

Quantum-Resilient Phishing Detection:

A Post-Quantum Secure Framework for Email URL Analysis
Using AI-Driven Models

By:

Under the Guidance of

Shamanth M Hiremath (1MS22CS128)
Sanchit Vijay (1MS22CS122)
Ashutosh Kumar (1MS22CS036)
Trijal Shinde (1MS22CS153)

Sangeetha V Soumya CS Het Joshi



Agenda

 Introduction Objectives Methodology Used Technology Stack Comparison to Existing Solution Results • References



Introduction

- Phishing Attacks: Phishing is a major cyber threat. It involves using deceptive URLs in emails to trick users.
- Novel Solution: Introduces a new method for detecting phishing attacks. Combines quantum-resilient encryption with AI-driven URL analysis.
- Secure Processing: URLs from emails are securely processed. Protects the detection model and its parameters from future quantum computing threats.
- High Predictive Accuracy: Maintains a high level of accuracy in predicting phishing threats.
- Research Contribution: Connects cybersecurity with advanced cryptographic methods. Aims to provide robust solutions in an evolving cyber threat environment.



Introduction

This research introduces a quantum-secure phishing detection framework that combines:

- 1. An AI-driven Random Forest model, trained on 30 critical URL features, to classify URLs as phishing or legitimate with high accuracy.
- 2. A post-quantum cryptographic layer to encrypt and secure the model's weights and parameters, ensuring resilience against future quantum threats.
- 3. A browser extension that extracts URLs from incoming emails, performs real-time phishing analysis, and safeguards user data.

This innovative integration addresses the dual challenge of effective phishing detection and future-proof cybersecurity.



Objectives

- To develop an AI-based phishing detection model using 30 features extracted from email URLs like Length of the URL, Number of special characters, Presence of IP address in URL, Count of subdomain, HTTPS usage, Length of the hostname, Age of domain.
- To secure the model's parameters and predictions using post-quantum cryptography.
- To demonstrate the feasibility of real-time deployment via an email-parsing browser extension.
- To compare the proposed solution with existing methods in terms of accuracy, security, and operational efficiency.
- To establish the model's resilience against both classical and quantum-based attacks.



Methodology

Dataset Preparation

Source: URLs sourced from repositories such as PhishTank, OpenPhish, and legitimate domains from Alexa Top 1 Million Sites.

Size: ~100,000 labeled URLs, evenly distributed between phishing and legitimate classes.

Features: Extracted 30 parameters from each URL, encompassing both lexical (URL structure) and host-based (domain-related) features.

Model Development

- Algorithm: Random Forest Classifier
- Optimization: GridSearchCV with:
 - \circ n_estimators = 100
 - max_features = log2
 - criterion = entropy
- Performance: Achieved an accuracy of 97.24% with balanced precision and recall metrics.



Methodology

Quantum-Secure Encryption

- 1. Encryption Methodology:
 - Model weights and parameters are encrypted using AES-256 for speed and efficiency.
 - AES keys are secured using Kyber, a post-quantum key encapsulation mechanism, ensuring quantum resistance.

2. Workflow:

- At runtime, the model is decrypted for predictions, and the parameters are re-encrypted after use.
- Encryption and decryption add a latency of ~50ms per operation without significant performance degradation.

3.5 Integration with Email Parsing

- Developed a browser extension to:
 - a. Parse email content.
 - b. Extract embedded URLs.
 - c. Send URLs for phishing prediction using the quantum-secure framework.



Technology Stack

- Programming: Python (backend), JavaScript (browser extension).
- AI/ML: scikit-learn, pandas, NumPy
- Cryptography: PyCryptodome (AES-256), pqcrypto (Kyber).
- Email Parsing: Python Email library, Gmail API, IMAP/SMTP protocols.
- **Deployment:** Google Colab for training, AWS cloud services for hosting, browser extension for realworld testing, VS Code for development.



Comparison to Existing Solution

Aspect	Existing Solutions	Proposed Solution
Features Analyzed	10–15 features in most cases	30 comprehensive URL-based features
Encryption	Standard (AES or RSA)	Post-quantum secure (Kyber + AES-256)
Quantum Resistance	Not addressed	Fully quantum-resilient encryption
Deployment	Cloud/local	Browser extension with real-time URL parsing
Accuracy	~80-93%	97.24% with enriched features and RFA optimization



Results

Model Performance:

a. Accuracy: 97.24%

b. Precision: 96.89%

c. Recall: 98.11%

d. F1-Score: 97.49%

• Feature Importance:

Key contributors: URL length, DNS record validity, and number of special characters.



Results

• Encryption Performance:

Encryption/decryption overhead: ~50ms per operation.

Ensured end-to-end model security with no compromise on prediction speed.

• Real-World Deployment:

Browser extension successfully parsed emails, extracted URLs, and delivered real-time predictions.



References

- 1. Bernstein, D. J., et al. Post-Quantum Cryptography: A NIST Perspective. NIST, 2022.
- 2. PQClean: A Collection of Clean Implementations of Post-Quantum Cryptographic Algorithms.
- 3. scikit-learn developers and documentation. Machine Learning in Python: Random Forest Implementation.
- 4. Kaggle for Dataset: Open Database of Phishing URLs.
- 5. PyCryptodome Library. Python Cryptography Toolkit.
- 6. YouTube
- 7. Google GMail Documentation
- 8. ChatGPT



Thank You