Appendix:

A.1 Data Extraction form

Table 1. Data type and item extracted from each study

| Data Type | ID | Data Item | Description |
|-----------|---------|---|---|
| Context | D1-D7 | Title, author, venue, publication year, publisher, summary, open challenges | Title, author, venue, publication year, publisher, summary including aim, strength, and weakness of the study and Open challenges to be resolved in future. |
| RQ1 | D8-D9 | Features, Feature Engineering Method | Features and feature engineering method: automatic or manual used to implement NLP-based HIDS |
| | D10-D12 | Learning Type, Classifier Type, Detection technique | Type of Learning method, Classifier type (e.g., Base, Ensemble), detection technique used for intrusion detection |
| | D13-D14 | HIDS type, Attack detection/classification | Type of HIDS (misuse, anomaly), attack detection (e.g., benign, malicious) or classification (detect specific attack) |
| RQ2 | D15 | Attacks | Attacks that are targeted to be detected |
| RQ3 | D16-D17 | Data Source, Dataset | Data source or dataset used for training or testing HIDS |
| RQ4 | D18 | Evaluation Metric | Metrics used for evaluating HIDS |

A.2 Feature types with mapped studies

Table 2. Feature types used in NLP-based HIDS with mapped studies

| Feature Type | Study Ref |
|------------------------------------|--|
| Statistical (22) | \$1, \$2, \$3, \$6, \$7, \$14, \$20, \$27, \$32, \$33, \$36, \$37, \$40, \$41, \$42, \$49, \$52, \$65, \$80, \$81, \$86, \$87 |
| Contextual (49) | \$4, \$5, \$8, \$9, \$10, \$11, \$13, \$15, \$16, \$17, \$19, \$21, \$22, \$24, \$25, \$26, \$28, \$29, \$34, \$35, \$39, \$43, \$44, \$45, \$47, \$50, \$54, \$58, \$59, \$60, \$62, \$63, \$64, \$67, \$68, \$70, \$71, \$72, \$73, \$74, \$76, \$79, \$82, \$84, \$88, \$93, \$94, \$95, \$97 |
| Attribute (1) | S18 |
| Temporal (3) | S12, S55, S66 |
| Statistical+Contextual (7) | S46, S61, S77, S78, S85, S96, S98, |
| Statistical+Attribute (4) | S30, S48, S53, S99 |
| Statistical+Attribute+Temporal (3) | S57, S75, S91 |
| Contextual+Attribute (4) | S23, S31, S51, S56 |
| Temporal+Contextual (3) | S38, S69, S83 |

| Statistical+Contextual+Attribute+Temporal | S89, S90, S92 |
|---|---------------|
| (3) | |

A.3 Detection techniques categories, sub-categories, instances mapped with reviewed studies

Table 3. Detection techniques categories, sub-categories, and instances mapped with the reviewed studies

| Detection Category | Sub-category | Technique Instances | Study Ref |
|--|--|--|--|
| Traditional ML (67) S1, S2, S3, S5, S6, S7, S11, S14, S18, S19, S21, S22, S23, | Bayesian (16) S3, S19, S21, S23, S30, S31, S32, S45, | Naive Bayes (6) | S19, S32, S60, S62, S65, S91 |
| \$24, \$25, \$26, \$27, \$28, \$30, | S48, S60, S62, S65, S69, S70, S91, S99 | BernoulliNB | S65 |
| S31, S32, S33, S34, S35, S36, S37, S39, S40, S41, S42, S44, | | MultinomialNB | S66 |
| S45, S46, S48, S52, S53, S57, | | GaussianNB | S65 |
| S59, S60, S61, S62, S63, S64, | | GaussianProcessClassifier | S65 |
| \$65, \$69, \$70, \$71, \$72, \$75, \$77, \$78, \$79, \$84, \$85, \$86, | | Gaussian Mixture Models (GMM) | S30, S48, S99 |
| \$87, \$88, \$89, \$90, \$91, \$92, | | ComplementNB | S65 |
| S93, S94, S95, S96, S97, S99 | | SC2/SC2.2/Markov-Bayes 5 | S21, S23, S31, S45, S70 |
| | | time Bayesian networks (CTBN) | S69 |
| | | Multi-variable Naïve Bayesian (MNB) | S3 |
| | Rule system (10) S3, S6, S7, S19, S53, S59, S60, S62, S65, S92 ("Decision Tree (8) S3, S6, S19, S53, S60, S62, S65, S92" "Rule system (4) S7, S59, S60, S62") Instance-based (13) S2, S3, S24, S32, S33, S36, S41, S52, S62, S65, S85, S87, S99 | ExtraTreeClassifier | S65 |
| | | Decision Tree/C4.5/C5 | S3, S6, S19, S53, S60, S62, S65, S92 |
| | | PART | S60 |
| | | RIPPER | S60, S62 |
| | | OneR | S62 |
| | | ZeroR | S62 |
| | | Rough Set Classification (RSC) | S7, S59 |
| | | K-Nearest Neighbors (KNN) | S2, S3, S24, S32, S33, S36, S41, S52, S62, S65, S85, S87, S99 |
| | | K-furthest neighbors (KFN) | S85 |
| | | K-centers | S87 |
| | Ensemble (12) S1, S3, S5, S21, S60, | Random Forest (RF) | S3, S60, S64, S65, S75, S94 |

| S63, S64, S65, S75, | Isolation Forest (IF) | S1, S5, S63 |
|---|--|---|
| S89, S90, S94 | ExtraTreesClassifier | S65 |
| | AdaBoost | S21. S65 |
| | Bagging Classifier | S65 |
| | GradientBoostingClassifier | S65 |
| | XGBoost | S65, S89, S90 |
| Statistical model (15) S11, S22, S25, S26, S28, S35, S39, S42, | Clustered Markov Networks (CMN) /CMN with Outlying Subspace (CMN-OS) | S42 |
| S44, S46, S53, S79, S88, S95, S96 | Markov chain | |
| 300, 393, 390 | CRF (conditional random fields) | S35 |
| | HMM/I-HMM | S11, S22, S25, S26, S28, S39, S44, S46, S53, S79, S88, S95, S96 |
| Clustering (12) S23, S24, S31, S36, | Harmony Search based K- means clustering | S72 |
| S42, S48, S62, S72, S77, S84, S93, S99 | k-means (9) | \$24, \$36, \$42, \$48, \$62, \$77, \$84, \$93, \$99 |
| | distance/RE based | S23, 31 |
| | Fuzzy clustering | S24 |
| Support Vector Machine (24) \$1, \$2, \$3, \$5, \$6, \$19, \$30, \$32, \$33, \$37, \$46, \$48, \$57, \$60, \$61, \$62, \$63, \$64, \$71, \$85, \$86, \$91, \$92, \$96 | Support Vector Machine (SVM) (10) One Class SVM (OCSVM) (12) | \$2, \$3, \$6, \$19, \$32, \$33, \$60, \$62, \$64, \$71 \$1, \$5, \$30, \$37, \$46, \$48, \$61, \$63, \$85, |
| | CMO (Sequential Minimal | \$91, \$92, \$96 \$62 |
| | SMO (Sequential Minimal Optimization) | |
| 277. (2) | SVDD | S57, S86 |
| NN (9) S3, S6, S14, S19, S27, S32, S34, S40, S97 | Multi-layer Perceptron (MLP)/ NN 8 | S3, S14, S19, S6, S32, S34 |
| | extreme learning machine (ELM) 4 | S27, S32, S40, S97 |
| | Logistic reg | S2, S65 |
| | axis aligned bounding box | S87 |

| | Miscellaneous (6) S1, S2, S18, S42, S65, S87 | Optimization Algorithm Based on Bee Stinging (OABBS) | S18 |
|--|---|--|--|
| | | Label Propagation Algorithm (LPA) | S42 |
| | | Calibrated Classifier CV | S65 |
| | | Linear Discriminant Analysis | S65 |
| | | Quadratic Discriminant Analysis | S65 |
| | | Local Outlier Factor (LOF) | S1 |
| DL (22) S4, S5, S8, S10, S11, S15, S17, S58, S63, S64, S65, S67, S68, S71, S73, S74, S78, S80, | seq2seq Language modeling (5) S5, S63, S64, S73, S94 | RNN/ RNN-VED /LSTM/GRU /BiLSTM/CuDNNLSTM 5 | S5 , S63, S64, S73, S94 |
| S82, S93, S94, S98 | Deep Neural network (18) S4, S8, S10, S11, S17, S58, S64, S65, S68, S71, S74, S78, S80, S82, S93, S94, S98 | Deep Multi-layer Perceptron/DNN | S65, S78, S98 |
| | | CNN/FCN/TCN(temporal convolutional neural network) (8) | S4, S8, S10, S17, S64, S71, S82, S94 |
| | | RNN/LSTM/GRU (8) | S4, S10, S11, S58, S64, S68, S74, S93 |
| | | Autoencoder /Variational Autoencoder (2) | S58, S93 |
| | | DBN (Deep belief network) | S80 |
| | Ensemble NN (3) S4, S67, S15 | LSTM-FCN, GRU-FCN | S4 |
| | | CNN-LSTM, CNN-GRU, CNN-LSTM-NN | S4, S67, S15 |
| Rule-based (27) S9, S12, S13, S16, S20, S25, | Semantic Ontology (4) | | S13, S29, S43, S55 |
| \$26, \$29, \$38, \$43, \$46, \$47, \$49, \$50, \$51, \$54, \$55, \$56, \$59, \$66, \$76, \$77, \$81, \$83, \$85, \$96, \$98 | Model/language-based (11) | | \$26, \$38, \$50, \$77, \$9, \$20, \$47, \$51, \$54, \$59, \$81 |
| | Sequence based (12) | | \$12, \$16, \$25, \$46, \$56, \$96, \$98, \$26, \$76, \$85, \$49, \$66, \$83 |

A.4 Data source, availability, dataset type, and instance mapped with reviewed studies

Table 4. Data source, availability, dataset type, and instance mapped with reviewed studies

| Data Source | Availability | Dataset Type | Instance | Study Ref |
|----------------|--------------|-----------------|------------|---|
| Sys | public | Real | AWSCTD | S4, S10 |
| call | | Sim | UNM | \$12, \$16, \$21, \$22, \$24, \$25, \$34, \$35, \$38, \$39, \$40, \$45, \$51, \$54, \$70, \$72, \$76, \$77, \$81, \$83, \$84, \$86, \$88, \$98 |
| | | | Firefox DS | S25, S38 |
| | | | FIT-UTK | S51 |
| | | Hyb | ADFA-LD | \$1, \$2, \$3, \$5, \$6, \$7, \$8, \$11, \$14, \$15, \$16, \$17, \$19, \$25, \$26, \$27, \$32, \$33, \$36, \$37, \$40, \$41, \$46, \$52, \$58, \$60, \$61, \$62, \$63, \$64, \$65, \$67, \$68, \$71, \$73, \$74, \$78, \$79, \$80, \$82, \$85, \$87, \$89, \$90, \$92, \$93, \$94, \$95, \$98 |
| | | | ADFA-WD | S1, S6, S21, S62, S78, S92 |
| | | | CANALI-WD | S46, S79, S96 |
| | private | Real | Customized | S42 |
| | | Sim | Customized | S22, S26, S28, S39, S44, S49, S50, S58, S59 |
| Audit | public | Real | Vergina | S30, S48, S99 |
| | | | thmmy | S30, S99 |
| | | | www_ee | S30, S99 |
| | | | PUS | S83 |
| | | Sim | DARPA | S23, S31, S32, S40, S51, S57, S69 |
| | | | NGIDS-DS | S1, S17 |
| | private | Real | Customized | S13, S29, S53 |
| | | Sim | Customized | S9, S47, S55, S56, S57 |
| | | Hyb | Customized | S75 |
| Sys | private | Real | Customized | S20, S91 |
| log | | Sim | Customized | S18, S43 |

A.5 Evaluation Metrics of intrusion detection with mapped studies

Table 5. Evaluation Metrics of intrusion detection with mapped studies

| | <u> </u> | |
|--------------------------|---|--|
| Detection Performance | Detection Rate (Recall, detection accuracy, TPR, true positive rate) | S2, S3, S4, S5, S6, S7, S8, S11, S12, S15, S17, S18, S19, S20, S21, S22, S23, S25, S27, S30, S31, S32, S33, S34, S35, S36, S37, S38, S39, S40, S41, S42, , S44, S45, S46, S47, S48, S52, S53, S57, S60, S61, S62, S63, S67, S68, S70, S71, S72, S73, S74, S76, S77, S79, S80, S81, S82, S85, S86, S87, S89, S90, S91, S92, S93, S98, S99 |
| | False Alarm Rate (FAR, FPR, false positive rate) | \$4, \$5, \$6, \$8, \$11, \$12, \$15, \$17, \$21, \$22, \$23, \$25, \$26, \$27, \$30, \$31, \$32, \$33, \$34, \$35, \$36, \$37, \$38, \$39, \$40, \$41, \$42, \$45, \$46, \$47, \$48, \$52, \$53, \$57, \$58, \$60, \$61, \$62, \$63, \$68, \$70, \$72, \$73, \$74, \$75, \$76, \$77, \$79, \$80, \$81, \$83, \$84, \$86, \$87, \$92, \$93, \$98, \$99 |
| | False Alarm Rate defined by S85 | S85 |
| | Receiver Operating Characteristic curve (ROC)/ area under the curve (AUC) | \$1, \$2, \$5, \$8, \$10, \$11, \$12, \$14, \$15, \$16, \$21, \$22, \$30, \$33, \$36, \$37, \$39, \$40, \$41, \$46, \$51, \$52, \$54, \$58, \$60, \$61, \$62, \$63, \$64, \$65, \$69, \$71, \$73, \$79, \$80, \$82, \$87, \$88, \$89, \$90, \$91, \$93, \$95, \$96, \$98 |
| | False Negative Rate (FNR) = Missing Rate | S4, S32, S33, S42, S52, S60, S63, S76, S86, S92 |
| | True Negative Rate (TNR) | S42, S52, S60, S83 |
| | Confusion matrix | S4, S21, S60, S62, S89, S90, S92 |
| | Classification Accuracy or Classification rate (CR) | S3, S4, S8, S10, S11, S18, S21, S23, S31, S45, S47, S49, S52, S53, S62, S63, S67, S70, S71, S78, S80, S82, S91, S92, S98 |
| | Precision | S3, S4, S5, S6, S7, S18, S19, S21, S50, S52, S53, S60, S62, S63, S67, S80, S82, S89, S90, S91, S92 |
| | F1, F2(S42) | S3, S4, S5, S6, S7, S18, S19, S21, S24, S42 (F2), S52, S60, S62, S65, S67, S80, S82, S91, S92 |
| | Classification Error | S4, S21 |
| | Matthews Correlation Coefficient (MCC) | S4 |
| Computation Performance | Time (training time/ testing time/ execution time) | S2, S3, S4, S5, S9, S12, S18, S22, S25, S28, S44, S49, S50, S51, S58, S63, S79, S80, S83, S85, S95, S96, S98 |
| | Resource utilization (Storage and computational overhead) | S22, S28, S50, S51 |