



Quantum Computing Applications Across Industrial Sectors

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Article Info:

DOI: 10.22399/ijcesn.4835
Received : 28 November 2025
Revised : 26 January 2026
Accepted : 28 January 2026

Keywords

Quantum Computing,
Enterprise Optimization,
Quantum Machine Learning,
Financial Risk Management,
Healthcare Simulation

Abstract:

Quantum computing serves as a revolutionary paradigm in computing, harnessing quantum mechanics principles such as quantum entanglement and superposition for optimization simulation, machine learning, and other problems that are a challenge to classical computer systems. The article will examine different usage opportunities in retail and supply chain operations, healthcare and life sciences domains, insurance operations, and financial service industries. Better computation can be achieved with quantum algorithms in discrete and continuous optimization, logistics network optimization, warehouse management, and product delivery optimized by quantum algorithms. Application domains in healthcare include simulation of molecules for medical breakthroughs, genomic analysis, and medical studies optimized by quantum eigenvalue algorithms and quantum eigensolver algorithms. The insurance industry portfolio targets catastrophic modeling, identification, and risk assessment optimized by quantum amplitude estimation algorithms and quantum machine learning algorithms. Financial service industries include derivative pricing, portfolio optimization, credit risk evaluation, and real-time risk calculation optimized in quantum algorithms. The path towards achieving quantum benefit in quantum technology involves using noisy intermediate quantum technology with hybrid quantum and classical designs and error management schemes, where integration opportunities for different industries arise from their common computation problem schemas in need of further scaling in hardware technology.

1. Introduction to Quantum Computing in Enterprise Software

Quantum computing varies from classical computation in its application of quantum mechanics principles, especially in the principles of superposition and entanglement, which make possible computations not achievable in classical systems. The mathematical model for quantum computation employs quantum gates that apply unitary transformations to states in a qubit in a manner that enables quantum algorithms to search spaces with an exponentially large dimensionality [1]. The theoretical model derived from quantum circuit theory shows that for some computations, such as search over an unstructured space, factorization of an integer, and simulation of a quantum system, a speedup proportional to an exponential or polynomial function of size relative to classical computation is achievable using a quantum computer. The nature of a multi-qubit state, which is basically a tensor product, gives

quantum computers an essential capability to encode a state space of exponentially large dimensionality in a manner suitable for solving optimization and machine learning problems common in business-oriented computations.

Current research in quantum machine learning has proven theoretical grounds for using quantum algorithms to improve conventional machine learning models in both supervised, unsupervised, and reinforcement learning settings [2]. Specifically, quantum algorithms in linear operations such as matrix inversion and eigendecomposition have proven computational superiority over their conventional machine learning counterparts in tasks reliant on such core operations. The application of quantum computing in artificial intelligence enables hybrid systems incorporating both classical and quantum systems, with quantum processing specialized in dealing with particular computer bottlenecks in computations, leaving classical systems to address associated data preprocessing, analysis, and control

algorithmic workflow processing. The implementation of quantum computing in enterprises demands a thorough evaluation of problem domains with a focus on domains with authentic potential gains in incorporating quantum computing over optimized classical processing algorithms. Domains with characteristic settings, including complex optimization problems, analysis in high-dimensional spaces, and simulation-driven processing, represent principal domains suitable for implementation with quantum computers for retail business operations, healthcare research studies, insurance risk analysis, and financial portfolio analysis, among others, where complexity in processing can render conventional processing inoperable.

2. Uses of Quantum Computers in Retailing and Supply Chain Management

Quantum computing provides solutions for basic optimization problems in retail and supply chain operations using advanced algorithms for handling discrete and continuous optimization mathematically modeled in logistics networks, inventory control models, and distribution planning tasks. The integration of quantum computing for solving large-scale optimization problems is based on decomposition methods, which divide complicated industrial problems into smaller quantum-computational versions assisted by classical optimization methods for handling coordination among different decomposed solutions [3]. Methods for solving supply chain network design with facility location, capacity constraint, and multiple demand requirements have been developed by using classical mixed-integer programming solvers in combination with approximate solutions from quantum algorithms for optimization in quantum computing. The quantum processing of supply chain network design uses a combination of quantum processors for searching in the solution space based on discrete decisions and classical methods for continuous decisions representing production output, transport, and inventory in a supply chain network.

A common characteristic of supply chain optimization tasks is the presence of objective functions with non-convex regions, local optima, and combinatorial complexity exponentially scaling with problem size, making them intractable using traditional heuristic and exact algorithms. Quantum annealing algorithms represent optimization problems in the form of Ising model Hamiltonian problems, such that their ground state solutions can be used to leverage quantum tunneling effects to overcome local optima where conventional

simulated annealing algorithms usually get stuck [3]. The combination of machine learning paradigms with quantum computing enables adaptive supply chain systems with continuous updating of optimization models based on real-time operating, forecasting, and disruption information. Additionally, warehouse automation systems benefit from quantum optimization methods, solving robotic path routing optimization, task optimization using machine learning, with consideration of dynamic constraints such as order priority, time requirements, and resource availability. The relative complexity benefit of quantum algorithms over classical solutions can be noted in retail business environments with large product offerings, geographically dispersed warehouse logistics, and real-time customer demand behavior requiring fast re-computation of logistics solutions in response to disruptions and demand variability.

3. Quantum Computing in Healthcare & Life Sciences

The power of quantum computing can bring revolutionary capabilities to healthcare and life sciences with advanced simulation tools for molecular systems, optimization algorithms for treatment protocols, and machine learning algorithms for processing pattern recognition in biomedical signals. The inherent strength of quantum algorithms in solving eigenvalue problems leads to efficient calculations of molecular energy levels and quantum mechanical properties governing chemical behavior, protein folding rates, and ligand-target binding affinities [5]. The implementation of quantum algorithms optimized for simulating sparse Hamiltonian matrices leads to an exponential speedup over conventional algorithms in calculating eigenvalue spectra of quantum systems; hence, quantum algorithms can bring algorithmic solutions for simulating molecular properties without resorting to exponentially scaling calculations in conventional quantum chemistry simulation algorithms. The implementation of quantum phase estimation algorithms to simulate molecular Hamiltonian matrices leads to an efficient calculation of ground state energies and excited state properties with polynomial scaling in computation, thereby resolving a critical bottleneck in drug discovery simulation algorithms, where accurate calculation of ligand-target binding affinities in molecular systems demands an efficient quantum mechanical simulation of electronic structures. The simulation of chemistry with quantum computers can thus overcome the problem of exponential complexity,

which limits classical computers from simulating chemistry in a realistic way beyond a small scale, allowing scientists to simulate pharmaceutical molecules, proteins, and biochemical reactions with quantum mechanical accuracy [6]. Quantum simulation algorithms exploit the intrinsic similarity between quantum chemistry and quantum computing by representing wavefunctions in a quantum computer with a direct correspondence to molecular wavefunctions, simulating time evolution using a sequence of quantum gates. The Variational Quantum Eigensolver algorithm enables a practical implementation of quantum simulation with noisy intermediate-scale quantum processors by using parameterized quantum circuits in combination with classical optimization algorithms to search for minimal energy expectations. The application of healthcare goes beyond novel therapeutic approaches in pharmaceutical discovery but integrates analysis of the human genome using quantum machine learning algorithms, interpreting genetic sequence information to detect disease-associated variants, patient-specific treatment outcomes, and molecular subtypes of diseases using a combination of multiple datasets in a variety of diseases. Clinical trials are optimized with quantum algorithms to create adaptive protocols with minimized sample requirements and optimized statistical validity, which have been seen to pose a challenge in terms of increased healthcare cost and time in pharmaceutical research and medical device evaluation.

4. Quantum Computing in Insurance Industry Operations

Quantum computing impacts the operations of the insurance sector by providing better risk evaluation methods, better methods for fraudulent claims identification, and better catastrophe modeling solutions to cope with the intensive computation involved in calculations performed by actuaries [13]. The application of quantum algorithms in solving risk analysis models utilizes amplitude estimation algorithms to speed up Monte Carlo simulation algorithms, which are in widespread usage in insurance companies for pricing, reserve, and capital allocation calculations [7]. The quantum algorithm for amplitude estimation offers a quadratic speed-up in calculating expectations and risks from a given probability distribution, allowing an insurance firm to calculate collective risk exposure in big books of business with less computation compared with classical Monte Carlo simulations. The algorithm builds a quantum state representing a probability distribution of loss outcomes, using quantum phase estimation to

recover classical statistics such as expectations of losses, risk, and value-at-risk measures, and conditional tail expectations used in capital requirements. In insurance, quantum computing can be applied in catastrophe modeling where "complex dependencies among natural hazards, infrastructure vulnerabilities, and insured exposures give rise to high-dimensional optimization problems that demand assessment of a large variety of correlated risk situations" [7]. Moreover, quantum algorithms can allow insurance firms to assess "systemic risk resulting from cascading failures, geometric correlations among natural hazards, and time-series dependencies in catastrophe event occurrence" in a manner not feasible classically in simulations. Furthermore, hybrid systems using quantum machine learning in insurance fraud analysis can improve "pattern analysis for atypical claims activity, dubious transaction trails, and coordinated fraudulent activity for multiple policies or claimants" [8]. In addition, quantum support vector machines can offer speedup in computation over classical machine learning algorithms in classification when dealing with datasets in "high-dimensional spaces, characterized by a combination of structured insurance policy data and unstructured text information from claims descriptions and investigation reports" in insurance datasets including demographic, behavioral, and transactional information. Generally, quantum machine learning can be applied in "computer vision in claims assessment, improving image processing for claims evaluation, such as image analysis in damage assessment, with enhanced processing speed and accuracy in claim assessment using advanced image processing capabilities not readily available in classical computation with standard machine learning algorithms based on large datasets where processing time increases significantly in a linear manner with increasing size based on classical algorithms, which can be a bottleneck in quick claims processing and assessment in insurance companies using standard machine learning algorithms without access to quantum machine learning capabilities especially with large datasets in processing speed and accuracy in claim evaluation based on image analysis capabilities" as indicated in [9].

5. Quantum Computing in Financial Services

Quantum computing solutions have computationally intensive domains in financial services such as derivative pricing, portfolio optimization, and risk management using quantum algorithms with speedups for Monte Carlo simulations, linear algebra computations, and

combinatorial optimization relevant to financial mathematical modeling [4]. The financial service industry encounters computational intensity in derivative pricing with path-dependent options, asset portfolio combinations, and exotic derivative clauses requiring intensive Monte Carlo simulations to approximately calculate fair derivative prices and their risk sensitivities in stochastic asset price models. The application of quantum algorithms in option pricing employs amplitude estimation algorithms with quadratic speedup benefits in Monte Carlo integration calculations of payoff expectations under risk-neutral Probability measures [9]. Such quantum algorithms estimate the payoff expectations via operators representing asset price evolution simulations and payoff functions, with financial risk analysis based on quantum algorithmic computation of derivative prices using operators interpreted to represent mathematical mappings of financial optimality in delta, gamma, and vega risk sensitivities under hedging and risk management regimes. A major application area where portfolio optimization studies exist is being tackled by quantum algorithms because of the complexity of mean-variance optimization under actual constraints such as transaction costs, positions, and regulatory issues in financial systems [10]. Approximate optimization algorithms have used portfolio optimization problem statements in terms of quadratic unconstrained binary optimization, where the problem is stated with binary variables indicating asset selection, with objective functions taking into account both return and variance of the portfolio and trading costs. The quantum algorithm consists of problem Hamiltonian evolution and mixing steps, where they improve towards solving portfolio allocation based on parameter refinement in a variational manner optimized using classical methods [4]. The financial application using quantum computing is in credit risk analysis, where machine learning algorithms using quantum computing analyze borrower information with higher dimensions to estimate the probability of default using non-linear relationships and interaction terms, improving over standard logistic regression and scorecard analysis. Real-time risk calculation systems can benefit from using quantum computing algorithms in calculating sensitivity analysis of large derivative portfolio calculations, where financial institutions can fulfill regulatory requirements for stress analysis of capital adequacy in a computationally efficient manner suitable for real-time intraday risk analysis and hedging strategies.

6. Future Implications and Cross-Sector Integration

The path towards the application of quantum computing in a practical manner in enterprises goes through a noisy intermediate-scale quantum stage with non-ideal devices with a small number of qubits, short coherence times, and gate fidelities not suitable for fault-tolerant quantum computation [11]. Variational quantum algorithms based on optimization methods, taking into consideration the available capabilities of current devices, utilize variational quantum circuits with parameterized quantum circuits optimized with classical optimization methods, thereby making possible meaningful computation despite noise and small circuit depths. Methods of error mitigation, such as zero noise extrapolation and probabilistic error cancellation, have advanced the limits of computation with noisy quantum devices, using post-processing of measurement outcomes to reduce error contributions without implying full quantum error correction overhead. Hybrid quantum-classical algorithms are basically the main approach towards achieving quantum benefit in the noise-sensitive devices of the current near-term quantum computers by dividing a problem into two parts: solutions obtained by quantum computers and solutions obtained with classical computers. Opportunities in cross-sector integration arise from similarities in problem structures amenable to computation across different industries, where optimization challenges in logistics, scheduling, and resource allocation share mathematical structures suitable for solution using quantum algorithms [12]. Evidence of quantum computational superiority in specialized sampling problems confirms the capability of quantum hardware to accomplish specific computational tasks in a classical sense, but solutions in a meaningful and commercial context have yet to be developed using algorithms targeting such tasks. Industrial utility of quantum computation awaits improvements in hardware performance, such as enhanced coherence times of qubits, gate fidelity, and inter-qubit connectivity sufficient for larger quantum operations. Development efforts in quantum software environments, cloud-infrastructure support for quantum computation, and educational programs in quantum computation can help improve access to quantum capabilities and hasten translation from research to industry. Cross-sector standardization in performance comparisons of quantum computers, algorithm validation, and guidelines for developing efficient quantum software will allow intelligence-driven evaluation and assessment of quantum utility in application domains such as retail, healthcare, insurance, and financial services.

Table 1: Quantum Computing Applications in Retail and Supply Chain Operations [3]

Application Domain	Optimization Challenge	Quantum Approach	Computational Advantage
Logistics Network Design	Facility location and capacity allocation	Hybrid quantum-classical decomposition	Exponential solution space exploration
Supply Chain Distribution	Non-convex optimization with local optima	Quantum annealing with Ising Hamiltonians	Quantum tunneling escape from local optima
Warehouse Automation	Robot routing and task assignment	Quantum optimization with dynamic constraints	Real-time re-optimization capability
Inventory Management	Multi-period demand satisfaction	Discrete-continuous variable processing	Parallel combinatorial search

Table 2: Quantum Computing Applications in Healthcare and Life Sciences [5, 6]

Application Domain	Computational Problem	Quantum Algorithm	Healthcare Impact
Drug Discovery	Molecular energy calculations	Quantum phase estimation	Polynomial scaling for binding affinity prediction
Chemical Simulation	Electronic structure modeling	Variational Quantum Eigensolver	Quantum mechanical accuracy for pharmaceuticals
Genomic Analysis	Disease variant identification	Quantum machine learning	Patient-specific treatment prediction
Clinical Trials	Protocol optimization	Adaptive quantum algorithms	Minimized sample size requirements

Table 3: Quantum Computing Applications in Insurance Industry Operations [7-9]

Application Domain	Risk Challenge	Quantum Technique	Operational Benefit
Catastrophe Modeling	Correlated risk scenarios	Quantum amplitude estimation	High-dimensional dependency assessment
Risk Analysis	Monte Carlo simulation	Quantum phase estimation	Quadratic speedup in expectation calculations
Fraud Detection	Anomalous pattern recognition	Quantum support vector machines	High-dimensional feature space classification
Claims Processing	Image analysis for damage assessment	Quantum machine learning	Enhanced processing speed and accuracy

Table 4: Cross-Sector Integration and Future Development Pathways [11, 12]

Development Area	Current Challenge	Solution Approach	Industry Impact
Hardware Limitations	Limited qubit coherence times	Error mitigation techniques	Extended computational reach
Algorithm Design	Noise-sensitive computation	Hybrid quantum-classical decomposition	Near-term quantum advantage
Cross-Sector Applications	Common optimization structures	Shared algorithmic frameworks	Multi-industry implementation
Standardization	Performance benchmarking	Algorithm validation guidelines	Informed technology adoption decisions

7. Conclusions

Integration of quantum computing in different sectors of enterprises marks a paradigm shift in computational power pertaining to optimization, simulation, and machine learning tasks in retail, healthcare, insurance risk analysis, and financial industries. The evolution of quantum computation from theoretical algorithms towards implementation focuses on hybrid quantum-

classical architectures, wherein computation is performed within quantum processors over constrained steps of computation, with classical systems controlling algorithmic steps. The short-term application benefit of quantum computation is established over domains such as supply chain optimization, molecular simulation studies, catastrophe modeling, and financial derivative pricing, where quantum algorithms show a tangible benefit over computation time despite hardware

shortcomings. The migration path towards fault-tolerant quantum computation in enterprises continues to demand progress in quantum processor coherence time and error-correcting algorithms in tandem with industry-oriented algorithmic designs over relevant commercially important computations. Opportunities in quantum computation over different sectors arise based on commonalities in computation models over optimization and machine learning domains, with cloud-accessible quantum computation software facilitating access to quantum computation power.

Author Statements:

- **Ethical approval:** The conducted research is not related to either human or animal use.
- **Conflict of interest:** The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper
- **Acknowledgement:** The authors declare that they have nobody or no-company to acknowledge.
- **Author contributions:** The authors declare that they have equal right on this paper.
- **Funding information:** The authors declare that there is no funding to be acknowledged.
- **Data availability statement:** The data that support the findings of this study are available on request from the corresponding author. The data are not publicly available due to privacy or ethical restrictions.
- **Use of AI Tools:** The author(s) declare that no generative AI or AI-assisted technologies were used in the writing process of this manuscript.

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