## **Online 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)	S1, S2, S3, S6, S7, S14, S20, S27, S32, S33, S36,
	S37, S40, S41, S42, S49, S52, S65, S80, S81,
	S86, S87
Contextual (55)	S4, S5, S8, S9, S10, S11, S12, S13, S15, S16,
	S17, S19, S21, S22, S24, S25, S26, S28, S29,
	S34, S35, S38, S39, S43, S44, S45, S47, S50,
	S54, S55, S58, S59, S60, S62, S63, S64, S66,
	S67, S68, S69, S70, S71, S72, S73, S74, S76,
	S79, S82, S83, S84, S88, S93, S94, S95, S97
Attribute (1)	S18
Statistical+Contextual (7)	S46, S61, S77, S78, S85, S96, S98,
Statistical+Attribute (7)	S30, S48, S53, S57, S75, S91, S99
Contextual+Attribute (4)	S23, S31, S51, S56
Statistical+Contextual+Attribute (3)	S89, S90, S92

# 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)	\$19, \$32, \$60, \$62, \$65, \$91
S24, S25, S26, S27, S28, S30,	\$48, \$60, \$62, \$65, \$69, \$70, \$91, \$99	BernoulliNB	S65
S31, S32, S33, S34, S35, S36,		MultinomialNB	S66
S37, S39, S40, S41, S42, S44, S45, S46, S48, S52, S53, S57,		GaussianNB	S65
\$59, \$60, \$61, \$62, \$63, \$64,		GaussianProcessClassifier	S65
\$65, \$69, \$70, \$71, \$72, \$75, \$77, \$78, \$79, \$84, \$85, \$86,		Gaussian Mixture Models (GMM)	S30, S48, S99
S87, S88, S89, S90, S91, S92,		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
	Instance-based (13) S2, S3, S24, S32, S33, S36, S41, S52, S62, S65, S85, S87, S99	K-Nearest Neighbors (KNN)	\$2, \$3, \$24, \$32, \$33, \$36, \$41, \$52, \$62, \$65, \$85, \$87, \$99
		K-furthest neighbors (KFN)	S85
	Ensemble (12) S1, S3, S5, S21, S60, S63, S64, S65, S75,	K-centers	S87
		Random Forest (RF)	S3, S60, S64, S65, S75, S94
	S89, S90, S94	Isolation Forest (IF)	S1, S5, S63
		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, S44, S46, S53, S79, S88, S95, S96	Clustered Markov Networks (CMN) /CMN with Outlying Subspace (CMN-OS)	S42
		Markov chain	
		CRF (conditional random fields)	S35
		HMM/I-HMM	S11, S22, S25, S26,

			\$28, \$39, \$44, \$46, \$53, \$79, \$88, \$95, \$96
	Clustering (12) S23, S24, S31, S36, S42, S48, S62, S72, S77, S84, S93, S99	Harmony Search based K- means clustering	S72
		k-means (9)	\$24, \$36, \$42, \$48, \$62, \$77, \$84, \$93, \$99
		distance/RE based	S23, 31
		Fuzzy clustering	S24
	ML-based Rule system	ExtraTreeClassifier	S65
	(10) S3, S6, S7, S19, S53, S59, S60, S62, S65, S92	Decision Tree/C4.5/C5	\$3, \$6, \$19, \$53, \$60, \$62, \$65, \$92
	("Decision Tree (8)	PART	S60
	\$3, \$6, \$19, \$53, \$60, \$62, \$65, \$92"	RIPPER	S60, S62
	"Rule system (4)	OneR	S62
	S7, S59, S60, S62")	ZeroR	S62
		Rough Set Classification (RSC)	S7, S59
	Support Vector Machine (24) S1, S2, S3, S5, S6, S19, S30, S32, S33, S37, S46, S48, S57, S60, S61, S62, S63, S64, S71, S85, S86, S91, S92, S96	Bin/multi class SVM (10)  One Class SVM (OCSVM)	\$2, \$3, \$6, \$19, \$32, \$33, \$60, \$62, \$64, \$71 \$1, \$5, \$30,
		(12)	\$37, \$46, \$48, \$61, \$63, \$85, \$91, \$92, \$96
		SMO (Sequential Minimal Optimization)	S62
		SVDD	S57, S86
	NN (9) S3, S6, S14, S19, S27, S32, S34, S40, S97 Miscellaneous (6) S1, S2, S18, S42, S65,	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
	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)	Deep Multi-layer Perceptron/DNN	S65, S78, S98
	S4, S8, S10, S11, S17, S58, S64, S65, S68, S71, S74, S78, S80, S82, S93, S94, S98  Ensemble NN (3) S4, S67, S15	CNN/FCN/TCN(temporal convolutional neural network) (8)	\$4, \$8, \$10, \$17, \$64, \$71, \$82, \$94
		RNN/LSTM/GRU (8)	S4, S10, S11, S58, S64, S68, S74, S93
		Autoencoder /Variational Autoencoder (2)	S58, S93
		DBN (Deep belief network)	S80
		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)		S12, S16, S25, S46, S56, S96, S98, S26, S76, S85, S49, S66, S83

# 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

Datastian	Detection Data	C2 C2 C4 C5 C6 C7 C0 C11 C12 C15 C17 C10 C10 C20 C21 C22
Detection Performance	Detection Rate (Recall, detection accuracy, TPR, true positive rate)	\$2, \$3, \$4, \$5, \$6, \$7, \$8, \$11, \$12, \$15, \$17, \$18, \$19, \$20, \$21, \$22, \$23, \$25, \$27, \$30, \$31, \$32, \$33, \$34, \$35, \$36, \$37, \$38, \$39, \$40, \$41, \$42, \$44, \$45, \$46, \$47, \$48, \$52, \$53, \$57, \$60, \$61, \$62, \$63, \$67, \$68, \$70, \$71, \$72, \$73, \$74, \$76, \$77, \$79, \$80, \$81, \$82, \$85, \$86, \$87, \$89, \$90, \$91, \$92, \$93, \$98, \$99
	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	\$3, \$4, \$5, \$6, \$7, \$18, \$19, \$21, \$50, \$52, \$53, \$60, \$62, \$63, \$67, \$80, \$82, \$89, \$90, \$91, \$92
	F1, F2(S42)	\$3, \$4, \$5, \$6, \$7, \$18, \$19, \$21, \$24, \$42 (F2), \$52, \$60, \$62, \$65, \$67, \$80, \$82, \$91, \$92
	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