

SNRG

Security and Networks
Research Group

DNS tricks and other nifty things

By: Etienne Stalmans

Supervisor: Mr Barry Irwin



Bright Ideas[®]
Projects 39



RHODES UNIVERSITY
Where leaders learn

Why DNS?

- It is everywhere
- Used by most (read: all) programs requiring internet connection
- This includes Malware
- Modern botnet structures rely on DNS

Botnets and DNS

- Fast-Flux
 - Avoid detection
 - Prevent shut down of Command and Control servers
 - Increase robustness
- If bots use DNS, can we use DNS to detect the bots?

Botnet structure

- Many hosts
- Receive instructions from C&C servers
- C&C servers in different locations
- Botnet controllers have limited control over availability and performance of C&C servers
- C&C servers constantly shifting to prevent botnet shutdown

Fast-flux DNS Query

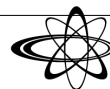
;; QUESTION SECTION:

;champiogogo.ru. IN A

;; ANSWER SECTION:

champiogogo.ru.	300	IN	A	60.13.74.23
champiogogo.ru.	300	IN	A	62.42.100.212
champiogogo.ru.	300	IN	A	148.217.94.55
champiogogo.ru.	300	IN	A	212.69.189.125
champiogogo.ru.	300	IN	A	217.217.199.129

IP Address	Net block	ASN	Country
60.13.74.23	60.13.64.0/18	4837	CN
62.42.100.212	62.42.0.0/16	6739	ES
148.217.94.55	148.217.0.0/16	6503	MX
212.69.189.125	212.69.160.0/19	8218	DE
217.217.199.129	217.216.0.0/15	6739	ES



Features

- Short TTL
- Multiple A records, different IP ranges
- Multiple Autonomous System Numbers (ASNs)
- Name-servers in different network ranges

Malicious or not?

- Manual inspection
 - Why when it can be automated?
- Heuristics
 - Need kept up to date
- C5.0 decision tree
 - Same as heuristics
- Naïve Bayesian classifier
 - Evolve along with botnets

Results

Domain	Safe Score	Malicious Score	Classification
gingerbucksea.com	0.005304578	0.3550235	Fast-flux
pearlrumor.ru	3.059976e-14	7.490562e-13	Fast-flux
wordpress.com	1.536894e-08	4.250896e-10	Legitimate
champiogogo.ru	3.395984e-09	1.723838e-06	Fast-flux
yahoo.com	1.940412e-15	1.509179e-69	Legitimate (CDN)

- Detects fast-flux domains missed by C5.0 classifier and rule based classifier
- Better distinguishing between Content Distribution Networks and fast-flux
- Fast

Extra goodies

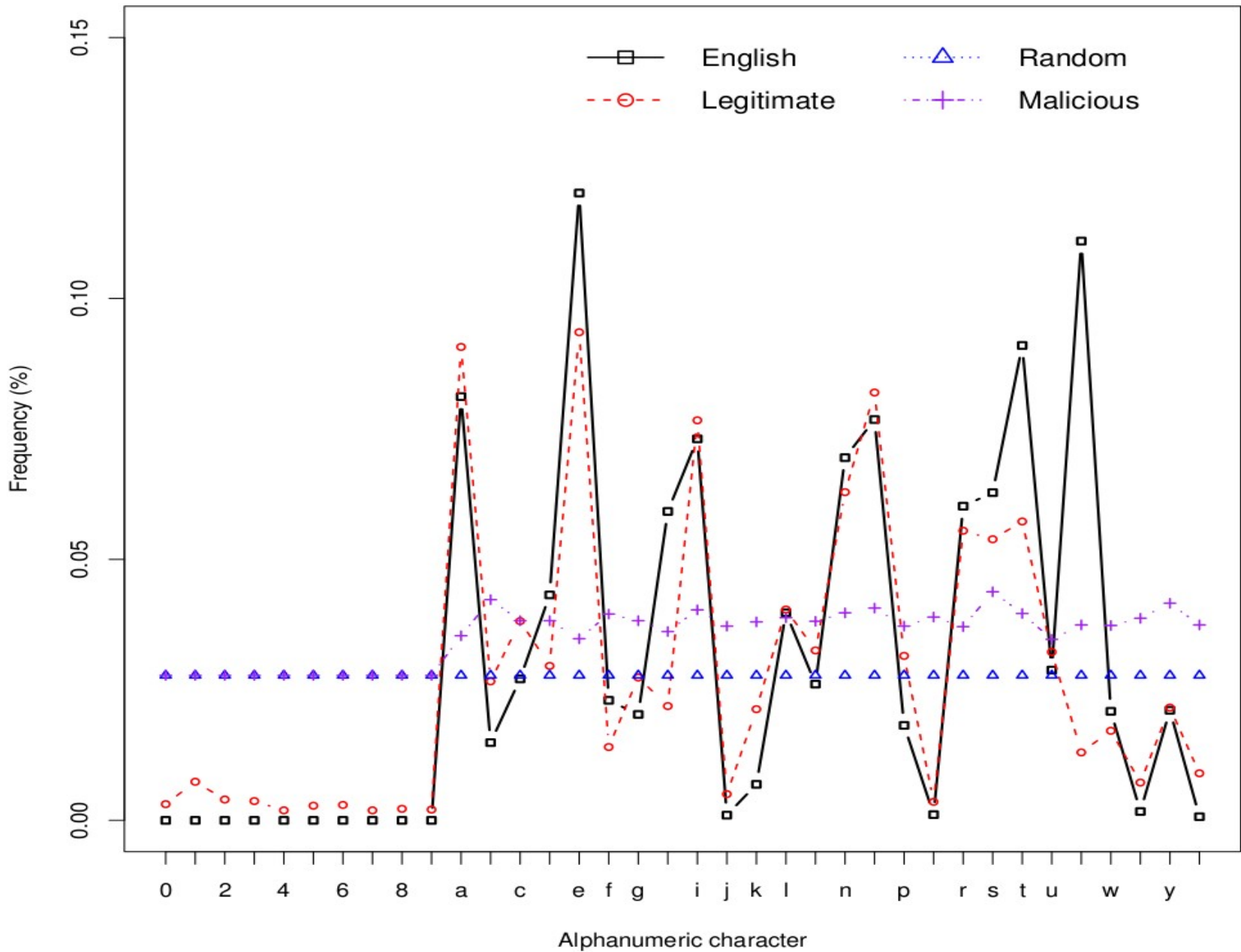
- Detection of “Domain fluxing” using stats
- Conficker, Kraken, Torpig all use algorithmically generated domain names for C&C servers
- Conficker-C generates 50 thousand names a day
- Not possible to block, shut down and pre-register so many domains

But how?

- Site name: www.facebook.com
- Malicious domain: bbhkxkjh.com.fj
- Easy to tell them apart, well as a human, yes.
- Computer needs to “learn”
- You have got to love stats...



Frequency distributions



Classifiers

- Naïve Bayesian Classifier
- Bayesian Classifier
- Probability distribution
- Variation distance

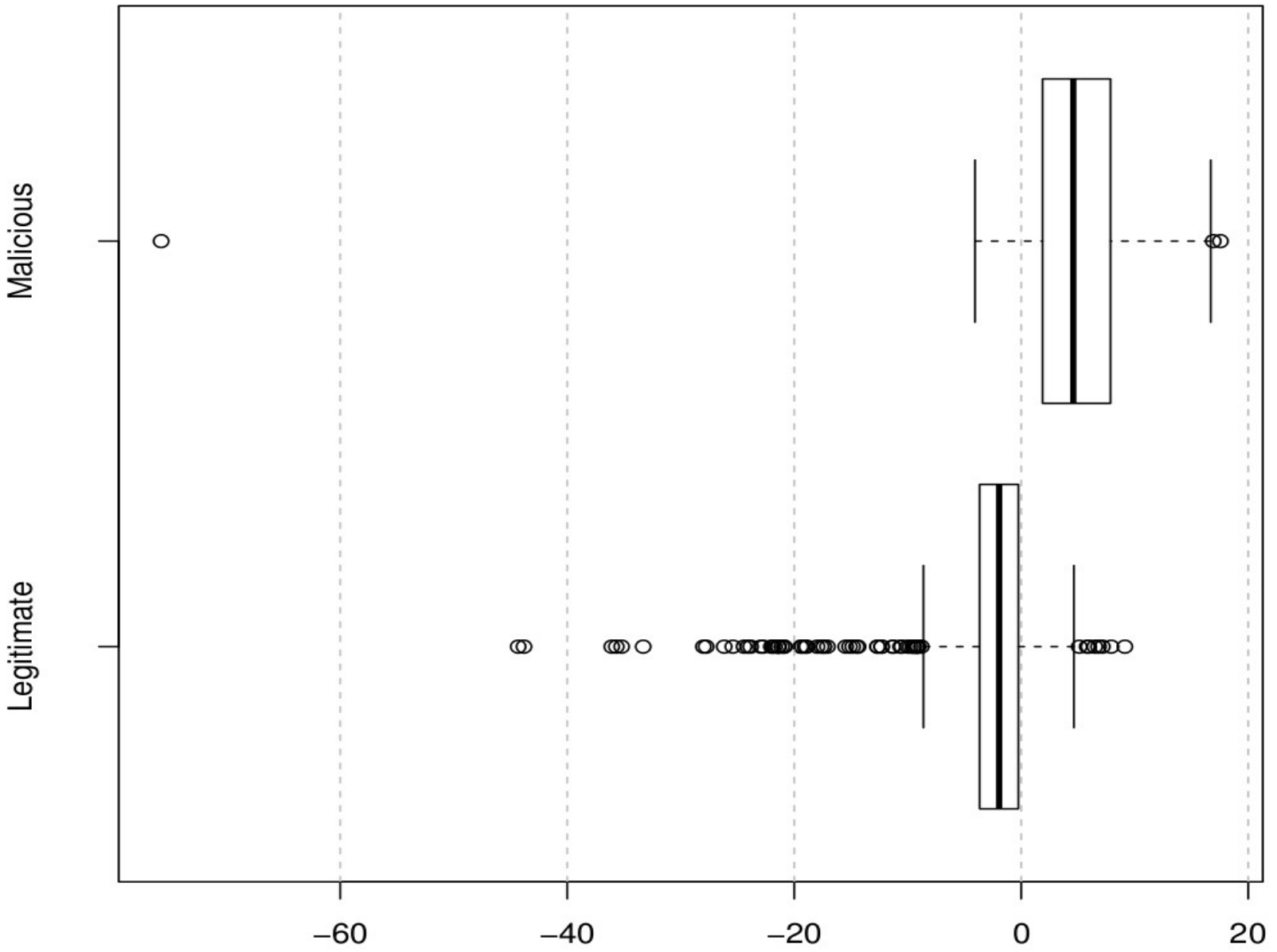
Classifier	Accuracy	TPR	FPR
Naive Bayesian	86.5%	82%	8.3%
Variation	82.4%	80%	17%
Probability	84.3%	86%	17%
Bayesian	85%	81.3%	11%

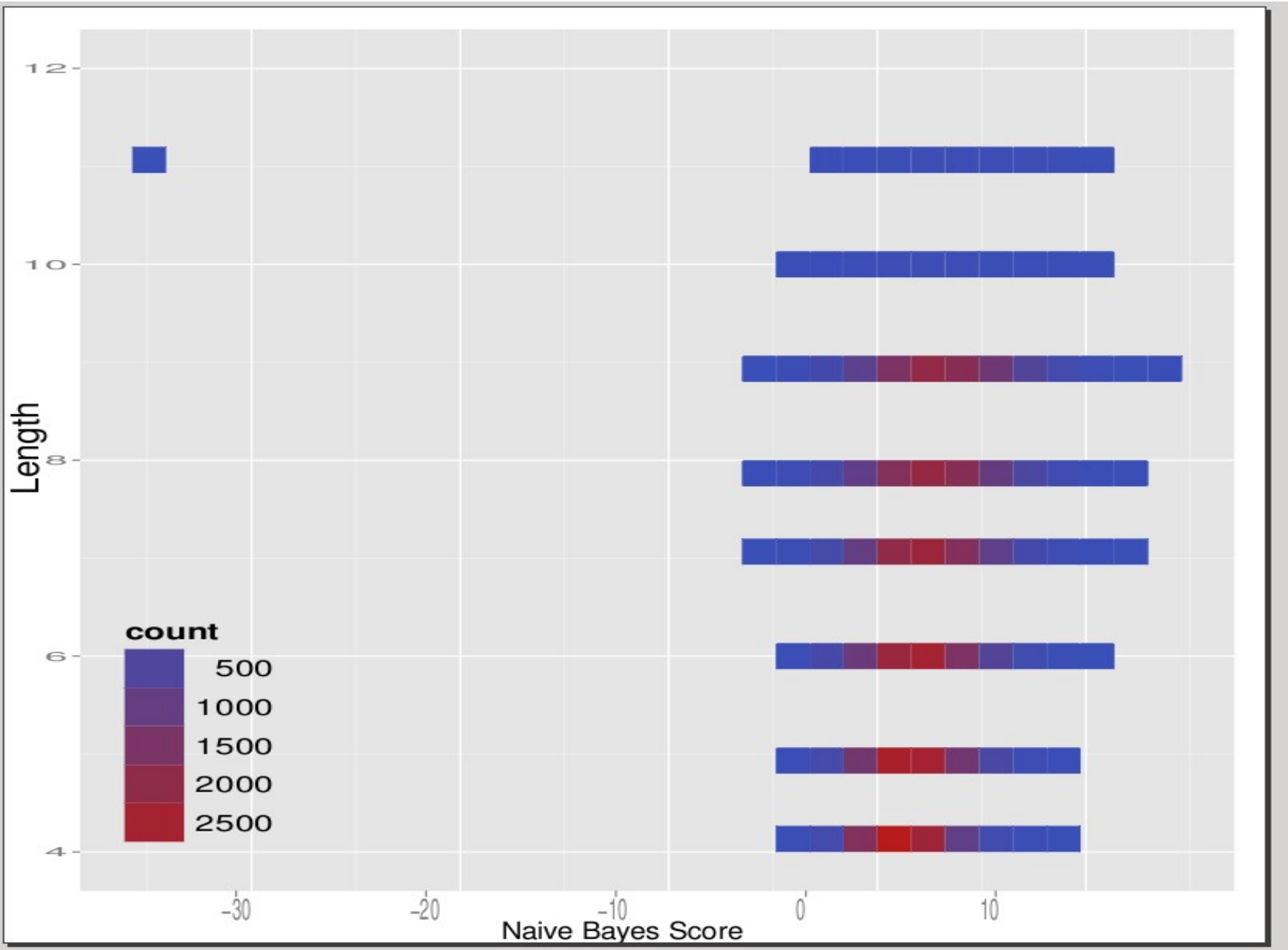
TPR: True Positive Rate

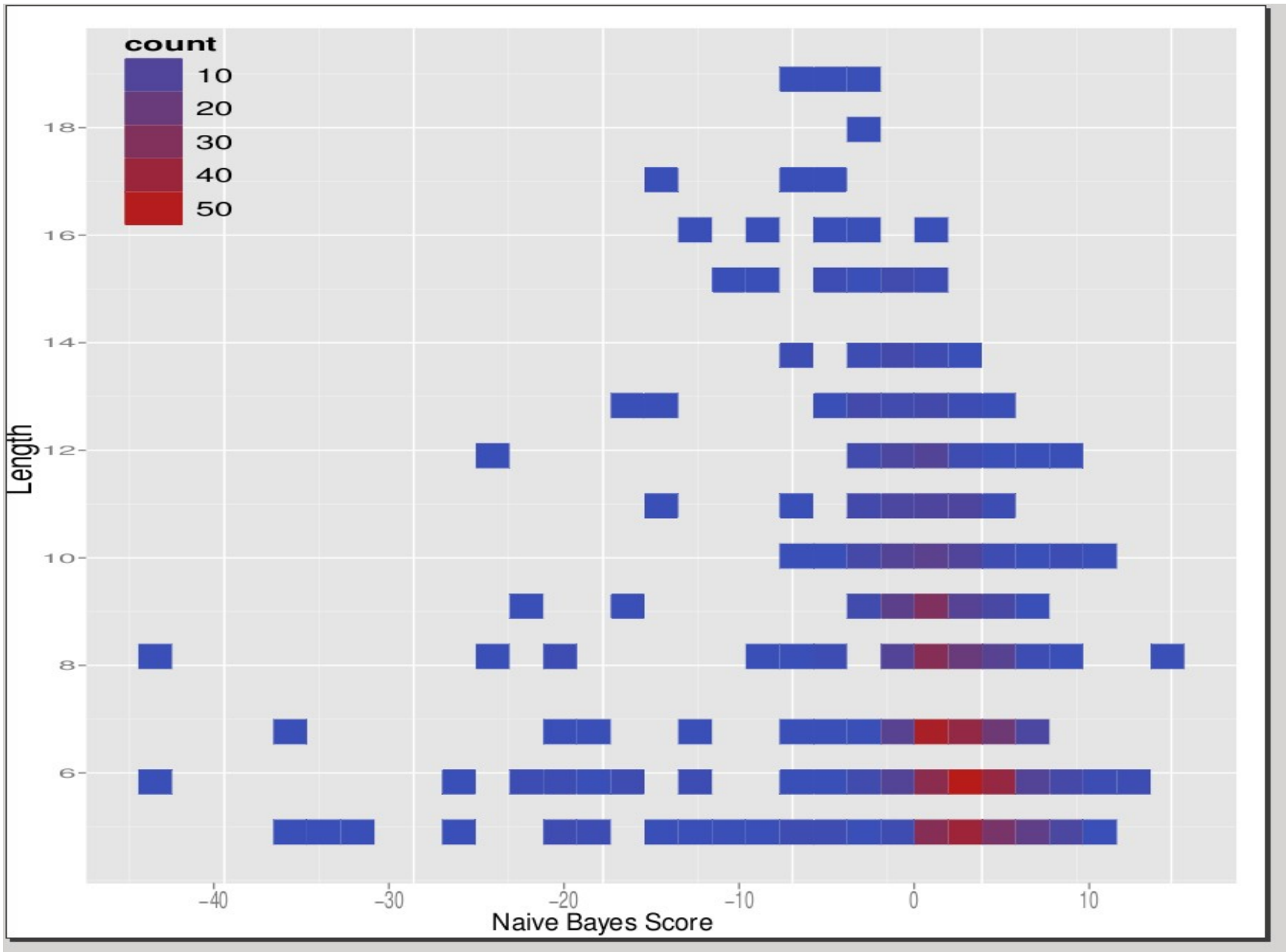
FPR: False Positive Rate



Box-plot, because it's sexy







Questions

