# Analysis of the KAIST Dooray Mailing System: Enhancing Security Against Phishing and Identity Attacks

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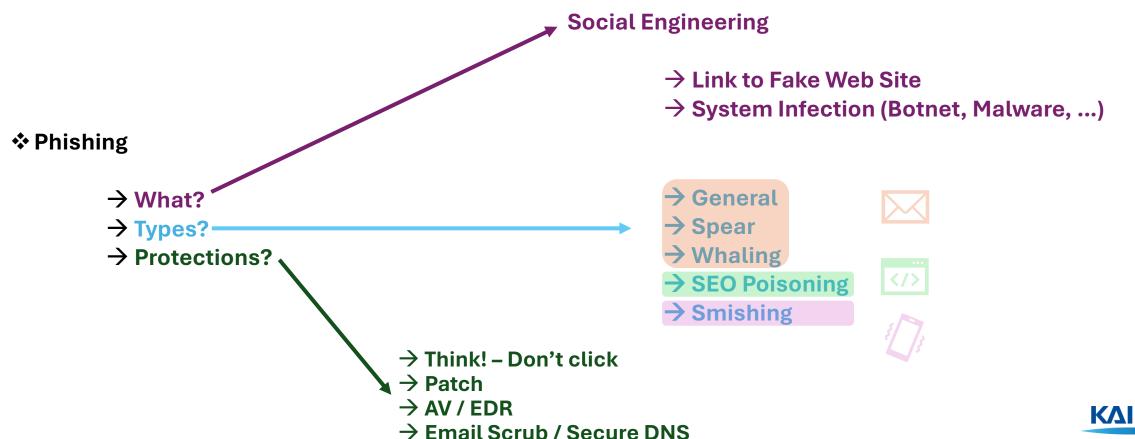
# Summary

- Introduction
- Motivation
- Background
- Research Work Contrib. Zphisher Dooray Security Mechanism
- Research Work Contrib. ML Phishing Detection Tool
- Technical Challenges
- Evaluation Results & Impact
- Limitation & Future Work



## Introduction

O What is Phishing?



## **Motivation**

Ubiquiti Networks victim of \$39 million social engineering attack

News Analysis

07 Aug 2015 • 5 mins

Cybercrime Data Breach Fraue

ov stug zoro + o mins

CRIME & COURTS

KU employees fall victim to phishing scam, lose paychecks

By Bryan Lowry

blowry@wichitaeagle.com

The Wichita Eagle

July 11, 2016 4:52 PM | 🖵

July 11, 2016 4:52 PM

blowry@wichitaeagle.com

Technology

# Austria's FACC, hit by cyber fraud, fires CEO

By Reuters

May 25, 2016 6:52 PM GMT+9 · Updated 9 years ago

May 25, 2016 6:52 PM GMT+9 · Updated 9 years ago



an IDG, Inc. company



## **Motivation**

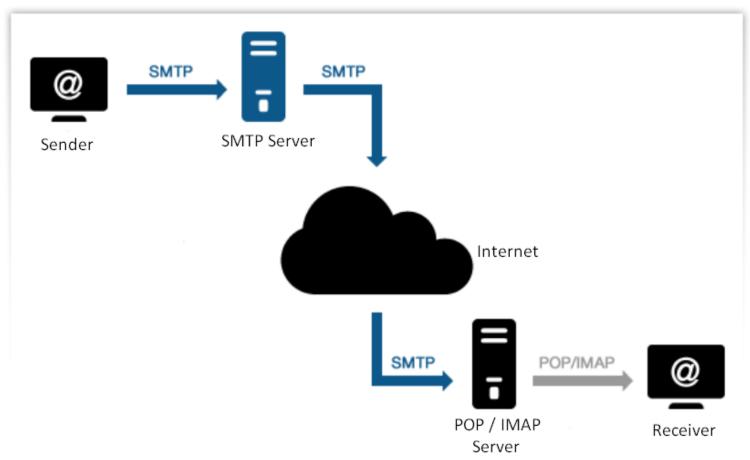
## Assess the Security Mechanisms of KAIST's Dooray Mailing System

- Critical Role of Email Security
  - → Over 90% of cyberattacks originate from Phishing emails
  - → Securing mailing systems ensure trusted communication
- Challenges in Phishing Detection
  - → Sophisticated attacks bypass traditional defenses
  - → Gaps in protocols like SMTP, DKIM, and DMARC can be exploited
- Opportunity for Improvement
  - → Analyzing Dooray can reveal vulnerabilities and enhance defenses





SMTP (Simple Mail Transfer Protocol)





### SMTP (Simple Mail Transfer Protocol)

#### Purpose

- > Standard protocol for sending emails between servers
- → Ensures email delivery across different domains

#### Features

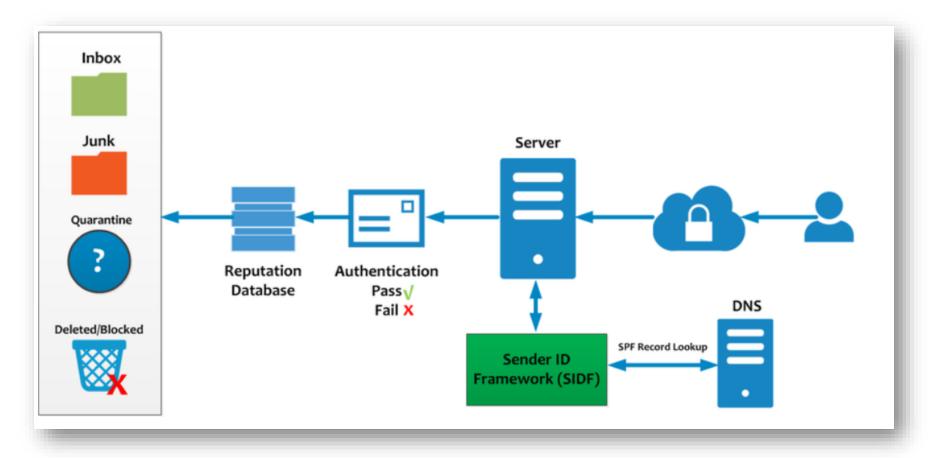
- → Relies on plaintext by default (can be encrypted with STARTTLS)
- → Specifies rules for email formatting and transfer

#### Limitations

- → Vulnerable to spoofing without additional layers (e.g., SPF, DKIM, DMARC)
- → Does not inherently verify sender authenticity

```
telnet smtp.---- 25
Connected to smtp.----.
220 smtp.---- SMTP Ready
HELO client
250-smtp.----
250-PTPFLINING
250 8BITMIME
MAIL FROM: <auteur@yyyy.yyyy>
250 Sender ok
RCPT TO: <destinataire@---->
250 Recipient ok.
DATA
354 Enter mail, end with "." on a line by itself
Subject: Test
Corps du texte
250 Ok
OUIT
221 Closing connection
Connection closed by foreign host.
```

SPF (Sender Policy Framework)





SPF (Sender Policy Framework)

#### Purpose

> Defines which mail servers are authorized to send emails on behalf of a domain

#### How It Works

- → Domain owners create SPF records in Public DNS
- → Receiving servers check the SPF record to verify if the sending IP is authorized



SPF (Sender Policy Framework)

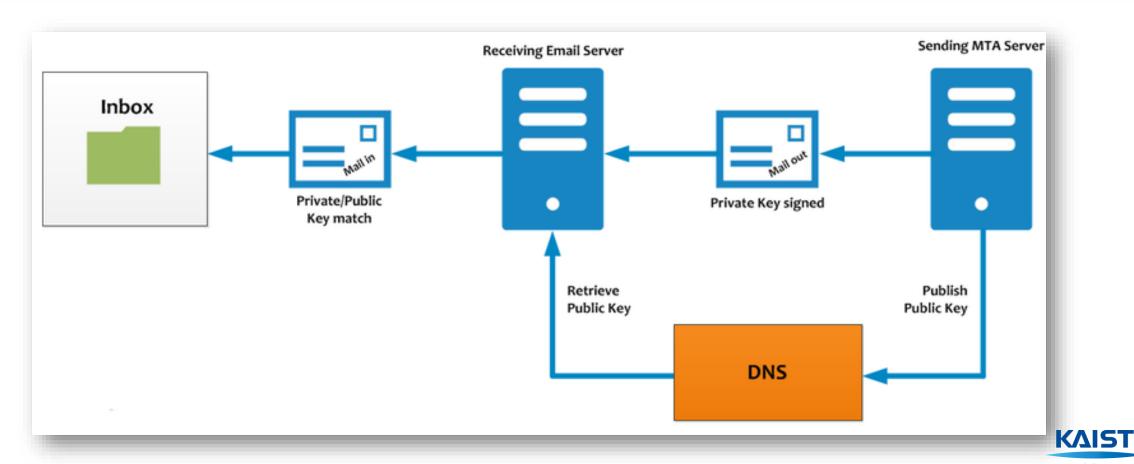
#### Benefits

- → Reduces email spoofing
- → Simple to set up compared to other protocols

#### Limitations

- → Does not work on forwarded emails
- → Cannot prevent phishing if the sender uses an authorized domain

DKIM (DomainKey Identified Mail)



DKIM (DomainKey Identified Mail)

#### Purpose

→ Adds a digital signature to outgoing emails to verify sender authenticity

#### How It Works

- → The sender's domain generates a cryptographic key pair
- → The private key signs the email, and the public key is published in DNS records
- → The recipient verifies the signature using the public key

DKIM (DomainKey Identified Mail)

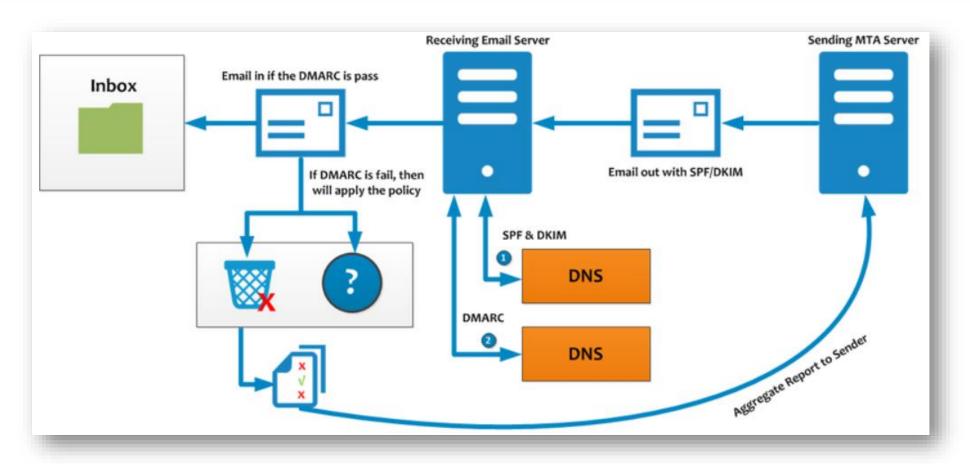
#### Benefits

- → Prevents email tampering during transit
- → Establishes trust in sender identity

#### Limitations

- → Does not directly prevent phishing or spoofing without DMARC
- → Relies on proper DNS configuration

o DMARC (Domain-based Message Authentication, Reporting & Conformance)





o DMARC (Domain-based Message Authentication, Reporting & Conformance)

#### Purpose

- → Builds on SPF and DKIM to provide policy enforcement for email authentication
- → Specifies how to handle unauthorized emails (reject, quarantine, or none)

#### How It Works

- → Domain owners publish a DMARC record in DNS
- → Incoming emails are checked against SPF & DKIM
- → A report is generated on email authentication outcomes



o DMARC (Domain-based Message Authentication, Reporting & Conformance)

#### Benefits

- → Combats phishing and spoofing more effectively than SPF/DKIM alone
- → Provides detailed reports on email abuse attempts

#### Limitations

- → Requires correct SPF/DKIM configuration to work
- → Complex to implement for organizations with multiple email-sending services



#### **Phishing Email Testing Methodology**



# Contribution of our Research Work (Part. 1)

Zphisher - Dooray's response to phishing emails from different domains



**Dooray Mail System** 

Domains: Gmail.com, ForvisMazars.com, Yopmail.com, Efrei.net

"nails Sent

Phishing Emails (ZPhisher)

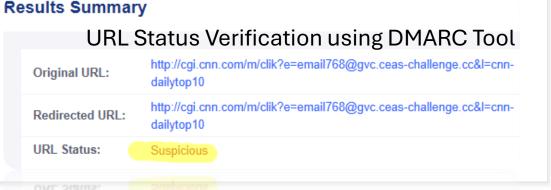
Zphisher - Dooray's response to phishing emails from different domains

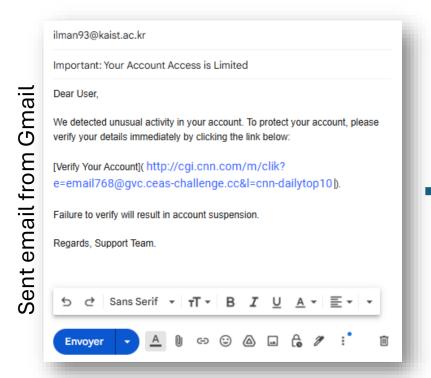
```
Dear User,
We detected unusual activity in your account. To protect your account, please verify
 Verify Your Account](http://phishing-example.com).
Failure to verify will result in account suspension.
Regards,
Support Team.
Dear User,
Your recent payment was declined. To avoid account deactivation, please update your
 Update Payment Info](http://phishing-example.com).
Thank you,
Billing Team.
Your account information needs verification. Click below to confirm your details:
 [Confirm Details](http://phishing-example.com).
Failure to verify may result in restricted access to your account.
Regards,
Technical Support.
Hello [User's Name],
Your account password has expired. For your security, please reset it now using the
[Reset Password](http://phishing-example.com).
This link will expire in 24 hours.
Best regards,
IT Department, Forvis Mazars.
```

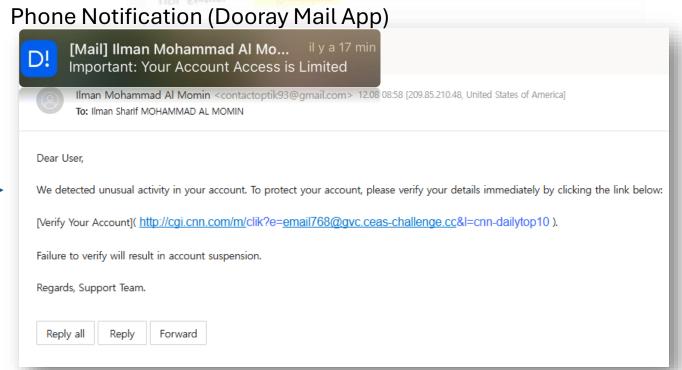
```
| Content of second 2.2.4 M. Income infose expect TOOLS (ZPhisher rzphisher roote enfose expect 2.2.4 M. Income infose expect TOOLS (ZPhisher rzphisher 100028) | 2.3.5 | | URL 1 : https://ieee-tc-purple-greece.trycloudflare.com | URL 2 : https://is.gd/H4MbQQ | | URL 3 : https://infosecpattesting.com@is.gd/H4MbQQ | | URL 3 : https://infosec
```

Zphisher is used to generate emails containing fraudulent URLs for phishing detection testing.

Result – Gmail -> Dooray







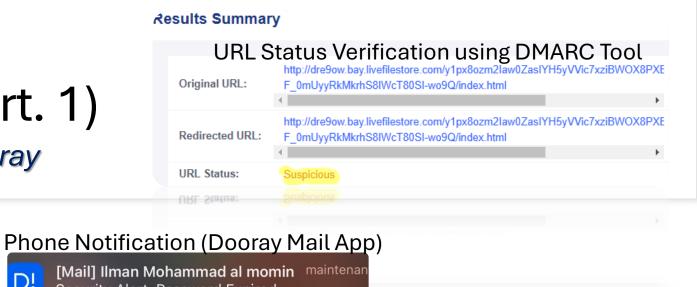
#### Email Reception on Mailbox (Dooray)

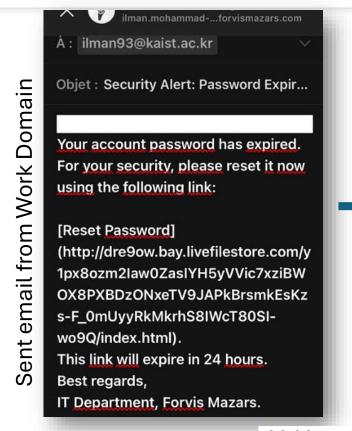
Phishing Email Test: Gmail → Dooray (Not Detected)

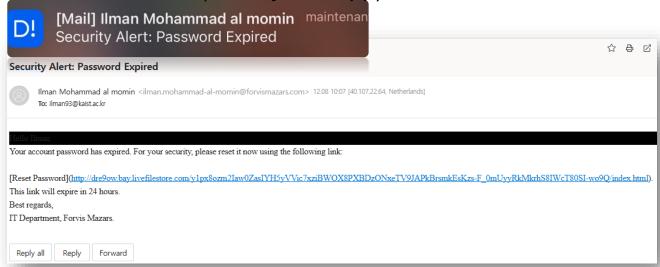




Result – Forvis Mazars -> Dooray







Email Reception on Mailbox (Dooray)

Phishing Email Test: Forvis Mazars Mail → Dooray (Not Detected)



Result – EFREI.net -> Dooray

#### Results Summary

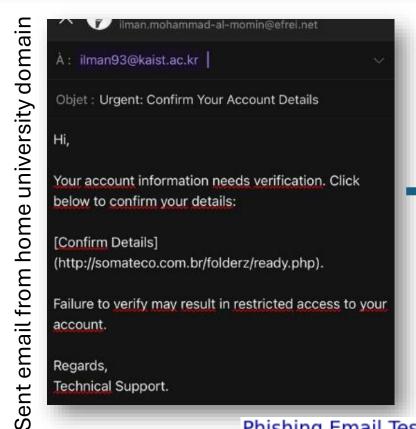
#### URL Status Verification using DMARC Tool

Original URL: http://somateco.com.br/folderz/ready.php

Redirected URL: http://somateco.com.br/folderz/ready.php

URL Status: Suspicious

RL Status: Suspicious



Phone Notification (Dooray Mail App)
[Mail] Ilman MOHAMMAD AL M... maintenant

**Urgent: Confirm Your Account Details** 

Ilman MOHAMMAD AL MOMIN <ilman.mohammad-al-momin@efrei.net> 12.08 11:57 [40.107.247.115, Netherlands]
To: ilman93@kaist.ac.kr

Your account information needs verification. Click below to confirm your details:

[Confirm Details]

(http://somateco.com.br/folderz/ready.php).

**Urgent: Confirm Your Account Details** 

Failure to verify may result in restricted access to your account.

Regards,

Technical Support.

Email Reception on Mailbox (Dooray)

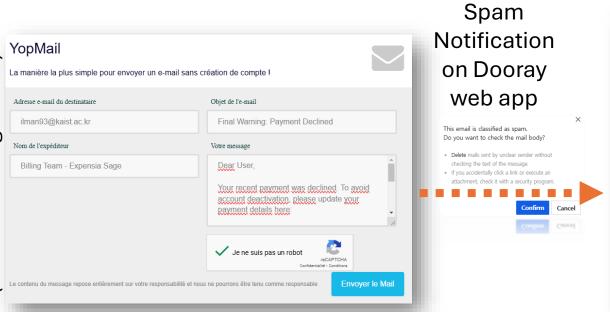
Phishing Email Test: EFREI.net Mail → Dooray (Not Detected)



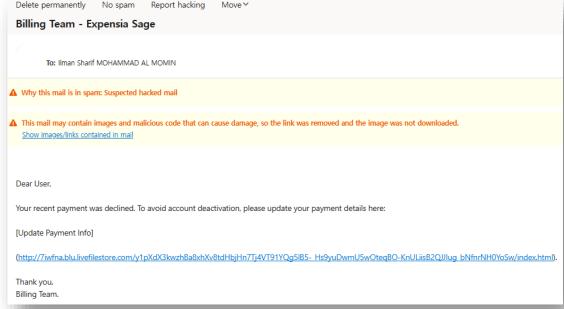
**EFREI.net Mail** 

Dooray (Not Detected

Result – YopMail -> Dooray







Email Reception on Spam Mailbox (Dooray)

Phishing Email Test: YopMail → Dooray (Detected)





## Contribution of our Research Work (Part. 2)

- Machine Learning Based Phishing Detection Tools
  - First Tool: Phishing Mail Detection
  - Second Tool: Phishing URL Detection
    - Evaluation of both models using an External Dataset (CEAS\_08.csv)

# Methodology

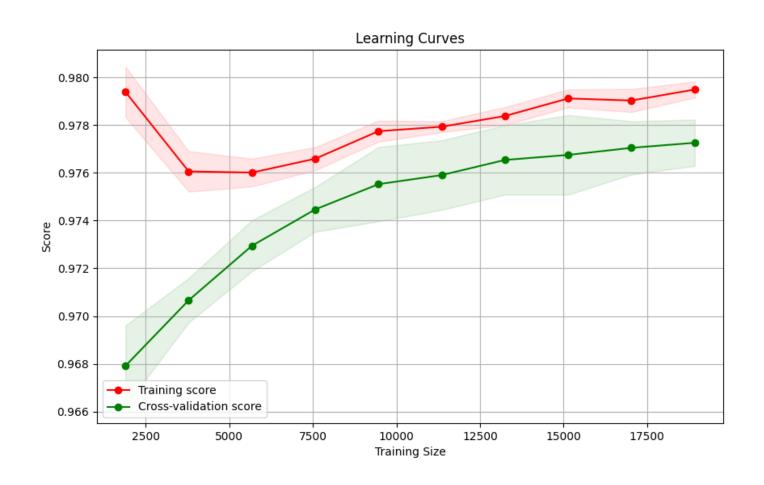
#### Mail Model In Detail

## Mail Model 🔀

Based on Logistic Regression Vectorizer: TF\_IDF Clean Data Features Remove\_NonASCII Data Vectorizer 🏋 subject, body, label Remove\_Spaces Training 🖶 Verify Model 🗎 Result Plot 🚎 TF\_IDF Remove\_NonAlpha-Num enron.csv except-from '@', '.', ':', '/', '&', '=', '?' Preprocessing **Model Training** Dataset

## Results

Learning Curves for Mail Model using Logistic Regression



#### **Comment:**

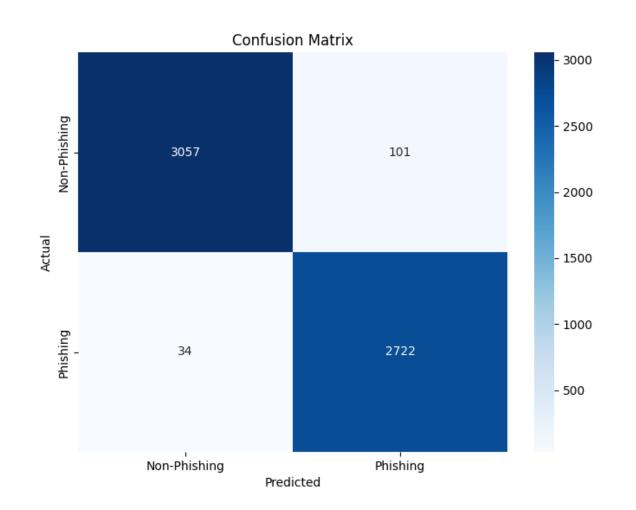
This graph shows us that our model is not Over / Under Fitting using the enron.csv dataset

Checking Learning Curves
ensure us that the Model is
capable to replicate the result on
external dataset



## Results

Confusion Matrix for Mail Model using Logistic Regression



#### Comment:

This Confusion Matrix shows us that the Model successfully identified most phishing and non-phishing emails, with 101 false positives and 34 false negatives, indicating areas for potential improvement.



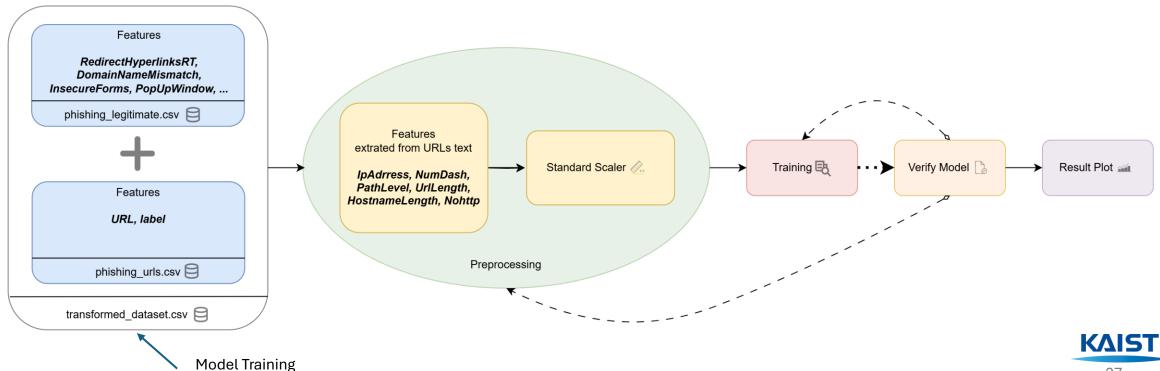
# Methodology

**Combined Datasets** 

URL Model In Detail

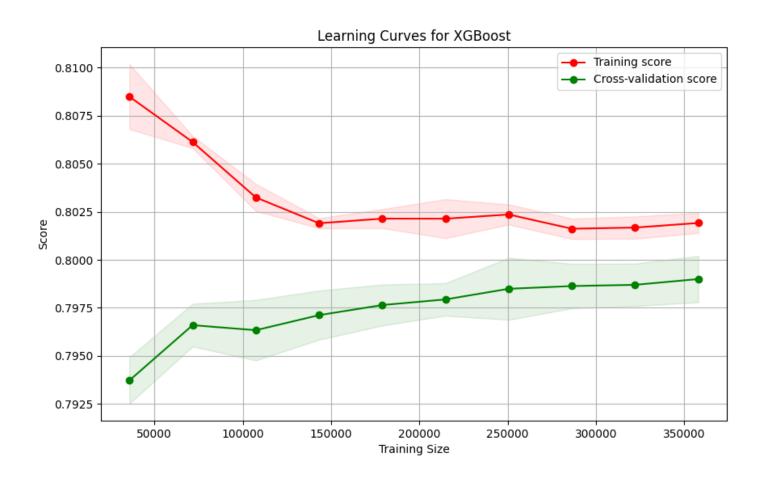
### URL Model &

Based on Boosting Trees (XGBoost)



## Results

Learning Curves for URL Model using Boosting Trees (XGBoost)



#### Comment:

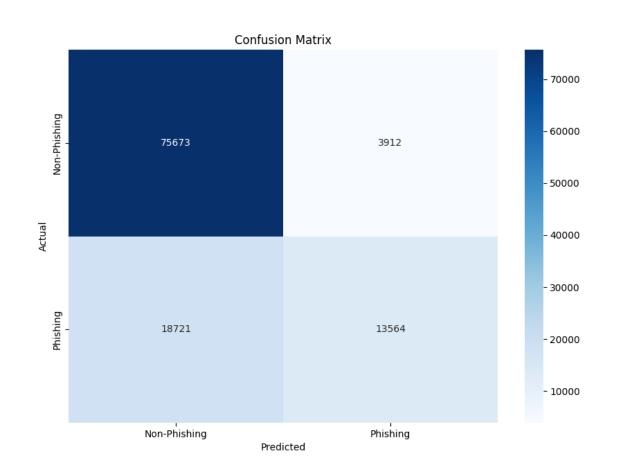
This graph shows us that our model is not Over / Under Fitting using the enron.csv dataset

**Checking Learning Curves** ensure us that the Model is capable to replicate the result on external dataset



## Results

Confusion Matrix for URL Model using Boosting Trees (XGBoost)

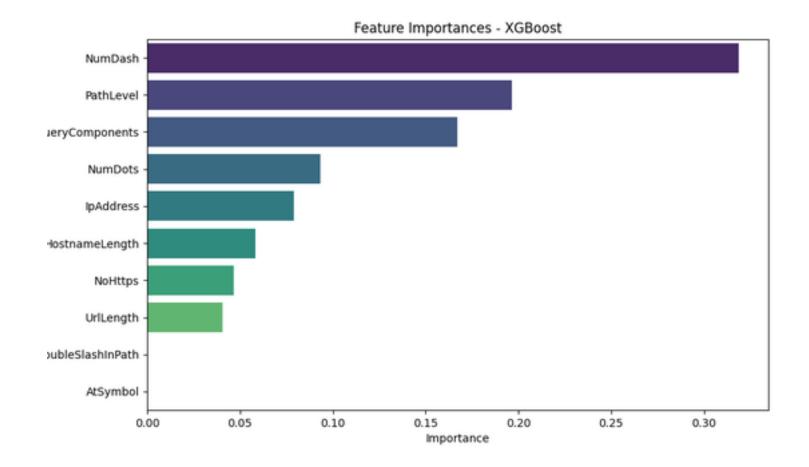


#### Comment:

The model performs well in detecting non-phishing URLs but performs poorly on detecting phishing emails, indicating a need for improved sensitivity. However, we used the ability of this model to detect non-phishing emails for the rest of our work.

## Features Exploitation

Extracted URL Features classified by Importance phishing\_legitimate.csv



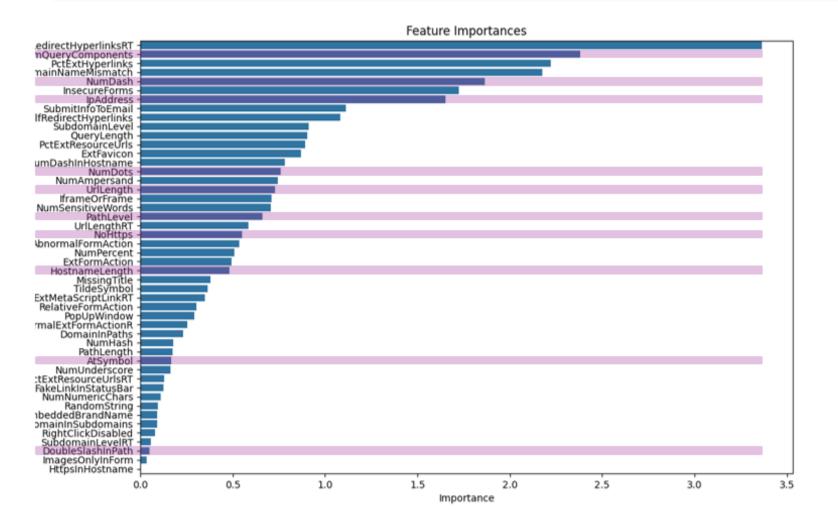
Features Extraction used to train our XGBoost model

Comment: This plot show the features used for training our model in order to classify the emails.



## Features Exploitation

Extracted URL Features classified by Importance phishing\_legitimate.csv



The **48 features** contained in the dataset vs the **10 features** we used in training our model



#### Comment:

The dataset used

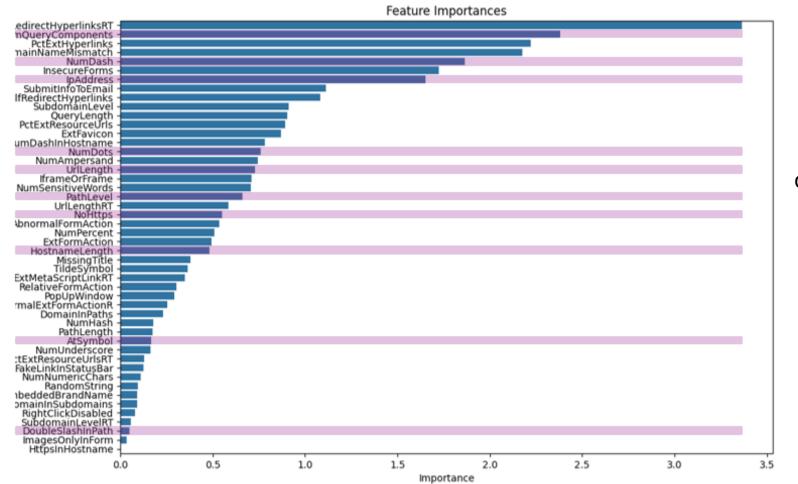
Selenium Web Driver ToolKit to
extract the features from each
URL.

KAIST

31

## Features Exploitation

Extracted URL Features classified by Importance phishing\_legitimate.csv



#### Comment:

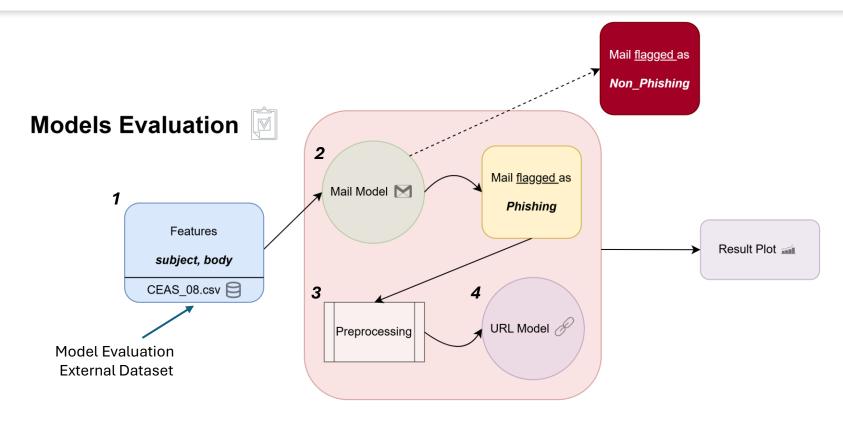
We didn't use the 48 features of the dataset because when testing our model, most of the websites from the URLs were down as the mails were old.

→ We focused only on the « persistent » features.



# Methodology

#### Models Evaluation In Detail



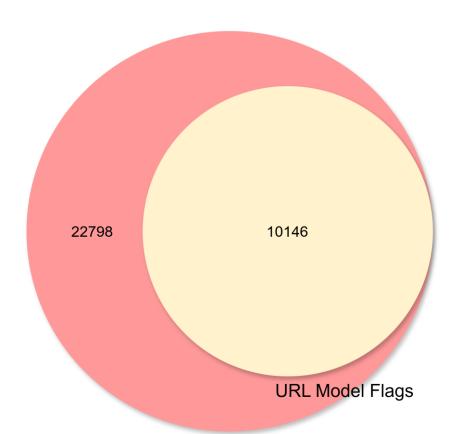
#### Comment:

- 1 We use an external dataset to test our framework.
- 2 We use the Mail Model to flag mails as *Phishing* or *Non-Phishing*.
- **3 -** We scrap all the URLs from the mail that are flagged as Phishing (Preprocessing)
  - 4 We use the URL Model to flag those URLs as Phishing / Non-Phishing.



Venn Diagram for Comparative Results

Overlap Between Email and URL Models



#### **Comment:**

This diagram shows the result of the evalution of our framework on the CEAS\_08 dataset:

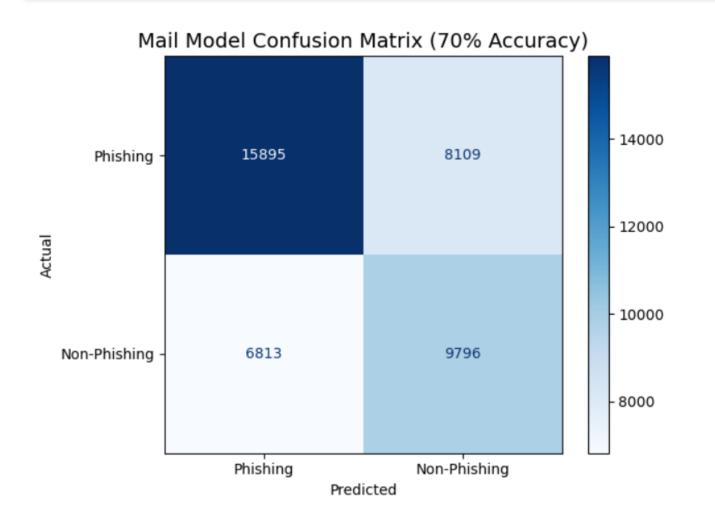
- → The first Mail Model flags 22798 emails as phishing
- We extract the URLs from all these emails, (if they have URLs, otherwise they are labelled as non-phishing)
  - We preprocess & transfer them into the second URL model, we separate **non-phishing** from **phishing** URLs

(the model has a high accuracy for detecting non-phishing URLs)

→ We find 10146 emails flagged as **phishing**.



Confusion Matrix for Mail Model using Boosting Trees (XGBoost) on CEAS\_08 Dataset

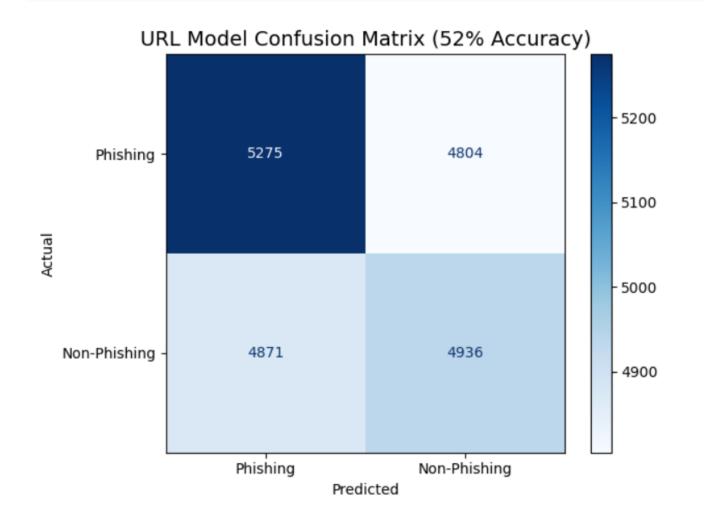


#### Comment:

Evaluation the Mail Model using CEAS\_08 gave us an accuracy of 70%.

The mail model shows a moderate performance with a clear distinction between true positives and true negatives, but a noticeable number of false positives and false negatives reduces its reliability.

Confusion Matrix for URL Model using Boosting Trees (XGBoost) on CEAS\_08 Dataset



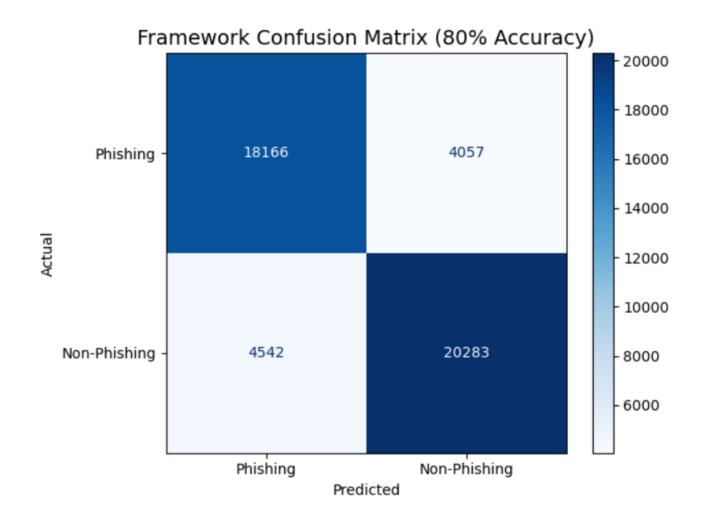
#### Comment:

Evaluation the URL Model using CEAS\_08 gave us an accuracy of 52%.

The URL model exhibits lower accuracy, indicating significant challenges in correctly predicting phishing URLs (feature scrapping issue), with a high count of both false positives and false negatives.

Confusion Matrix for the Double Model Framework

using Boosting Trees (XGBoost) on CEAS\_08 Dataset



#### Comment:

The overall accuracy increased significantly, demonstrating better performance in detecting phishing URLs.

The combined framework demonstrates improved accuracy, effectively reducing false negatives and false positives, showing the benefit of integrating both models.



## Overall Technical Challenges

Selection & Data Prep

- Finding the most exploitable dataset
  - → enron.csv, phishing\_legitimate.csv, CEAS\_08.csv
- Missing & Mismatched Data
  - → Mixing datasets: phishing\_legitimate.csv + phishing\_urls.csv & Standardization

## Technical Challenges

o for Mail Model

- Model Training
  - → Choosing the Learning Method (Logistic Regression, Boosting Trees, Random Forest)
- Over / Underfitting Issues
  - → Adjusting Hyperparameters & Optimizing Learning Curves

## Technical Challenges

#### o for URL Model

- URLs and features extraction
  - → Variety of URL formats (http, https, IPs, ...)
  - → Adapting the Extraction Function
- Model Training
  - → Choosing the Learning Method (Logistic Regression, Boosting Trees, Random Forest)
- Datasets
  - → Lack of diversity in the dataset for the training
  - → Limited features exploitability



# Limitations of our Methodology

o Dependency on standard installations & real-world Variances

Limitation	Description
Poor Performance on External Datasets	The model performs poorly on external datasets like <b>CEAS-08</b> due to mismatched data and class imbalance, affecting its generalization ability.
Insufficient Feature Complexity	Simple features (e.g., NumDots , PathLevel ) do not capture complex patterns in phishing URLs, resulting in <b>low detection accuracy</b> .
Missed URLs in Attachments or Scripts	The model only detects URLs in the body of emails, failing to identify URLs in attachments, JavaScript, or hidden elements within emails.

## **Future Work**

#### **❖** Fine Tune Double-Evaluation Models

- → Extend Research to Non-Phishing Flagged Mail
- → Enhance Models accuracy
- → Detection of Phishing Emails with Images URLs and Tracking artifacts

#### Browser Automation Kit

- → Usage of a Web Driver Tool Kit like Selenium could be effective to scrap URL content
- → Result -> More Features could be exploited for Model Training

#### Integration with Dooray API

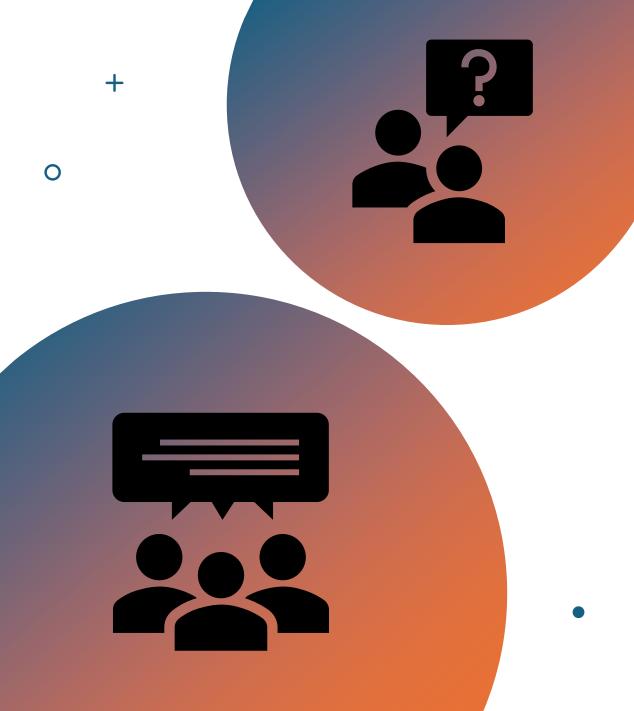
→ Do real-time phishing detection



## Conclusion

- Research on Mailing Protocols
  - → DMARC, DKIM, SPF, SMTP
- Analyzed Dooray's Security Mechanism
  - → Zphisher & Several Mailboxes
- **❖** Tool Framework
  - → ML-based Phishing Detection Tools
  - → Evaluation on public Dataset (CEAS\_08.csv)





# Discussion