Model Research

MODEL RESEARCH BY THREAT TYPE

| Threat Type | Dataset | Problem Type | Recommende d Models | Key Features / Inputs | Notes |
|-------------------------------|-----------------------------|-------------------------------------|---|--|---|
| Malware Classificatio n | EMBER | Tabular (binary / multiclass) | XGBoost, LightGBM, Autoencoder | 2,381 static PE features (entropy, imports, sections, strings) | Tree models handle sparse + non-linear numeric features well. |
| Phishing URL Detection | UCI Phishing Websites | Tabular (binary) | Random Forest, | Lexical + host features (URL length, HTTPS, @ symbol, IP in URL) | Feature importance interpretable; can also use ANN for non- linearities. |
| Spam / Malicious Email | Enron Email | NLP (binary / multi- class) | Naive Bayes (baseline), TF- IDF + SVM, LSTM / BERT | Subject, body text, headers | Classical NB for baseline transformer (BERT) for context semantics. |

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| Static Code Vulnerabilit y | Juliet Test Suite | Source Code analysis | CodeBERT / | Tokenized AST or code embedding s | GNN captures data flows; CodeBERT for semantic understandin g. |
| Malware Image Classificatio n | Malimg | Image classificatio n | CNN (ResNet, EfficientNet) | | Use transfer learning for fast training on GPU (Colab). |
| Intrusion / DDoS / Port Scan | CICIDS201 7, CSE-CIC- IDS2018 | Network flow classificatio n | XGBoost, LSTM, 1-D | flow features (bytes, pkts. | Tabular ML for baseline → deep sequence for time features. |
| IoT / Botnet Threats | ВоТ-ІоТ | Network traffic classificatio n | LSTM, GRU, CNN-LSTM, Autoencoder | Time- series IoT features (packet rates, flows) | Capture temporal patterns; handle class imbalance. |
| Network Anomaly | UNSW- NB15 | Tabular binary / multi-class | Isolation Forest, Autoencoder, | 49 network features | Try unsupervised first → |

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| | | | Forest | • | supervised with labels. |
| Real-Time Threat Intel | AlienVault OTX | Stream matching / ranking | lookup + ML | reputation, frequency, | Combine threat feed scores + ML confidence. |

Model Families Overview

☐ Classical ML (fast baselines)

- Random Forest / XGBoost / LightGBM: Best for tabular network or static
 PE data.
- Logistic Regression / SVM: Good interpretable baseline for phishing URLs.

NLP Models

- TF-IDF + SVM / Naive Bayes for spam.
- Transformer Models (BERT, RoBERTa) fine-tuned on email or phishing text.

☐ Deep Learning Models

- **CNN / ResNet / EfficientNet** malware image classification.
- LSTM / GRU / CNN-LSTM sequential network traffic.
- Autoencoder / Variational AE unsupervised anomaly detection.
- Graph Neural Networks (GNN) IP-domain-file relationship graphs.

Preprocessing & Feature Design

| Data Type | Steps | Typical Features | |
|------------------|---|--|--|
| Network flows | Standardize numeric features (0–1 scaler), encode categoricals (proto, service) | bytes, packets, duration, ports, ratios, entropy | |
| URLs | Extract lexical tokens, domain age, length, special chars | binary flags, lengths, host entropy | |
| Emails | Clean text, tokenize, TF-IDF / word embeddings | word freqs, topic vectors | |
| Code | Tokenize AST / CFG graphs | function calls, flows, data dependencies | |
| Images | Normalize pixels (0–1) / resize | image arrays | |