







Please help me find five research papers to add to my citation. Topic: Revolutionizing Network Security: Stacked Generalization for Malicious Traffic Detection. Prioritize open-source databases.

Thought for 4 seconds >

Here are five solid, open-access research papers focusing on stacked generalization—or stacking ensembles —for malicious traffic detection and network security. Each is from reputable sources and contributes unique insights relevant to your topic:

1. Malicious Traffic Detection in IoT and Local Networks Using Stacked Ensemble Classifier

- Authors: Pubudu et al. (2022)
- Highlights: Proposes "Extra Boosting Forest" (stacked ensemble combining Extra Trees, Gradient Boosting, and Random Forest) applied to merged IoT and local traffic datasets (UNSW-NB15 and IoTID20). Achieved ~98.5% accuracy using PCA for feature reduction researchoutput.csu.edu.au +10
- Why include it: Directly aligns with stacking methodologies on relevant datasets; open access.

2. Web Attacks Detection Using Stacked Generalization Ensemble for LSTMs

- Authors: Lamrani Alaoui & Nf aoui (2023)
- Highlights: Utilizes Word2Vec embeddings and stacked LSTM ensembles for HTTP web request classification (CSIC-2010 dataset). Offers strong performance and efficiency with a deep-learning stacking approach researchgate.net +1.
- Why include: Extends stacking to deep learning architectures—valuable for modern, application-layer malicious traffic detection.

3. Effective Network Intrusion Detection Using Stacking-Based Ensemble

- Authors: Alamri et al./Springer (2023)
- Highlights: Demonstrates ~98.24% weighted F1-score on a new CIPMAIDS2023-1 dataset and CICIDS2017 using stacking (base models: SVM, KNN, RF; meta: XGBoost) linkspringer.com +15
- Why include: Highlights the adaptability of stacking across diverse datasets and high performance with classical ML approaches.

4. SEDAT: A Stacked Ensemble Learning-Based Detection Model for Multiscale Network Attacks

- Authors: Feng et al. (2024)
- Highlights: Introduces a stacked ensemble combining autoencoders and Random Forests to detect
 multiscale (light/medium/heavy, concealed) attacks. Demonstrates solid accuracy enhancement versus
 baselines on CIC-IDS2017 and UNSW-NB15 [mdpi.com +4].
- Why include: Unique focus on multiscale attack patterns and stacked learning—including unsupervised components—ideal for detecting sophisticated vious network behavior.

5. A Stacking Ensemble for Network Intrusion Detection Using Real-Time

- Authors: Jain & Kumar (2020, Wiley)
- **Highlights**: Applies stacking ensembles in a real-time intrusion detection scenario; meta-learning improves base classifier predictions on real-world network flows onlinelibrary.wiley.com +15.
- Why include: Offers empirical evidence for stacking's efficacy in practical, real-time NIDS environments.

% How to Use These in Your Work

Paper	Contribution	Relevance
Pubudu et al. (2022)	Ensemble of tree-based models	Real-world datasets, high accuracy
Lamrani & Nf aoui (2023)	LSTM stacking for HTTP web attacks	Deep learning adaptation
Alamri et al. (2023)	Stacking with XGBoost meta learner	Data generalization and meta-model design
Feng et al. (2024)	Autoencoder-enabled stacking for multiscale attacks	Unsup + sup ensemble structure
Jain & Kumar (2020)	Real-time stacked NIDS	Practical deployment validation