



SCIENTIST PRESENTATION

# DETECTION OF PHISHING WEBPAGES WITH SUPERVISED LEARNING



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# AGENDA

**PHISHING DETECTION PIPELINE**

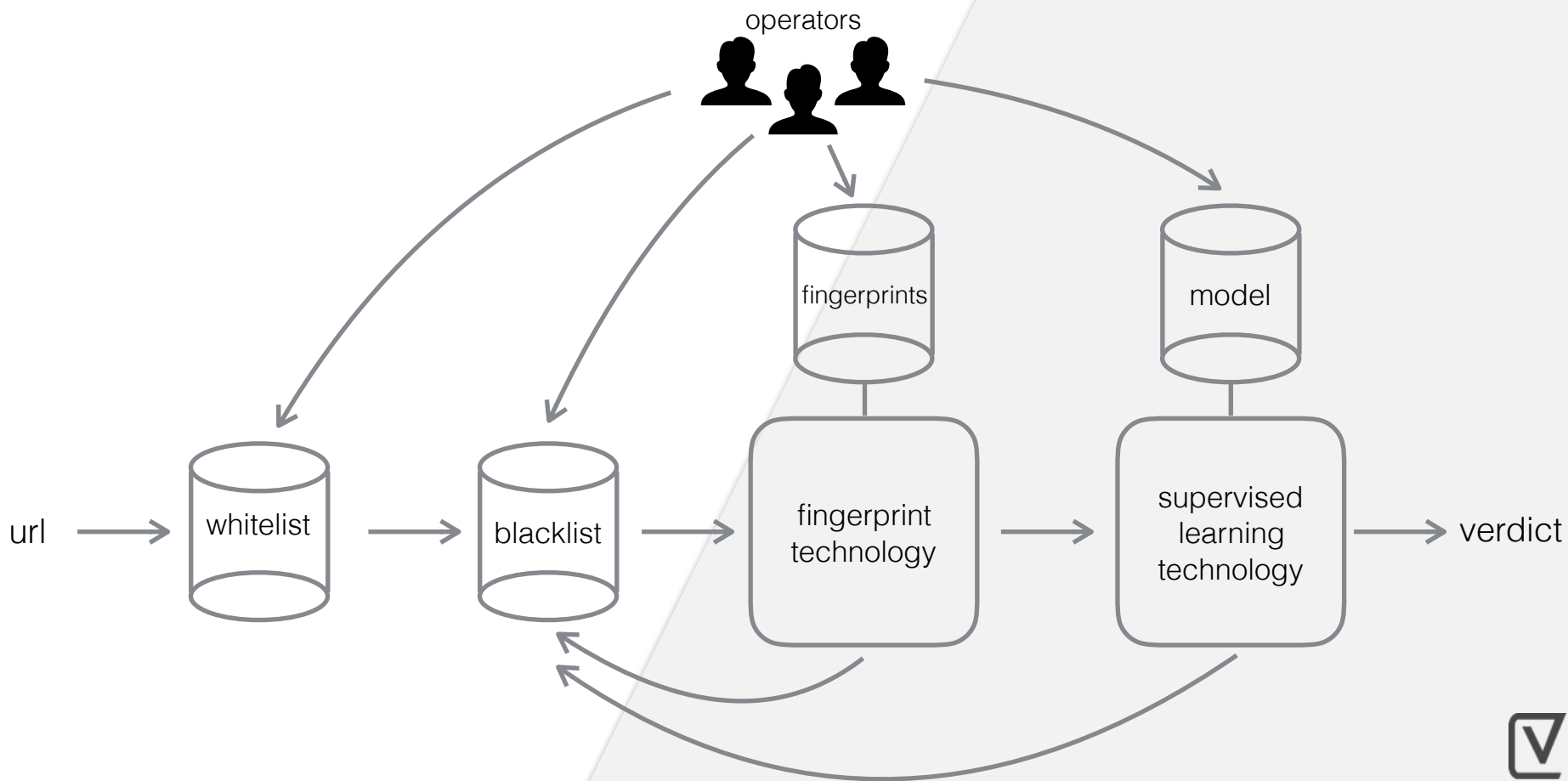
**BRAND DETECTION**

**SUPERVISED LEARNING TECHNOLOGY**

**RESULTS**



# PHISHING DETECTION PIPELINE



## BRAND DETECTION - A UNIQUE APPROACH

- A phishing attack always impersonates a brand (PayPal, Apple...)
- Pros:
  - Improve catch rate with brand specific features
  - Alert brand when a URL is detected



# SUPERVISED LEARNING TECHNOLOGY

- Collect data
- Build features vector
- Classify vector

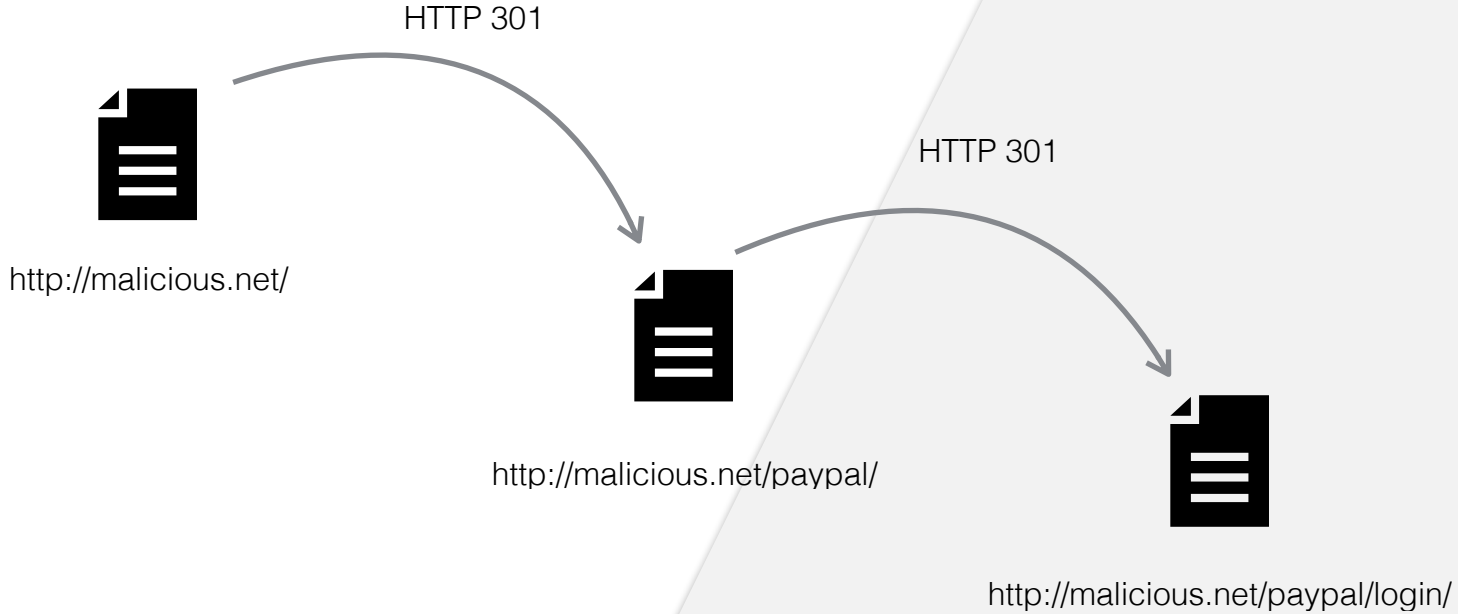


# SUPERVISED LEARNING TECHNOLOGY - COLLECT DATA

- **Input:** url
- **Output:** final url and final document
- Redirections are followed
  - HTTP 4xx status code
  - Meta refresh
  - Javascript



# SUPERVISED LEARNING TECHNOLOGY - COLLECT DATA



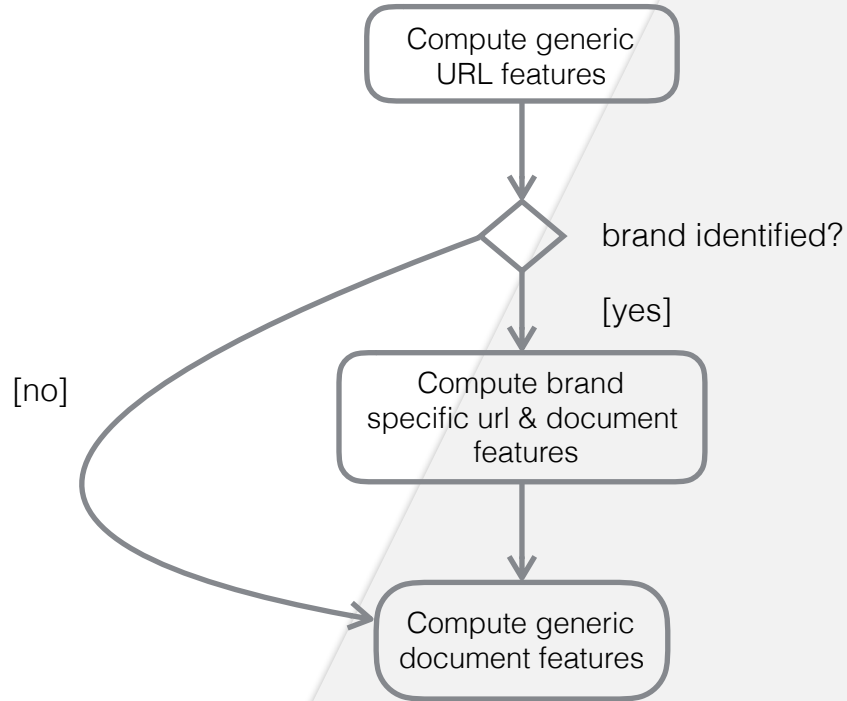
# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR

- Binary features
- A feature relates either to a URL or a document
- Feature type can be:
  - Generic: common to most phishing webpages
  - Brand specific: specific to a brand (Paypal, Apple...)





# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR



# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR

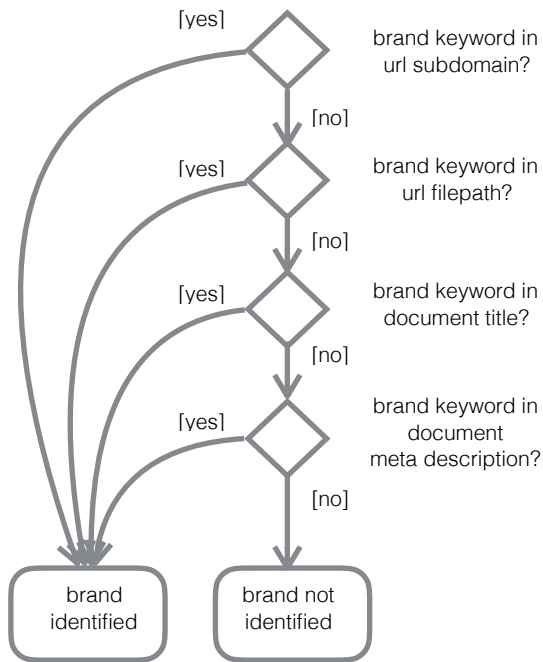
## Generic URL features

| Feature                               | Exemple   |
|---------------------------------------|---|
| URL_HOSTNAME_IPV4                     | <a href="http://75.125.43.213/itunees">http://75.125.43.213/itunees</a>   |
| URL_MANY_SUBDOMAINS                   | <a href="http://login.paypal.com.arest.net/websrc">http://login.paypal.com.arest.net/websrc</a>   |
| URL_WORDPRESS_PATH_COMPONENT_OR_TILDE | <a href="http://econergy.info/wp-includes/ppl/">http://econergy.info/wp-includes/ppl/</a><br><a href="http://13.185.121.59/~cadas507/index1.php">http://13.185.121.59/~cadas507/index1.php</a>  |
| URL_ACTION_KEYWORD_SUSPECT            | <a href="http://zgcake.com/wellsfargo/signon.htm">http://zgcake.com/wellsfargo/signon.htm</a><br><a href="http://odn.zce.szcecin.cz/login/">http://odn.zce.szcecin.cz/login/</a><br><a href="http://ausgreesoar.com.au/logn.php">http://ausgreesoar.com.au/logn.php</a> |



# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR

## Identify brand



```
<brand name="apple">
  <keyword>apple</keyword>
  <keyword>icloud</keyword>
  <keyword>itunes</keyword>
  [...]
</brand>
```

brand keyword in url subdomain,  
match can be fuzzy

<http://75.125.43.213/itunes>



# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR

## Brand specific features

| Feature                                   | Description   |
|---|---|
| URL_SUBDOMAIN_SUSPECT                     | One the URL subdomain matches one of the brand <b>keyword</b> elements. Match can be fuzzy.                               |
| URL_PATH_SUSPECT                          | One the URL path element matches one of the brand <b>keyword</b> elements. Match can be fuzzy.                            |
| DOCUMENT_TITLE_OR_METADESCRIPTION_SUSPECT | Document title (resp. meta description) matches at least one of the brand <b>title</b> (resp. <b>meta_desc</b> ) elements |
| DOCUMENT_ICON_OR_CSS_OR_JS_SUSPECT        | Document icon (resp. CSS and JS) matches at least one of the brand <b>icon</b> (resp. <b>css</b> and <b>js</b> ) elements |
| DOCUMENT_HIGH_DOMAIN_RATE                 | At least 50% of document links have a domain matching the brand <b>domain</b> elements                                    |
| DOCUMENT_DATA_SUSPECT                     | There is data in the document that matches at least one of the brand <b>data</b> elements                                 |



# SUPERVISED LEARNING TECHNOLOGY - BUILD FEATURES VECTOR

## Brand description

```
<brand name="apple">  
  <keyword>apple</keyword>  
  <keyword>icloud</keyword>  
  <keyword>itunes</keyword>  
  <domain>apple.com</domain>  
  <domain>icloud.com</domain>  
  <domain>cdn-apple.com</domain>  
  <title>Apple</title>  
  <title>iCloud</title>  
  <title>iTunes</title>  
  <meta_desc>Apple ID</meta_desc>  
  <meta_desc>iCloud</meta_desc>  
  <css>signin.css</css>  
  <js>signin.js</js>  
  <js>apple.js</js>  
  <data>Apple ID</data>  
</brand>
```

A brand can have multiple services or products

Most resources (images, CSS, JS...) are stored on these domains



# SUPERVISED LEARNING TECHNOLOGY - GENERIC DOCUMENT FEATURES

## Feature

## Description

DOCUMENT\_FORM\_SUSPECT

A form contains a keyword in its attribute that is relevant of a required user action within the phishing process. Example:

```
<form id="form" action="login.php" method="post">  
<form method="POST" id="signIn" name="ConnectForm" action="u-  
send.php">
```

DOCUMENT\_CREDENTIAL\_FIELD

An input field is a password field or a credit card security code field. Example:

```
<input autocomplete="off" type="password" id="loginpwd"  
name="loginpwd" value="">  
<input name="cvc" maxlength="4" id="cvc"  
autocomplete="off" style="width:47px;" type="text">
```

DOCUMENT\_SUSPECT\_TECHNIC

There is a suspect technic.



# SUPERVISED LEARNING TECHNOLOGY - SUSPECT TECHNICS

| Technic               | Example   |
|-----------------------|---|
| Homoglyph attack      | <u>Google</u><br><u>PayPal</u>  |
| HTML hex encoding     | <code>&amp;#065;&amp;#112;&amp;#112;&amp;#108;&amp;#101</code>  |
| JS hex encoding       | <code>document.write(unescape('%53%69%67%6E%20%49%6E'))</code>  |
| JS AES 256 encryption | <pre>var hea2p = ('0123456789[...]nopqrstuvwxyz'); var hea2t = 'XwHq/n3w[...]yFz/191Q=='; var output = Aes.Ctr.decrypt(hea2t, hea2p, 256); document.write(output)</pre> |



## SUPERVISED LEARNING TECHNOLOGY - CLASSIFY VECTOR

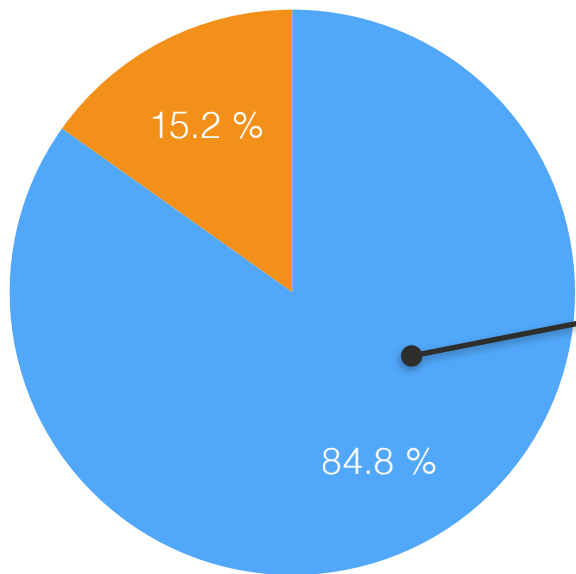
- Binary classifier: SVM with Gaussian RBF kernel
- Training & test corpus:
  - Phishing: fingerprint technology, operators decisions
  - Clean: urls matching whitelist, operators decisions
- Goal : 0 false positive





# RESULTS

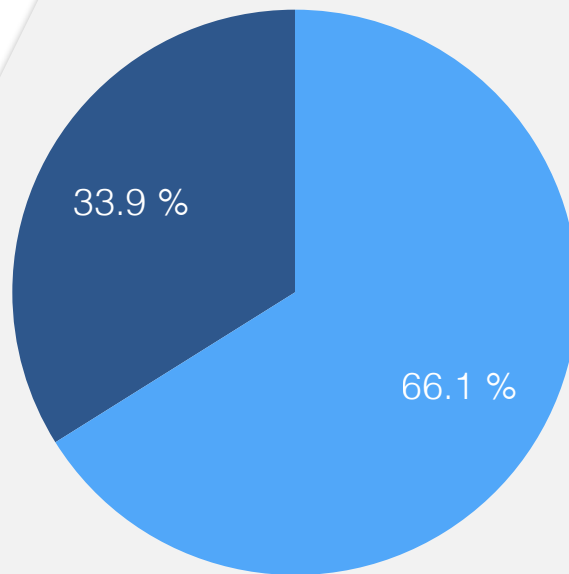
## Phishing catch rate



● catch rate

● false negative rate

## Brand detection



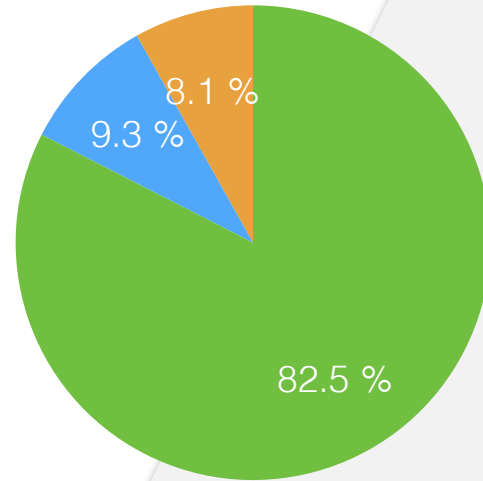
● brand detected

● no brand detected



# RESULTS

## Technologies together



- fingerprint technology
- supervised learning technology
- false negative rate



## CONCLUSION

|                                | Pros                                       | Cons   |
|--------------------------------|--|--|
| Fingerprint technology         | High catch rate<br>Fast<br>Flexible        | Binary output<br>Fails to detect unknown threats |
| Supervised learning technology | High catch rate<br>Output is a probability | Slow<br>Difficult to scale brands                |



# Merci

Feel free to exchange with our lead scientist



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If you want to use it: [IsItPhishing.org](https://IsItPhishing.org)

Free for Researchers and Students

