

| Model Type | Dataset(s) Used | Accuracy/F1 | AUC-ROC | Main Strengths/Use Cases |
|--------------------------------------|---------------------------|------------------------|---------------------|--|
| Multi-scale CNN | CIC-IDS, UNSW, Edge IIoT | High (≥ 0.90) | High-90s | Excels in spatial features; often paired with RNN |
| LSTM, BiGRU RNN | CIC-IDS, Edge IIoT | F1 ≈ 0.90 | High | Models sequential flow sessions; dual attention aids focus |
| Transformer Attention Models | CIC-IDS, Edge IIoT | High precision | High | Long-range dependency modeling, telemetry fusion |
| Graph/Hypergraph Encoders | CIC-IDS, UNSW, multi-set | Strong generalization | Not always reported | Encodes multi-view relations, boosts robustness |
| GANs (WGAN-GP, CGAN, hybrids) | CIC-IDS synthetic traffic | Not direct accuracy | Robustness | Traffic augmentation, adversarial robustness |
| Contrastive Self-Supervised Learning | Phishingweb, CIC-IDS | Improved F1, Proxy AUC | | Reduces label dependency, increases domain transferability |

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| Autoencoders/Isolation Forest | CIC-IDS, UNSW | High anomaly detection | AUC \approx 0.95 | Unsupervised normal behavior modeling |
| Ensemble Trees (RF, XGBoost) | Tabular (CIC-IDS, etc.) | Accuracy \approx 0.90 | Moderate | Baseline model, interpretable, useful for risk scoring |