap.alliancekl.com | NaN | 4 |
| 1 | 2017-11-06 | VMZeus | bestframingnailerreview.com | 104.243.42.202 | 2 |
| 2 | 2017-10-23 | Citadel | flex.comonwealthplc.com | 37.49.225.131 | 4 |
| 3 | 2017-10-22 | VMZeus | kjkdndskjl.info | NaN | 4 |
| 4 | 2017-10-21 | Citadel | 193.0.178.27 | 193.0.178.27 | 4 |

| | Status | Files Online | SBL | Country | AS number | Uptime |
|---|---------|--------------|------------|---------|-----------|-----------|
| 0 | offline | 0 | Not listed | - | NaN | 50:10:43 |
| 1 | online | 0 | Not listed | NaN | AS20473 | 74:01:57 |
| 2 | online | 0 | SBL374926 | NaN | AS50673 | 408:04:02 |
| 3 | offline | 0 | Not listed | - | NaN | 123:58:31 |
| 4 | unknown | 0 | Not listed | NaN | AS57062 | - |

```
In [20]: #Dataset 3, requires a bit more cleaning
```

```
Ransomware = pd.read_csv('https://ransomwaretracker.abuse.ch/feeds/csv/',
iplistsplt = Ransomware['IP address(es)'].astype(str).str.split('|')
Ransomware['IP address(es)'] = [item[0] for item in iplistsplit]
countrytsplt = Ransomware['Country'].astype(str).str.split('|')
Ransomware['Country'] = [item[0] for item in countrytsplt]
Ransomware = Ransomware.iloc[: -1, :]
Ransomware.tail()
```

```
Out [20]:
```

| | # | Firstseen (UTC) | Threat | Malware | Host | \ |
|-------|------------|-----------------|--------|------------|-----------------|---|
| 13717 | 2015-04-15 | 14:00:53 | C2 | CryptoWall | marcortes.com | |
| 13718 | 2015-03-07 | 13:46:58 | C2 | CryptoWall | drdigitalmd.com | |
| 13719 | 2015-03-04 | 12:10:59 | C2 | CryptoWall | rajsima87.com | |
| 13720 | 2015-03-02 | 04:14:59 | C2 | CryptoWall | jauregia.net | |
| 13721 | 2015-03-02 | 01:18:48 | C2 | CryptoWall | lzclient.com | |

| | URL | Status | \ |
|-------|---------------------------------|---------|---|
| 13717 | http://marcortes.com/img5.php | offline | |
| 13718 | http://drdigitalmd.com/img1.php | offline | |
| 13719 | http://rajsima87.com/img2.php | offline | |
| 13720 | http://jauregia.net/img5.php | offline | |
| 13721 | http://lzclient.com/img4.php | offline | |

| | Registrar | IP address(es) | ASN(s) | \ |
|-------|---|-----------------|--------|---|
| 13717 | OVH | 91.134.158.216 | 16276 | |
| 13718 | PDR LTD. D/B/A PUBLICDOMAINREGISTRY.COM | 192.210.234.140 | 36352 | |
| 13719 | eNom, Inc. | nan | NaN | |
| 13720 | TECNOCRATICA CENTRO DE DATOS, S.L. | 185.92.246.115 | 201446 | |
| 13721 | eNom, Inc. | nan | NaN | |

| | Country |
|-------|---------|
| 13717 | FR |
| 13718 | US |
| 13719 | nan |

```
13720      ES
13721      nan
```

In [21]: *#Create a list of random IP addresses for comparison*

```
Random_Ips = []
n = 15000
for i in range (0,n):
    Random_Ips.append(socket.inet_ntoa(struct.pack('>I', random.randint(1,
    print(' Drawing IP no:' + str(i), end='\r')
print("Drew " +str(n)+ " random IP addresses")
Random_Ips = pd.DataFrame({'ip':Random_Ips})
```

Drew 15000 random IP addresses

In [22]: *#Print out the ten most commons IP addresses in the datasets*

```
cybercrime['-::IP'].value_counts()[0:10], Zeus['IP address'].value_counts
```

```
Out[22]: (62.109.9.188      83
198.105.221.5      71
174.127.78.72      66
108.162.199.107    58
198.176.28.49      45
66.45.253.74       33
172.93.106.18      33
108.170.51.58      31
108.61.47.186      31
198.1.80.203       31
Name: -::IP, dtype: int64, 104.238.158.106      17
FastFlux Botnet    14
141.8.226.58       9
198.54.117.212     3
5.9.107.19         3
144.76.115.36      3
66.45.245.150      2
80.78.250.26       2
186.250.244.100    2
195.20.44.100      1
Name: IP address, dtype: int64, 184.105.192.2      242
127.0.0.1          239
213.205.40.169     170
195.157.15.100     84
204.11.56.48       71
80.150.6.138       65
95.211.144.65      47
112.140.42.29      47
52.71.185.125      46
208.100.26.234     38
Name: IP address(es), dtype: int64)
```

```
In [23]: #Print out the ten most commons hosts in the datasets
Zeus['Host'].value_counts()[1:10],Ransomware['Host'].value_counts()[0:10]
```

```
Out[23]: (lisovfoxcom.418.com1.ru      1
prtscrinsertcn.net                  1
120.63.157.195                      1
atmape.ru                          1
gyodundena.hotmail.ru              1
trust-s-b.com                      1
dl.dropbox.com                     1
nicktung.com                       1
noonepa.tk                         1
Name: Host, dtype: int64, bolizarsospos.com      84
mafianeedsyoudqq.com                22
gutentagmeinliebeqq.com             21
itsyourtimeqq.su                    21
goonwithmazerqq.com                 19
helloyoungmanqq.com                 18
facerecognition.com.ba              13
lenovowantsyouqq.com                12
invoiceholderqq.com                 11
thisisyourchangeqq.com              11
Name: Host, dtype: int64)
```

```
In [24]: #Print out the ten most commons hosts in the datasets
Zeus['Country'].value_counts()[0:10],Ransomware['Country'].value_counts()
```

```
Out[24]: (-      284
Name: Country, dtype: int64, nan      4462
US      3253
DE      634
RU      547
NL      412
IT      404
GB      353
FR      273
CN      225
PL      225
Name: Country, dtype: int64)
```

```
In [25]: #Print out the ten most common malware types in each dataset
cybercrime['-::TYPE'].value_counts()[0:10], Zeus['Malware'].value_counts()
```

```
Out[25]: (Pony      2698
ZeusS      2325
Citadel    1175
Blackhole   850
WSO         400
IceIX       299)
```

```

Stealer          255
Mailer           193
Betabot          182
Solar            178
Name: -::TYPE, dtype: int64, Zeus          192
Citadel          163
VMZeus           69
KINS             46
Ice IX           10
Name: Malware, dtype: int64, Locky          11303
Cerber           1573
TeslaCrypt       477
CryptoWall       203
TorrentLocker    120
PayCrypt         21
Sage              9
CTB-Locker       6
PadCrypt         4
FAKBEN           3
Name: Malware, dtype: int64)

```

In [37]: *#Venn plot*

```

a = Zeus['IP address']
b = Ransomware['IP address(es)']
c = cybercrime['-::IP']

s = (
    len(a),      # Abc
    len(b),      # aBc
    len(set.intersection(set(a), set(b))),    # ABc
    len(c),      # abC
    len(set.intersection(set(a), set(c))),    # AbC
    len(set.intersection(set(b), set(c))),    # aBC
    len(set.intersection(set(a), set(b), set(c))),    # ABC
)
v = venn3(subsets=s, set_labels=('Zeus', 'Ransomware', 'Cybercrime'))

# Subset labels
v.get_label_by_id('100').set_text(str(len(a)))
v.get_label_by_id('010').set_text(str(len(b)))
v.get_label_by_id('110').set_text(str(len(set.intersection(set(a), set(b)))))
v.get_label_by_id('001').set_text(str(len(c)))
v.get_label_by_id('101').set_text(str(len(set.intersection(set(a), set(c)))))
v.get_label_by_id('011').set_text(str(len(set.intersection(set(b), set(c)))))
#v.get_label_by_id('111').set_text(str(len(set.intersection(set(a), set(b)))))

# Subset colors

```

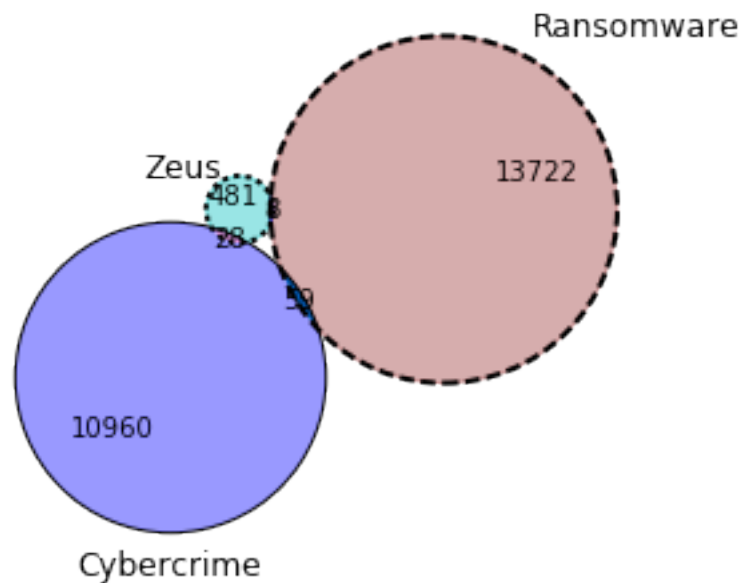
```

v.get_patch_by_id('100').set_color('c')
v.get_patch_by_id('010').set_color('#993333')
v.get_patch_by_id('110').set_color('blue')

# Subset alphas
v.get_patch_by_id('101').set_alpha(0.4)
v.get_patch_by_id('011').set_alpha(1.0)
#v.get_patch_by_id('111').set_alpha(0.7)

# Border styles
c = venn3_circles(subsets=s, linestyle='solid')
c[0].set_ls('dotted') # Line style
c[1].set_ls('dashed')
c[2].set_lw(1.0) # Line width
plt.figure(figsize=(15,15))
plt.show()

```



<matplotlib.figure.Figure at 0xec88290>

```

In [38]: def fracx(a):
          #Creates x values for plotting cumulative fractions
          a = np.array(list(range(0,len(a))))/len(a)
          return a

In [39]: a = Zeus['IP address'].value_counts(normalize=True)
          b = Ransomware['IP address(es)'].value_counts(normalize=True)

```

```

c = cybercrime['-::IP'].value_counts(normalize=True)
d = Random_Ips['ip'].value_counts(normalize=True)

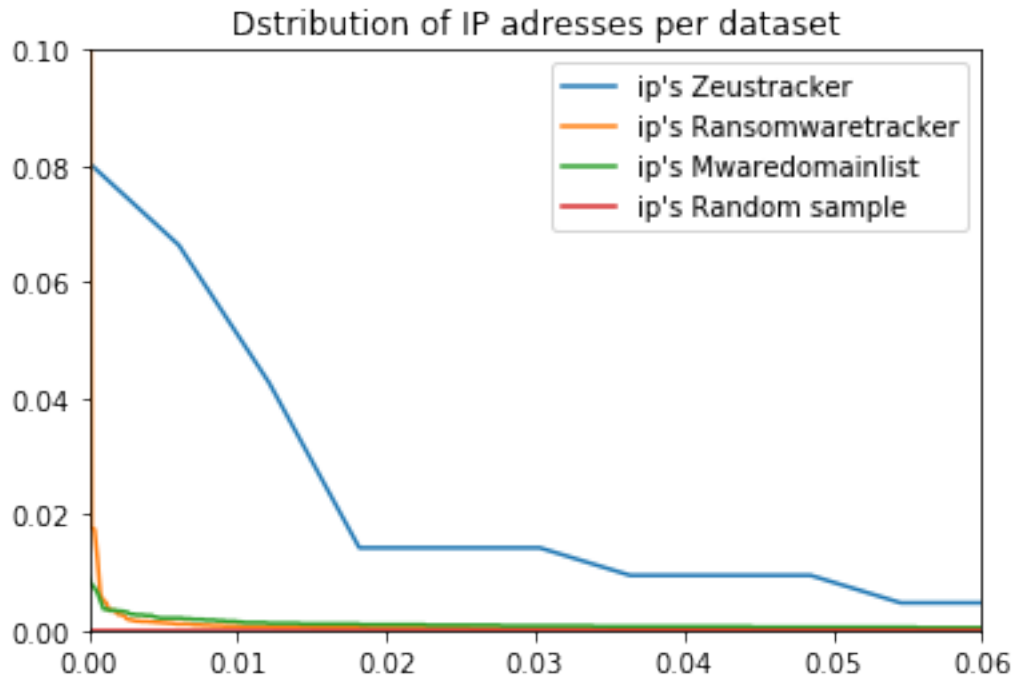
ax = plt.subplot()

plt.plot(fracx(a),a, label="ip's Zeustracker")
plt.plot(fracx(b),b, label="ip's Ransomwaretracker")
plt.plot(fracx(c),c, label="ip's Mwaredomainlist")
plt.plot(fracx(d),d, label="ip's Random sample")

ax.set_xlim(0,0.06)
ax.set_ylim(0,0.1)
plt.title('Dstribution of IP adresses per dataset')

plt.legend()
plt.show()

```



```

In [40]: a = Zeus['IP address'].value_counts(normalize=True).cumsum()
b = Ransomware['IP address(es)'].value_counts(normalize=True).cumsum()
c = cybercrime['-::IP'].value_counts(normalize=True).cumsum()
d = Random_Ips['ip'].value_counts(normalize=True).cumsum()

ax = plt.subplot()

plt.plot(fracx(a),a, label="ip's Zeustracker")

```

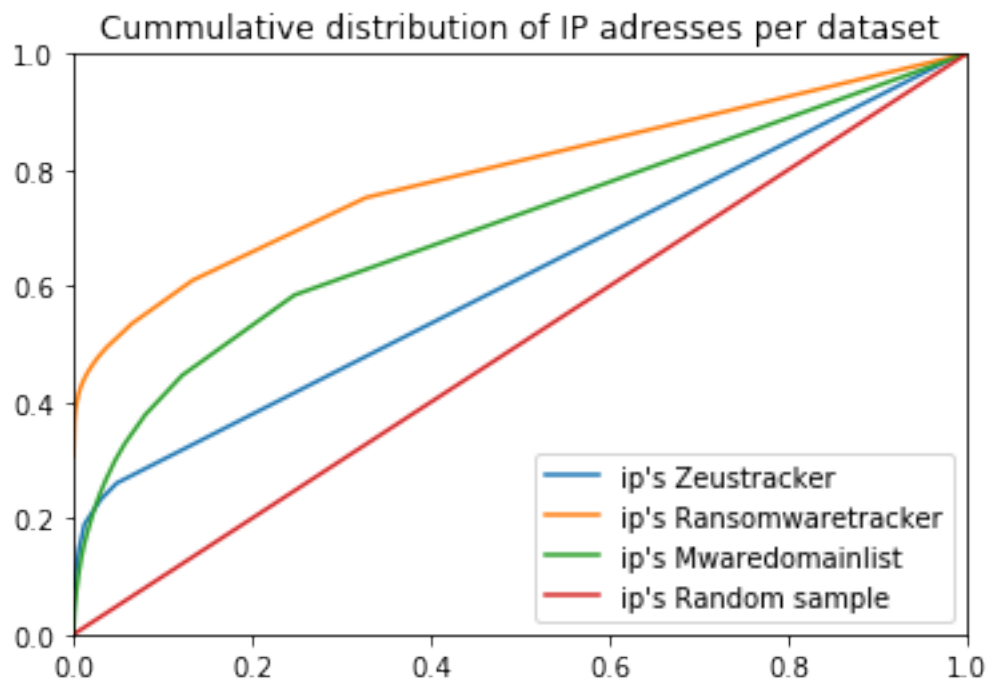
```

plt.plot(fracx(b),b, label="ip's Ransomwaretracker")
plt.plot(fracx(c),c, label="ip's Mwaredomainlist")
plt.plot(fracx(d),d, label="ip's Random sample")

ax.set_xlim(0,1)
ax.set_ylim(0,1)
plt.title('Cummulative distribution of IP adresses per dataset')

plt.legend()
plt.show()

```



```

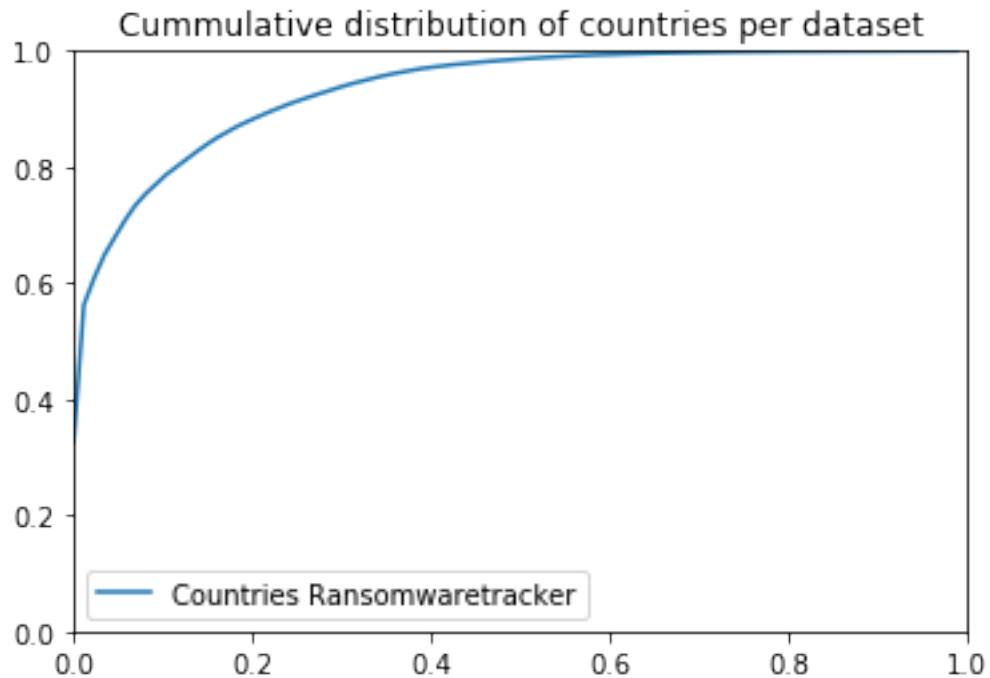
In [41]: # Plot the cumulative distribution of countries
b = Ransomware['Country'].value_counts(normalize=True)
ax = plt.subplot()

plt.plot(fracx(b),b.cumsum(), label="Countries Ransomwaretracker")
plt.title('Cummulative distribution of countries per dataset')

ax.set_xlim(0,1)
ax.set_ylim(0,1)

plt.legend()
plt.show()

```

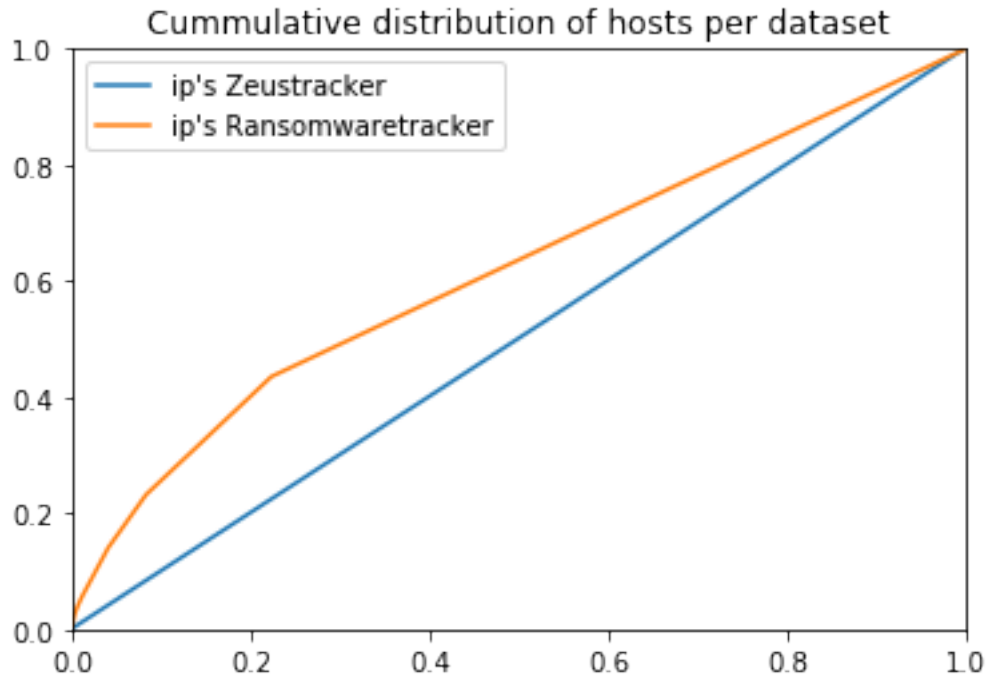
```
In [42]: a = Zeus['Host'].value_counts(normalize=True).cumsum()
        b = Ransomware['Host'].value_counts(normalize=True, dropna=False).cumsum()

        ax = plt.subplot()

        plt.plot(fracx(a),a, label="ip's Zeustracker")
        plt.plot(fracx(b),b, label="ip's Ransomwaretracker")

        ax.set_xlim(0,1)
        ax.set_ylim(0,1)
        plt.title('Cummulative distribution of hosts per dataset')

        plt.legend()
        plt.show()
```



```
In [43]: # This function prints out the outcomes of our statistics test
def ks_samp(a,b):
    print(stats.ks_2samp(a,b))
```

```
In [44]: a = Zeus['IP address'].value_counts(normalize=True).cumsum()
b = Ransomware['IP address(es)'].value_counts(normalize=True).cumsum()
c = cybercrime['-::IP'].value_counts(normalize=True).cumsum()
d = Random_Ips['ip'].value_counts(normalize=True).cumsum()
```

```
In [45]: # Perform the statistical tests
ks_samp(a,b) #Accept 0 hyp (from same dist.) if p>0.05
ks_samp(b,c) #Accept 0 hyp (from same dist.) if p>0.05
ks_samp(a,c) #Accept 0 hyp (from same dist.) if p>0.05
ks_samp(a,d) #Accept 0 hyp (from same dist.) if p>0.05
ks_samp(b,d) #Accept 0 hyp (from same dist.) if p>0.05
ks_samp(c,d) #Accept 0 hyp (from same dist.) if p>0.05
```

```
Ks_2sampResult(statistic=0.36752090639330998, pvalue=1.4830729305974341e-19)
Ks_2sampResult(statistic=0.22438511318167625, pvalue=1.4387054376873975e-110)
Ks_2sampResult(statistic=0.22104204385435527, pvalue=2.4255996810306513e-07)
Ks_2sampResult(statistic=0.21211515151515151, pvalue=6.221236023578169e-07)
Ks_2sampResult(statistic=0.47685143422354104, pvalue=0.0)
Ks_2sampResult(statistic=0.33757304365318852, pvalue=0.0)
```

```
In [46]: #Calculate the % up per country metric
a = Ransomware[Ransomware['Status'] == 'offline'].Country.value_counts()
b = Ransomware[Ransomware['Status'] == 'online'].Country.value_counts()
(b/(a+b)).dropna().sort_values(ascending=False)[0:10]
```

```
Out[46]: BE      0.285714
        FI      0.250000
        GB      0.246459
        IL      0.166667
        IR      0.105263
        SK      0.090909
        AR      0.086957
        KR      0.083333
        TH      0.080882
        ID      0.072917
        Name: Country, dtype: float64
```

```
In [47]: #Calculate the % up per host metric
a = Ransomware[Ransomware['Status'] == 'offline'].Host.value_counts()
b = Ransomware[Ransomware['Status'] == 'online'].Host.value_counts()
#c = a + b
(b/(a+b)).dropna().sort_values(ascending=False)[0:10]
```

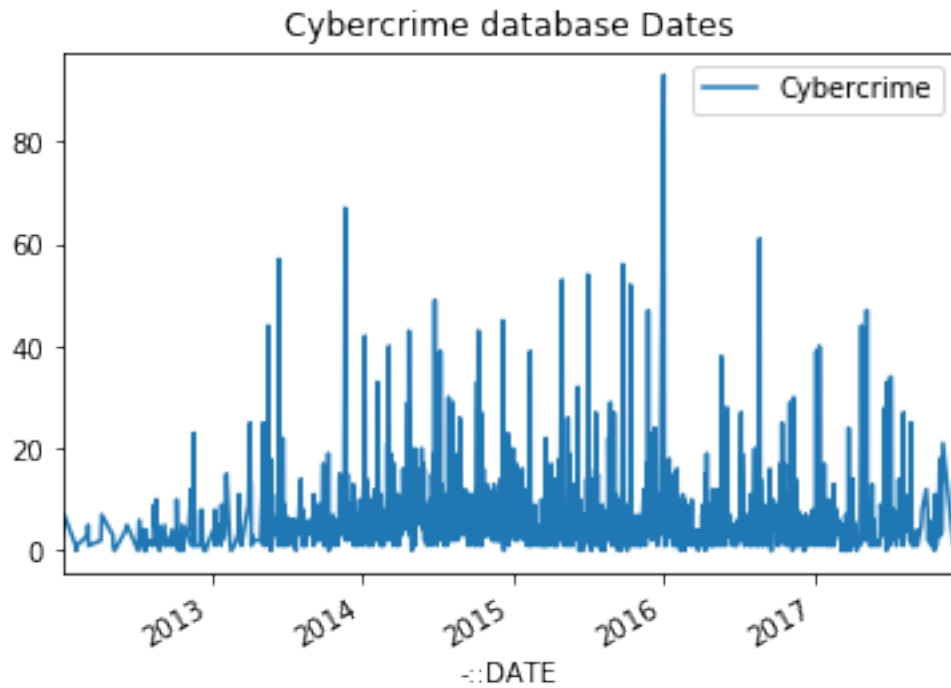
```
Out[47]: bolizarsospos.com      0.964286
        wt7dzbn78.homepage.t-online.de  0.666667
        galeona.com      0.666667
        greatgoods2.bravepages.com      0.666667
        www.resumebuddy.net      0.500000
        3e.com.pt      0.500000
        bptpm.sragenkab.go.id      0.500000
        demo.evgesha.ru      0.500000
        flax-fiber.com      0.500000
        hjhqmbxyinislkkt.15u3kg.top      0.500000
        Name: Host, dtype: float64
```

```
In [48]: # Prepare the kruskal test to see if number of ransomware per country is c
b = pd.DataFrame(Ransomware['Country'].value_counts())
needed = ['US', 'DE', 'RU', 'NL', 'CH', 'CN', 'CA', 'IN', 'BR', 'TH']
lista = b[b.index.isin(needed)].Country
listb = [36,27,17,15,10,6,5,5,5,5]
stats.kruskal(lista,listb) # p<0.5, so not simmilar
```

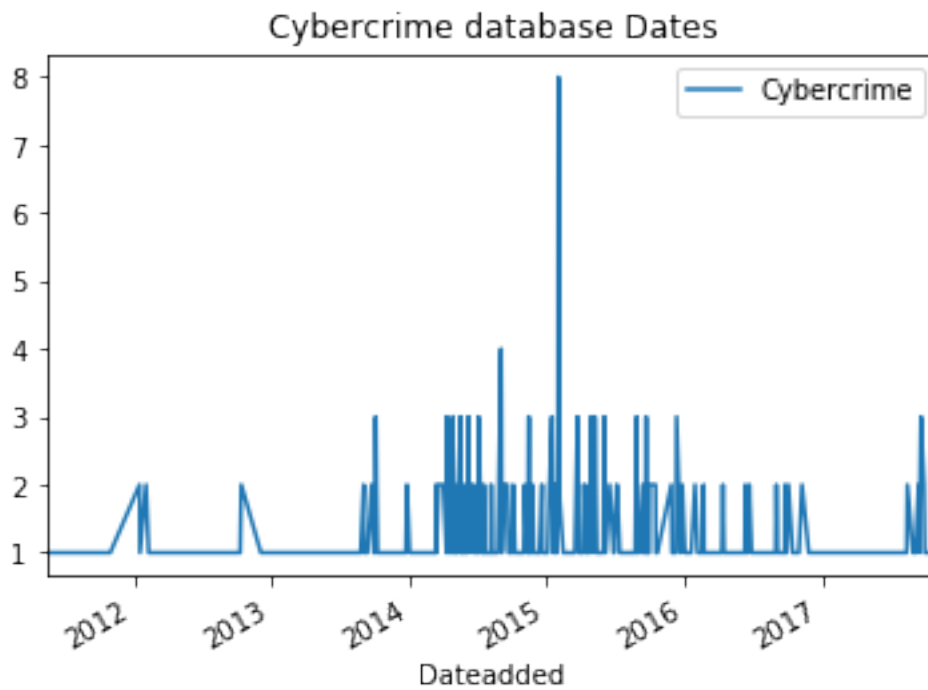
```
Out[48]: KruskalResult (statistic=14.393939393939387, pvalue=0.00014827877260963348)
```

```
In [49]: #Make dates plotable
cybercrime['-::DATE'] = pd.to_datetime(cybercrime['-::DATE'])
Zeus['Dateadded'] = pd.to_datetime(Zeus['Dateadded'])
Ransomware["# Firstseen (UTC)"] = pd.to_datetime(Ransomware["# Firstseen (UTC)"])
```

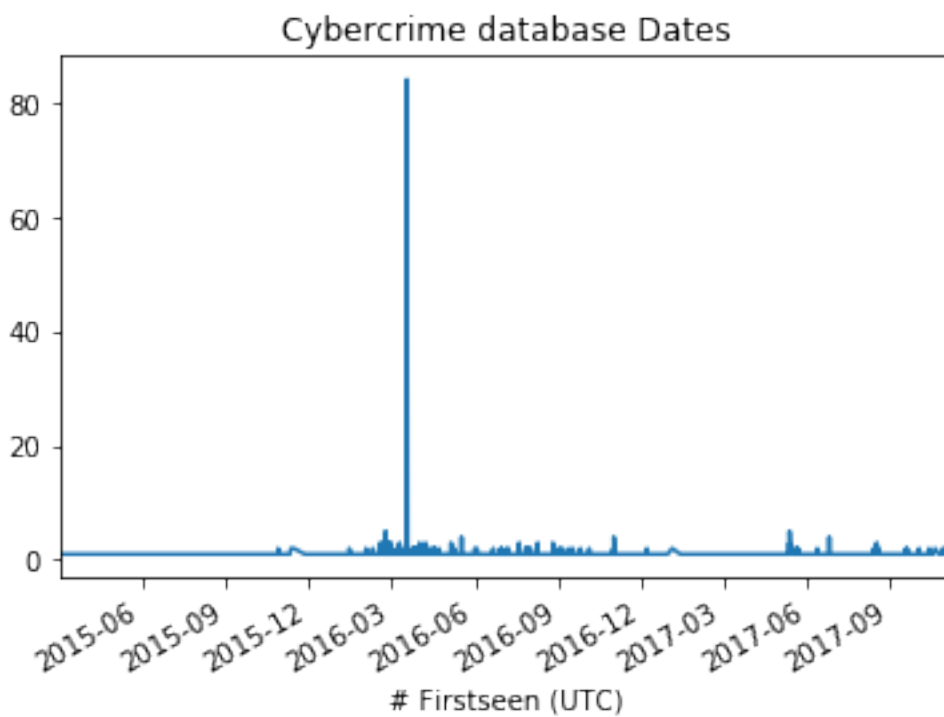
```
In [50]: #Plot entries per date for Cybercrime
cybercrime.groupby('-::DATE').count().iloc[1:,1].plot(label='Cybercrime',
plt.legend()
plt.show()
```



```
In [51]: #Plot entries per date for Zeus
Zeus.groupby('Dateadded').count()['Status'].plot(label='Cybercrime', title
plt.legend()
plt.show()
```

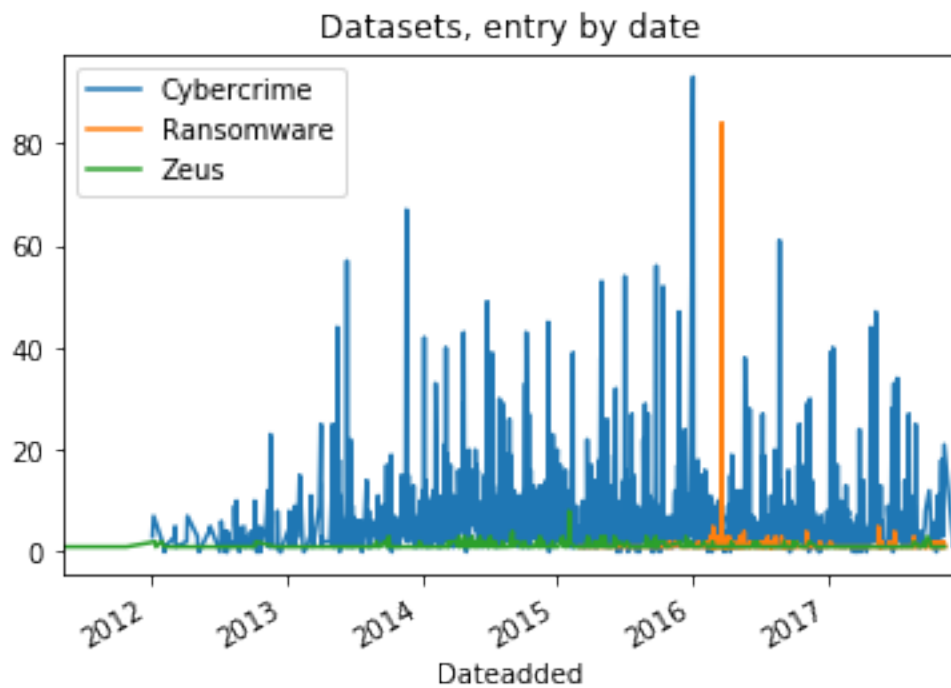


```
In [52]: #Plot entries per date for Ransomware
Ransomware.groupby("# Firstseen (UTC)").count().Status.plot(label='Cybercrime')
plt.show()
```



```
In [53]: #Plot for all
cybercrime.groupby('-::DATE').count().iloc[1:,1].plot(label='Cybercrime')
Ransomware.groupby("# Firstseen (UTC)").count().Status.plot(label='Ransomv')
Zeus.groupby('Dateadded').count()['Status'].plot(label='Zeus', title='Data

plt.legend()
plt.show()
```



```
In [ ]:
```