

# Limin Yang

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## EDUCATION

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University of Illinois (UIUC), Ph.D. in Computer Science, advisor: Gang Wang	Aug. 2019 – May 2023
Virginia Tech, Ph.D. in Computer Science, advisor: Gang Wang	Aug. 2018 – Jul. 2019
East China Normal University, M.S. Study in Computer Science	Sep. 2015 – Jun. 2018
East China Normal University, B.Eng. in Computer Science	Sep. 2011 – Jun. 2015

## INTERNSHIPS

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<b>IBM Research, Visiting Scholar (Research Intern), Hybrid Cloud Team</b>	May 2022 – Aug. 2022
<ul style="list-style-type: none"><li>Cleaned a noisy real-world network intrusion dataset (25 million traffic/day) from National Supercomputing Center</li><li>Semi-automatically labeled the dataset, defined 65 features for network logs, and summarized 279 security incidents</li><li>Built anomaly detection models (per host) with LSTM autoencoder and applied on 2 past real-world attacks</li></ul>	
<b>TikTok, Security Engineering Intern, Threat XDR Team</b>	May 2021 – Aug. 2021
<ul style="list-style-type: none"><li><b>Lark Email Spam Rule System:</b> detected 5,000 more spam/week by adding 25 patterns and 20 rules</li><li><b>Protected 50 client companies</b> and 1 million emails/week by allowlisting 300 domains with semi-automation</li><li>Clustered similar emails based on user actions and increased the size of user labeled ground-truth by 100%</li><li>Contributed to 2 urgent incidents response: a vendor offline and a spam campaign</li></ul>	

## SELECTED PROJECTS

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<b>Selective ML Backdoor Attack</b>   Submitted to IEEE S&P'23	Sep. 2020 – Jan. 2022
<ul style="list-style-type: none"><li>Proposed a new backdoor attack against Android malware classifiers: only a specific malware family is misclassified</li><li>Achieved 80%–98% attack success rates by alternate optimization with a customized loss function</li><li>Defeated 4 recent backdoor defenses while traditional backdoor cannot bypass the detections</li></ul>	
<b>Concept Drift Detect and Explain</b>   Published in <b>USENIX Security'21</b>	Jun. 2019 – Jun. 2020
<ul style="list-style-type: none"><li>Leveraged contrastive learning and autoencoder to detect concept drift samples from previously unseen classes</li><li><b>Explainable AI:</b> proposed a novel distance-based explanation to find 45/1000 features making a sample outlier</li><li>Increased detection rate to <math>F_1 = 96\%</math> versus state-of-the-art (<math>F_1 \leq 80\%</math>) on malware and network datasets</li><li>Identified 161/165 unseen families on a company Blue Hexagon's Windows PE malware database</li></ul>	
<b>VirusTotal Reliability</b>   Published in <b>IMC'19</b> and <b>USENIX Security'20</b>	Feb. 2019 – Nov. 2019
<ul style="list-style-type: none"><li>Controlled 66 phishing sites to measure the label inconsistencies and dynamics between vendors and VirusTotal</li><li>Measured the label dynamics of 14,000+ PE malware from 65 vendors via daily snapshots over one year</li><li>Offered insights and suggestions on a more proper use of VirusTotal (received <b>100+ citations</b> in total)</li></ul>	

## SELECTED PUBLICATIONS (CITATIONS: 300+, H-INDEX: 9)

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- [Submit to **USENIX Security'23**] **1st author**. *Title anonymized for double-blind submission.*
- [Submit to **IEEE S&P'23**] **1st author**. “Jigsaw Puzzle: Selective Backdoor Attack to Subvert Malware Classifiers”.
- [**USENIX Security'21**] **1st author**. “CADE: Detecting and Explaining Concept Drift Samples for Security Applications”. Artifact Evaluated.
- [Deep Learning and Security'21] **1st author**. “BODMAS: An Open Dataset for Learning based Temporal Analysis of PE Malware”. Dataset requested by **62 institutions across 22 countries**.
- [**USENIX Security'20**] 3rd author. “Measuring and Modeling the Label Dynamics of Online Anti-Malware Engines”.
- [**IMC'19**] 2nd author. “Opening the Blackbox of VirusTotal: Analyzing Online Phishing Scan Engines”.
- [**USENIX Security'18**] 3rd author. “Understanding the Reproducibility of Crowd-reported Security Vulnerabilities.”.

## TECHNICAL SKILLS

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<b>Languages:</b>	Python, C++, C, Shell, SQL (Postgres, Hive), Ruby
<b>Security:</b>	ML-based malware detection, Network IDS, Spam, Phishing, Bug reproduction
<b>Deep Learning:</b>	Keras, Tensorflow, PyTorch
<b>Frameworks:</b>	Hadoop, PySpark, MongoDB, Flask, Scrapy, Alexa Skills, Elasticsearch, Ruby on Rails
<b>Developer Tools:</b>	Linux, Git, VS Code, Jupyter Notebook, tmux, AWS, Vim, Docker, VirusTotal, Nmap
<b>Libraries:</b>	Scikit-learn, LightGBM, NumPy, pandas, Matplotlib