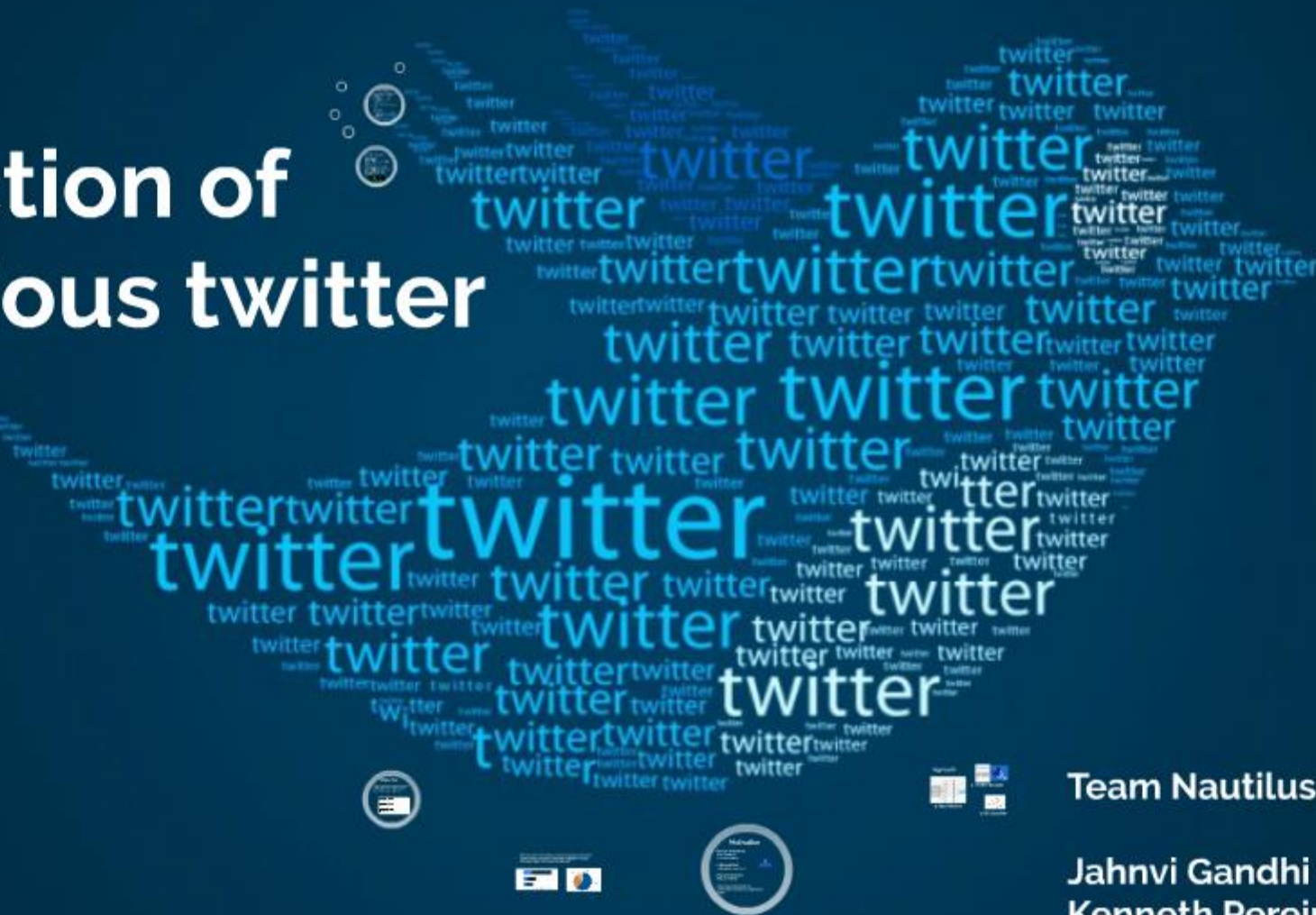


# Detection of Malicious twitter feeds



Team Nautilus

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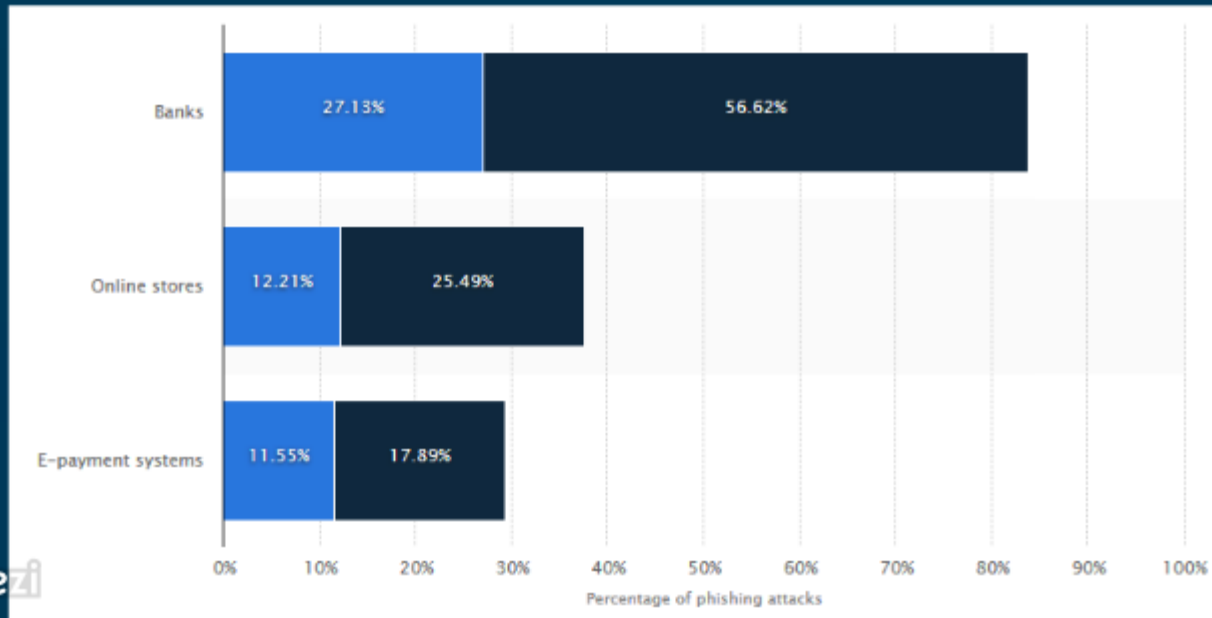
# Objective

Build a system that scrapes real-time twitter feeds and classifies shared URLs as phishing or legitimate

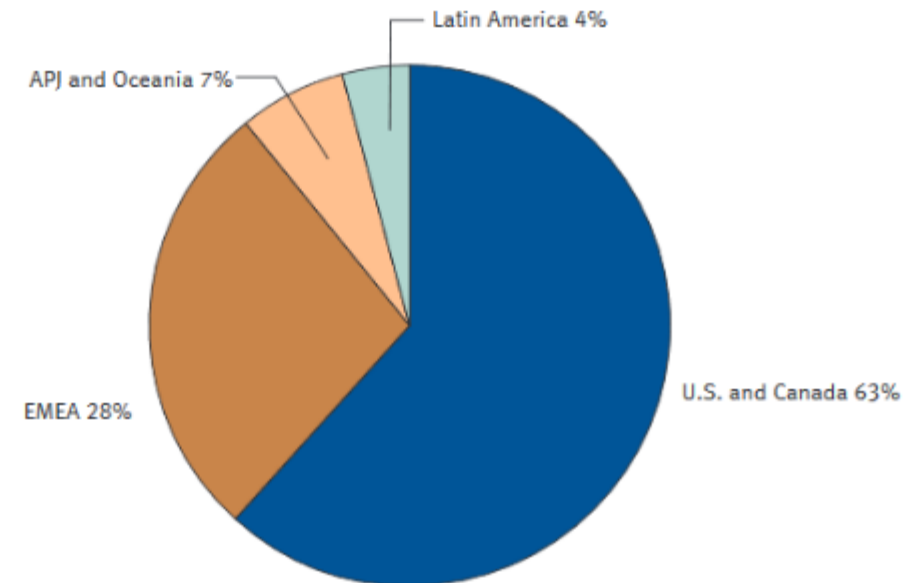


# Why?

- More than 50% of phishing cases lead to Financial losses
- Organizations encounter more than \$3 Billion in losses
- Increase in Data Theft and Identity theft



## PHISHING BY GEOGRAPHY



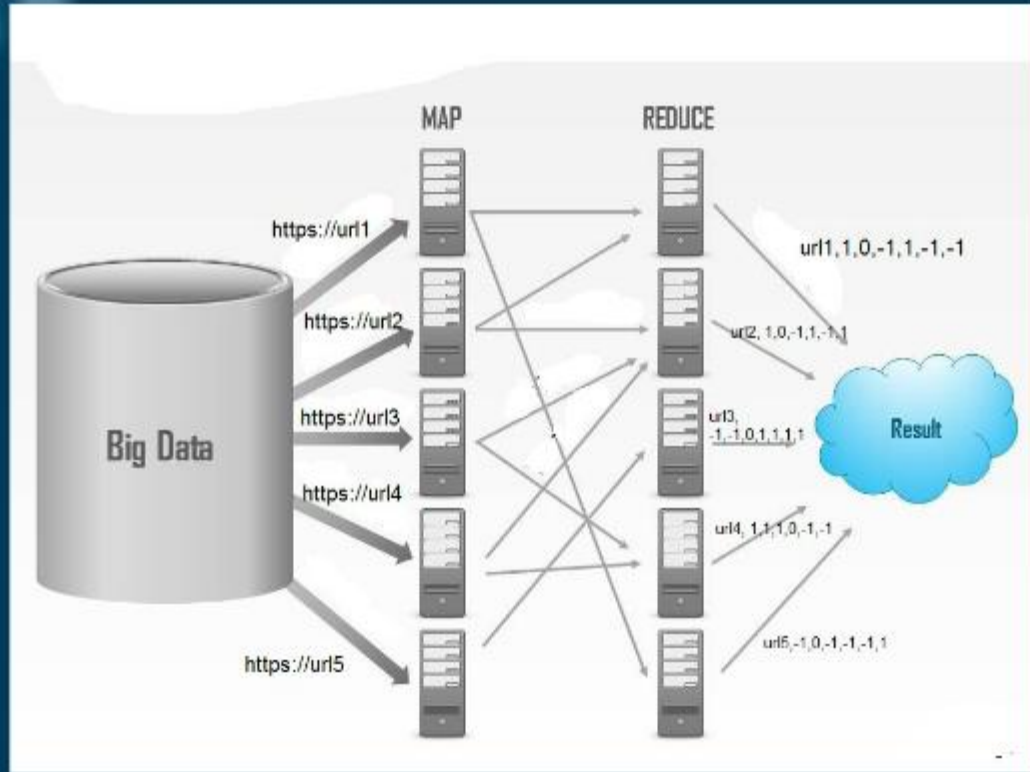
# Motivation

- Rami M. Mohammad  
Fadi Thabtah  
Lee McCluskey
- Collected Data  
using their own tool
- Phishing Websites  
Dataset (2015)
- Only rules discussed, no  
instruction as how to implement  
them





# Approach

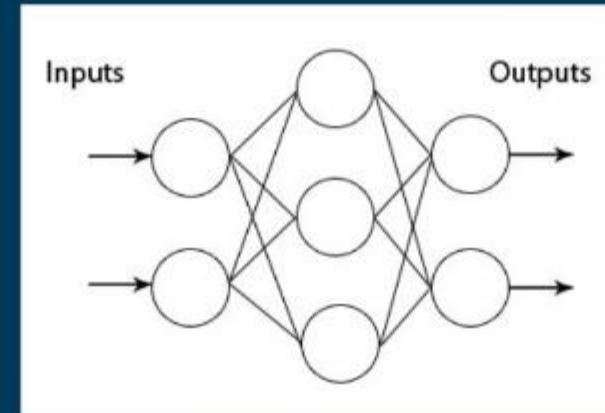


## 2. Map Reduce

URL (66)
<a href="https://twitter.com/coneyrice/lists/music/members">https://twitter.com/coneyrice/lists/music/members</a>
<a href="https://twitter.com/DARTparker/lists/music-writer/members">https://twitter.com/DARTparker/lists/music-writer/members</a>
<a href="https://twitter.com/luke_moton/lists/music/members">https://twitter.com/luke_moton/lists/music/members</a>
<a href="https://twitter.com/hedgioia/lists/music-critics-bloggers/members">https://twitter.com/hedgioia/lists/music-critics-bloggers/members</a>
<a href="https://twitter.com/sabinekindel/lists/music/members">https://twitter.com/sabinekindel/lists/music/members</a>
<a href="https://twitter.com/magnoliaartists/lists/music-mage-journos/members">https://twitter.com/magnoliaartists/lists/music-mage-journos/members</a>
<a href="https://twitter.com/AtistopiaMusic/lists/music-news-events/members">https://twitter.com/AtistopiaMusic/lists/music-news-events/members</a>
<a href="https://twitter.com/peterauh/lists/music-peeps/members">https://twitter.com/peterauh/lists/music-peeps/members</a>
<a href="https://twitter.com/parakoveets/lists/music-critics/members">https://twitter.com/parakoveets/lists/music-critics/members</a>
<a href="https://twitter.com/ZinziLive/lists/music/members">https://twitter.com/ZinziLive/lists/music/members</a>
<a href="https://twitter.com/jneezys/lists/music-bloggers-etc/members">https://twitter.com/jneezys/lists/music-bloggers-etc/members</a>
<a href="https://twitter.com/culture_czar/lists/music-junkies/members">https://twitter.com/culture_czar/lists/music-junkies/members</a>
<a href="https://twitter.com/jnathansbire/lists/music/members">https://twitter.com/jnathansbire/lists/music/members</a>
<a href="https://twitter.com/misakikey/lists/music/members">https://twitter.com/misakikey/lists/music/members</a>
<a href="https://twitter.com/sobercool/lists/music-artists/members">https://twitter.com/sobercool/lists/music-artists/members</a>



## 1. Twitter Scraper



## 3. ML classifier

# Twitter Scraper

- Built using a twitter API
- API gets feeds
- Output file (Tweet data, userDetails, timestamp, location...)

## *Roadblocks:*

- API calls restricted to 180 requests every 15 minutes
- Missing data from the Twitter feeds due to User Account Settings
- Shortend URLs

```
Output URLs: 165122  
Waiting 15 minutes to continue  
Tweets read: 310968  
Output URLs: 165122  
ubuntu@ip-172-31-20-54:~$ █
```



# Map Reduce

- MR Model takes in Twitter feed
- Reducer checks and applies URL rules
- MR Output is a file with 6 attributes
- ex: URL, 1, -1, 0, 1, 1, 1

Adv: Easy detection of repeated URLs



*Roadblocks:*

- Consistent data required, if any scrapped field has missing data, MR job fails
- Ex: location

# MapReduce

- ML output

```
http://1.usa.gov/1Xd5hV1      ,1,1,0,1,1,1,10  
http://10.Weather             ,1,1,0,1,1,1,56  
http://100-singalong-songs-for-kids.i9-news.co  
utm_source=dlvr.it&utm_medium=twitter ,-1,1,0,  
http://1000oldies.de          ,1,1,0,1,1,1,8  
http://1000phones.com/phone/1-617-830-7579    ,  
http://1000rockhits.de        ,1,1,0,1,1,1,18  
http://1001.tl/122813         ,1,1,0,1,1,1,2  
http://100resilientcities.org/blog/entry/bosto
```

Instance State	Status Checks	Public DNS (IPv4)
 running	 2/2 checks ...	ec2-52-15-123-175.us-...



# Reducer

## URL rules:

- Using the IP Address

`"http://0x58.0xCC.0xCA.0x62/2/paypal.ca/index.html"`

*Rule: IF*  $\begin{cases} \text{If The Domain Part has an IP Address} \rightarrow \text{Phishing} \\ \text{Otherwise} \rightarrow \text{Legitimate} \end{cases}$

- Long URL to Hide the Suspicious Part

`http://federmacedoadv.com.br/3f/aze/ab51e2e319e51502f416dbe46b773a5e/?cmd=_home&disp  
atch=11004d58f5b74f8dc1e7c2e8dd4105e811004d58f5b74f8dc1e7c2e8dd4105e8@phishing.website.  
html`

*Rule: IF*  $\begin{cases} \text{URL length} < 54 \rightarrow \text{feature} = \text{Legitimate} \\ \text{else if URL length} \geq 54 \text{ and } \leq 75 \rightarrow \text{feature} = \text{Suspicious} \\ \text{otherwise} \rightarrow \text{feature} = \text{Phishing} \end{cases}$

- URL's having "@" Symbol

*Rule: IF*  $\begin{cases} \text{Url Having @ Symbol} \rightarrow \text{Phishing} \\ \text{Otherwise} \rightarrow \text{Legitimate} \end{cases}$

# Reducer

## URL rules (cont):

- Redirecting using “//”

<http://www.legitimate.com/http://www.phishing.com>

Rule: IF  $\begin{cases} \text{ThePosition of the Last Occurrence of "://" in the URL} > 7 \rightarrow \text{Phishing} \\ \text{Otherwise} \rightarrow \text{Legitimate} \end{cases}$

- Adding Prefix or Suffix Separated by (-) to the Domain

<http://www.Confirme-paypal.com/>

Rule: IF  $\begin{cases} \text{ThePosition of the Last Occurrence of "/" in the URL} > 7 \rightarrow \text{Phishing} \\ \text{Otherwise} \rightarrow \text{Legitimate} \end{cases}$

- The Existence of “HTTPS” Token in the Domain Part of the URL

<http://https-www-paypal-it-webapps-mpp-home.soft-hair.com/>

Rule: IF  $\begin{cases} \text{Using HTTP Token in Domain Part of The URL} \rightarrow \text{Phishing} \\ \text{Otherwise} \rightarrow \text{Legitimate} \end{cases}$

# MapReduce findings

- Mapper Input: Input file (includes Urls, location, etc)
- Mapper Output: URLs , Count
- Reducer input: URLs, Count
- Reducer output: URLs, Attributes Values
  
- Six URL rules out of 30 were achievable for predicting the phishing websites
- Ex for non-achievable rules: Favicon, HTTPS certificate, age of Domain, website traffic
  
- Parallelization achieved
- Map-Reduce helped in observing the duplicate URLs.



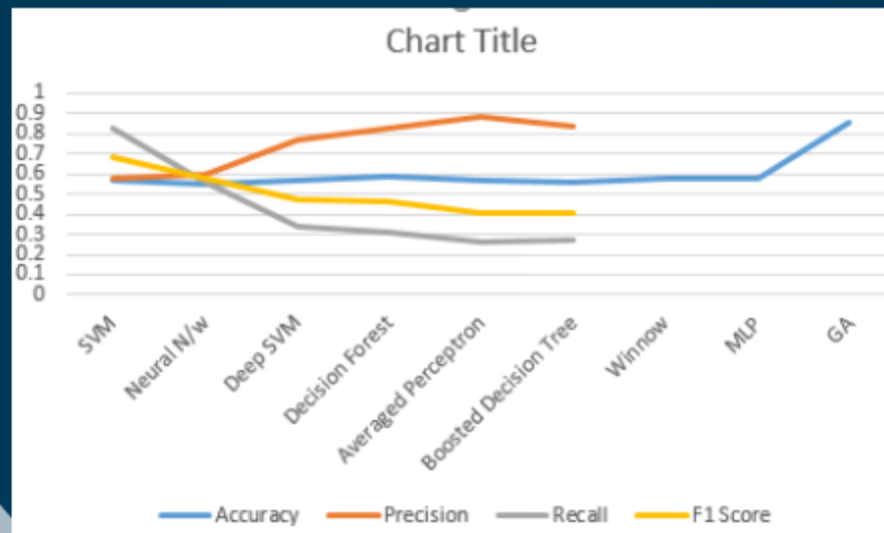
# ML classifier

- Two class classification
  - 0 = Phishing
  - 1 = Legitimate
- Input = URL and 6 attributes  
ex: url, 1, -1, 0, 1, -1, 1
- Output = URL, 6 attributes, target Value  
ex: url, 1, -1, 0, 1, -1, 1, 1

```
https://www.youtube.com/watch?v=4di9SQ90KAU&feature=share,0,1,0,1,1,1,1  
https://www.youtube.com/watch?v=4xAjrQ7wb3Q,1,1,0,1,1,1,1  
https://www.youtube.com/watch?v=56qdgDRDdUM,1,1,0,1,1,1,1  
https://www.youtube.com/watch?v=6EPvRdVg5Ug&feature=share,0,1,0,1,1,1,1  
https://www.youtube.com/watch?v=6I1bNZ5OdvQ,1,1,0,1,1,1,1  
https://www.youtube.com/watch?v=6X0jxo3jlok&feature=share,0,1,0,1,1,1,1  
https://www.youtube.com/watch?v=6X0jxo3jlok&feature=share,0,1,0,1,1,1,1
```

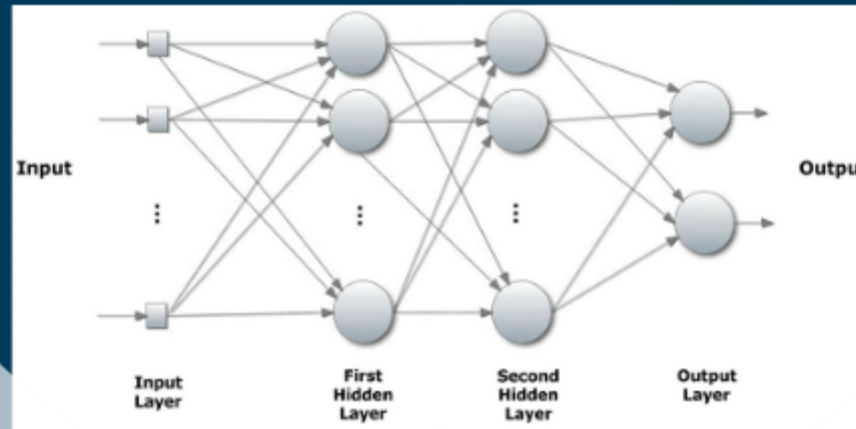
# ML findings

- MLP perceptron (deepLearning4j)
- Microsoft Azure ML
- 6 algorithms applied and tested
- Winnow, Logistic Classifier, GA



# MLP

- Feedforward artificial neural network
- Each layer fully connected to the next one
- Each node is a neuron
- Nonlinear activation function
- Supervised learning technique called backpropagation
- Modification of linear perceptron and can distinguish data that are not linearly separable





# ML charts

Twitter Data\_Final\_Algos > Score Model > Scored dataset

rows	columns				
98046	10				
double_slash_redirecting	Prefix_Suffix	HTTPS_token	Count	Scored Labels	Scored Probabilities
1	1	1	10	1	0.990307
1	1	1	56	1	0.990307
-1	1	1	2	1	0.999087
1	1	1	8	1	0.990307
1	1	1	2	1	0.990307
1	1	1	18	1	0.990307

**2 class Neural Nw**

Twitter Data\_Final\_Algos > Score Model > Scored dataset

rows	columns				
98046	10				
double_slash_redirecting	Prefix_Suffix	HTTPS_token	Count	Scored Labels	Scored Probabilities
1	1	1	10	1	0.958993
1	1	1	56	1	0.958993
-1	1	1	2	1	0.819083
1	1	1	8	1	0.958993
1	1	1	2	1	0.958993
1	1	1	18	1	0.958993

**2 class Support  
Vector Machine**

## Future Scope

- Automate twitter API to look for new feeds and classify URLs as and when new feeds are available
- Look for more sophisticated algorithms (deep learning techniques) to get better accuracy
- URL advisor
- ChatBot - scrap data from page being viewed and inform user about URLs not to be clicked
- Sanitize the phishing URLs

**Thank You**