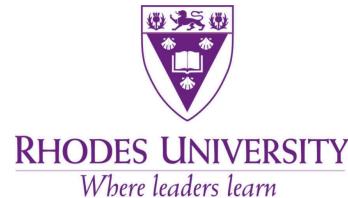


DNS Based Botnet C2 Server Detection

Spatial Statistics as a detection metric





@kamp_staaldraad



etienne@sensepost.com

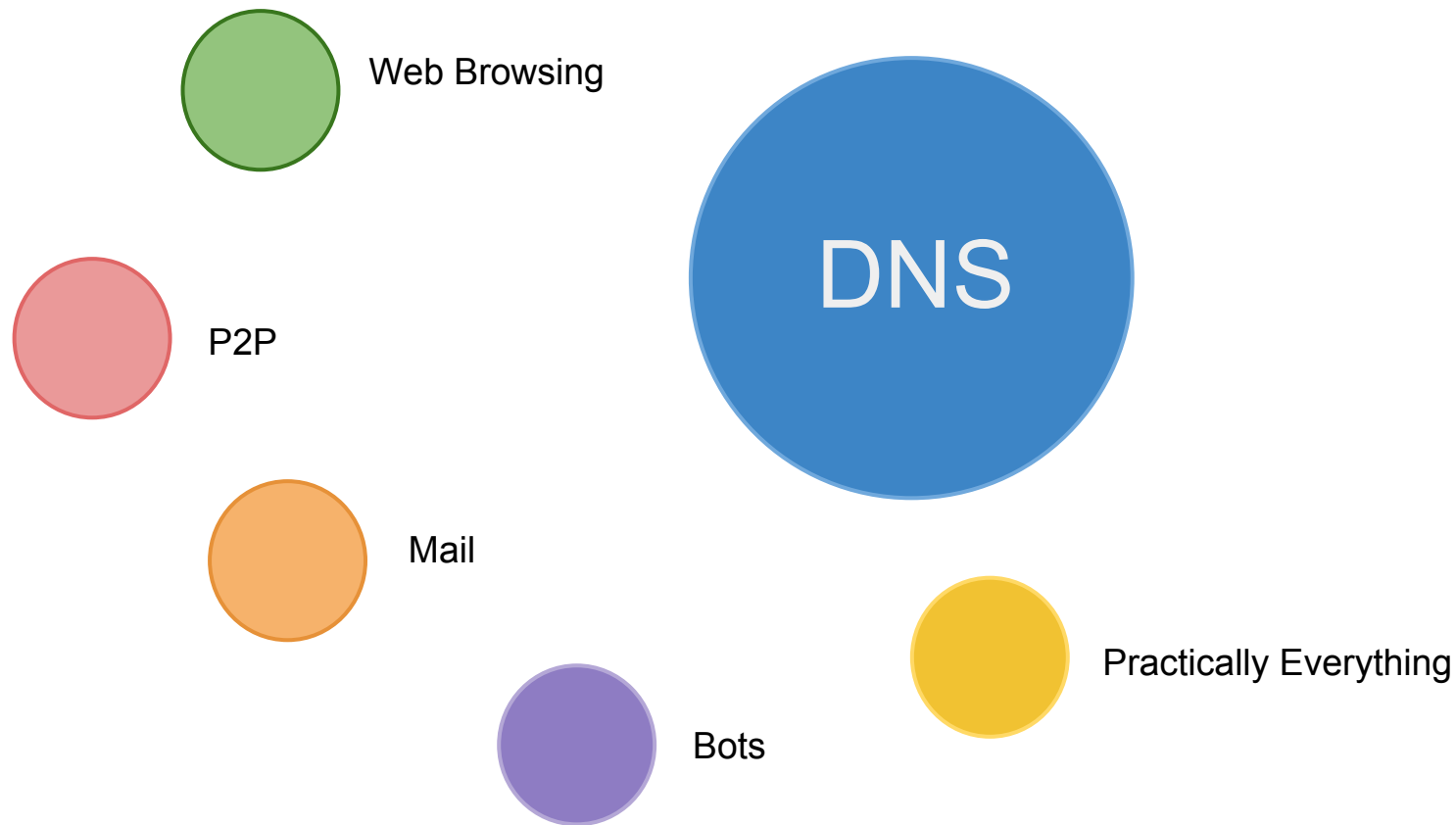


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Geographic Analysis

Research Goals

- Accurately detect botnet traffic
 - Assume no prior knowledge
 - Lightweight
 - Fast
 - Adaptable
- Early detection



Examining DNS

DNS Fast-Flux

```
; <<>> DiG 9.8.1-P1 <<>> lovenewgirl.com
;; global options: +cmd
;; Got answer:
;; ->>HEADER<- opcode: QUERY, status: NOERROR, id: 12744
;; flags: qr rd ra; QUERY: 1, ANSWER: 5, AUTHORITY: 5, ADDITIONAL: 0

;; QUESTION SECTION:
;lovenewgirl.com.                IN      A

;; ANSWER SECTION:
lovenewgirl.com.        600     IN      A      67.190.124.84
lovenewgirl.com.        600     IN      A      78.60.45.112
lovenewgirl.com.        600     IN      A      118.14.224.139
lovenewgirl.com.        600     IN      A      194.90.36.187
lovenewgirl.com.        600     IN      A      222.106.31.112

;; AUTHORITY SECTION:
lovenewgirl.com.        86400   IN      NS      ns5.oldcitiesmaps.com.
lovenewgirl.com.        86400   IN      NS      ns2.oldcitiesmaps.com.
lovenewgirl.com.        86400   IN      NS      ns3.oldcitiesmaps.com.
lovenewgirl.com.        86400   IN      NS      ns4.oldcitiesmaps.com.
lovenewgirl.com.        86400   IN      NS      ns1.oldcitiesmaps.com.

;; Query time: 493 msec
;; SERVER: 127.0.0.1#53(127.0.0.1)
;; WHEN: Tue Aug  7 09:08:11 2012
;; MSG SIZE rcvd: 217
```

- Short TTL
- Multiple A Records
- Different IP Ranges

DNS Fast-Flux

```
;; ANSWER SECTION:
lovenewgirl.com.      600      IN       A        67.190.124.84
lovenewgirl.com.      600      IN       A        78.60.45.112
lovenewgirl.com.      600      IN       A        118.14.224.139
lovenewgirl.com.      600      IN       A        194.90.36.187
lovenewgirl.com.      600      IN       A        222.106.31.112
```

Lat: 39.7437 Lon: -104.9793
UTM: 37Z
MGRS: 13SED0177499311

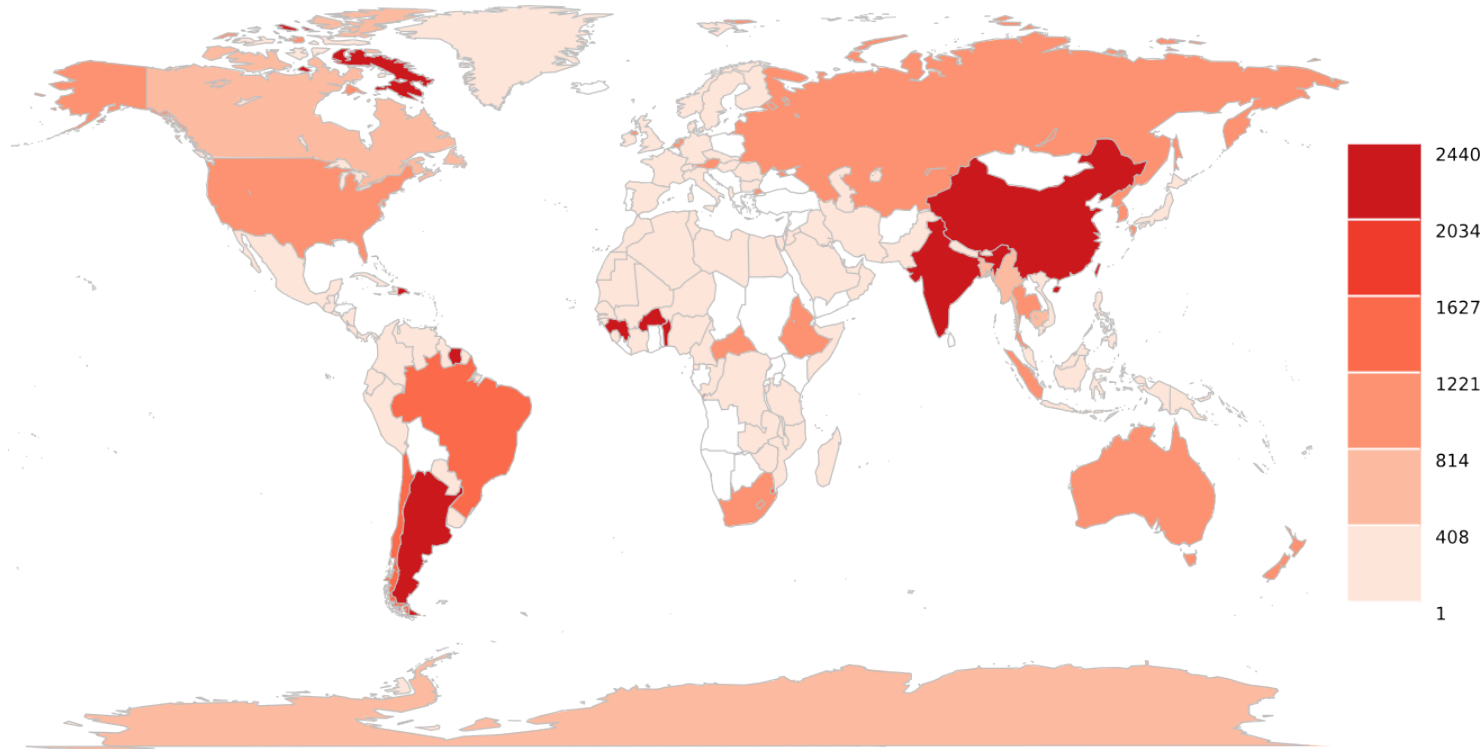
Lat: 54.6833 Lon: 25.3167
UTM: 40R
MGRS: 35ULA9148060851

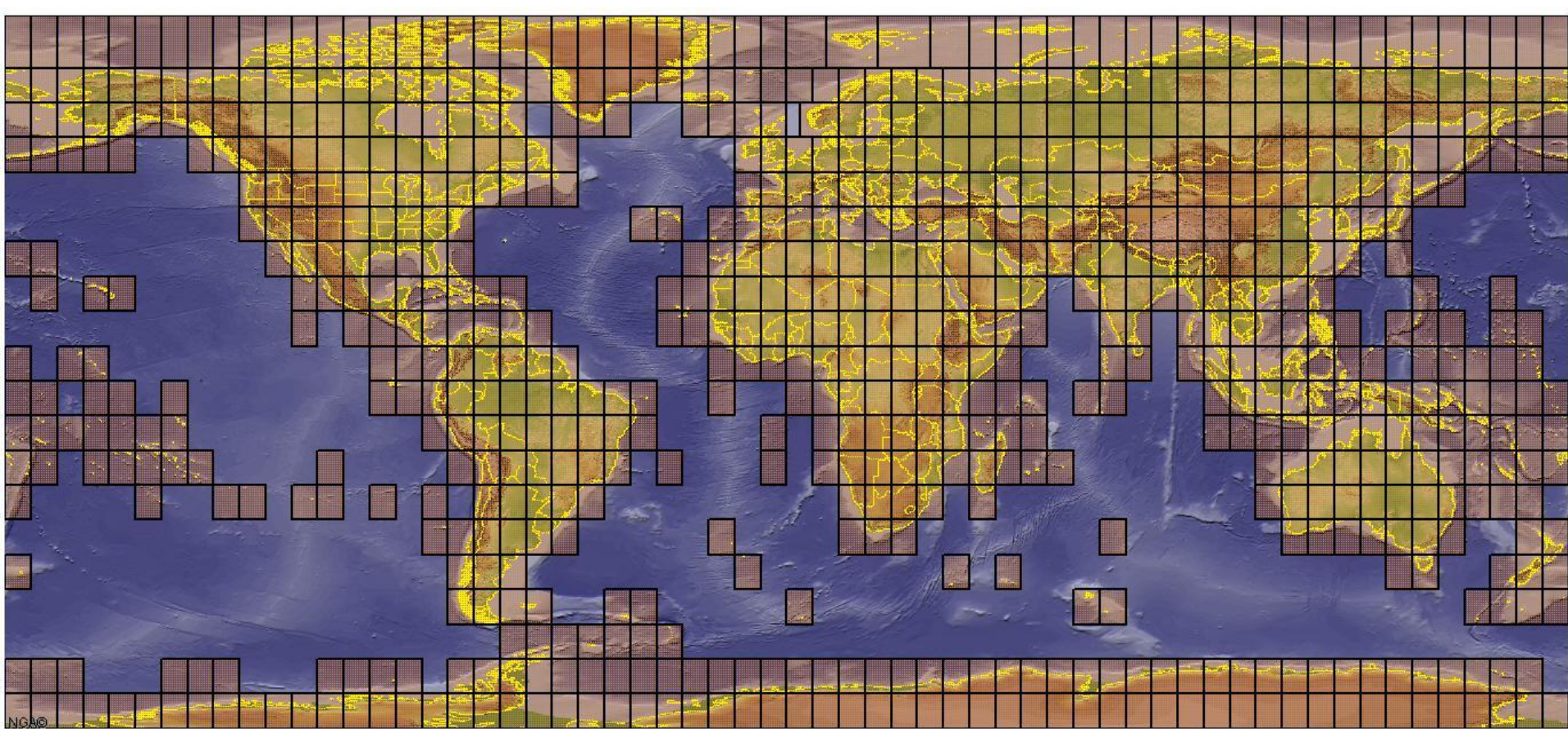
Lat: 34.6833 Lon: 135.8333
UTM: 36Z
MGRS: 53SNU7633338239

IP	ASN	Country	Timezone
67.190.124.84	AS7922	US	America/Denver
78.60.45.112	AS8764	LT	Europe/Vilnius
118.14.224.139	AS4713	JP	Asia/Tokyo
194.90.36.187	AS1680	IL	Asia/Jerusalem
222.106.31.112	AS4766	KR	Asia/Seoul

- Multiple ASNs
- Multiple Countries
- Multiple Timezones
- Multiple Unique Location Identifiers

Widely Dispersed Networks





Spatial Measures

http://earth-info.nga.mil/GandG/coordsys/images/MGRS_1km_Polygon_Shapefiles_Coverage.jpg

Spatial Measures

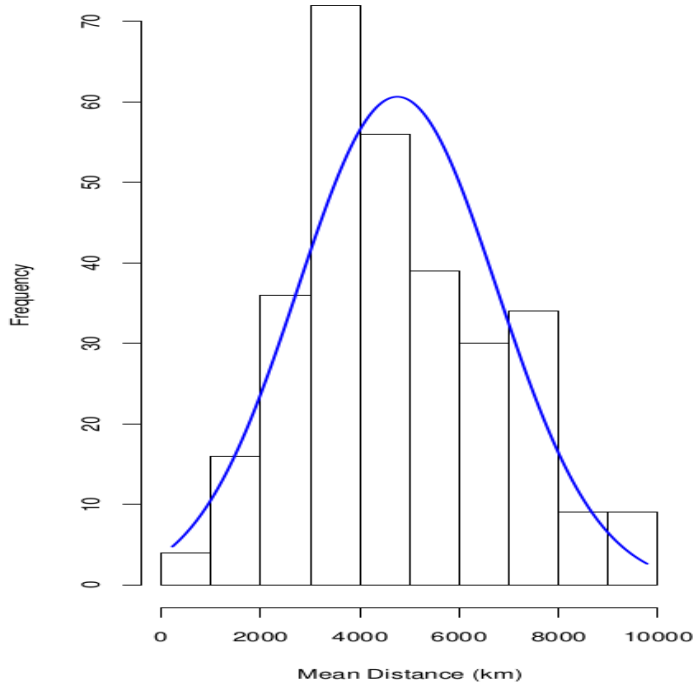
[08:32]etienne@null

↳ ~/Documents/Rhodes/Masters/Prototype > python presentationData.py

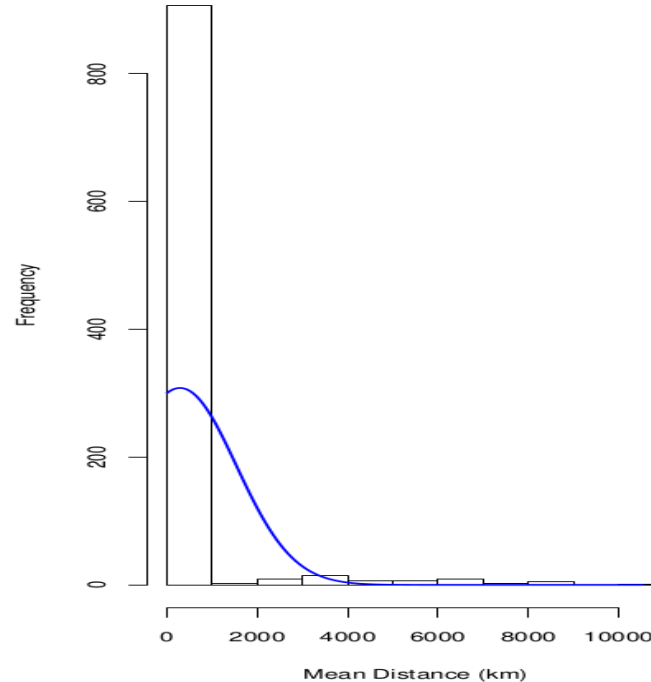
IP	ASN	Country	Timezone	UTM	MGRS
67.190.124.84	AS7922	US	America/Denver	37Z	13SED0177499311
78.60.45.112	AS8764	LT	Europe/Vilnius	40R	35ULA9148060851
118.14.224.139	AS4713	JP	Asia/Tokyo	36Z	53SNU7633338239
194.90.36.187	AS1680	IL	Asia/Jerusalem	36S	36SXB8622432597
222.106.31.112	AS4766	KR	Asia/Seoul	37Z	52SCG2334159589
79.108.149.71	AS6739	ES	Europe/Madrid	37M	30SYH0125936055
79.139.110.20	AS41740	PL	Europe/Warsaw	39Q	34UFA2837416063
79.139.110.20	AS41740	PL	Europe/Warsaw	39Q	34UFA2837416063
88.132.63.164	AS35311	HU	Europe/Budapest	38Q	34TDT0755809583
79.108.149.71	AS6739	ES	Europe/Madrid	37M	30SYH0125936055
124.6.3.225	AS24165	TW	Asia/Taipei	34Z	51QTF2762705352
89.229.214.126	AS21021	PL	Europe/Warsaw	39Q	34UCE6257855864
124.6.3.225	AS24165	TW	Asia/Taipei	34Z	51QTF2762705352
68.119.57.22	AS20115	US	America/New_York	36Z	17SKT4153099735

Nearest Neighbours

Fast-Flux Domains



Benign Domains



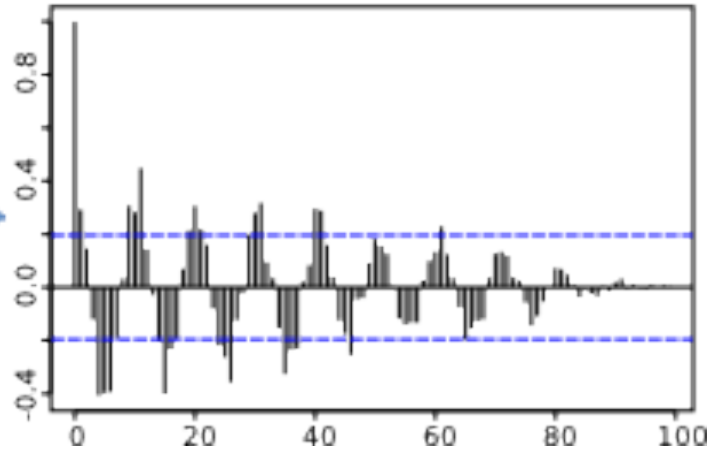
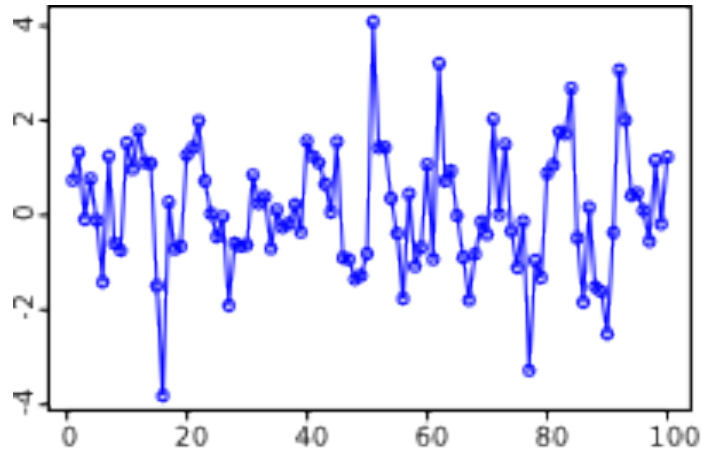


Spatial Statistics

First Law of Geography

"All things are related, but near things are more related than far things." - W. Tobler

Autocorrelation



Moran's Index

$$I = \frac{N}{\sum_i \sum_j w_{ij}} \frac{\sum_i \sum_j w_{ij} (X_i - \bar{X})(X_j - \bar{X})}{\sum_i (X_i - \bar{X})^2}$$

Geary's Coefficient

$$C = \frac{(N - 1) \sum_i \sum_j w_{ij} (X_i - X_j)^2}{2W \sum_i (X_i - \bar{X})^2}$$


```

54 def calcHaverDistance(self,lat1,lat2,lon1,lon2):
55     '''
56     Calculate the Distance between two locations using the Haver method
57     @param latitude of location 1
58     @param latitude of location 2
59     @param longitude of location 1
60     @param longitude of location 2
61     @return the distance between the two locations
62     '''
63     dLat = math.radians(lat2-lat1)
64     dLon = math.radians(lon2-lon1)
65
66     a = math.sin(dLat/2) * math.sin(dLat/2) + math.sin(dLon/2) * math.sin(
67     c = 2 * math.atan2(math.sqrt(a), math.sqrt(1-a))
68     d = self.R * c
69     return d
70
71 def geary(self,matrix,values,meanv,N):
72     vx1 =0
73     vx2 =0
74     vx3 =0
75     W = 0
76     for i,x1 in enumerate(values):
77         for j,x2 in enumerate(values):
78             #print matrix[i][j],x1,x2
79             vx1 += matrix[i][j]*((x1-x2)*(x1-x2))
80
81             W += matrix[i][j]
82             vx3 += (x1-meanv)*(x1-meanv)
83     vx2 = 2*W
84
85     if vx3 ==0 or vx2==0:
86         Ix = 0
87     else:
88         Ix = ((N-1)*vx1)/(vx2*vx3)
89
90     return Ix
91
92 def getMGRSVal(self,lat,lon):
93     m = mgrs.MGRS()
94     c = m.toMGRS(lat,lon)
95     ind = 1
96     for i in range(1,len(c)):
97         if c[i].isalpha():
98             ind = i
99             break
100     v1 = int(c[:ind])*(ord(c[ind:ind+1])+ord(c[ind+1:ind+2]))
101     v2 = int(c[ind+3:])
102
103     return (v1*v2,c)

```

Building the Classifiers

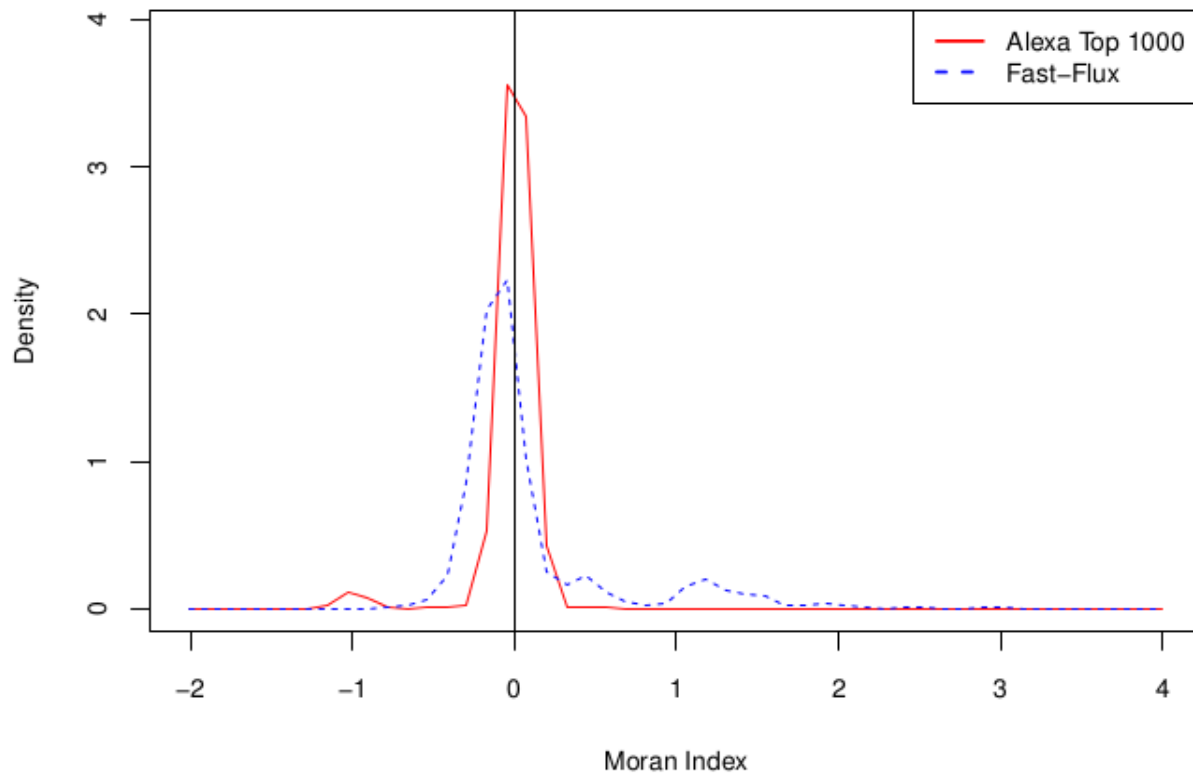
Classifier Training

 **Alexa** Benign Dataset

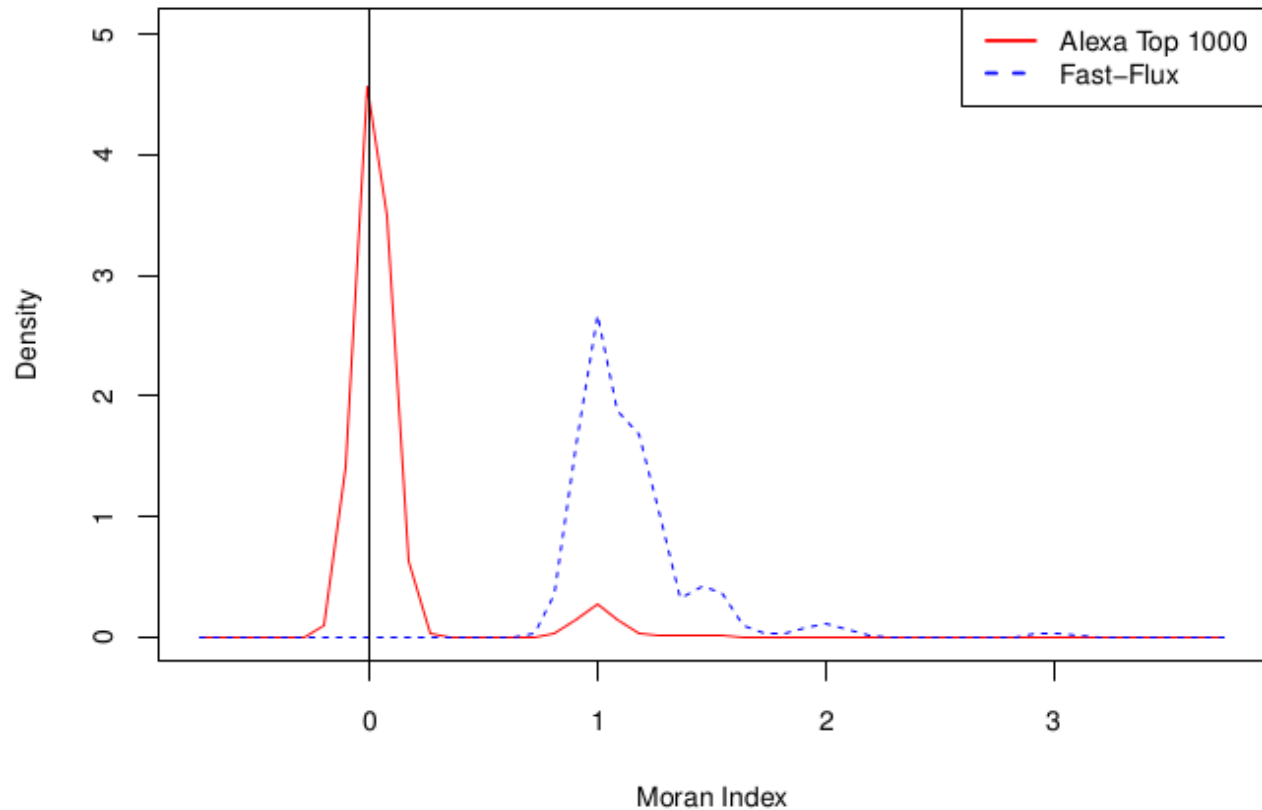
 **ARBOR**
NETWORKS

Fast-Flux
Dataset

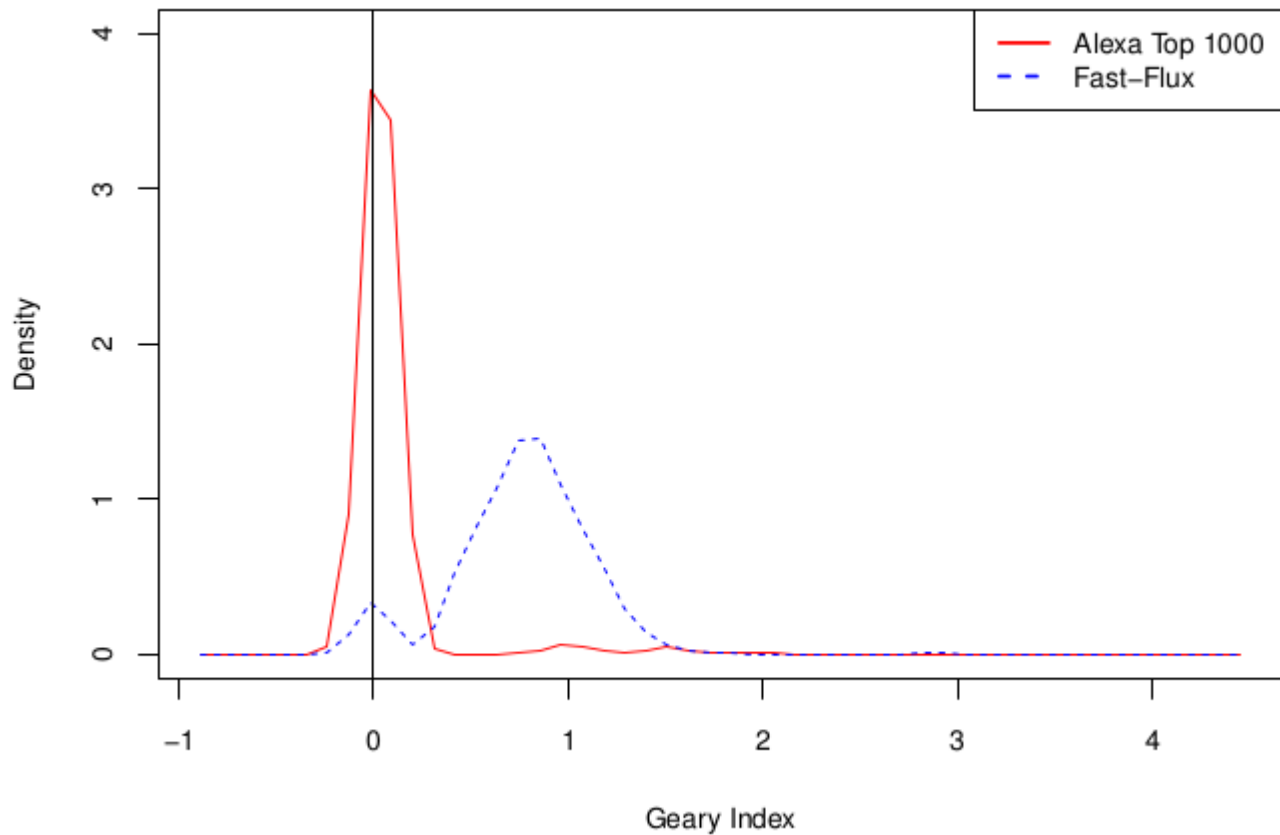
abuse.ch Zeus Tracker



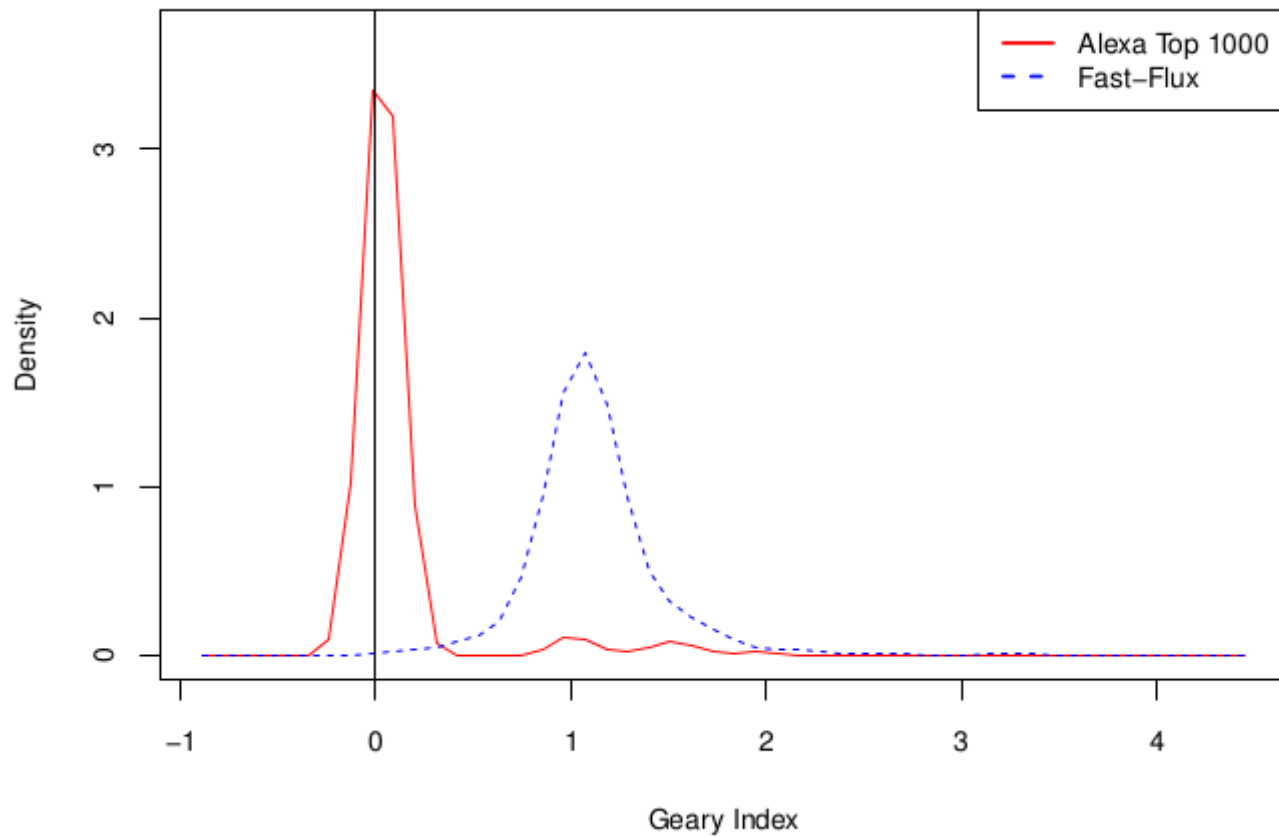
Moran's I: Timezones



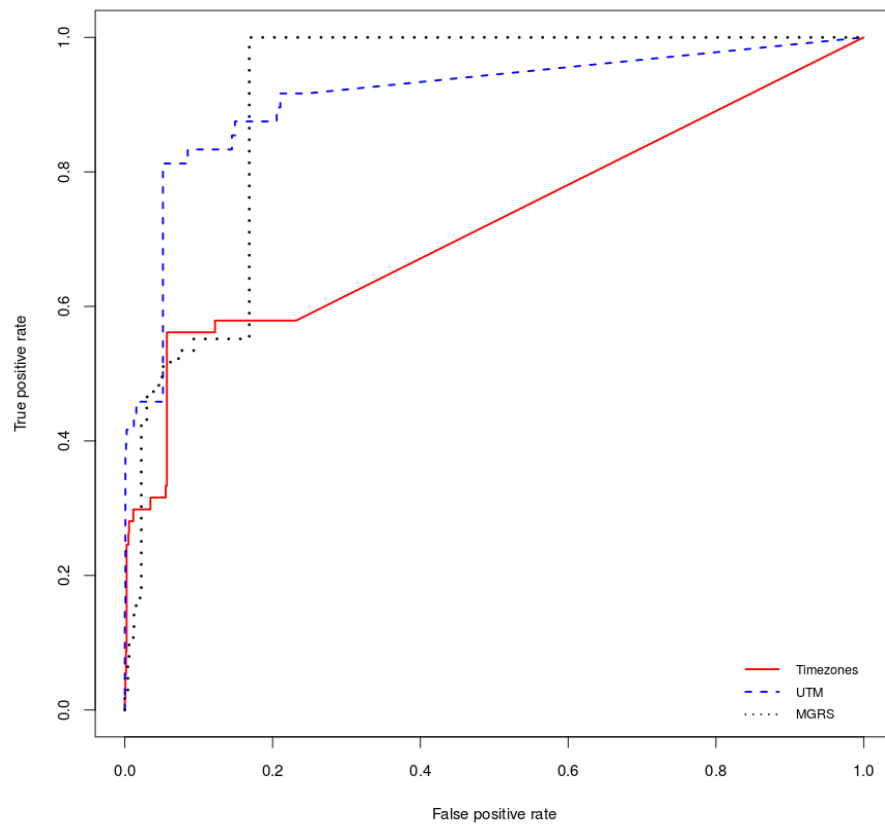
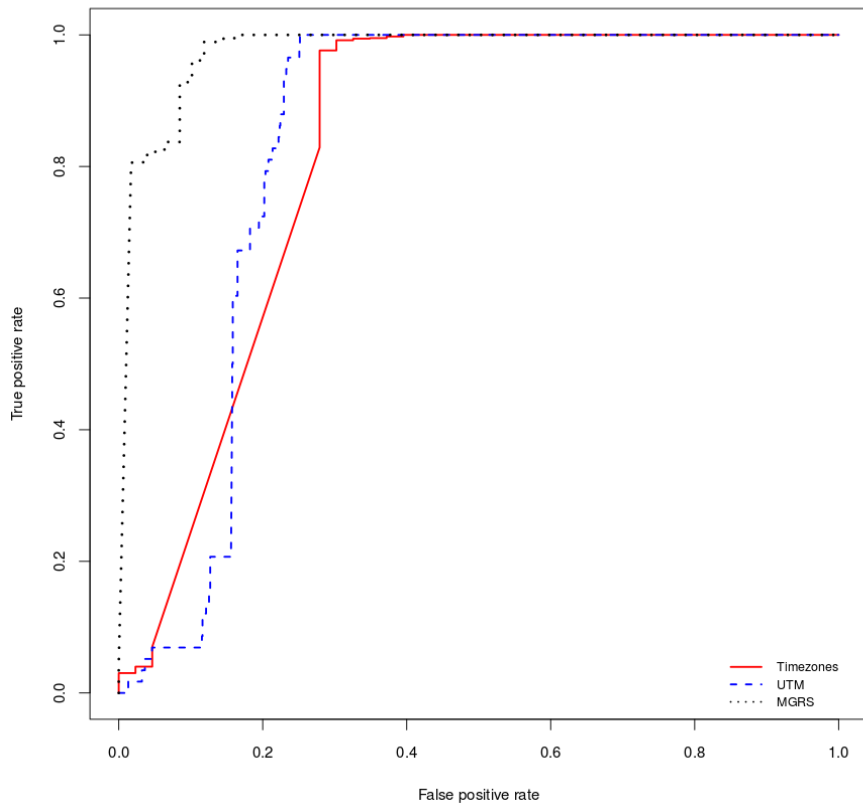
Moran's I: UTM



Geary's C: UTM



Geary's C: MGRS



Classifier Results

Moran Classifier Results

97% Timezones

UTM 95%

95% MGRS

Accuracy

Geary Classifier Results

95% Timezones

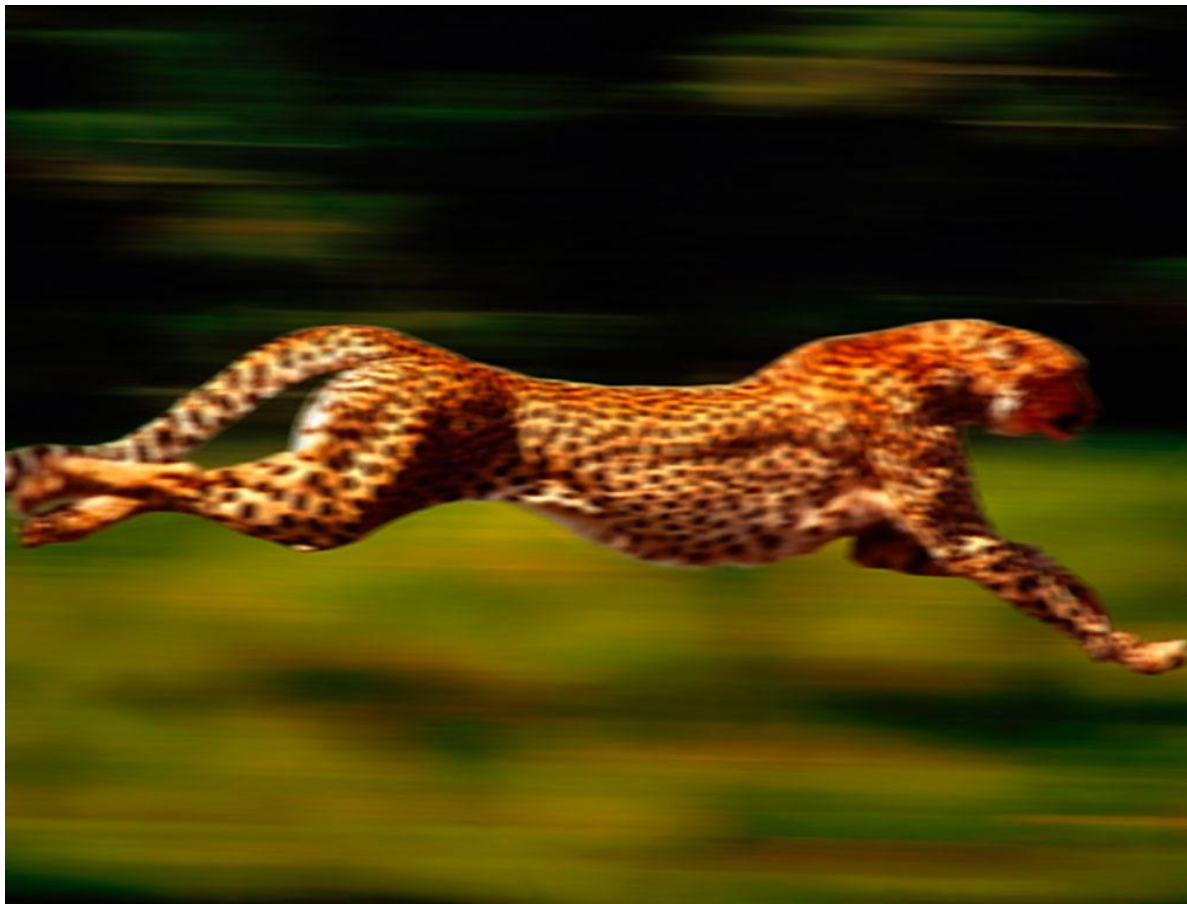
UTM 96%

95% MGRS

Accuracy

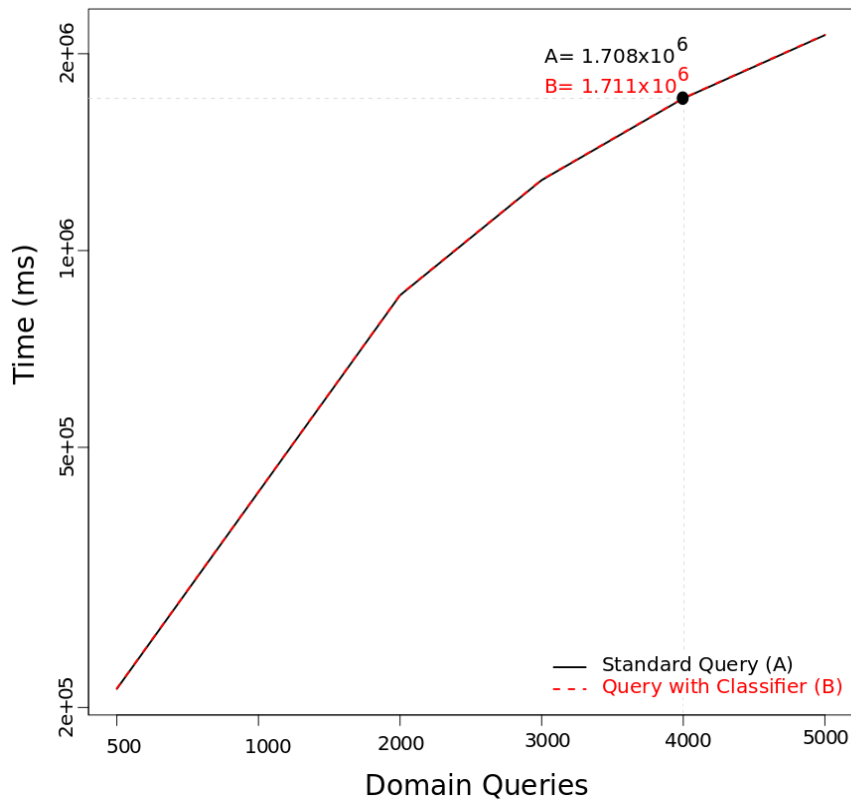
Evaluating Performance

- Determine resource usage
- Impact on normal network performance
- Scalability



Classifier Performance Impact

<http://beyond.customline.com/wp-content/uploads/2012/04/Cheetah-performance.jpg>



Measured Performance

Measured Performance

20,000 domain lookups

Processed in 13 seconds

6.501×10^{-4} seconds per domain

Benefits

Fast

Small

Low maintenance

Scalable

Future Work

- Combine classifiers into stand-alone solution
- Combine detection and blocking
- Increase accuracy of geo-location

Conclusion



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