



جامعة
Teknologi
MARA

CSP760

*PREDICTING VULNERABILITY SUSCEPTIBILITY IN
MALAYSIAN BANK USING SUPERVISED MACHINE
LEARNING*

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UPDATE WEEK 7

Paper	Dataset Source	Algorithm	Key Finding	Justification
Oliveira et al., 2023 – <i>Hybrid ML for Vulnerability Detection in Web Apps</i>	Web app vulnerability logs (CVE-related)	Random Forest, SVM, Naive Bayes	RF performs strongly for vulnerability detection; hybrid supervised models increase accuracy.	Supports RF as a strong classifier for structured vulnerability data.
Williams et al., 2020 – <i>Vulnerability Analysis & Prediction Framework</i>	Historical vulnerability metadata & severity scores	Artificial Neural Networks (ANN), Regression models	ANN delivers strong performance in predicting vulnerability severity.	Directly supports the use of Neural Network Regression for predictive scoring.
Asmar & Tuqan, 2024 – <i>ML for Cybersecurity in Digital Banks</i>	Digital banking cybersecurity datasets (financial institutions)	SVM, RNN, HMM, LOF	ML improves detection accuracy in banking security environments.	Financial institution context – ML is suitable for Malaysian banks.
Yosifova et al., 2021 – <i>Predicting Vulnerability Type in CVE Database</i>	CVE Details dataset - 122,774 CVE records (expanded to 162,789 after exploding categories)	Random Forest, Naive Bayes, Linear SVM	RF performs strongly on CVE classification.	Directly supports RF for CVE-based training, matching my dataset (NVD + Tenable + ExploitDB).
Sharadhi et al., 2022 – <i>ML Algorithms for Software Defect Prediction</i>	PROMISE defect dataset (NASA) - More than 40 datasets (sizes vary by module)	ANN, RF, SVM, Decision Trees	RF & ANN repeatedly outperform other ML algorithms.	Supports choosing RF + NN as my primary models due to strong precision in defect/cybersecurity prediction tasks.

Adani Kamal, 2025 – <i>Predicting Vulnerability Susceptibility in Malaysian Banks Using Supervised ML (Your Study)</i>	Merged dataset: NVD + Tenable + ExploitDB	Random Forest + Neural Network Regression	Combines global vulnerability metadata with exploit information + scanner data to build predictive model for banks.	Extends prior work by integrating multi-source real-world datasets to target Malaysian banking vulnerability susceptibility.
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Paper / Framework	Asset Types Listed
FFIEC IT Handbook (Financial Sector)	Transaction systems, Settlement & clearing systems, Customer information databases, Identity systems, Network & security infrastructure
BNM RMiT (Bank Negara Malaysia) (Regulatory Framework)	Core Banking System (CBS), Payment systems (FPX, IBG, RENTAS, DuitNow), Internet/mobile banking, Fraud monitoring systems, ATM & card systems, Customer information systems (CIF), Third-party vendor systems, Cloud workloads used for banking
NIST SP 1800-25 (Financial Services)	Treasury systems, Fund transfer systems, Card processing systems, API gateways, IAM / MFA systems, network device such as (Firewalls, routers, switches)
SWIFT CSCF (Banking Payment Network)	SWIFT interface servers, HSM (Hardware Security Modules), Payment messaging servers, Secure communication nodes
ISO/IEC 27001 (General Framework)	Information assets, Servers & databases, Security appliances, Network devices, End-user devices



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Problem	Research Questions	Hypothesis	Objectives	Scope	Significance
<p>Malaysian banks rely heavily on Vulnerability Assessment (VA) tools such as Nessus, Qualys, and OpenVAS that only detect <i>existing vulnerabilities</i> but cannot <i>predict future threats</i>. This leaves financial institutions unprepared for advanced cyberattacks.</p>	<p>RQ1: What are the limitations of current vulnerability assessment tools in predicting future attack trends?</p> <p>RQ2: To what extent can machine learning models improve vulnerability prioritization and remediation by forecasting attack patterns?</p> <p>RQ3: How effective is the proposed ML model compared to traditional VA methods in predicting vulnerability susceptibility?</p>	<p>H1: Machine Learning algorithms can enhance the predictive capabilities of vulnerability assessments.</p> <p>H2: Random Forest (RF) and Neural Network Regression (NNR) are suitable models for forecasting vulnerability susceptibility.</p>	<ol style="list-style-type: none"> To design and propose a machine learning model for predicting cyberattack susceptibility using vulnerability Assessment data. To develop and implement a supervised ML-based predictive system using NVD + ExploitDB + Tenable datasets. To evaluate the accuracy and performance of the model in forecasting potential vulnerabilities. 	<ul style="list-style-type: none"> Data: Vulnerability from NVD CVE vulnerability, exploitDB, Tenable datasets. Tools: Orange Data Mining, Python (Jupyter Notebook). Models: Random Forest (classification), Neural Network Regression (regression). Evaluation: Accuracy, Precision, Recall, F1-score, MAE, RMSE. 	<p>The study enhances the cybersecurity posture of Malaysian banks by shifting from reactive to predictive vulnerability management. It supports AI-driven, enables faster remediation prioritization.</p>