

Supplementary Materials

Table 1. Characteristics of Sources of Evidence.

| First Author | Year | Publication Type | Study Type | Country | AI Technique | QKD Security Focus |
|-------------------|------|------------------------|----------------------------|--------------|---------------------------|---|
| H. A. Al-Mohammed | 2024 | Journal Article | Simulation-Based | Qatar | Autoencoder | Enhancing error correction and scalability for high-data-rate QKD environments. |
| W. Wang | 2020 | Thesis | Experimental & Theoretical | Canada | Neural Networks | Optimizing QKD parameters in free-space environments with atmospheric turbulence. |
| S. R. Sihare | 2024 | Journal Article | Experimental & Theoretical | India | Error Prediction | Improving error resilience against signal noise and imperfections in guided/unguided QKD. |
| K. Durak | 2022 | Journal Article | Experimental | Turkey | Deep Neural Network (DNN) | Detecting side-channel RF fingerprints for physical security of QKD. |
| M. Ahmadian | 2022 | Journal Article | Experimental | Spain | DNN for SOP Prediction | Reducing QBER and enhancing Key Exchange Rate in polarization encoded QKD. |
| M. Nozari | 2024 | Journal Article | Experimental & Theoretical | UAE & Turkey | DNN | Optimizing underwater QKD under intersymbol interference and background noise. |
| T. Johann | 2023 | Conference Proceedings | Simulation-Based | Germany | LSTM Neural Network | Optimizing routing and reducing denial-of-service rates in meshed QKD networks. |
| H. Zhang | 2022 | Journal Article | Simulation-Based | China | Autoencoder | Compensating for nonlinear distortion in high-speed CV-QKD. |
| R. M. Bommi | 2023 | Conference Proceedings | Experimental & Simulation | India | Various ML Techniques | Enhancing QKD resilience against noise and quantum attacks[4] . |

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|---------------|------|------------------------|----------------------------|--------------|----------------------------|--|
| P. Mehdizadeh | 2024 | Journal Article | Simulation-Based | Iran & Spain | KNN, NN, XGB | Optimizing frequency selection for QKD in multi-band optical transmission. |
| B. Colombier | 2023 | Journal Article | Experimental & Theoretical | France | Random Forest | Analyzing power consumption vulnerabilities relevant to QKD security. |
| J. Xu | 2024 | Journal Article | Experimental & Theoretical | China | Random Forest | Detecting device imperfections and eavesdropping threats in QKD. |
| J. Y. Liu | 2019 | Journal Article | Experimental | China | LSTM Neural Network | Real-time phase modulation stabilization in QKD. |
| Zi. Ren | 2021 | Journal Article | Experimental & Theoretical | China | RF, SVM, CNN | Dynamically selecting optimal QKD protocols based on security needs. |
| Y. Mao | 2020 | Journal Article | Experimental & Theoretical | China | Artificial Neural Network | Detecting quantum attacks in CV-QKD by analyzing pulse variations. |
| J. L. Kang | 2023 | Journal Article | Experimental & Theoretical | China | Various Neural Networks | Optimizing parameter generation rates in Twin-Field QKD[20] . |
| Y. Yi | 2021 | Conference Proceedings | Experimental & Theoretical | China | Improved Random Forest | Enhancing transmission probability and optimizing signal strength. |
| W. Wang | 2019 | Journal Article | Experimental & Theoretical | China | Feedforward Neural Network | Enhances QKD performance by optimizing parameters for secure and efficient key generation. |
| Hong-fu Chou | 2024 | Journal Article | Experimental & Theoretical | Luxembourg | GMM, Bayes Classifier | Detecting Trojan-horse attacks with risk-aware ML in QKD networks[21] . |
| Hua-Jian Ding | 2020 | Journal Article | Experimental & Theoretical | China | Random Forest | QKD parameter prediction to optimize secure key generation rates |

Table 2. Data Charting Form. Variables and Definitions.

| Variable | Definition/Description |
|---------------------------------|--|
| Study Details | Information about each study, including: |
| - First Author | Name of the first author or lead researcher in the study. |
| - Year | Year of publication. |
| - Publication Type | Type of publication (e.g., journal article, conference proceedings, thesis). |
| Machine Learning (ML) Technique | Type of ML algorithms used in the study (e.g., neural networks, random forests) and their purpose in QKD security. |
| QKD Security Applications | Security challenges addressed by ML, such as mitigating side-channel attacks, noise reduction, or error rate optimization. |
| Outcome Metrics | Metrics reported to measure effectiveness, including accuracy, response time, adaptability, and scalability. |
| Country of Study | Country where the research was conducted, typically based on author affiliations. |
| Study Type | Classification of the study approach, such as simulation-based, experimental, theoretical, or a combination. |

Table 3. Sample Data Charting Form Entry.

| Study Details | ML Technique | QKD Security Applications | Outcome Metrics | Country | Study Type |
|---------------------------------|-----------------|--|---------------------------------------|---------|----------------------------|
| Author: Hasan Abbas Al-Mohammed | Autoencoder | Enhancing error correction in high-data-rate QKD | Accuracy > 99%, scalability | Qatar | Simulation-Based |
| Year: 2024 | Neural Networks | Optimizing QKD parameters in free-space | Adaptability to environmental factors | Canada | Experimental & Theoretical |

Table 4. Outcomes of Individual Sources of Evidence.

| First Author & Year | ML Technique Applied | QKD Security Challenge Addressed | Outcome Metric(s) | Key Findings | Limitations | Future Work |
|-------------------------------|--|--|---|--|---|--|
| Hasan Abbas Al-Mohammed, 2024 | Autoencoder for nonlinear compensation | Error correction, scalability, high data rates, resource optimization in QKD systems | Prediction accuracy > 99%, MSE 17% | Improved QBER and key rate prediction accuracy with scalability benefits | Memory limitations; high computational needs | Integrate with decoy-state methods, apply to different QKD protocols |
| Wenyuan Wang, 2020 | Neural network for parameter | Atmospheric turbulence, channel asymmetry in | 2-4 orders of magnitude speedup in parameter optimization | Enhanced efficiency and key rate under asymmetric | Limited real-world testing, dependency on simulation data | Explore multi-user QKD networks; extend to free-space QKD |

| | optimization | free-space QKD | | channel conditions | | implementations |
|------------------------|--|--|---|--|--|--|
| Shyam R. Sihare, 2024 | Error prediction model | Quantum error resilience, noise impact, signal processing in guided and unguided media | QER prediction accuracy | Improved system robustness through accurate error prediction | Reliance on theoretical models; need for experimental validation | Adaptive error correction techniques; ML-based error prediction for system adjustments |
| Kadir Durak, 2022 | Deep Neural Network (DNN) | Side-channel attack through RF fingerprinting | Classification accuracy > 99% | Effective detection of RF fingerprints in QKD devices, highlighting side-channel vulnerabilities | Limited to lab settings; dependency on antenna gain and APD separation | Optimize for real-world settings; assess eavesdropping distances and antenna gain |
| Morteza Ahmadian, 2022 | DNN for State of Polarization (SOP) prediction | Environmental SOP fluctuations, high costs in polarization-encoded QKD | QBER < 0.5%, KER improvement up to 89% | Enhanced key rate and security by reducing SOP fluctuation effects | Performance drops in extreme environmental conditions; high reliance on SOP prediction | Optimize DNN architectures; expand to real-world and additional QKD protocols |
| Mostafa Nozari, 2024 | DNN for system parameter optimization | High QBER due to underwater conditions (ISI, background noise) | QBER improvement by 2 orders of magnitude | Real-time optimization for underwater QKD; improved performance under underwater noise | Limited by parameter estimation accuracy; complex underwater conditions | Extend to multi-hop QKD; explore noise reduction strategies |
| Tim Johann, 2023 | Long Short-Term Memory (LSTM) Neural Network | Key depletion and routing inefficiencies in QKD networks | Key store depletion reduction, latency reduction | Enhanced routing efficiency; reduced key depletion and denial-of-service risk | Simulated network basis limits generalizability to real-world QKD networks | Test on real-world networks; explore alternative ML models for higher accuracy |
| Hang Zhang, 2022 | Autoencoder for nonlinear | Nonlinear distortions in high-speed CV-QKD, | Excess noise reduction to 10^{-3} level, improved | Effective real-time nonlinear distortion | Dependency on neural network model | Real-time implementation, adapt to other |

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|------------------------|--|--|---|---|--|---|
| | compensation | affecting key rates | secure key rate | compensation in CV-QKD | accuracy; generalization issues across hardware | impairments, apply to experimental setups |
| Bommi R.M., 2023 | Deep Learning, Unsupervised Learning, Reinforcement Learning | Noise, eavesdropping, quantum attacks, and hardware imperfections in QKD systems | Key generation rate accuracy, eavesdropping detection rate | Enhanced efficiency and security in QKD through dynamic adjustment and optimization | High data requirements; vulnerability to adversarial attacks | Extend to other protocols; improve robustness in real-world conditions |
| Pouya Mehdizadeh, 2024 | KNN, NN, and XGB | Impact of classical traffic load and spectrum usage on QKD secure key rates | High accuracy in optimal frequency prediction | Improved frequency selection for QKD and classical coexistence | Limited classical communication rate and quantum distance | Explore larger datasets and real-world applications |
| Brice Colombier, 2023 | Random Forest for power analysis | Side-channel vulnerabilities in cryptographic systems affecting QKD | Success rate in message recovery | Demonstrated vulnerabilities in post-quantum cryptographic systems | Effectiveness depends on implementation and environmental factors | Test across broader datasets; optimize for lower computing complexity |
| Jiaxin Xu, 2024 | Random Forest | Detection of device imperfections and eavesdropping in QKD | 98% accuracy for imperfection and attack classification | High accuracy in real-time imperfection and attack detection | Potential misjudgments with 2% error rate; limited security strategy compatibility | Implement advanced strategies to reduce error rates; apply to other QKD protocols |
| Jing-Yang Liu, 2019 | Long Short-Term Memory (LSTM) Neural Network | Phase modulation inefficiencies in QKD systems | Improved transmission efficiency (duty ratio from 50% to 83%) | Effective real-time phase modulation control for QKD systems | Requires continuous model updates and monitoring for accuracy | Apply to large-scale QKD networks; optimize LSTM for diverse conditions |
| Wenyuan Wang, 2019 | Feedforward Neural Network | Parameter optimization inefficiencies in QKD for low-power devices | Secure key rate achievement (95-99% of optimal) | Enhanced efficiency in QKD parameter optimization | May require retraining for different conditions or protocols | Expand to additional optimization tasks, such as polarization control |
| Zi-Ang Ren, 2021 | RF, SVM, KNN, MNB, CNN | Real-time protocol selection for secure | RF accuracy 98%, high AUC | Enabled real-time protocol selection in | Limited to studied protocols; potential | Expand to more protocols and complex QKD networks |

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|---------------------|---|--|--|--|---|--|
| | | communication | | QKD for improved security and efficiency | inaccuracies with imbalanced data | |
| Yiyu Mao, 2020 | Artificial Neural Network (ANN) | Detection of quantum attacks in CVQKD systems | High precision and recall for attack detection accuracy | Effective real-time quantum attack detection in CVQKD | Slight reduction in secret key rate and transmission distance | Apply to additional attack types; improve model efficiency |
| Jia-Le Kang, 2023 | Neural Networks: BPNN, RBFNN, GRNN | Inefficiency in Twin-Field QKD parameter optimization | Key rates comparable to traditional methods, faster computation | Improved secure key generation rates through neural network-based parameter optimization | May require retraining for different QKD setups | Expand model for other protocols and large-scale networks |
| Yuling Yi, 2021 | Random Forest with data preprocessing | Signal strength and transmission probability optimization | High prediction accuracy for QKD parameters | Improved efficiency and key rate accuracy in QKD | Dependent on simulated data quality | Refine Random Forest for improved accuracy; test on diverse QKD protocols |
| Hua-Jian Ding, 2020 | Random Forest for parameter optimization | Real-time QKD parameter optimization for large-scale networks | 99% optimal key rate with reduced computation time | Enhanced QKD key rates with reduced computational demands | May vary across different QKD configurations | Extend to additional protocols and network conditions |
| Hong-fu Chou, 2024 | Gaussian Mixture Model (GMM) and Bayes Classifier | Detection of Trojan-horse attacks and time-variant vulnerabilities in QKD networks | Improved Trojan-horse attack detection and risk-aware thresholding | Enhanced trust in QKD networks by mitigating Trojan-horse attack risks through risk-aware ML | Model effectiveness varies with quantum channel conditions; quality dependent on empirical data | Explore risk-aware reinforcement learning for real-time detection and enhance detection mechanisms in QKD networks |