



# Measuring the IQ of your Threat Intelligence Feeds (#TIQtest)

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# whoami(s)

## Alex Pinto

- Science guy at MLSec Project
- ML trainer
- Network security aficionado
- Tortured by SIEMs as a child
- Hacker Spirit Animal™:  
CAFFEINATED CAPYBARA



([https://secure.flickr.com/photos/kobashi\\_san/](https://secure.flickr.com/photos/kobashi_san/))

## Kyle Maxwell

- Researcher at [REDACTED]
- Math Smuggler
- Recovering Incident Responder
- GPL zealot
- Hacker Spirit Animal™:  
AXIOMATIC ARMADILLO



(<http://www.langorigami.com/art/gallery/gallery.php?tag=mammals&name=armadillo>)

# Agenda

- Threat Intel 102
- Measuring Intelligence
- Data Preparation
- Testing the Data
- Tools:
  - COMBINE
  - TIQ-TEST
- Some parting ideas



(<http://www.savagechickens.com/2008/12/iq-test.html>)

# Threat Intel 102: Capability and Intent

- What are they able to do?
- What are they intending to do?

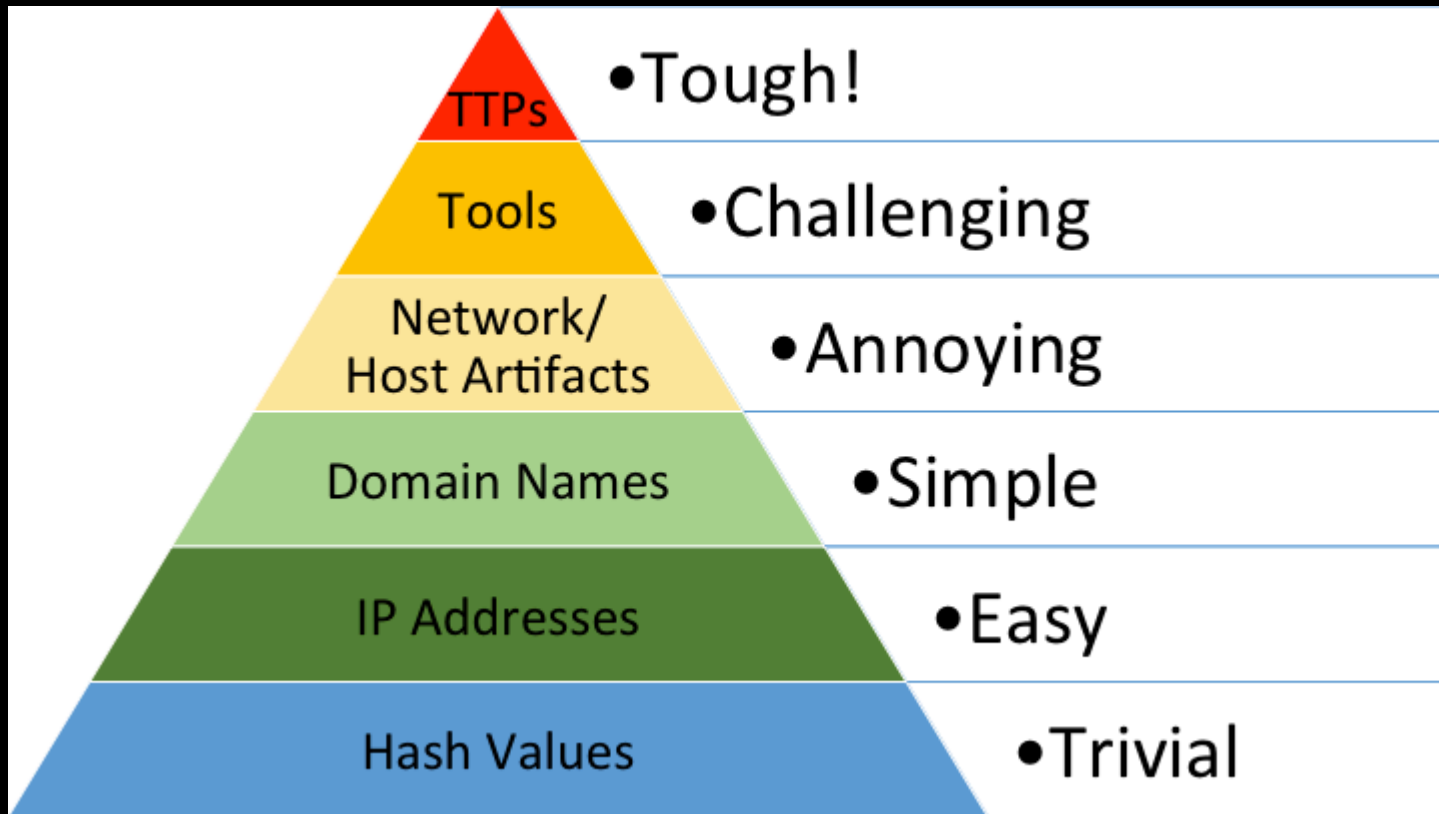


# Threat Intel 102: Cage Matches

- Signatures vs Indicators
- Data vs Intelligence
- Tactical vs Strategic
- Atomic vs Composite



# Threat Intel 102: Pyramid of Pain

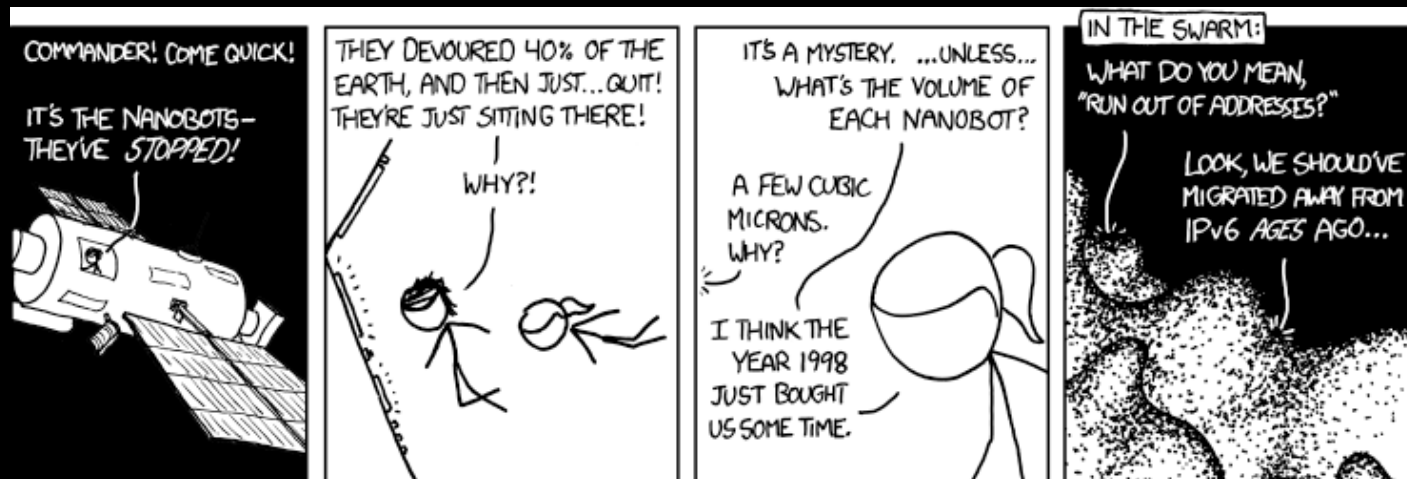


(David Bianco – Pyramid of Pain)

“Simple” and “easy” aren’t always

# What about IP addresses?

- Approximately same value as hostnames (APT vs DGA)
- Finite resource (until IPv6, that is)
  - Managed / controlled by orgs
  - Difficulty / economic incentives / implied “cost”
  - Also, recyclable



*Given IP addresses harvested from  
TI feeds, can we measure how  
much they “help” our defense  
metrics?*

**STAND BACK**



**I'M GOING TO TRY  
SCIENCE**



# Introducing TIQ-TEST

- All these tests are available as R functions at
  - <https://github.com/mlsecproject/tiq-test>
  - Have fun, prove me wrong, suggest stuff
- Tools that implement those tests
- Sample data + R Markdown file
- The excuse to learn a statistical language you were waiting for!

# Data Sources – Types of data

- Extract the “raw” information from indicator feeds
- Both IP addresses and hostnames were extracted

```
outbound.ti = tiq.data.loadTI("raw", "public_outbound", "20140701")
outbound.ti[, list(entity, type, direction, source, date)]
```

```
##           entity type direction      source      date
##  1:      1.224.163.26 IPv4  outbound alienvault 2014-07-01
##  2:      1.242.99.155 IPv4  outbound alienvault 2014-07-01
##  3:        1.85.2.118 IPv4  outbound alienvault 2014-07-01
##  4:        1.93.1.162 IPv4  outbound alienvault 2014-07-01
##  5:        1.93.161.204 IPv4  outbound alienvault 2014-07-01
##  ---
## 16298:      winscoft.com FQDN  outbound      zeus 2014-07-01
## 16299:          wmzbase.ru FQDN  outbound      zeus 2014-07-01
## 16300:      zhabademon.net FQDN  outbound      zeus 2014-07-01
## 16301: zhangleetranding.com FQDN  outbound      zeus 2014-07-01
## 16302:          znatnydom.by FQDN  outbound      zeus 2014-07-01
```

# Data Sources – Feeds Selected

- Data was separated into “inbound” and “outbound”

```
inbound.ti = tiq.data.loadTI("raw", "public_inbound", "20140701")
unique(inbound.ti$source)
```

```
## [1] "alienvault"      "autoshun"        "blocklistde"
## [4] "bruteforceblocker" "charleshaley"   "ciarmy"
## [7] "dragonresearch"  "dshield"        "honeypot"
## [10] "openbl"         "packetmail"     "virbl"
```

```
outbound.ti = tiq.data.loadTI("raw", "public_outbound", "20140701")
unique(outbound.ti$source)
```

```
## [1] "alienvault"      "botscout"        "malcode"
## [4] "malcode_zones"  "malwaredomainlist" "malwaredomains"
## [7] "malwaregroup"   "palevotracker"   "spyeye"
## [10] "zeus"
```

# Data Preparation and Cleansing

- Convert the hostname data to IP addresses:
  - Active IP addresses for the respective date (“A” query)
  - Passive DNS from Farsight Security (DNSDB) 👍
- We removed non-public IPs from the dataset (RFC1918)
  - Yeah, we know it is a “parking technique”



(<https://xkcd.com/742/>)

# Data Preparation and Cleansing

- For each IP record (including the ones from hostnames):
  - Add asnumber and asname (from MaxMind ASN DB)
  - Add country (from MaxMind GeoLite DB)
  - Add rhost (again from DNSDB) – most popular “PTR”
- The experiments will be around ASNs and Geolocation

# Data Preparation and Cleansing

- However, we will NOT be using maps. Just let it go.



# Data Preparation and Cleansing

- Small enriched sample:

```
enrich.ti = tiq.data.loadTI("enriched", "public_outbound", "20140710")
enrich.ti = enrich.ti[, notes := NULL]
enrich.ti[c(2,22264, 22266)]
```

```
##           entity type direction      source      date asnumber
## 1:  1.224.163.26 IPv4  outbound alienvault 2014-07-10      9318
## 2: 95.181.178.177 IPv4  outbound      zeus 2014-07-10      57311
## 3: 98.131.185.136 IPv4  outbound      zeus 2014-07-10      32392
##
##                   asname country
## 1:                   Hanaro Telecom Inc.      KR
## 2: FOP ILIUSHENKO VOLODYMYR OLEXANDROVUCH      GB
## 3:                   Ecommerce Corporation      US
##
##                   host      rhost
## 1:                   NA      NA
## 2:                   newdomaininfo.ru host178-177.neohost.net
## 3: projects.globaltronics.net      NA
```

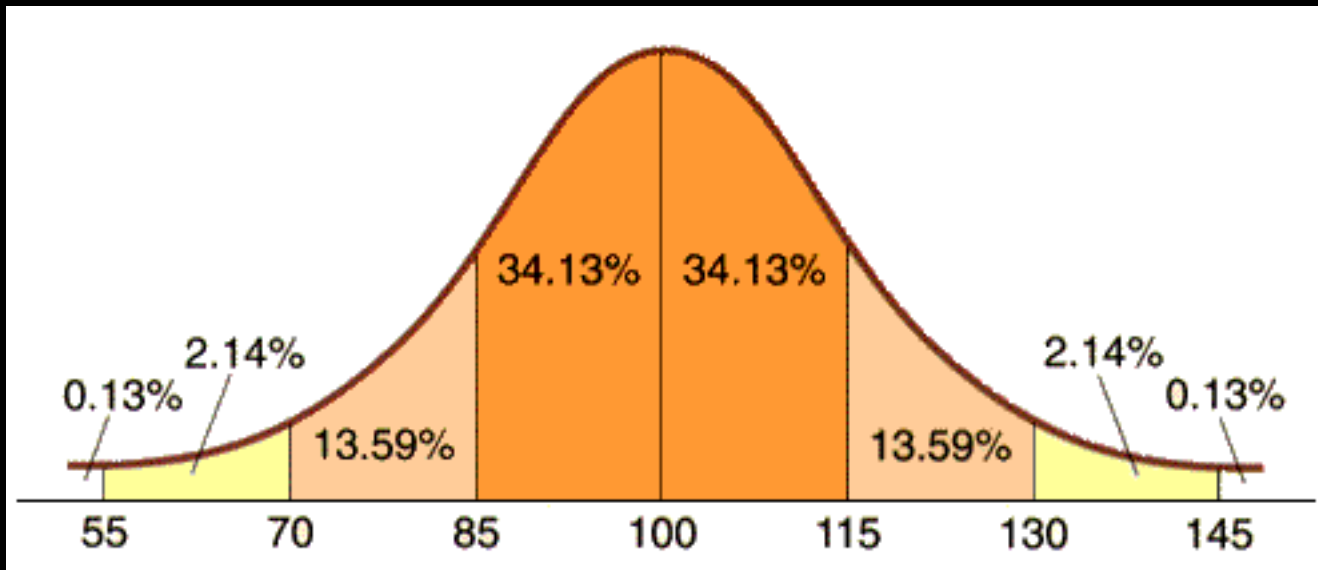
# Testing the Data

- Let's generate some interesting metrics:
  - NOVELTY – How often do they update themselves?
  - OVERLAP – How do they compare to what you got?
  - POPULATION – what is in them anyway?
- Population is tricky:
  - Could mean the entire world (all IPv4 space)
  - Should ideally mean YOUR world



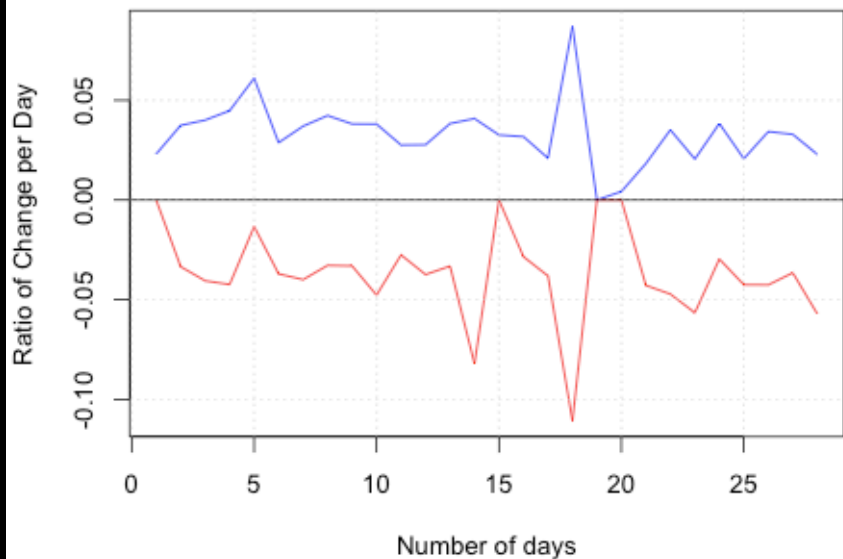
# But WHAT IS THE IQ?!??1?

- We will withhold judgment
- The best data composition is the best one for you
- We will do our best to explain results so you can decide.
- Maybe on further (or more private) research...

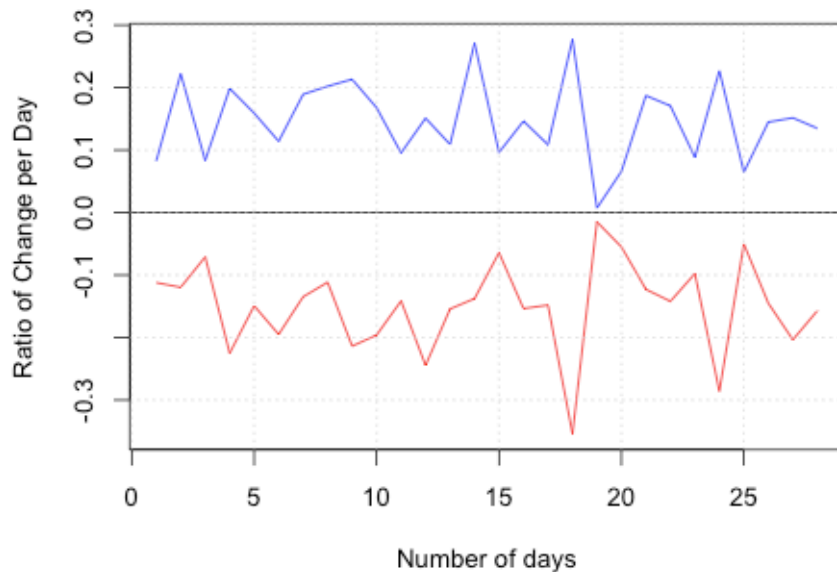


*Novelty Test* – measuring added  
and dropped indicators

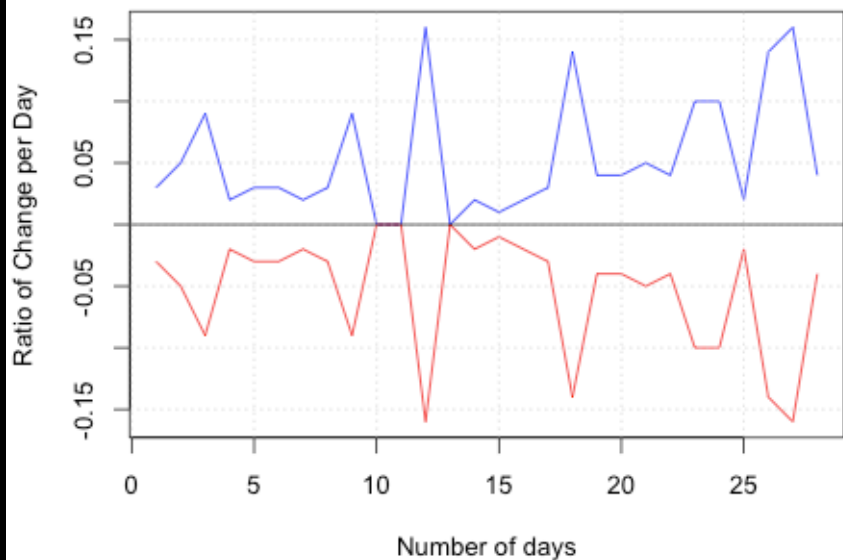
**Source Name: alienvault**  
**Avg. Size: 180870**



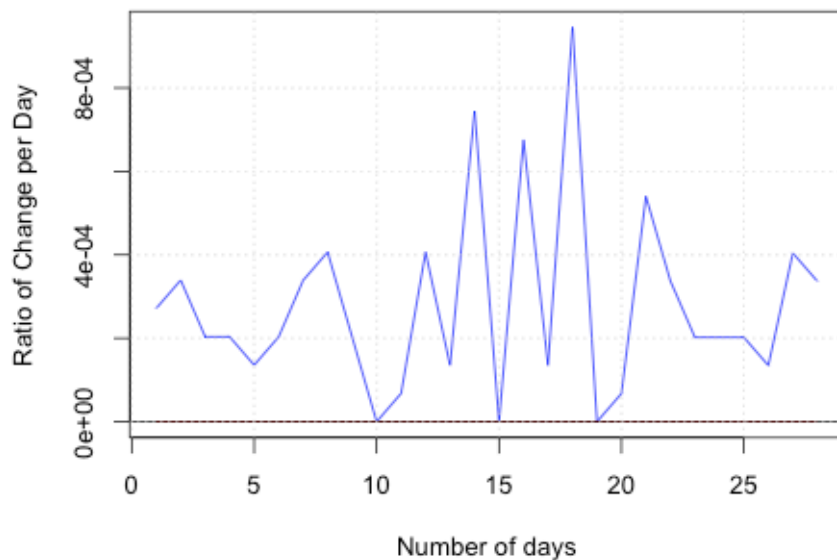
**Source Name: blacklistde**  
**Avg. Size: 20193**



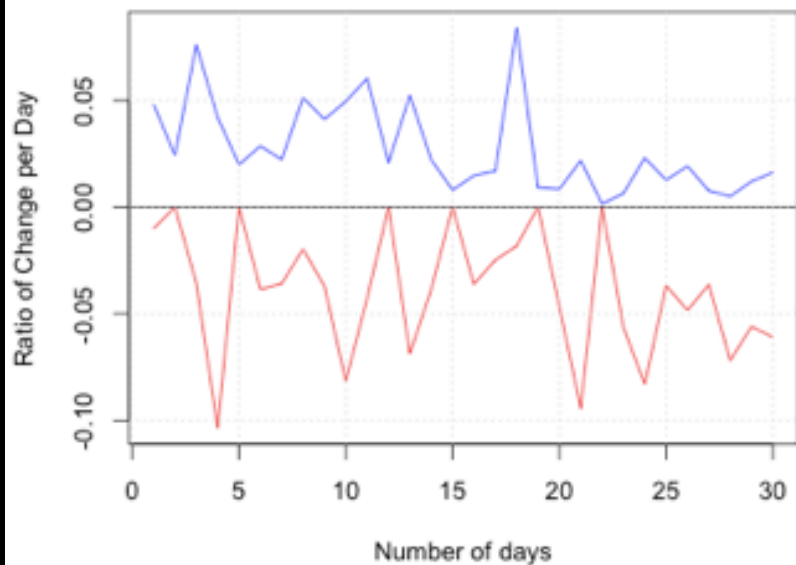
**Source Name: dshield**  
**Avg. Size: 100**



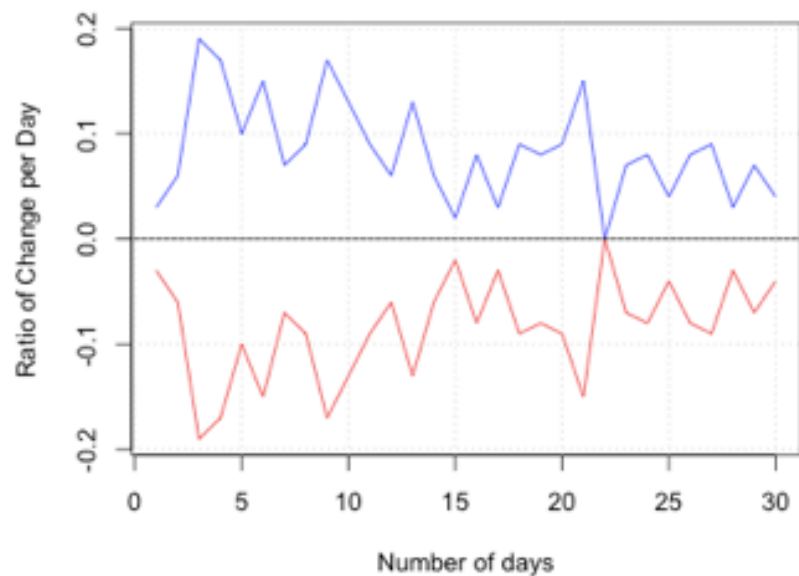
**Source Name: charleshaley**  
**Avg. Size: 14786**



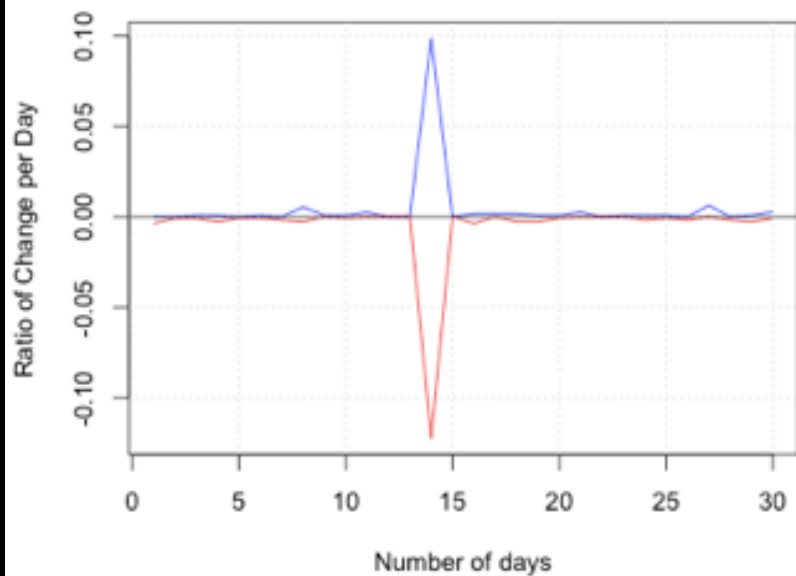
**Source Name: alienvault**  
**Avg. Size: 4573**



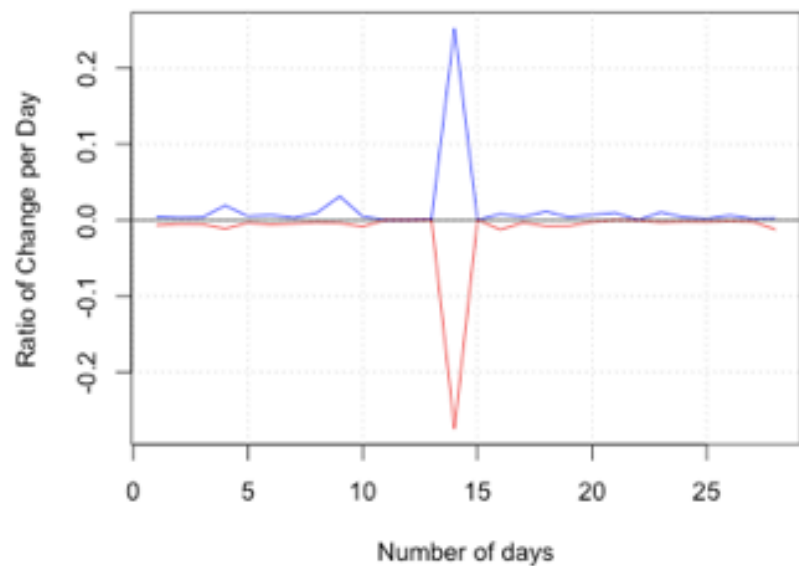
**Source Name: malwaregroup**  
**Avg. Size: 100**



**Source Name: malwaredomainlist**  
**Avg. Size: 1015**



**Source Name: malwaredomains**  
**Avg. Size: 7291**

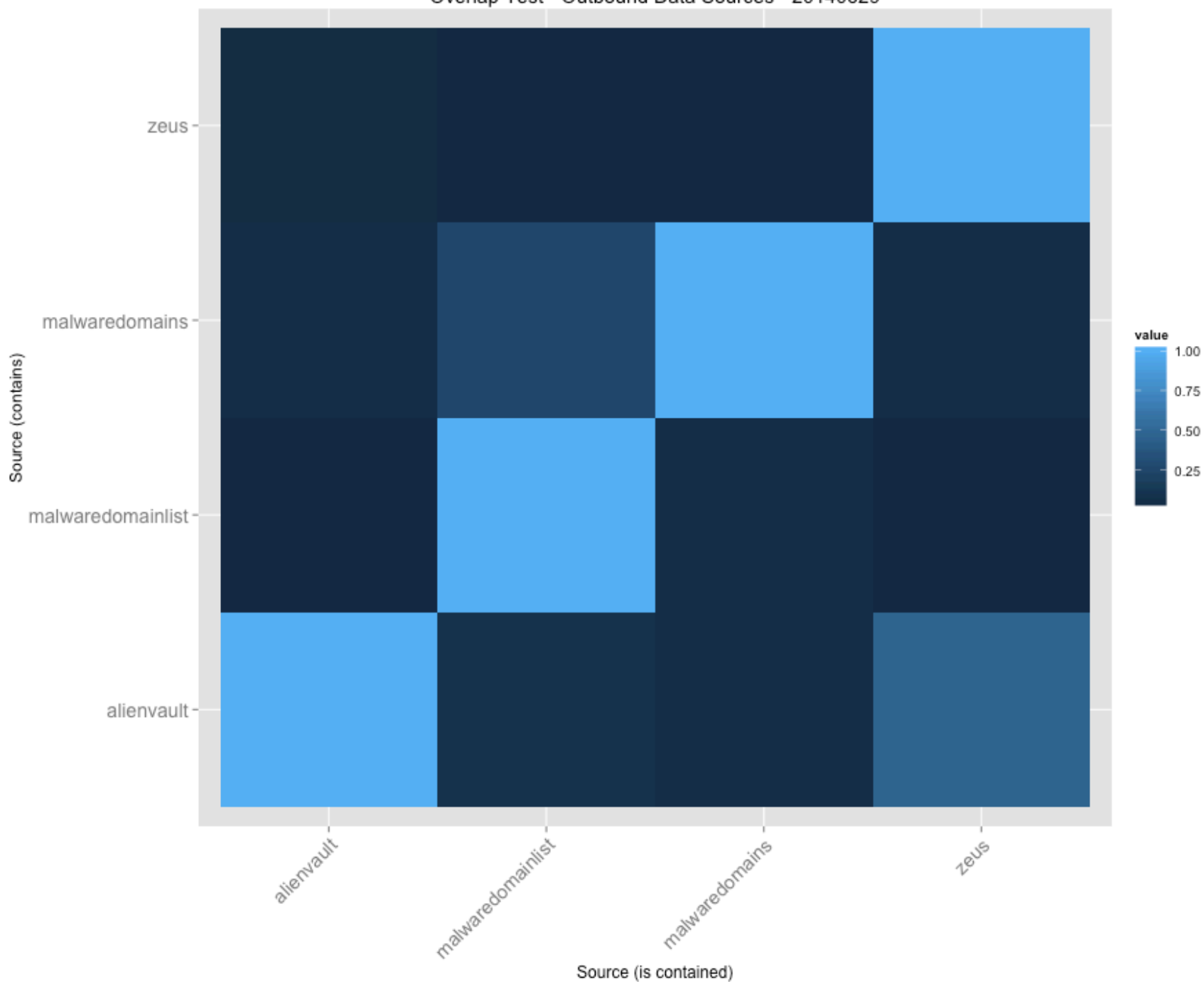


*Overlap Test* – More data is better,  
but make sure it is not the same  
data





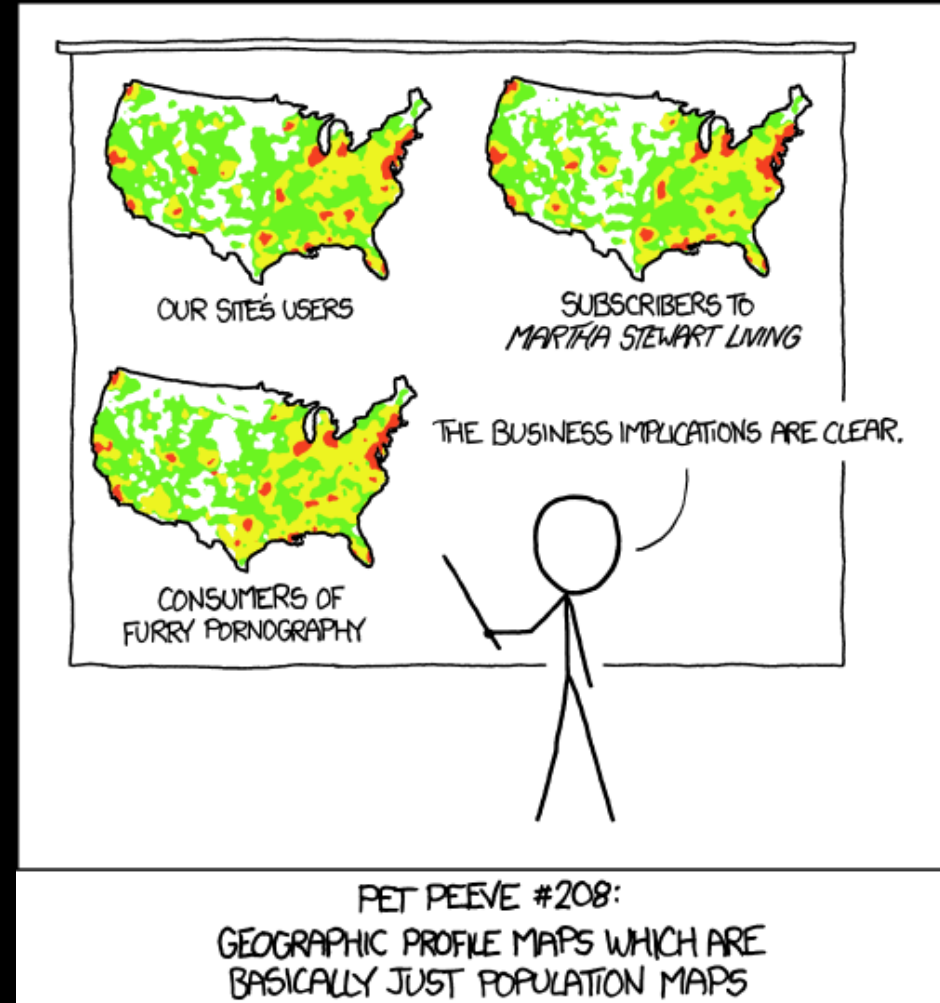
Overlap Test - Outbound Data Sources - 20140629

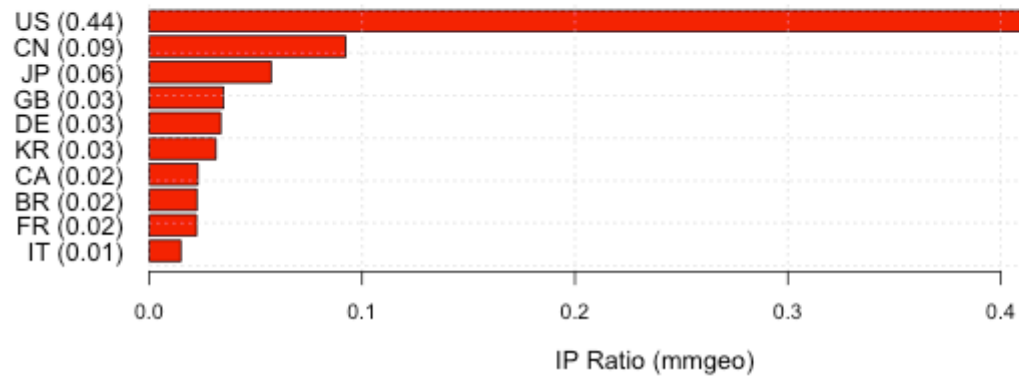
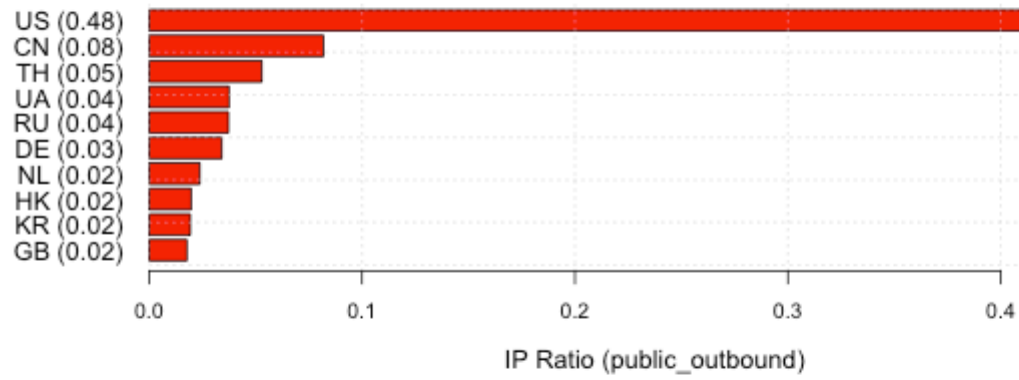
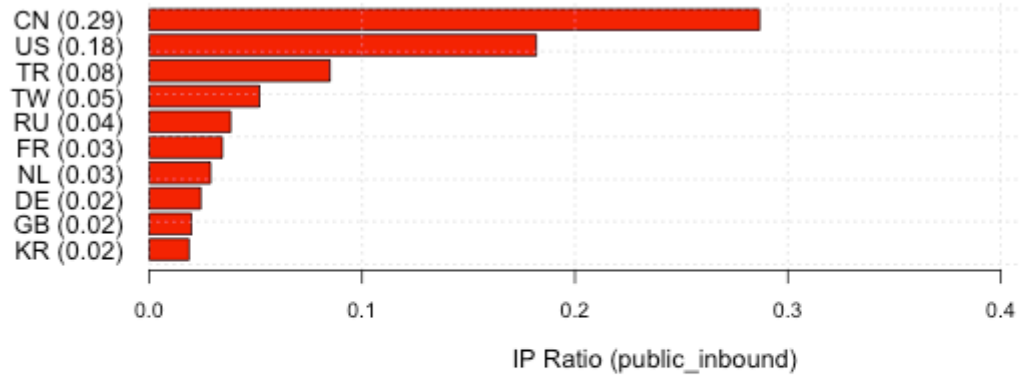


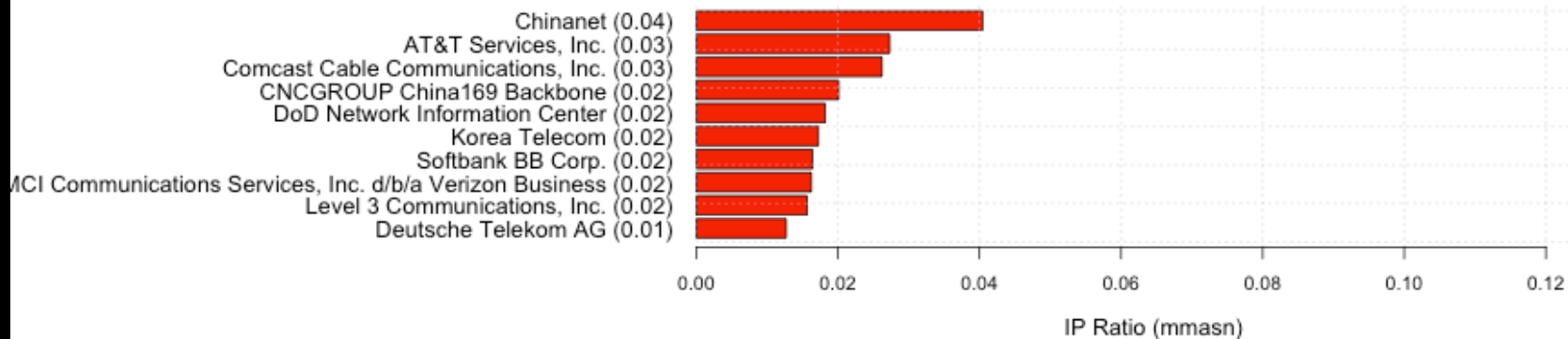
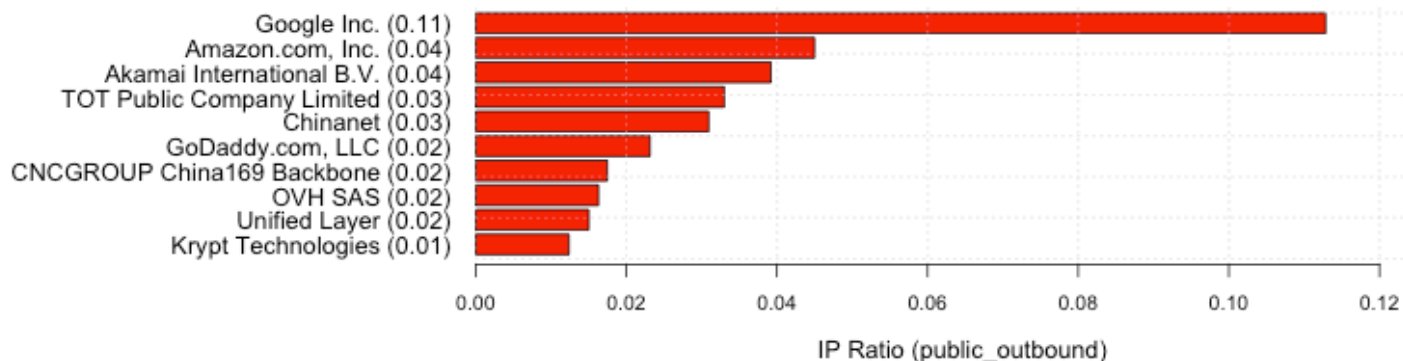
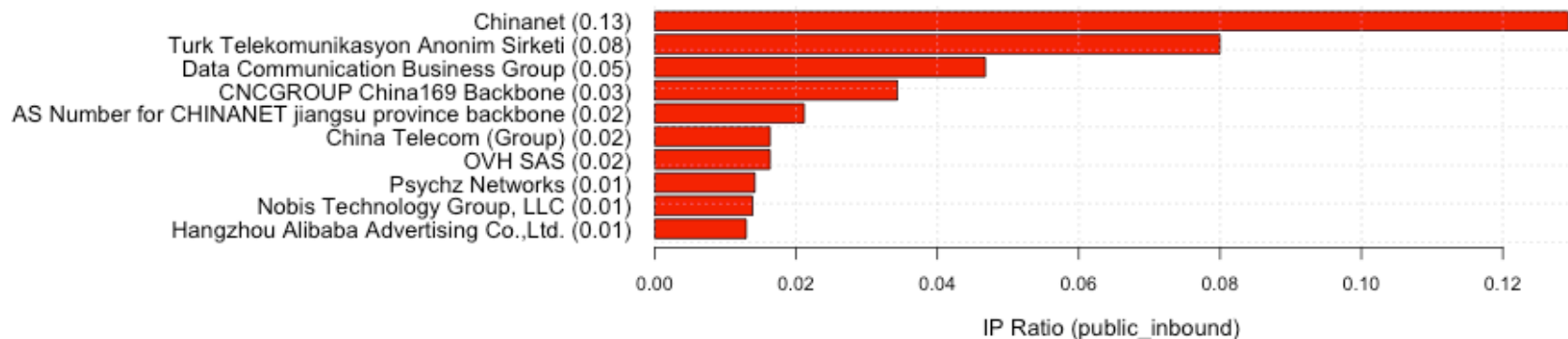


# Population Test

- Let us use the ASN and GeoIP databases that we used to enrich our data as a reference of the “true” population.
- But, but, human beings are unpredictable! We will never be able to forecast this!







# Can we get a better look?

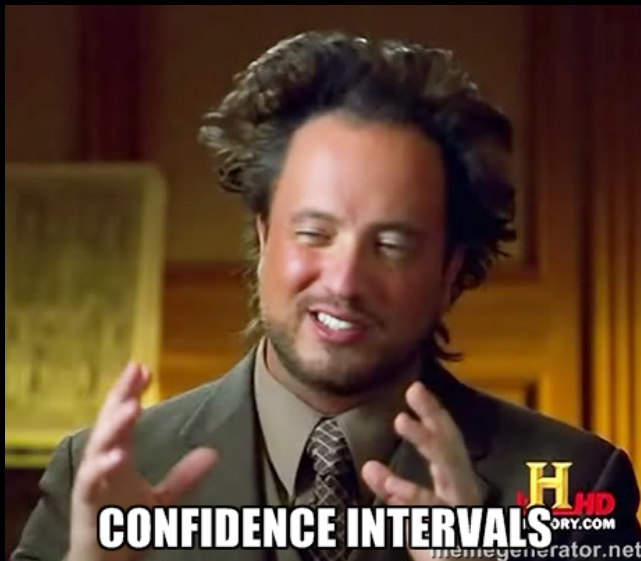
- Don't like squinting either
- Statistical inference-based comparison models (hypothesis testing)
  - Exact binomial tests (when we have the “true” pop)
  - Chi-squared proportion tests (similar to independence tests)

$$\left(\sqrt{(-things)}\right)^2$$

**THINGS JUST GOT REAL.**

# Can we get a better look?

- We can better estimate, with confidence intervals, our measures of error.
- Also, p-values! (with apologies to Alex Hutton)
- We promise to be very conservative in using them.



## Statistics Professors HATE Him!



*Doctor's discovery revealed the secret to learning any problem with just 10 training samples. Watch this shocking video and learn how rapidly you can find a solution to your learning problems using this one sneaky kernel trick! Free from overfitting!*

<http://www.oneweirdkerneltrick.com>

```
tests = tiq.test.populationInference(complete.pop$mmgeo,
                                     outbound.pop$public_outbound, "country",
                                     exact = TRUE, top=10)

# Whose proportion is bigger than it should be?
tests[p.value < 0.05/10 & conf.int.end > 0][order(conf.int.end, decreasing=T)]
```

```
##      country conf.int.start conf.int.end  p.value
## 1:      TH      0.047044      0.05415  0.000e+00
## 2:      US      0.025335      0.04111  9.406e-17
## 3:      UA      0.031252      0.03730  0.000e+00
## 4:      RU      0.021363      0.02739  1.198e-105
## 5:      HK      0.014238      0.01868  2.412e-128
## 6:      NL      0.007818      0.01268  4.091e-23
```

```
# Whose is smaller?
tests[p.value < 0.05/10 & conf.int.start < 0][order(conf.int.start, decreasing=F)]
```

```
##      country conf.int.start conf.int.end  p.value
## 1:      GB      -0.01926      -0.015040  8.988e-38
## 2:      CN      -0.01469      -0.005996  5.356e-06
## 3:      KR      -0.01411      -0.009713  3.809e-20
```

```
# And whose is the same? ~\_(\ツ)\_/~
tests[p.value > 0.05/10]
```

```
##      country conf.int.start conf.int.end p.value
## 1:      DE      -0.002366      0.003411  0.7553
```

# Hacker Spirit Animal™ Guide

- US – Eagle
- CA – Moose
- FR – Frog
- GB – Bulldog
- AU – Koala
- BR – Capybara / Toucan
- Texas – Armadillo



- Disclaimer: we do not endorse Geolocation-based attribution

```

outbound.pop2 = tiq.test.extractPopulationFromTI("public_outbound", "country",
                                                date = "20140712",
                                                select.sources=NULL,
                                                split.ti=FALSE)
tests = tiq.test.populationInference(outbound.pop$public_outbound,
                                     outbound.pop2$public_outbound, "country",
                                     exact = F, top=10)

# Whose proportion is bigger than it should be?
tests[p.value < 0.05/10 & conf.int.end > 0][order(conf.int.end, decreasing=T)]

```

```

##   country conf.int.start conf.int.end  p.value
## 1:      TH      0.008892   0.01949 1.312e-07

```

```

# Whose is smaller?
tests[p.value < 0.05/10 & conf.int.start < 0][order(conf.int.start, decreasing=F)]

```

```

## Empty data.table (0 rows) of 4 cols: country,conf.int.start,conf.int.end,p.value

```

```

# And whose is the same? ~\_(ツ)_/~
tests[p.value > 0.05/10]

```

```

##   country conf.int.start conf.int.end p.value
## 1:      CN     -0.008903   0.003230 0.3652
## 2:      DE     -0.005626   0.002421 0.4461
## 3:      GB     -0.003826   0.002055 0.5753
## 4:      HK     -0.004286   0.001887 0.4612
## 5:      KR     -0.004004   0.002129 0.5682
## 6:      NL     -0.004471   0.002308 0.5484
## 7:      RU     -0.005538   0.002877 0.5489
## 8:      UA     -0.005500   0.002947 0.5675
## 9:      US     -0.009315   0.012858 0.7613

```



```
complete.pop = tiq.data.loadPopulation("mmasn", c("asnumber", "asname"))
tests = tiq.test.populationInference(complete.pop$mmasn,
                                     outbound.pop$public_outbound,
                                     c("asname", "asnumber"),
                                     exact = TRUE, top=10)

# Whose proportion is bigger than it should be?
tests[p.value < 0.05/10 & conf.int.end > 0][order(conf.int.end, decreasing=T)]
```

```
##           asname conf.int.start conf.int.end  p.value
## 1:         Google Inc.      0.10756      0.11758 0.000e+00
## 2:         Amazon.com, Inc.  0.04015      0.04673 0.000e+00
## 3: Akamai International B.V.  0.03534      0.04151 0.000e+00
## 4: TOT Public Company Limited 0.03019      0.03588 0.000e+00
## 5:         GoDaddy.com, LLC  0.02052      0.02532 0.000e+00
## 6:         OVH SAS          0.01397      0.01802 1.046e-302
## 7:         Unified Layer     0.01292      0.01682 7.411e-323
## 8:         Krypt Technologies 0.01049      0.01404 8.007e-265
```

```
# Whose is smaller?
tests[p.value < 0.05/10 & conf.int.start < 0][order(conf.int.start, decreasing=F)]
```

```
##      asname conf.int.start conf.int.end  p.value
## 1: Chinanet      -0.01216    -0.006648 4.903e-10
```

```
# And whose is the same? ~\_(ツ)_/~
tests[p.value > 0.05/10]
```

```
##           asname conf.int.start conf.int.end p.value
## 1: CNCGROUP China169 Backbone  -0.004651  -0.0004625 0.01762
```

```
outbound.ti = tiq.data.loadTI("enriched", "public_outbound", "20140711")
outbound.ti[asname %like% "Google", list(entity, type, source, asname, host)]
```

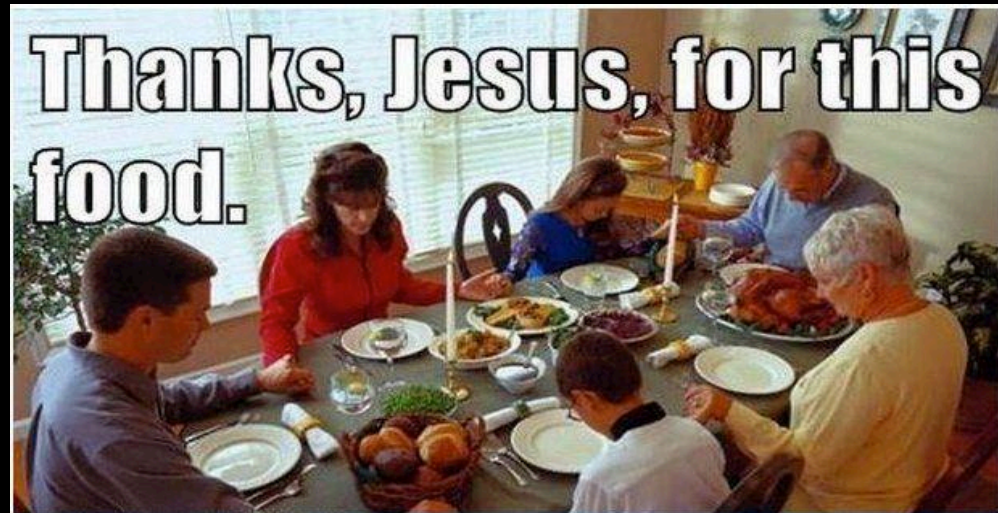
```
##           entity type           source      asname           host
##  1: 74.125.228.43 IPv4      malcode Google Inc.           NA
##  2: 74.125.228.75 IPv4      malcode Google Inc.           NA
##  3: 173.194.115.16 IPv4 malcode_zones Google Inc. googleapis.com
##  4: 173.194.115.17 IPv4 malcode_zones Google Inc. googleapis.com
##  5: 173.194.115.18 IPv4 malcode_zones Google Inc. googleapis.com
## ---
## 1964:      8.8.8.8 IPv4 malwaredomains Google Inc. revlister.com
## 1965:      8.8.8.8 IPv4 malwaredomains Google Inc.  statalyze.net
## 1966:      8.8.8.8 IPv4 malwaredomains Google Inc.  statisticbench.net
## 1967:      8.8.8.8 IPv4 malwaredomains Google Inc.  webdestinct.net
## 1968:      8.8.8.8 IPv4      spyeye Google Inc. futuretelefonica.com
```

```
outbound.ti[asname %like% "Google" & entity != "8.8.8.8" & entity != "1.1.1.1",
list(entity, type, source, asname, host)]
```

```
##           entity type           source      asname           host
##  1: 74.125.228.43 IPv4      malcode Google Inc.           NA
##  2: 74.125.228.75 IPv4      malcode Google Inc.           NA
##  3: 173.194.115.16 IPv4 malcode_zones Google Inc. googleapis.com
##  4: 173.194.115.17 IPv4 malcode_zones Google Inc. googleapis.com
##  5: 173.194.115.18 IPv4 malcode_zones Google Inc. googleapis.com
## ---
## 1950: 74.125.70.101 IPv4 malwaredomains Google Inc. chrome.google.com
## 1951: 74.125.70.102 IPv4 malwaredomains Google Inc. chrome.google.com
## 1952: 74.125.70.113 IPv4 malwaredomains Google Inc. chrome.google.com
## 1953: 74.125.70.138 IPv4 malwaredomains Google Inc. chrome.google.com
## 1954: 74.125.70.139 IPv4 malwaredomains Google Inc. chrome.google.com
```

# Introducing COMBINE

- Harvesting feeds takes some work.
- Most of us let somebody else do it without thinking about what it actually takes.



# Introducing COMBINE



<https://github.com/mlsecproject/combine>

# Introducing COMBINE

- Components:
  1. **Reaper** gathers the threat data directly from feeds.
  2. **Thresher** normalizes it into a simplistic data model.
  3. **Winnower** optionally performs basic validation or enrichment.
  4. **Baler** transforms the data into CybOX, CSV, JSON, and CIM. (Only CSV and JSON work right now). Could also write others fairly easily. (nudge nudge, wink wink)

# Introducing COMBINE

- Always trying to feed it more. Lots of possibilities, including your own data sources.
- We clearly do NOT endorse any included feeds.



# Introducing COMBINE

- Enrichments - think metadata.
  - AS, geolocation
  - DNS resolutions courtesy of Farsight DNSDB
  - Ask them for an API key to test it, tell them Alex Pinto sent you ;)



# MLSec Project

- Both projects have been released as GPLv3 by MLSec Project
- Will replace the internal versions we have on the main code
- Looking for participants and data sharing agreements
- Liked TIQ-TEST? We can benchmark your private feeds using these and other techniques
- Visit <https://www.mlsecproject.org> , message @MLSecProject or just e-mail me.





# Take Aways

- Analyze your data.
- Extract value from it!
- Try before you buy! Different test results mean different things to different orgs.
- Use the tools! Suggest new tests!
- Share data with us! We take good care of it, make sure it gets proper exercise.

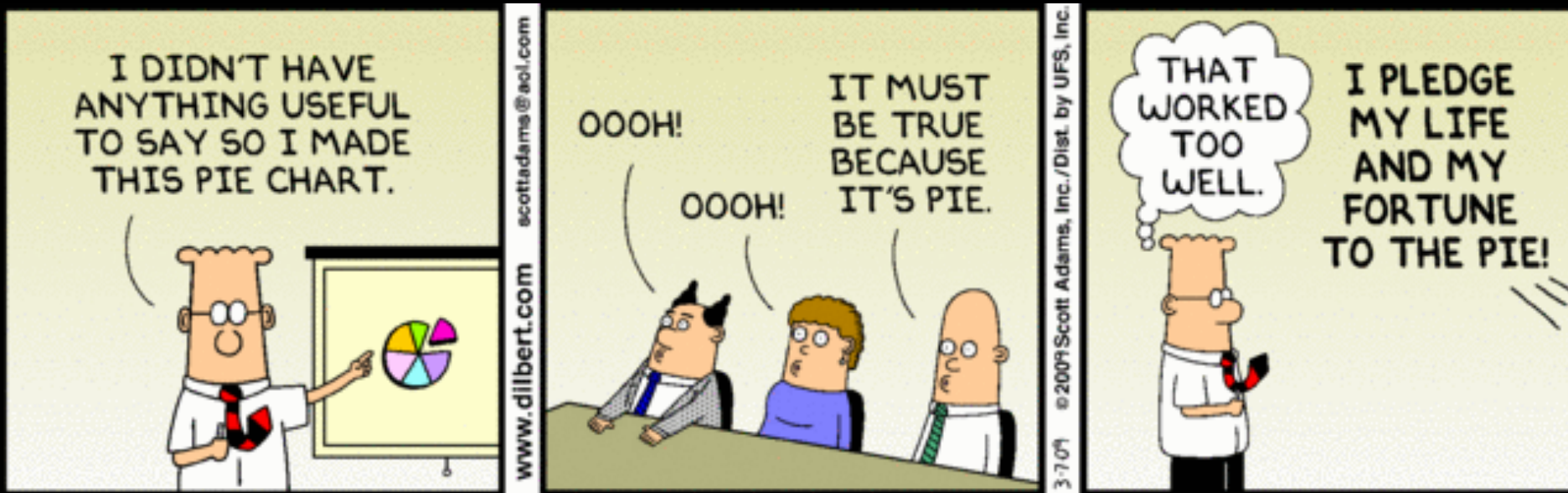


# Thanks!

- Q&A?
- Feedback!

Alex Pinto  
@alexcpsec  
@MLSecProject

Kyle Maxwell  
@kylemaxwell



"The measure of intelligence is the ability to change."

- Albert Einstein