

AI-Driven Optimization of Green Hydrogen Production Networks for Saudi Arabia's Industrial Cities Under Vision 2030

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Abstract: Green hydrogen is becoming a strategic industrial fuel for Saudi Arabia because it can convert world-class solar and wind resources into exportable molecules, decarbonized feedstocks, and resilient energy services for industrial cities. This review paper develops an AI-driven optimization perspective for green hydrogen production networks in NEOM, Yanbu, Jubail, Ras Al-Khair, Dammam and related logistics corridors under Vision 2030. The aim is to explain how machine learning, digital twins, stochastic optimization, and explainable decision systems can reduce cost, water stress, renewable curtailment, electrolyzer degradation, and offtake uncertainty while strengthening industrial competitiveness. A structured review methodology was used to synthesize recent literature, policy reports, and applied hydrogen system studies published between 2020 and 2025. The paper proposes an integrated framework that links renewable forecasting, desalinated water allocation, electrolyzer scheduling, storage sizing, ammonia conversion, port export planning, and ESG reporting. Findings show that AI is most valuable when it is embedded across the full network rather than limited to plant-level control. The study concludes that Saudi industrial cities can accelerate green hydrogen deployment by adopting interoperable data standards, bankable optimization models, cybersecure control rooms, and phased implementation roadmaps aligned with Vision 2030.

Keywords: Green Hydrogen, Artificial Intelligence (AI), Saudi Arabia, Vision 2030, Renewable Energy, Machine Learning.

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1. INTRODUCTION

Saudi Arabia is repositioning its energy system from an oil-export model toward a diversified industrial economy in which clean electricity, green hydrogen, advanced manufacturing, logistics, and data-driven infrastructure reinforce each other. Vision 2030 provides the national direction for this transition by emphasizing economic diversification, renewable energy, local content, industrial competitiveness, environmental stewardship, and global investment attraction. Green hydrogen is especially relevant because it allows renewable

electricity to be transformed into hydrogen, ammonia, synthetic fuels, process heat, and low-carbon feedstocks that can serve both domestic industries and international markets. The Kingdom's industrial geography is well suited to this opportunity: coastal regions provide port access and desalination capacity, the north-west offers strong renewable resources, and established industrial cities already host petrochemicals, mining, metals, fertilizers, transport and energy-intensive manufacturing.

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However, a hydrogen economy is not created by building electrolyzers alone. It requires coordinated decisions across renewable generation, grid connection, water supply, desalination, compression, liquefaction, ammonia synthesis, storage, transport, port logistics, industrial offtake, certification, safety, and finance. These decisions are uncertain because renewable output varies hourly, hydrogen demand grows in stages, electrolyzer performance changes with operating conditions, water availability is location-specific, and export market prices may fluctuate. Conventional project planning often treats these elements separately, which can produce oversized assets, stranded storage, avoidable curtailment, slow maintenance response, and weak investor confidence. AI-driven optimization offers a more integrated alternative because it can learn from operational data, simulate alternative futures, and recommend decisions that balance cost, carbon, reliability, water use, and industrial value.

The topic therefore has strong academic and practical importance. At the academic level, recent hydrogen research increasingly recognizes the importance of machine learning, digital twin simulation, techno-economic assessment, and life-cycle analysis, but there remains a gap in connecting these methods to industrial city networks in arid economies. At the policy level, Saudi Arabia requires planning approaches that make green hydrogen competitive without overlooking water-energy constraints or social benefits such as skilled employment. At the business level, developers need transparent models that convert technical optimization into bankable investment cases. This paper addresses these needs through a review-based framework tailored to Saudi Arabia's industrial cities, while following a structure inspired by recent Springer-style review articles that combine context, literature synthesis, methodology, conceptual framework, tables, figures, implications and future research directions.

The aim of the study is to develop a review-based framework for AI-driven optimization of green hydrogen production networks in Saudi industrial cities under Vision 2030. The objectives are fourfold: first, to identify the operational and strategic variables that determine green hydrogen network performance; second, to review how AI, machine learning, digital twins and mathematical optimization can improve planning and control; third, to map the relationship between hydrogen infrastructure, renewable energy, desalinated water, industrial clusters and export corridors; and fourth, to propose implementation recommendations for policymakers, developers, industrial city authorities and investors. The central research question is: how can AI optimize

green hydrogen production networks so that Saudi industrial cities achieve lower cost, higher reliability, reduced emissions, responsible water use and stronger industrial diversification?

2. LITERATURE REVIEW

Green hydrogen literature between 2020 and 2025 shows a rapid shift from technology feasibility toward system integration. Early discussions often focused on electrolyzer types, renewable electricity costs, and hydrogen storage pathways. More recent studies consider whole-system design, including power-to-X integration, ammonia exports, industrial clusters, certification systems, and cross-border trade. This shift matters for Saudi Arabia because industrial cities do not need isolated demonstration units; they need scalable networks that can connect renewable energy zones with refineries, petrochemical plants, mining complexes, ports, transport corridors and future clean-fuel markets. Research on hydrogen techno-economics demonstrates that electricity price, capacity factor, electrolyzer utilization, financing cost, degradation, water treatment, and storage configuration strongly influence the levelized cost of hydrogen. These variables are dynamic rather than fixed, which makes AI-supported forecasting and optimization particularly useful (IRENA, 2020; IEA, 2023; Osman *et al.*, 2022).

The renewable energy dimension is central. Saudi Arabia has high solar irradiation and growing interest in wind resources, but the interaction between generation variability and electrolyzer operation is complex. Running electrolyzers only when renewable electricity is abundant can reduce electricity cost and emissions, yet it may lower utilization and increase unit costs. Operating continuously can improve utilization, yet it may require storage, grid balancing or hybrid renewable portfolios. AI can support this trade-off through probabilistic weather forecasting, renewable output prediction, dynamic dispatch, and scenario analysis. Machine learning models such as gradient boosting, recurrent neural networks, reinforcement learning, and hybrid physics-informed models can predict resource availability and equipment performance more accurately when sufficient data governance is in place. Digital twins can then test how dispatch choices affect hydrogen output, stack degradation, water demand, storage pressure and contractual delivery obligations.

Water is another critical theme in arid-region hydrogen planning. Electrolysis needs purified water, and coastal Saudi industrial cities can use desalinated seawater, but desalination adds energy demand, brine management responsibilities and infrastructure costs. The water requirement per

kilogram of hydrogen is often small compared with many industrial uses, yet the network-level implications are significant when projects scale to export volumes. AI-driven optimization can coordinate renewable electricity, desalination schedules, water storage, electrolyzer operation and brine management so that water constraints are not treated as an afterthought. This is important for Vision 2030 because environmental performance must include both carbon reduction and responsible resource management. The literature also shows that life-cycle assessment is essential, since low-carbon claims depend on electricity source, equipment manufacturing, water treatment, transport pathways, and certification boundaries.

Industrial offtake research highlights that hydrogen has different value propositions across sectors. In petrochemicals and refining, green hydrogen can reduce emissions from existing hydrogen use. In fertilizers, hydrogen can be converted into green ammonia for domestic use and export. In steel, mining and heavy transport, hydrogen can provide a pathway for hard-to-abate decarbonization. In ports, hydrogen derivatives can support bunkering and clean logistics. For Saudi Arabia, the industrial city network is therefore not merely a production map; it is a demand and conversion ecosystem. AI can support matching between supply and demand by forecasting industrial load, optimizing contracts, identifying the best conversion route, scheduling storage withdrawals, and prioritizing high-value uses during periods of renewable scarcity. This network thinking aligns with international hydrogen hub strategies, where co-location and shared infrastructure are used to reduce cost and accelerate adoption (Hydrogen Council, 2023; IEA, 2024).

The digitalization literature provides the technological foundation. AI and digital twins are increasingly used in renewable energy forecasting, predictive maintenance, electrolyzer performance monitoring, anomaly detection, safety management, and optimal control. Reinforcement learning can explore operational strategies under uncertain electricity prices and renewable output. Mixed-integer linear programming and stochastic optimization can support investment decisions for capacity sizing, pipeline routing, storage design and port logistics. Explainable AI is becoming important because hydrogen networks involve safety-critical assets, regulated infrastructure and high-value investment decisions; black-box recommendations may not be acceptable to operators, financiers or regulators. Cybersecurity is also essential because hydrogen control rooms will depend on sensors, industrial IoT, SCADA integration, cloud analytics and cross-organizational data exchange. In this context, optimization must be trustworthy, auditable and resilient rather than purely mathematically efficient.

Despite these advances, three research gaps remain. First, many studies optimize individual plants, while fewer evaluate multi-city hydrogen networks that include renewable supply, water, storage, ports and industrial offtakers. Second, arid-region water-energy interactions are often underdeveloped in optimization models, even though they are central to Saudi deployment. Third, Vision 2030 alignment is sometimes discussed at a policy level but not translated into operational metrics such as local job creation, emissions avoided, industrial localization, export readiness, resilience, safety, and investor bankability. This paper contributes by integrating these dimensions into a single review-based framework for Saudi industrial cities.

Table 1: Comparative AI optimization functions across the green hydrogen value chain

Value-chain stage	AI/optimization tool	Expected contribution	Saudi industrial relevance
Renewable supply	Weather ML, probabilistic forecasting	Higher renewable matching and lower curtailment	NEOM, Tabuk, Red Sea wind-solar zones
Water and desalination	Schedule optimization and digital twins	Reduced energy-water conflict and improved storage planning	Coastal industrial cities and desalination assets
Electrolysis	Predictive maintenance and optimal dispatch	Higher availability, safer operation and lower degradation	Large electrolyzer clusters and modular plants
Storage and conversion	Stochastic sizing and inventory models	Reliable delivery despite renewable variability	Ammonia, compressed hydrogen and port export hubs
Industrial offtake	Demand forecasting and contract optimization	Better matching of hydrogen to high-value users	Jubail, Yanbu, Ras Al-Khair and Dammam corridors

3. METHODOLOGY

This study adopts a structured narrative review methodology supported by elements of systematic search logic. A full systematic review was not selected because the topic spans engineering,

energy economics, AI, industrial policy, water systems, logistics and Saudi development strategy; a narrow database protocol would miss important policy reports and applied industry evidence. The review therefore followed a transparent multi-stage

process. First, the scope was defined around green hydrogen production networks, AI optimization, Saudi industrial cities and Vision 2030. Second, recent peer-reviewed articles and institutional reports from 2020 to 2025 were screened for relevance to electrolysis, renewable forecasting, digital twins, techno-economic analysis, life-cycle assessment, desalinated water, hydrogen hubs, ammonia conversion and industrial decarbonization. Third, the evidence was coded into thematic categories: resources, technologies, optimization methods, infrastructure, governance, finance, ESG and implementation risks. Fourth, insights were synthesized into a conceptual framework and practical roadmap.

The inclusion criteria were designed to ensure relevance and quality. Sources were included when they addressed green hydrogen, renewable energy integration, AI or optimization methods, hydrogen techno-economics, industrial clusters, desalination-energy interaction, Saudi energy transition, or global hydrogen policy. Sources were excluded when they were outdated, purely promotional, unrelated to industrial-scale hydrogen, or focused only on fossil-based hydrogen without transition relevance. Preference was given to papers and reports published between 2020 and 2025, consistent with the user's requirement for recent citations. The attached Springer-style reference paper was used as a formatting and structural model, particularly in its use of clear objectives, integrated conceptual diagrams, comparative tables, sustainability metrics and policy-oriented discussion. The present paper differs in topic and content by focusing specifically on AI-enabled green hydrogen networks rather than smart housing or model-based engineering for urban systems.

The analytical framework combines three review lenses. The first is the techno-economic lens, which examines capital expenditure, operating cost, electrolyzer utilization, renewable capacity factor, storage requirement, water treatment, transport cost and levelized cost of hydrogen. The second is the systems optimization lens, which examines how AI and operations research can coordinate renewable generation, electrolysis, desalination, storage, industrial offtake and export logistics under uncertainty. The third is the Vision 2030 impact lens, which evaluates how the network supports diversification, local content, clean industry, export competitiveness, environmental protection and high-skilled employment. Combining these lenses allows the paper to move beyond a descriptive review and present a decision-oriented framework.

Because the study is a review paper, it does not claim to produce new empirical measurements

from field operations. Instead, it creates a synthesis that can guide future empirical work and project planning. This limitation is appropriate for a fast-moving sector where many Saudi hydrogen assets are still under development. The value of the review lies in clarifying how AI should be embedded in network design before large-scale assets are locked into fixed configurations. Future studies can extend this work by using real operational data from electrolyzers, desalination plants, renewable farms and industrial offtakers to validate the proposed optimization logic.

4. AI-Driven Optimization Framework

The proposed framework has six interacting layers. The first layer is resource intelligence, which uses satellite data, weather models, historical renewable output and sensor streams to forecast solar and wind generation at multiple time horizons. Short-term forecasts support hourly electrolyzer dispatch, while seasonal forecasts support maintenance planning, hydrogen inventory management and contract scheduling. The second layer is water-energy coordination. It links desalination, purified water storage, brine management and electricity availability so that water supply is optimized with renewable generation rather than treated as a fixed utility input. The third layer is electrolyzer performance management. AI models monitor stack efficiency, temperature, pressure, membrane condition, degradation and fault signals. Predictive maintenance can reduce downtime and avoid unsafe operating states, while dynamic control can protect equipment during variable renewable operation.

The fourth layer is network optimization. This layer determines how much hydrogen should be produced at each location, when it should be stored, where it should be converted into ammonia or other derivatives, and which industrial users should receive priority under different demand and price scenarios. Mixed-integer optimization can support capacity planning, while reinforcement learning and stochastic programming can improve real-time decisions. The fifth layer is logistics and market integration. It coordinates pipeline, truck, port, ship, ammonia, methanol or synthetic fuel pathways depending on distance, volume, safety requirements and customer specifications. The sixth layer is governance intelligence. This layer converts technical outputs into dashboards for carbon intensity, water intensity, cost, safety, local content, employment, certification and Vision 2030 contribution. Explainable AI is essential here because regulators, investors and industrial customers need to understand why a decision is recommended.

For Saudi industrial cities, the framework can operate as a federated digital twin. Each

production or demand node maintains local operational control, while an integrated optimization layer exchanges standardized data on renewable forecasts, electrolyzer availability, water inventory, storage capacity, port congestion, offtake commitments and maintenance plans. This approach respects organizational boundaries while still enabling network-level efficiency. For example, if wind output in the north-west exceeds expectations, the system can increase electrolysis, store hydrogen, schedule ammonia conversion, and prepare export slots. If port congestion or equipment maintenance reduces export capacity, production can be redirected toward domestic industrial demand or stored if capacity is available. If water stress increases, desalination schedules can shift toward periods of renewable surplus.

The framework also supports resilience. Hydrogen networks will face dust storms, heat waves, equipment faults, grid constraints, cyber threats, market volatility and shipping delays. AI-based scenario simulation can test these disruptions before they occur. Digital twins can run stress tests to

identify bottlenecks in storage, desalination, compression, pipeline capacity or port handling. Optimization models can then recommend redundancy, flexible contracts, alternative routing, backup power, maintenance buffers and safety stock. Resilience is not merely a technical issue; it is a bankability issue because financiers and offtakers need confidence that delivery obligations can be met under variable conditions.

A key principle is that AI should not replace engineering judgement or regulatory oversight. Instead, it should augment decision-making through transparent, auditable and human-supervised systems. Operators should be able to inspect assumptions, override recommendations, trace data sources and test alternative scenarios. This is especially important in hydrogen because safety risks require conservative operating envelopes. Therefore, the recommended model is a human-in-the-loop optimization architecture, where AI provides predictive insight and decision support while certified engineers retain accountability for critical actions.

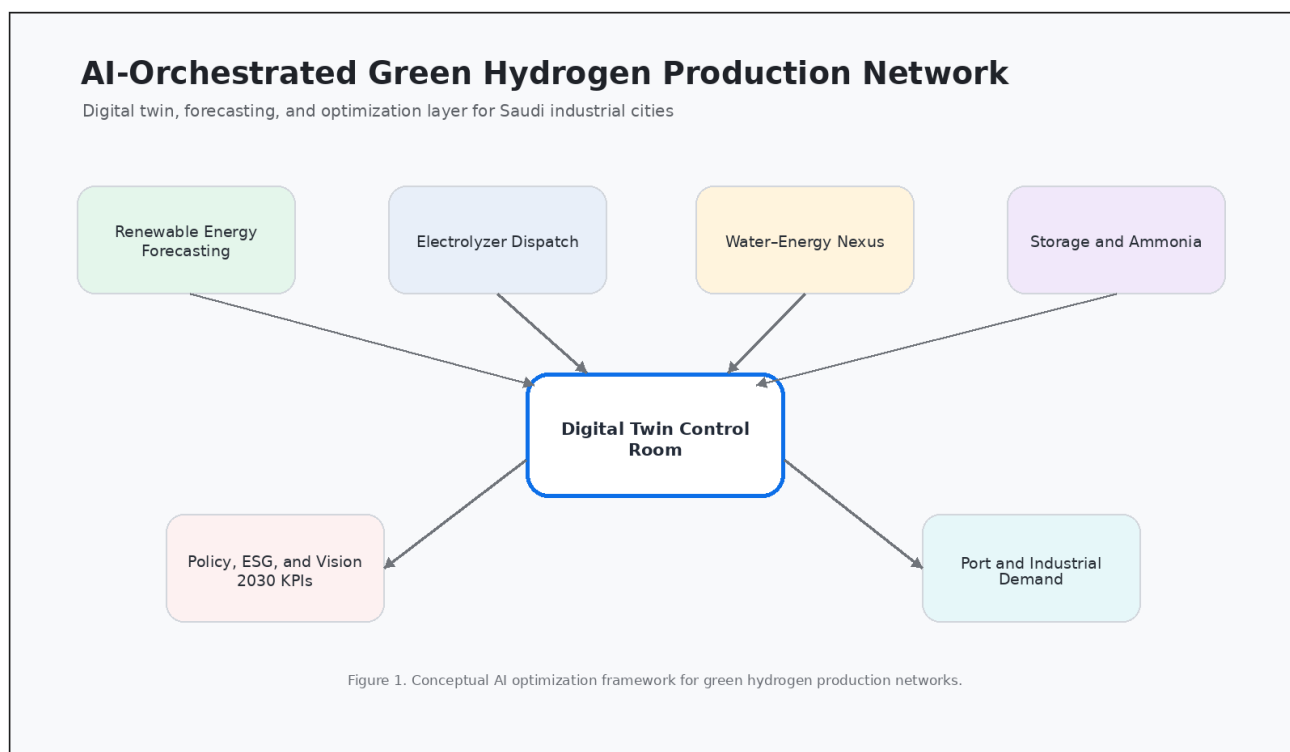


Figure 1: Conceptual AI optimization framework for green hydrogen production networks in Saudi industrial cities.

5. Saudi Industrial City Applications

NEOM is the flagship production and export case because it combines large renewable ambitions, coastal access, green hydrogen investment and a global branding platform for clean energy. AI optimization in NEOM should focus on hybrid renewable dispatch, electrolyzer utilization,

ammonia synthesis scheduling, storage sizing, seawater desalination integration and export reliability. The objective is not only to produce hydrogen at scale, but to prove that Saudi Arabia can operate a digitally advanced green molecule supply chain. Lessons from NEOM can be transferred to

other industrial cities through standardized data models and shared certification protocols.

Yanbu offers a western industrial corridor with refining, petrochemicals, port facilities and Red Sea logistics. Green hydrogen in Yanbu could support refinery decarbonization, ammonia production, clean fuels, and industrial heat applications. AI would be valuable for matching hydrogen supply with complex industrial demand profiles, integrating renewable electricity, and optimizing port movements. Jubail represents an eastern petrochemical and export powerhouse. Its hydrogen opportunity is linked to chemicals, fertilizers, refining, shipping fuels and potential carbon management integration. AI can support industrial symbiosis by identifying where hydrogen, oxygen by-products, heat, water and logistics assets can be shared across companies. Ras Al-Khair provides mining and metals relevance, where hydrogen may support low-carbon mineral processing and future direct reduction pathways. Dammam and nearby logistics corridors can support distribution, port services, transport demand and eastern export connectivity.

The network perspective is crucial because each city has different strengths. NEOM may lead in

renewable hydrogen production, Jubail in industrial conversion and petrochemical integration, Yanbu in Red Sea export and fuels, Ras Al-Khair in mining decarbonization, and Dammam in logistics and regional demand. AI can coordinate these differences by recommending whether investment should prioritize production, storage, conversion, transport or offtake development in each phase. This avoids a one-size-fits-all approach and allows Saudi Arabia to build a portfolio of hydrogen hubs with complementary roles.

Industrial city authorities can use the proposed framework to evaluate project proposals. Instead of assessing each project only on installed electrolyzer capacity, they can request digital evidence on expected utilization, renewable matching, water sourcing, carbon intensity, offtake security, safety systems, cyber resilience, local content and integration with other assets. This would encourage developers to design bankable and system-compatible projects. It would also support transparent reporting to Vision 2030 stakeholders and international buyers who increasingly demand proof of low-carbon production.

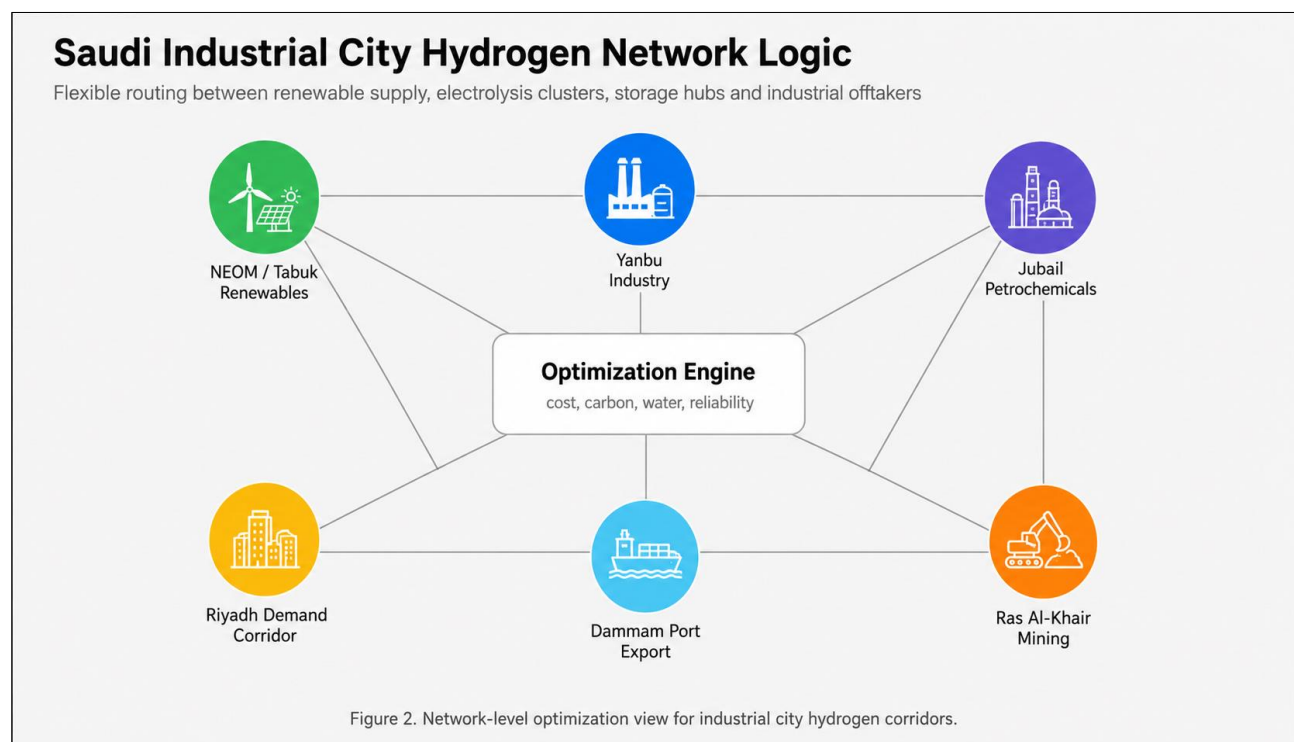


Figure 2: Network-level optimization view for Saudi industrial city hydrogen corridors

6. DISCUSSION

The review indicates that AI creates value at three levels: operational efficiency, investment optimization and strategic governance. Operationally, AI improves forecasting, dispatch,

maintenance and safety monitoring. Investment-wise, optimization models help determine the right mix of renewable capacity, electrolyzer size, storage, conversion and transport. Strategically, dashboards and scenario tools help policymakers understand

how hydrogen contributes to national goals. These levels must be connected. A highly efficient electrolyzer will not deliver national value if it is poorly located, underutilized, water-constrained or disconnected from offtakers. Similarly, an ambitious export project may struggle if certification, port scheduling, storage and supply reliability are weak.

The strongest opportunity for Saudi Arabia is not simply low-cost renewable electricity; it is the ability to design industrial hydrogen ecosystems from the beginning. Many countries must retrofit hydrogen into congested infrastructure. Saudi Arabia can build new digital control systems, renewable zones, hydrogen hubs and industrial corridors with fewer legacy constraints. This creates a chance to adopt common data standards, cybersecure interoperability, real-time ESG measurement and AI-enabled planning from the first major deployment phases. Such an approach would align with the Kingdom's wider ambition to become a leader in advanced technologies and high-value industrialization.

The main risks are fragmentation, data weakness, underdeveloped offtake, water governance, technology lock-in and public trust. Fragmentation occurs when renewable developers, electrolyzer operators, desalination providers, ports

and industrial users optimize separately. Data weakness occurs when sensors, contracts, maintenance records and environmental metrics are not standardized. Offtake risk arises when production assets are built before customers and certification pathways are secure. Water governance risk emerges if desalination and brine management are not integrated into hydrogen planning. Technology lock-in occurs when early projects adopt closed systems that cannot scale. Public trust can weaken if hydrogen is perceived as a branding exercise rather than a measurable sustainability transition. AI can reduce these risks, but only if data governance, institutional coordination and transparent reporting are built into the system.

From a Scopus/Q1 review-paper perspective, the topic is suitable because it connects global debates on green hydrogen, AI, digital twins, energy transition and industrial decarbonization with a specific national development context. The contribution is conceptual rather than empirical, but it provides a structured framework that future studies can test using simulation models, case data, optimization experiments and stakeholder interviews. The paper also contributes to Vision 2030 research by translating national transformation goals into operational variables that can be measured and optimized.

Table 2: Proposed Vision 2030 KPI structure for AI-enabled hydrogen networks

KPI dimension	Example metric	Optimization relevance	Governance use
Cost competitiveness	Levelized cost of hydrogen and ammonia	Optimizes electricity, utilization and logistics	Investor bankability and tariff design
Carbon performance	kg CO ₂ e per kg H ₂ or NH ₃	Tracks renewable matching and life-cycle emissions	Export certification and ESG reporting
Water responsibility	Liters of purified water per kg H ₂	Coordinates desalination, storage and brine management	Environmental permitting and local trust
Reliability	Delivery performance and asset availability	Improves maintenance, storage and contingency planning	Offtake security and resilience
Industrial diversification	Local jobs, local content and offtake value	Prioritizes projects with wider economic spillovers	Vision 2030 progress monitoring

7. Policy and Managerial Implications

Policymakers should establish a national hydrogen data architecture that defines common metrics for renewable matching, water intensity, electrolyzer performance, carbon intensity, safety incidents, cybersecurity, local content and export certification. Without common data, AI models will remain isolated and difficult to audit. Industrial city authorities should create hydrogen sandbox zones where developers can test AI-enabled dispatch, digital twins, smart metering, safety analytics and certification tools under supervised regulatory conditions. These sandboxes would reduce innovation risk while protecting safety and environmental standards.

Developers should design projects around flexibility rather than maximum nameplate capacity. Flexible assets can shift production, store hydrogen, adjust conversion schedules and respond to market changes. Investors should require scenario-based financial models that show project performance under renewable variability, price volatility, water constraints, equipment degradation and delayed offtake. Operators should invest in workforce capabilities, including data engineering, hydrogen safety, AI model validation, cybersecurity and digital maintenance. Universities and training providers can support this requirement by developing programs that combine chemical engineering, renewable energy, data science and industrial systems management.

For international buyers, Saudi Arabia should emphasize traceability. Green hydrogen exports will be judged not only by price but also by proof of renewable origin, emissions intensity, water responsibility and delivery reliability. AI-enabled certification systems can provide this proof through auditable records and real-time monitoring. For domestic industries, the value proposition should focus on competitiveness: low-carbon products, future export access, reduced regulatory exposure and participation in new supply chains. The managerial message is clear: hydrogen must be treated as a network business, not a single-plant commodity.

8. Future Research Directions

Future research should develop quantitative optimization models using Saudi-specific data on solar irradiation, wind profiles, desalination costs, industrial demand, port capacity, storage technologies and financing assumptions. Comparative simulations could test whether hydrogen should be produced close to renewable resources, near industrial demand, or in hybrid hub-and-corridor configurations. Studies should also examine the trade-offs between hydrogen pipelines, ammonia conversion, trucking and shipping for different volumes and distances. Another important area is electrolyzer degradation under variable renewable operation in hot and dusty environments, since performance data from temperate regions may not fully apply.

Social and governance research is also needed. Hydrogen deployment will require skilled workers, safety awareness, community acceptance, emergency planning and transparent environmental communication. Future studies can explore how industrial city stakeholders perceive hydrogen risks and benefits, and how explainable AI dashboards influence trust. Researchers should also evaluate cybersecurity frameworks for hydrogen digital twins, because networked control systems may become critical infrastructure. Finally, life-cycle assessment should be expanded to include desalination, brine management, construction materials, storage vessels, shipping, conversion losses and end-use pathways. These studies would strengthen evidence for Saudi Arabia's hydrogen strategy and improve its credibility in global clean-energy markets.

9. CONCLUSION

AI-driven optimization can play a decisive role in transforming Saudi Arabia's green hydrogen ambitions into resilient industrial networks. The review shows that the most valuable applications of AI are not limited to predicting renewable output or controlling electrolyzers. They extend across the entire hydrogen value chain, including desalinated

water scheduling, storage sizing, industrial offtake matching, ammonia conversion, port logistics, maintenance planning, ESG reporting and investment scenario analysis. For Saudi industrial cities, this integrated approach is essential because hydrogen projects must operate within complex relationships among renewable resources, water systems, heavy industry, export markets and Vision 2030 priorities.

The proposed framework positions Saudi industrial cities as complementary nodes within a national hydrogen network. NEOM can demonstrate large-scale renewable hydrogen and export capability; Yanbu and Jubail can connect hydrogen with refining, chemicals and port logistics; Ras Al-Khair can support mining and metals decarbonization; and Dammam can strengthen distribution and eastern industrial demand. AI can coordinate these roles by making decisions transparent, adaptive and evidence-based. The main recommendation is to build hydrogen projects as digital, interoperable and certifiable systems from the beginning. If Saudi Arabia combines renewable resources with strong data governance, explainable AI, cybersecure digital twins, responsible water management and phased industrial offtake, it can create a green hydrogen ecosystem that supports decarbonization, diversification and global competitiveness under Vision 2030.

10. Integrative Synthesis for Network Readiness

This orientation is particularly important for industrial cities because production assets, port infrastructure, water systems and customer contracts mature at different speeds. A phased network model allows planners to protect early investment while keeping later expansion technically and commercially flexible. It also encourages practical collaboration between energy developers, industrial tenants, regulators, technology vendors and financial institutions. Such collaboration is essential because the hydrogen value chain crosses organizational boundaries that conventional project governance rarely manages well. When optimization results are explained clearly, decision-makers can compare cost savings with resilience, emissions performance and water stewardship. This transparency improves investor confidence and supports credible reporting to both national stakeholders and international buyers. The Saudi case is therefore an opportunity to demonstrate how digital infrastructure can become part of clean industrial strategy. Rather than treating AI as an optional software layer, industrial planners should treat it as core enabling infrastructure. The strongest projects will combine engineering discipline, reliable data, safety culture and commercially realistic offtake pathways. This combination can help Saudi Arabia avoid fragmented hydrogen pilots and move toward

scalable industrial transformation. A careful balance between automation and human oversight will remain necessary in every high-pressure hydrogen environment. Engineers, operators and regulators should be able to test model assumptions before relying on operational recommendations. The review therefore supports gradual deployment, continuous validation and measurable performance improvement across each industrial hub. In this way, AI becomes a disciplined planning method rather than a vague innovation slogan. The same logic can guide procurement, workforce training, environmental permitting and long-term infrastructure financing. This orientation is particularly important for industrial cities because production assets, port infrastructure, water systems and customer contracts mature at different speeds. A phased network model allows planners to protect early investment while keeping later expansion technically and commercially flexible. It also encourages practical collaboration between energy developers, industrial tenants, regulators, technology vendors and financial institutions. Such collaboration is essential because the hydrogen value chain crosses organizational boundaries that conventional project governance rarely manages well. When optimization results are explained clearly, decision-makers can compare cost savings with resilience, emissions performance and water stewardship. This transparency improves investor confidence and supports credible reporting to both national stakeholders and international buyers. The Saudi case is therefore an opportunity to demonstrate how digital infrastructure can become part of clean industrial strategy. Rather than treating AI as an optional software layer, industrial planners should treat it as core enabling infrastructure. The strongest projects will combine engineering discipline, reliable data, safety culture and commercially realistic offtake pathways. This combination can help Saudi Arabia avoid fragmented hydrogen pilots and move toward scalable industrial transformation. Precise data governance strengthens safe scalable bankable hydrogen deployment across Saudi industrial corridors.

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