




Can Blockchain Strengthen the Energy Internet?

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Abstract: Emergence of the Energy Internet (EI) demands restructuring of traditional electricity grids to integrate heterogeneous energy sources, distribution network management with grid intelligence and big data management. This paradigm shift is considered to be a breakthrough in the energy industry towards facilitating autonomous and decentralized grid operations while maximizing the utilization of Distributed Generation (DG). Blockchain has been identified as a disruptive technology enabler for the realization of EI to facilitate reliable, self-operated energy delivery. In this paper, we highlight six key directions towards utilizing blockchain capabilities to realize the envisaged EI. We elaborate the challenges in each direction and highlight the role of blockchain in addressing them. Furthermore, we summarize the future research directive in achieving fully autonomous and decentralized electricity distribution networks, which will be known as Energy Internet.

Keywords: Energy Internet, Smart Grid 2.0, Blockchains, 6G, Key directions, Limitations and challenges

1. Introduction

The concept of smart grids emanated with the adoption of Internet of Things (IoT) devices and technologies, such as Internet connected advanced sensors and smart meters in electricity grids [1,2]. This facilitated bi-directional information flow to achieve near real-time grid operations while incorporating dynamic electricity pricing and effective Demand Response (DR) mechanisms [3]. Meanwhile, the energy requirement of the world is expected to grow continuously [4]. Catering to this requirement demands an increasing number of grid interconnections of Distributed Energy Resources (DER) contributing synergistically with one another. However, conventional smart grids, which are governed by a centralized authority, are not capable of facilitating this requirement. This urges the need for a more scalable, flexible and distributed grid architecture [4].

In response to this demand, the next iteration of the conventional smart grids, Smart Grid 2.0 (identified as the Energy Internet (EI) in this paper), is being realized to establish bi-directional energy and information transfer. This uses the electricity grid infrastructure and Internet Protocol (IP) based features of the Internet respectively [5]. This includes Distributed Network Protocol (DNP3) for maximizing the utilization of DERs and Transmission Control Protocol (TCP) based protocols, such as, IPv6 over Low-Power wireless Personal Area Networks (6LowPAN), to facilitate communication between compact, inexpensive, low-power, embedded devices and IEEE 802.15.4 networks. This is to enhance security features of the existing smart grids. Furthermore, EI facilitates real-time information exchange including energy usage data, dynamic pricing information and control signals through such implementations.

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IoT devices, including smart meters and sensors, communicate real-time measurement data of the large-scale participation of the Distributed Generation (DG) [6]. This is envisaged to facilitate autonomous operation of energy grids, benefiting seamless integration of DG without the involvement of a third party compared to the conventional counterpart. Implementation of EI grids is proposed as an overlay of four layers; namely, physical layer, communication and control layer, application layer and data analysis layer [7]. The former two comprise of IoT devices and beyond 5G communication technologies, enabled through edge computing respectively. The latter two layers of the novel architecture incorporate the applications of the envisaged EI grid and data analysis technologies supported through big data management [8]. Applications of EI span beyond offering dynamic energy prices to the consumers and obtaining their contribution in DR initiatives. They also exhibit prospects in multi-dimensional aspects, including: 1) Peer-to-Peer (P2P) energy trading; 2) plug-and-play interfacing for DERs; 3) microgeneration; 4) Demand Side Integration (DSI); 5) automation and management of distribution networks; and 6) management of energy data. Figure 1 illustrates the interrelationship of these applications.

Thus, EI grid architecture facilitates the paradigm shift from the monopoly vertical hierarchy towards a decentralized network configuration with bi-directional energy and information exchange across the grid and internet respectively. A comparison elaborating the significant differences of the conventional and next generation smart grids is presented in Table 1.

Together with these applications, EI envisages autonomous grid operation where the central authority governing the grid under the current context will be overlooked [4]. Further, this would be protruding as consumers begin to gain liberalization in the energy market and actively participate in power production. Human intervention in the decision-making process will be automated through smart contracts, facilitated by Artificial Intelligence (AI) and Machine Learning (ML) algorithms. However, as a consequence of delegation of authority among stakeholders and alleviating the contribution of the intermediary, trust establishment would become a key consideration regarding EI grids. Additionally, the cyber-physical system created by the increasing number of stakeholders connecting to the grid through the diverse applications of EI would result in innumerable access points and large data sets, which would elevate its vulnerability towards malicious

Table 1: Comparison of smart grids and EI grids [6,9,10]

Conventional Smart Grids	EI Grids / Smart Grid 2.0
Centralized power distribution	Decentralized power distribution
Integration of limited energy resources (i.e., conventional generation and a few Renewable Energy Sources (RES))	Integration of heterogeneous DER including RES, EVs and ESS
Utility-centric operations where utility governs the ownership of the grid Distribution System Operator (DSO)	Consumer-centric operations with equal level playing field for all participating stakeholders
Energy traded and information exchanged between the DSO and the customers	Energy traded and real-time information shared in a peer-to-peer manner
Closed proprietary and non-interoperable Information and Communication Technology (ICT) is the disruptive technology enabler	Open and inter-operable technologies based on IP is the disruptive technology enabler
Impedes autonomous grid operation	Integrates AI and ML technologies to enable autonomous grid operations

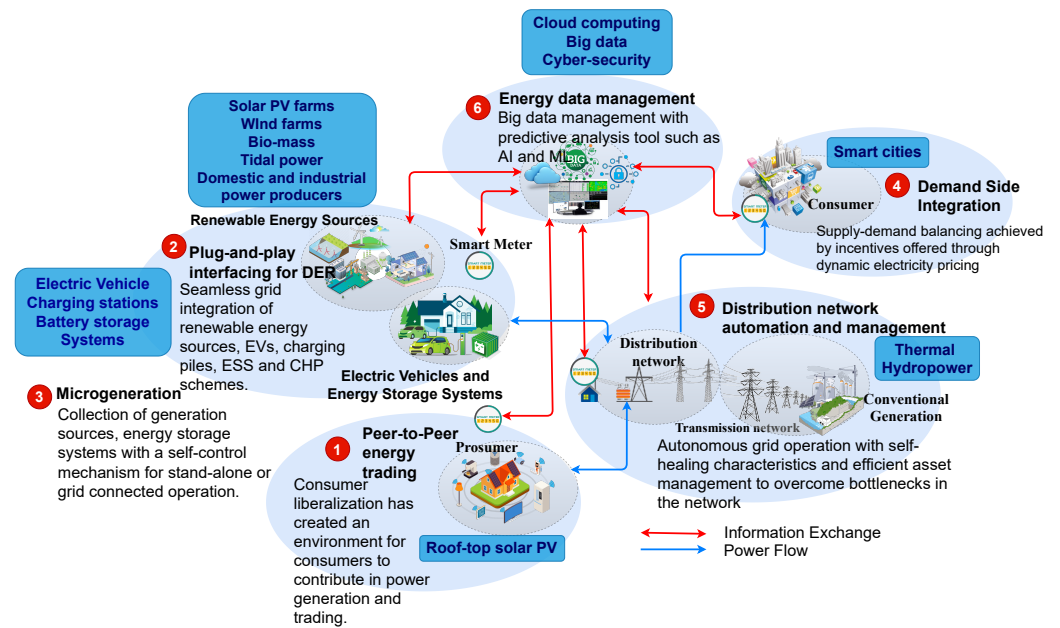


Figure 1. Overview of the bi-directional energy and information flows in the applications of the envisaged EI grids

68 attacks. Performing grid operations with a large number of heterogeneous access points
 69 will be challenging. This would require secure and reliable communication channels for
 70 control implementation using the aggregated data, which requires to be prevented from
 71 information leakage [4].

72 EI will be enabled through blockchain, which is a Distributed Ledger Technology
 73 (DLT) with inherent features including immutability, transparency, distributed verifi-
 74 cation/storage and decentralized authority over a peer-to-peer network [11]. Security
 75 features and privacy-preserving techniques incorporated with blockchains offer solu-
 76 tions to mitigate cyber-physical attacks and privacy violations within the operations
 77 of the EI grid. Smart contracts enable autonomous operations of the EI grid with the
 78 execution of programmed scripts upon the fulfilment of the defined pre-requisites [12].
 79 Some processes could be automated through the utilization of smart contracts imple-
 80 mented upon blockchain platforms. These include billing for the energy consumption,
 81 invocation and revocation of certificates to authorize heterogeneous DER integration,
 82 authorizing payments upon energy trading, dynamic price signalling and monitoring
 83 IoT devices to identify node tampering [13–15]. Realisation of EI is also envisaged to be
 84 facilitated through the developments of beyond 5G and 6G communication networks
 85 through inherent features. These include DLT/blockchain, ultra-massive machine-type
 86 communication, extremely low-power communication, extremely reliable low-latency
 87 communication, AI and ML, big data management and distributed processing through
 88 edge intelligence [16–21].

89 Even though blockchain is expected to become a key enabler of EI, the integration of
 90 blockchain platforms with EI has not been investigated to a considerable extent [4,9,22].
 91 This offers research directives in abundance and to address the identified research gap,
 92 this paper presents six key directions of blockchain utilization in EI grid realization.
 93 These are 1) energy sustainability through heterogeneity; 2) improved trust, security and
 94 privacy; 3) ultimate grid reliability and stability; 4) decentralized scalability; 5) advanced
 95 big data management; and 6) grid intelligence. The significance of inherent features
 96 of the blockchain, utilized towards realization of the next generation of smart grids
 97 in the identified directions, have been illustrated in Figure 2. Challenges pertaining
 98 in each identified direction and the role of blockchain utilization have been discussed,
 99 summarizing the future research directive in achieving fully autonomous and distributed

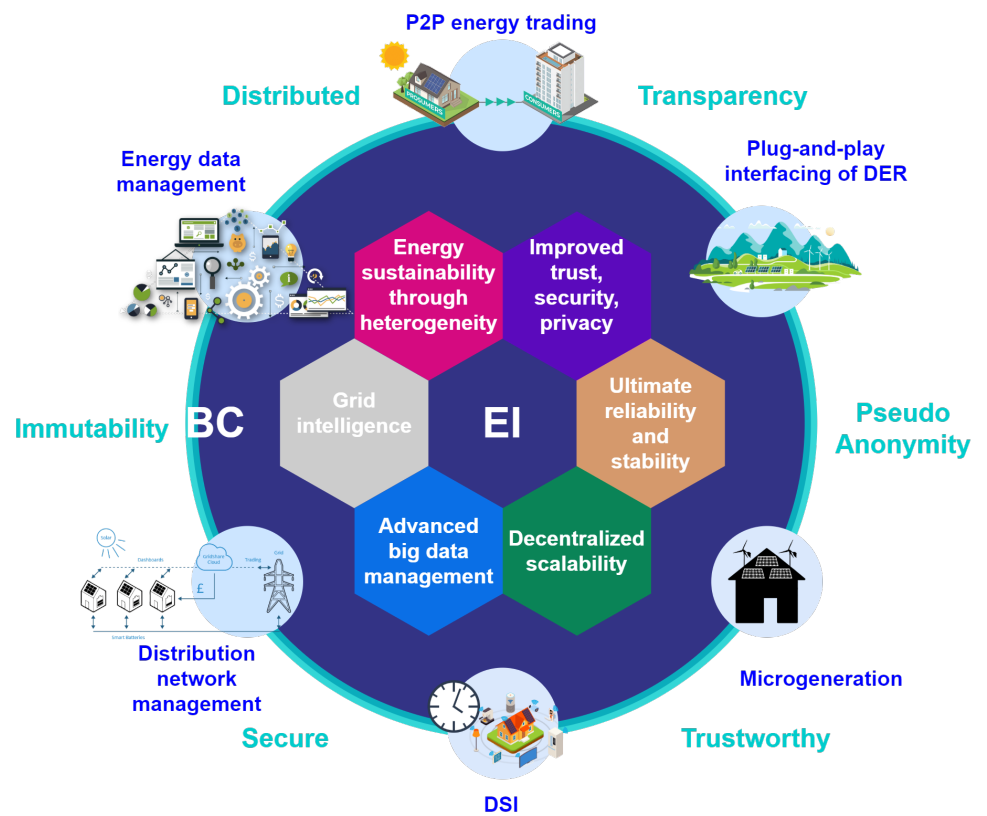


Figure 2. Key directions of blockchain utilization in EI realization

100 electricity distribution networks [16]. Hence, this paper aims to present a clear insight
 101 of blockchain utilization for strengthening EI grids to deliver maximum benefits to all
 102 participating stakeholders.

103 The rest of the paper is organized as follows: Section 2 provides an insight regarding
 104 how blockchains can facilitate futuristic electricity distribution to be sustainable through
 105 heterogeneity. Section 3 highlights the improvement achieved in security, privacy and
 106 trust establishment without the involvement of a third party, related to operations
 107 of the future EI grids. The benefits identified through the integration of blockchains
 108 include mitigating cyber-physical attacks leading to disruptions in the electricity supply,
 109 preventing breaches of privacy and instating trust in distributed operation. Section 4
 110 signifies the contribution of blockchain-based EI grids towards achieving grid stability
 111 and reliability, while section 5 discusses the benefits offered through improved scalability
 112 to cater the increasing number of stakeholders connecting. Section 6 and section 7
 113 elaborate on the role of blockchain to facilitate predictive data analysis integrated with
 114 EI data management. Each of these sections highlight the role of blockchains in this
 115 progression and the way forward in each individual aspect.

116 2. Energy Sustainability through Heterogeneity

117 Conventional power grids rely mostly on limited variants of energy sources for
 118 the fulfilment of the demand [23]. Renewable energy generation, intermittent in nature,
 119 geographically dispersed and located in the close proximity of the load centres, is
 120 replacing fossil fuel-based generation with a high carbon footprint [5,24]. Energy Storage
 121 Systems (ESS) are utilised with the intention of maximizing the benefits of the renewable
 122 sources [9]. Energy security is diversified through the implementation of Combined
 123 Heat and Power (CHP), using fuel cells where the waste heat from electricity production
 124 is utilized for space heating. Consumers who have gained more authority under the
 125 envisaged grid context can trade the excess power production from their domestic
 126 installations, achieving the benefit of the dynamic real-time energy prices. Electric

127 Vehicles (EV) are receiving attention as a green transportation alternative while the
128 charging stations are located across a large geographical area to facilitate their usage [4].
129 Incorporating these heterogeneous energy sources from various stakeholders into a
130 single platform would improve the sustainable energy usage.

131 However, the following challenges can be identified in the road map of integrating
132 heterogeneous energy sources to enable envisaged EI grid operations.

133 2.1. Key Challenges

- 134 1. **Seamless grid integration:** How can we facilitate seamless grid integration of
135 DERs in order to achieve a decentralized, customer-centric grid architecture [24]?
- 136 2. **Decentralized marketplace:** How is it possible to achieve a decentralized market-
137 place with dynamic price signaling and maximized consumer satisfaction [25]?
- 138 3. **Secure communication:** How can we manage secure communication links and
139 storage for large energy data aggregation arising from the increasing number of
140 grid interconnections in a decentralized and secure platform?
- 141 4. **Transient and dynamics:** How can we mitigate the transient over voltages and un-
142 desirable dynamics resulting from bi-directional energy routing and uncoordinated
143 grid interconnections of DG, ESS and EV? [26]
- 144 5. **Improving interoperability:** How can we achieve interoperability of heteroge-
145 neous energy sources that adopt different grid interconnection standards?

146 2.2. Role of Blockchain

147 Smart contracts can be utilized to ensure seamless integration of DERs through
148 invocation of certificates and revocation of them upon request. Blockchain establishes a
149 trusted environment in a decentralized marketplace, which provides confidence for the
150 prosumers to engage in P2P energy trading while avoiding the risk of non-repudiation
151 and double spending [13]. Microgrid implementations such as Brooklyn microgrid [27]
152 and Power Ledger [28] in Australia could be identified as promising, decentralized
153 blockchain-based solutions facilitating P2P trading. Furthermore, Share & Charge in
154 Germany and Juice Net of North America are milestone projects in decentralized energy
155 markets for P2P EV charging [14]. Inherent cryptographic encryption of blockchain
156 would enable user authentication without third-party intervention. Blockchain could
157 further function as a secure and reliable data communication link and storage platform
158 for efficient management of large data sets associated with these diverse EI applications.
159 Energy sustainability is assured with transparent exchange of heterogeneous forms of
160 energy [13].

161 However, mitigation of transient overvoltages and issues related to interoperability,
162 which arise while the implementation of heterogeneous EI grids cannot be facilitated
163 through a blockchain platform itself. Adequate standards need to be implemented with
164 the intention of sustaining desirable grid operations.

165 2.3. Future Directions

166 The incentive-based benefits obtained through real-time dynamic electricity pricing
167 are receiving attention in the grid integration of DERs beyond small scale capacities at
168 different voltage levels of the grid [29]. This will be facilitated through 5G and beyond 5G
169 technologies, offering ultra-reliable, low-latency communication and edge intelligence
170 supporting remote communication for intermittent connectivity respectively. Blockchain
171 would provide an overlay with distributed storage facility. Precision decision-making,
172 incorporating AI along with reliable communication and data processing through edge
173 devices, as envisaged with 6G, would enable the realization of next-generation electricity
174 networks [30].

175 **Lessons:** Decentralized energy trading is a well-established research area with several
176 approaches proposed and real-world scenarios being implemented [27,28,31–35]. Ex-
177 amples include P2P trading of renewable energy [34,36,37]. Communication links are

178 expected to be made secure through proposed blockchain integrated architecture, which,
179 however, has room for improvement with novel Smart Grid 2.0-specific security threats
180 to be addressed [38]. Meanwhile, facilitating seamless grid integration of DGs [13] and
181 their interoperability [9,22] are the challenges with future research prospects consid-
182 ering the existing work, which would require collaborative technological approaches
183 facilitated through blockchain.

184 3. Improved Trust, Security and Privacy

185 Future energy grids have envisaged seamless peer-to-peer connectivity with the
186 autonomous operation in contrast to conventional smart grid context. This is where
187 the Distribution System Operator (DSO) governs the ownership and authority over the
188 management of infrastructure, certificate invocation and revocation for DER integration
189 and supervision of energy exchange with the grid [27,39]. Consumers liberating in the
190 envisaged EI grids would promote microgeneration where energy is traded between
191 two nodes of the network without the involvement of a third party, reducing the losses
192 and additional cost incurred. A mechanism which could establish the trust factor
193 with minimal middle-man involvement would drive the energy grids towards the
194 expectations.

195 Furthermore, EI grid operation aggregates data related to real-time energy con-
196 sumption through smart meters, electrical measurements obtained by IoT sensors, bids
197 to trade excess energy, requests to fulfil demand deficit and control signals for grid
198 regulation. The key considerations that strengthen the future energy network would be
199 privacy-preserving protocols to eliminate the risk of revealing individual energy usage
200 patterns, exposing consumer identity and disclosing information to a third party without
201 the consent of the user. Additional key considerations would be secure operations
202 through reduced vulnerabilities of the grid towards physical attacks, software attacks,
203 network attacks, control-related attacks and encryption attacks [40]. The increase in the
204 number of access points connected to the network and the heterogeneity of the devices
205 observed would have a direct impact over the management of trust, security and privacy
206 issues of the speculated EI grids [41].

207 The following challenges were identified, impeding the secure and privacy-protected
208 operations of EI grids for which a blockchain platform would be a promising solution.

209 3.1. Key Challenges

- 210 1. **Device tampering:** How can we prevent tampering and unauthorized accessing
211 of smart meters and smart sensors to ensure integrity of the obtained energy
212 measurements [41]?
- 213 2. **Man-in-the-Middle attacks in EI grids:** How can we establish a secure communi-
214 cation link between the prosumer and the consumer during energy trading and
215 prevent Man-in-the-Middle attacks causing data manipulation?
- 216 3. **DDoS attacks in EI grids:** How will it be possible to detect Distributed Denial of
217 Service (DDoS) attacks causing deliberate traffic of energy requests and depriving
218 the legitimate users from consuming energy [42]?
- 219 4. **Privacy issues:** Can a consumer participate in DSI initiatives while preserving
220 the privacy of energy consumption data which can trace back to the behavioural
221 patterns of the user?
- 222 5. **Authentication:** How can the identity of a node in the energy grid be verified in a
223 decentralized architecture without revealing the connection between the energy
224 signature and the owner's name and location?
- 225 6. **AI and ML related-attacks:** How can we mitigate data poisoning attacks related
226 to integration of AI and ML techniques in predictive data analysis [43,44]?

227 3.2. *Role of Blockchain*

228 Blockchain platform inherently establishes trust with minimal external interven-
229 tions while offering a secure and transparent mechanism to create a reliable link between
230 the nodes participating in energy trading. Transactions are recorded in an immutable
231 and transparent format while each node holds a copy of the current ledger [22], pre-
232 venting data modification and false data injection. Smart contracts could automate
233 processes such as billing and finance settlement without the requirement of a trusted
234 third-party intervention at a cost, while blockchains with the inherent use of Public
235 Key Infrastructure (PKI) would enable identity authentication with pseudo-anonymity,
236 privacy preservation of the participating nodes and protection of data integrity [45]. The
237 Lightning Network and Smart Contract (LNSC) model proposed in [46] offers a security
238 model comprising of registration, scheduling, authentication and charging phases. This
239 integrates security options for user authentication, facilitating secure mechanisms for
240 charging and discharging EVs. Guardtime, a US-funded project, has utilized a keyless
241 authentication scheme for scalable EI grids with hash-function cryptography and digital
242 signature authentication [22].

243 Cryptanalysis, in which breach of encryption algorithm is observed, can be ad-
244 dressed through the digital signatures incorporating private-public key pair, which is
245 unique to each stakeholder. AI and ML models introduce a new set of adversaries,
246 including data poisoning attacks, model evasion, extraction and inversion-related ML
247 techniques utilized in EI grid realization [43]. Data poisoning can be mitigated through
248 the incorporation of blockchain distributed data storage, while alternatives such as
249 adversarial machine learning, moving target defence and defensive distillation would
250 provide resilience against adversaries identified in ML models [43].

251 Even though tampering of IoT devices and undesirable data traffic in communica-
252 tion channels cannot be fully addressed through blockchain initiatives, such platforms
253 can be utilized to monitor the scenario and execute corrective measures to minimize the
254 damage. Further, the existing security and privacy-preserving mechanisms incorporat-
255 ing cryptography and 51% attacks on the blockchain-based applications are vulnerable
256 to advancements in quantum computing, thus demanding for quantum-resilient security
257 alternatives [47–49].

258 3.3. *Future Directions*

259 Under the current blockchain context, immutability of the distributed ledger has
260 been exploited to establish the trust factor and verify information security. However,
261 the developments emerging with 6G technology facilitate distributed computing uti-
262 lizing edge devices, which could further enable consolidation of resources to achieve
263 computational efficiency [50]. Such collaborations would pose a risk of accumulating 51
264 % authority over the peer nodes, thereby gaining capabilities for modification of the past
265 records. Such adversaries should be addressed in the envisaged grid operation.

266 Further, integrity and confidentiality of the information, along with the identity of
267 the user, could be compromised through the revealing of the public and private keys
268 used in PKI. This could be mainly due to the prolonged usage of the keys and as a result
269 of the malicious attempts to reveal these cryptographic text patterns utilizing quantum
270 computing, which is the most recent development enabling extensive computational
271 capabilities [9]. Ensuring security and privacy with technological progressions would be
272 challenging in future grid implementations [51].

273 **Lessons:** Blockchain integration with Smart Grid 2.0 has facilitated in mitigating soft-
274 ware and network related attacks, including Man-in-the-Middle and DDoS adver-
275 saries [38]. Further, the existing work has proposed different user authentication and
276 privacy-preserving approaches, which have been implemented through cryptographic
277 techniques used in blockchain [46,52–57]. The most widely adapted approach could be
278 identified as cryptographic encryption-based digital signatures for user verification [38].
279 However, modern smart grids, which are to incorporate predictive data analytic tools

280 for intelligent decision, are vulnerable to AI and ML-related attacks [16,17]. These
281 adversaries would overlay the security threats governing Smart Grid 2.0, which would
282 require accelerating the existing research initiatives.

283 4. Ultimate Reliability and Stability

284 Future consumers would heavily rely on electricity through the utilization of smart
285 appliances with the onset of smart buildings, while the future electricity grids are envis-
286 aged to rely more on intermittent generation, including renewable power production
287 incorporated with ESS [1]. Grid stability and reliability of the power supply become
288 vital factors of the speculated EI architecture [58].

289 The intermittent operation of the renewable generation and the energy consumption
290 patterns of the dynamic loads are difficult to predict. Stability achieved through the grid
291 surveillance performed using IoT devices, facilitated by Deep Learning (DL) techniques,
292 would drive the expectations of the future electricity grids while managing the dynamic
293 nature of both generation and loads [59].

294 Delivering a reliable power supply with the stochastic nature of the electricity
295 generation and consumer demands real-time monitoring through IoT devices, precise
296 decision-making capabilities empowered through AI and ML-enabled big data manage-
297 ment, efficient information exchange, data processing through advanced communication
298 protocols and distributed computing. These would further be the pillars that strengthen
299 the future EI grids [17].

300 However, the following challenges have to be addressed in order to ensure stable
301 and reliable grid operations.

302 4.1. Key Challenges

- 303 1. **Supply-demand balancing:** How can we facilitate seamless integration of DG
304 to achieve dynamic response in power output to rectify supply-demand mis-
305 match [24]?
- 306 2. **Intermittent generation:** What will the possibility be of securing stable and reliable
307 grid operation in a decentralized architecture with no third-party involvement and
308 heterogeneous grid interconnections?
- 309 3. **Secure communication:** How can we facilitate secure communication links to
310 improve the exchange of energy data and control signals between peers to improve
311 stability and reliability of a distributed grid?
- 312 4. **Intelligent decision-making:** How do we arrive at intelligent decisions for optimal
313 generation allocation to improve grid stability management ?
- 314 5. **Energy theft:** How can we prevent energy theft, ensuring the consumer a reliable
315 energy supply?
- 316 6. **Power quality management:** How can we mitigate the issues related to non-
317 compliance of power quality standards by the prosumer, DG owner and consumer?

318 4.2. Role of Blockchain

319 Blockchain and smart contracts can be utilized to invoke certification for seam-
320 less grid integration of renewable energy generation upon fulfilment of the prerequi-
321 sites. This would decrease the latency and improve grid stability [60,61]. Lightweight
322 blockchain platforms would further provide means of increasing the transaction through-
323 put of the energy grid. Efficient correction of supply-demand mismatch through pre-
324 cision decisions arrived upon predictive analysis would facilitate reliable future grids.
325 Blockchains would be the means for secure storage and broadcasting of energy bids
326 and demand requests, enabling the predictive analysis on aggregated EI data. The in-
327 tegrity of data sets used for the application of AI and ML technologies could be ensured
328 through the cryptographic encryption methods incorporated with blockchain [30]. The
329 Spanish renewable initiative Iberdrola is utilizing blockchain for tracking of wind power
330 generation and is expected to contribute in seamless grid integration of DER by issuing

331 origin certification [14]. The decentralized solution eliminates the need for a third party
332 as a middle man.

333 Blockchain, however, cannot be identified as the ultimate solution for ensuring
334 stability and reliability of futuristic decentralized grid architecture with no central au-
335 thority. Advanced control strategies need to be proposed in this aspect, with blockchain
336 facilitating them from the rear end by providing secure data transmission and storage.
337 Deployment of smart contracts would enable autonomous execution of the control
338 strategies for securing grid stability.

339 4.3. Future Directions

340 Grid stability and reliability are expected to progressively advance through AI
341 and ML algorithms, enhancing the capabilities of the future grids beyond expectations.
342 Deployment of Virtual Power Plants (VPP) and Autonomous Vehicles (AV) contributing
343 in Vehicle-to-Grid (V2G) and Vehicle-to-Vehicle (V2V) interactions match the supply
344 with the demand. Meanwhile, smart cities that integrate smart buildings with intelligent
345 appliances would reshape the future grid architecture, with autonomous stability and
346 reliability management approaches being a requisite [4,62].

347 **Lessons: Elimination of energy theft through double auction and other mechanisms has**
348 **been presented in the existing literature [13,16,52,57,63]. However, securing commu-**
349 **nication channels to offer a reliable electricity supply while addressing the challenges**
350 **related to intermittent generation of DGs have research prospects for maximization of**
351 **renewable energy utilization in Smart Grid 2.0 [13,38]. This would enable matching the**
352 **supply with the demand; however, maintaining the power quality within the allowable**
353 **limits [52] would be another challenge to overcome in the envisaged grid architecture,**
354 **with few research being carried out related to this aspect.**

355 5. Decentralized Scalability

356 EI envisages millions of seamless interconnections of independent producers, mi-
357 crogeneration comprising of solar PV, wind, fuel cells, ESS for maximizing the benefits
358 of intermittent renewable generation and EVs with their charging stations dispersed in
359 a large geographical area [3]. IoT devices connected to this network would, thereby,
360 aggregate large volumes of transaction and energy information. These access points
361 increase in their numbers, owing to the attention received by the EI concept in the future
362 grid architecture where decentralized scalability would be a role player. 6G-enabled dis-
363 tributed processing by edge computing, cloud data storage, AI and ML-based competent
364 control algorithms, information security and privacy-protected data can be identified as
365 the facilitators of decentralized scalability of the envisaged EI grids [17,58].

366 Challenges to be addressed in terms of achieving decentralized scalability, with the
367 onset of increasing numbers of grid interconnections, have been discussed below.

368 5.1. Key Challenges

- 369 1. **Scalable, decentralized EI grids:** How can we offer scalable solutions for de-
370 centralized energy grids which would facilitate integration of large numbers of
371 heterogeneous generation sources and extending the customer base [4]?
- 372 2. **Low-latency grid synchronization:** How can we achieve low-latency decision-
373 making for the synchronization of large numbers of connected DGs [61]?
- 374 3. **Scalable data management:** How can we manage the large energy data set gener-
375 ated by the continuously expanding consumer base of future EI grids?

376 5.2. Role of Blockchain

377 Inherent features of blockchain enable the delegation of processing to edge nodes
378 and cloud storage platforms through a trustless trust establishment. Low latency and
379 high throughput can be achieved through distributed processing which further provides
380 solutions to intermittent connectivity in remote areas in a store and forward manner.

381 PETCON, a secure P2P trading mechanism for plug-in hybrid EVs [64], and Guardtime,
382 a permissioned blockchain-based approach, are considered to be scalable solutions for
383 complex data exchange in EI grids.

384 Further, with the incorporation of off-chains, side-chains, sharding and edge com-
385 puting to offload the computational burden, better results have been obtained to cater
386 for the increase of nodes connected with minimum trust issues. This would be further
387 empowered with the speculated advancements in communication infrastructure by de-
388 ployment of 6G technologies, enabling edge intelligence [30]. DL and big data analysis
389 applied on the data aggregated in cloud storage would be controlled and managed
390 through a secure mechanism utilizing blockchain platforms. Leakage of information
391 would be shielded and better control over the data can be achieved through such ap-
392 proaches [17].

393 Scalability issues related to EI grid implementations, however, have not been
394 fully overcome through the blockchain platforms. These can be identified as future
395 research prospects for the improvement of the applicability of blockchain for energy grid
396 management.

397 5.3. Future Directions

398 Achieving distributed scalability at the cost of compromising information security
399 and privacy of the user respectively defies the expectations of the future grids. The
400 blockchain trilemma would receive attention in the pathway towards successful imple-
401 mentation of the futuristic grid infrastructure [65]. The vulnerabilities in grid security
402 would expand beyond the current extent with the scalable, decentralized operations.
403 Thus, proper measures would be a necessity to instate the trust in the EI grids.

404 Further, connectivity has a major impact in reaching at the scalability goals, which
405 can be addressed through advancements such as 5G and 6G technologies. The former
406 would enable low-latency communication channels while the latter would address the
407 intermittent connectivity in rural geographical locations by utilizing edge devices [30].
408 **Lessons: Scalability of EI grids are partly addressed through off-chain and side-chain**
409 **implementations [66] as well as suitable selection of the blockchain platform [22,33,67].**
410 This includes scalable initiatives such as NRG Xchange [33] and analytical selection
411 of the blockchain platform such as HyperLedger over Ethereum [65]. Energy data
412 management with the increasing number of consumer connections would be a challenge
413 to be addressed, while facilitating low-latency grid synchronization of DERs has not
414 been discussed in the existing literature [13,66,68].

415 6. Advanced Big Data Management

416 Big data management facilitates performing of predictive analysis on aggregated
417 large data sets using AI and ML algorithms [69]. Such data set will be of utmost
418 value to a large community of stakeholders. In the instance of the energy sector, this
419 includes power producers, consumers, utilities and non-power participants such as
420 policymakers, investors and financial institutions. **The data** set can be utilized in various
421 applications in order to facilitate EI grid transformation [4]. Grid stability and reliability
422 of power supply are secured through the predictive analysis of this aggregated data
423 using AI and ML approaches and, further, by stimulating autonomous operations [17].
424 Energy sustainability achieved through the heterogeneity of the sources is managed and
425 coordinated using the effective utilization of measurements obtained from the connected
426 IoT devices.

427 However, the aggregated data should be protected against malicious attacks which
428 could manipulate the information, sabotaging the grid operations while preventing leak-
429 age of information to external parties. Big data management would, thus, be considered
430 as an important entity, driving the practical realization of the future grids [4].

431 6.1. *Key Challenges*

- 432 1. **Data silos:** How can we overcome data silos and establish trust between prosumers,
433 microgrids and large power plants for better coordination?
- 434 2. **Secure communication:** How can we achieve secure communication channels
435 between smart meters/smart sensor nodes and the Energy Management System
436 (EMS) [41]?
- 437 3. **Secure data storage:** How can we provide secure, privacy-preserving and scalable
438 storage for the aggregated large data sets containing generation and consumption
439 patterns of consumers and prosumers respectively?
- 440 4. **Data integrity protection:** How can we ensure the integrity of the stored energy
441 data utilized for AI model development, training, validation through ML tech-
442 niques, testing and deployment [41]?
- 443 5. **Data ownership:** How can we ensure ownership of the aggregated energy con-
444 sumption/production pattern data to prevent privacy-violations arising from unau-
445 thorized trading of these sensitive data to a third party?
- 446 6. **Scalable grids:** How **can we** facilitate the management of large data volumes while
447 offering scalability for grid expansion with numerous grid integration of prosumers,
448 microgrids, EVs and collaborative consumers participating in DSIs?

449 6.2. *Role of Blockchain*

450 Communication technologies are progressing towards delivering an efficient, ultra-
451 reliable, low-latency service to the consumer through 5G, thereby facilitating the data
452 aggregation process. Further, 6G network operation enables delegation of the compu-
453 tational capabilities for the processing of information to multiple edge devices [21,30].
454 Blockchain will be an integral part of both of these scenarios and its integration would
455 enable secure data, control signal transmission and trust establishment for distributed
456 processing using edge computing respectively [70].

457 Data aggregation further involves secure data storage on cloud-based platforms,
458 in which blockchain would ensure cyber-security, information security and network
459 security, preventing malicious attacks that could modify data. The hash function in-
460 corporated in blockchain ensures the integrity of the stored data. Moreover, privacy
461 could be enacted, gaining control over the information through the deployment of smart
462 contracts for data sharing while maintaining anonymity [71,72].

463 Predictive analysis upon the aggregated data will be facilitated through AI and ML
464 , in which blockchains could contribute as a trusted mediator. This would ensure the
465 integrity of the data incorporated and the algorithms compiled [17] while deploying
466 autonomous operations through smart contracts.

467 Blockchain alone, however, cannot facilitate scalable platforms for big data man-
468 agement. This demands alternative distributed storage platforms supported by the
469 blockchain from the back end, which include utilization of off-chain storage and Inter-
470 Planetary File System (IPFS).

471 6.3. *Future Directions*

472 Blockchains in collaboration with smart contracts could extend the utilization of the
473 aggregated data set where individual stakeholders (consumer, prosumer and individual
474 power producers) would trade the information to potential investors, asset managers
475 such as financial institutions, utilities and policy makers to obtain financial benefits.
476 The secure and transparent link will be established through 6G architecture and the
477 blockchain platform [73].

478 **Lessons:** Future EI grids will be integrated with AI and ML for predictive data anal-
479 ysis, giving rise to a new set of challenges which were not encountered in previous
480 generations of smart grids [74]. Blockchain integration with EI grids have facilitated
481 data integrity protection through cryptographic hashing [38] and is well addressed in
482 the existing literature. However, addressing the challenges, including data silos [75],

483 facilitating secure, scalable data communication and storage with privacy-preserving
484 data ownership [22,76], would require further research attention.

485 7. Grid Intelligence

486 The trends of the future energy grids have been speculated as autonomous power
487 delivery operation, self-healing fault recovery, efficient fault location identification to
488 reduce system downtime, minimized human interactions in the decision-making process
489 and accurate demand forecasting [62]. AI and ML have been recognised as prospective
490 candidates in these domains, which would stimulate the emergence of intelligent elec-
491 tricity grids. The former would facilitate automated network control through Zero-touch
492 network and Service Management (ZSM) [43] while predictive models will be enabled
493 through ML techniques. Process automation includes automated billing, renewable
494 integration through certificate invocation upon request and revocation owing to non-
495 compliance with the grid pre-requisites, supply-demand balancing, speculation of the
496 load patterns for the scheduling of the DERs, asset management and fault resilience [77].

497 However, AI itself has security and privacy issues which can be a potential instru-
498 ment for launching intelligent attacks.

499 7.1. Key Challenges

- 500 1. **Data manipulation:** How can we mitigate manipulation of energy input data
501 (electricity consumption and production data obtained through smart meters) and
502 validate the authenticity of the information?
- 503 2. **ML:** How can we prevent the model inversion, poisoning pertaining to training
504 and deployment of ML models, used for adaptive decision-making processes in
505 automated generation allocation of EI grids [43]?
- 506 3. **Ethical data aggregation:** How can we ensure ethical use of aggregated energy
507 production/consumption data for AI model training and prevent unauthorized
508 data sharing with compliance to privacy preservation?
- 509 4. **Transparency:** How can we improve transparency in model development, train-
510 ing, testing and deployment, resulting in algorithms that are reliable for diverse
511 applications with grid integration of heterogeneous energy sources?
- 512 5. **Automation:** How can we assure security in AI-based automation of network
513 control and orchestration with it? [43]?
- 514 6. **Trust management:** How can we establish trust among stakeholders participating
515 in energy trading in EI grids and improve transparency in process automation
516 through the deployment of AI models [78]?
- 517 7. **Accountability:** How do we ensure the accountability of the AI algorithms for
518 automated decision-making processes responsible for generation coordination,
519 distribution network management and fault recovery?

520 7.2. Role of Blockchain

521 Intelligent grids arrive at decisions based on the aggregated data set obtained
522 through IoT devices and incorporation of AI and ML techniques [79]. The integrity of
523 the data used in ML models and preventing data poisoning would be the key consid-
524 eration, which directly impacts the decision-making process. With blockchain being
525 an immutable DLT, this would ensure data integrity by preventing data manipulation,
526 injection and corruption [71,80]. Privacy preserving of the large data sets can be achieved
527 through distributed edge computing, facilitated by blockchain platforms where raw data
528 will remain closer to its origin, assuring confidentiality [43].

529 Nevertheless, the authenticity of the algorithms generated through the incorpo-
530 ration of AI and ML cannot be verified through the blockchain platform. Security
531 and privacy preservation in AI based systems, however, cannot be fully facilitated by
532 blockchain platforms and, thus, will require AI-resilient measures. At the same time,
533 this would affect the control decisions obtained for the stable and reliable operations of

534 the decentralized architecture. Standardization and regulatory enforcement are required
535 to overcome such issues arising in the envisaged electricity grids.

536 7.3. Future Directions

537 The advancements in big data and computing technologies have facilitated the
538 emergence of DL techniques for pattern recognition from the aggregated information [81].
539 This would, thereby, enable accurate demand forecasting for optimal load scheduling
540 and load balancing to reduce peak electricity demand, facilitate energy trading and
541 sharing, state estimation of the power grid and perform grid diagnostics for the detection
542 of energy theft. The higher precision achieved in the energy demand speculation would
543 benefit in maximizing the utilization of DER to cater for the demand requirements while
544 maintaining grid stability [82]. Further, fast blockchain-based data-feeding models need
545 to be introduced to facilitate the development of efficient and accurate ML models.

546 Initiatives towards collaborative model development approaches using AI and
547 ML techniques have given rise to Federated Learning (FL), in which individual data
548 storage on a decentralized network is encouraged [83–85]. This facilitates predictive data
549 analysis through training of a shared model using different data sets, offering capabilities
550 of generalized model development with the benefit of privacy preservation of the user
551 data. Blockchain facilitates trust establishment and prevents data poisoning, improving
552 transparency in training, validation, testing, deployment and storage of training data
553 sets for such shared models. Smart contracts enable the automation of iterative processes,
554 improving the flexibility of model development and data analysis. However, challenges
555 raised through the blockchain architecture relating scalability and smart contract security,
556 need to be addressed further to expand the capabilities of FL techniques [85].

557 xAI (explainable AI) is gaining attention in the current context of AI integration with
558 IoT for predictive data analysis. This could improve the transparency of the AI models
559 associated with demand forecasting, evaluation of energy usage patterns, renewable
560 energy modelling and automated grid operation with optimum generation allocation and
561 load scheduling. xAI unravels the black box AI models, thereby improving transparency
562 in predictive data analysis and establishing trust in the decisions arrived through such
563 approaches[78,86].

564 **Lessons:** Grid intelligence would dominate the future autonomous and the challenges
565 arising from AI-integrated smart grids are seldom addressed through the existing lit-
566 erature [74]. Improving transparency to ensure accountability of ML models [16,87]
567 and trust management in the decisions arrived through the models [22,87] would be EI
568 grid-specific challenges to be addressed in future research.

569 8. Discussion

570 The envisaged EI grids intend to integrate a diversity of distributed resources in
571 order to achieve decentralized, autonomous operations. This facilitates consumer liber-
572 alization while enhancing reliability, stability through a secure and privacy-preserving
573 platform. Blockchain is identified as an eminent factor driving the transformation from
574 the hierarchical architecture towards an open market. Blockchain facilitates the realiza-
575 tion of the intelligent grid architecture at the benefit of real-time operation with minimal
576 involvement of an intermediary [97].

577 The paper discusses six directions of future EI architecture realization through
578 the integration of blockchain platforms. Enhanced energy diversity catered with an
579 improved security and privacy-preserving mechanism utilizes the inherent features of
580 the blockchain-based architecture. Grid intelligence along with the EI data management,
581 facilitate the near real-time decision making through predictive data analysis [79]. The
582 ultimate goal of the blockchain integrated EI grids would be to ensure a reliable and
583 stable power supply while maximizing consumer participation in electricity distribu-
584 tion [98]. Further, it is envisaged to achieve delegated authority with the distributed
585 operation, which eliminates the threat of single-point failure.

Table 2: Research direction for blockchain integrated EI grids [88]

Challenges	Blockchain Features						Related Work	
	Decentralization	Traceability	Immutability	Consensus Mechanism	Digital Currency	Smart Contracts	Citations	Contribution
Energy sustainability through heterogeneity								
Seamless grid integration	H	L	L	M	L	H	[13]	Low
Decentralized marketplace	H	H	H	M	H	H	[27,28,31–35]	Very High
Secure communication	M	H	H	L	L	L	[57,89]	High
Transient overvoltages and dynamics	L	L	L	M	L	H	[70,90]	High
Improving interoperability	L	L	L	H	L	H	[9,22]	Low
Improved trust, security and privacy								
Device tampering	L	H	H	L	L	L	[16,38,89,91]	Low
Man-in-the-Middle attacks	H	H	H	L	L	H	[89]	Medium
DDoS attacks in EI grids	H	H	L	L	L	H	[13,38,52]	Low
Privacy issues	H	H	H	M	L	H	[46,52–57]	High
Authentication	L	H	H	L	L	H	[46,53,56,63]	High
AI and ML related attacks	L	H	H	L	L	M	[16,17]	Very Low
Ultimate reliability and stability								
Supply-demand balancing	H	M	L	H	H	H	[13,57,75]	Medium
Intermittent generation	L	H	L	H	L	H	[27,92,93]	Medium
Secure communication	L	H	H	H	L	M	[13,38]	Low
Intelligent decision making	H	M	M	H	L	H	[13,16,75]	Low
Energy theft	L	H	H	H	L	M	[13,16,52,57,63]	High
Power quality management	L	H	H	M	H	H	[52]	Low
Decentralized scalability								
Scalable, decentralized EI grids	H	H	H	H	L	H	[22,33,67]	Medium
Low-latency grid synchronization	L	H	L	H	L	H	[4,9]	Low
Scalable data management	H	H	H	M	L	M	[13,66,68]	Low
Advanced big data management								
Data silos	H	H	H	L	L	M	[22]	Low
Secure communication	M	H	H	M	L	H	[22]	Medium
Secure data storage	H	H	H	M	L	L	[9]	Low
Data integrity protection	H	H	H	H	L	L	[53,63]	High
Data ownership	H	H	H	H	L	L	[94]	Low
Scalable grids	H	H	H	H	L	H	[76]	High
Grid intelligence								
Data manipulation	H	H	H	H	L	H	[52,89]	High
ML attacks	H	H	H	M	L	L	[74]	Very Low
Ethical data aggregation	H	H	H	H	L	L	[63]	High
Transparency	H	H	H	H	L	L	[16,87]	Low
Automation	H	H	H	H	H	H	[35,95]	Medium
Trust management	H	H	H	H	L	H	[22,87]	Low
Accountability	H	H	H	H	L	M	[22,87,95,96]	Low

H High impact

M Medium impact

L Low impact

586 The key challenges related to each direction discussed in the previous sections are
587 summarised in Table 2. The applicability of inherent blockchain features to address these
588 challenges is visually represented. The contribution of the existing work in addressing
589 the key challenges identified for each driving trend has been categorised according to its
590 level of impact. Among the identified challenges, establishment of a decentralized, P2P
591 marketplace is a well established research area having many real-world implementations.
592 Securing communication links, adversarial handling related to privacy, authentication,
593 data manipulation and energy theft, attention given in establishing scalable grids having
594 data integrity preserved are addressed through the existing work up to a reasonable
595 extent. However, further developments are required to fully overcome the challenges
596 present. Challenges which require significant research consideration include facilitating
597 seamless grid integration for DERs enabled with low-latency grid synchronization,
598 ensuring interoperability, power quality management with high penetration of DG,
599 improving accuracy of intelligent decision making, efficient smart grid data management,
600 accountability and trust management. AI and ML related attacks require attention in
601 future grids in the way forward, facilitating predictive data analysis tools.

602 In addition, Table 3 summarizes the level of impact of each identified directions on
603 broad applications of EI grids. Further, the benefits of integrating blockchains in each
604 instance have been elaborated. Practical implementations and examples related to EI
605 grid applications have been considered in this analysis, which verifies the benefits of
606 blockchain integration in future energy grid realizations and signifies the way forward.

607 However, the attention received by these novel electricity grids has attracted more
608 stakeholders where scalability becomes effective. Decentralized scalability is a requi-
609 site to cater the demands of the distributed grid operation with minimal third-party
610 involvement; hence trust establishment mechanism should be transparent. Existing
611 blockchain platforms have not reached the level of maturity to cater the extensive num-
612 bers of stakeholders integrating with the EI grid [65]. Secure and privacy-preserving
613 techniques, which do not trade-off scalable blockchain solutions, need to be considered
614 for future electricity distribution networks to attain maximum benefits. Decentralization,
615 scalability and security/privacy trilogy should be equalized for successful grid transfor-
616 mation from conventional top-down hierarchical architecture to open electricity markets
617 offering the benefit of consumer liberalization.

618 Further, blockchain specific security attacks which are identified during the opera-
619 tions should be eliminated for interruption-free power delivery. The distribute operation
620 and increased consumer participation observed in the electricity trading would lead
621 towards the development of a 51% authority of collective stakeholder contribution,
622 thereby assume charge over the blockchain platform [100]. AI and ML integration in
623 blockchain-based EI grids would trigger security threats related to massive data set [71].

624 The latency observed in consensus-based transaction verification incorporated in
625 blockchains will create adverse impacts on the real-time operations of the envisaged
626 EI networks [101]. Further, the limitations in the storage availability of the distributed
627 ledger platform have to be managed to align with the increase of stakeholder integration
628 with Smart Grid 2.0 [76].

629 Maintenance of blockchains to overcome such challenges related to extensive in-
630 volvement of diversified stakeholders would require customized platforms, which is
631 seldom discussed in the research literature. An energy grid-specific blockchain plat-
632 form could better contribute towards efficient implementation of EI grids, ensuring the
633 decentralization, security and scalability trilogy. Integration of big data management
634 with AI and ML based predictive data analysis leads the way forward towards demand-
635 operative, autonomous, real-time grid operations where trust establishment would be
636 facilitated through a customized blockchain platform.

Table 3: Benefits of blockchains in each EI application [4,10].

EI application	Energy sustainability through heterogeneity	Improved trust, security and privacy	Ultimate reliability and stability	Decentralized scalability	Advanced big data management	Grid intelligence	Benefits attained through blockchain integration	Practical implementations/Examples
P2P energy trading	H	H	M	H	H	H	Reliable, low-latency communication links Less DSO intervention in trust establishment and supply-demand balancing	NRGcoin [33] Bankymoon [99] Pylon [9]
Plug-and-play interfacing of DER	H	H	H	H	H	H	Low-latency grid stability management Trusted, secure and privacy-preserving trading environment Automated certificate invocation and revocation for seamless integration	Greeneum [35] WePower [9] DAJIE blockchain platform [9] Iberdrola [14] Share & Charge [36] Juicenet [34]
Microgeneration	H	H	H	H	H	H	Autonomous supply-demand balancing for standalone operation Trustworthy, secure and privacy-assured P2P trading mechanism Maintaining stability and reliability with minimal DSO participation	Brooklyn microgrid [27] Powerledger [28]
Demand Side Integration (DSI)	L	H	L	L	H	H	Data silos preventing optimal load/generation scheduling Transparency in dynamic electricity pricing and incentive schemes	PETCON [13]
Distribution network automation and management	L	H	H	M	H	H	Asset management and optimal generation scheduling with less DSO participation Efficient and secure communication links to reduce system downtime during a fault Accurate fault diagnostics for automated protection schemes	PROSUME [91] PONTON [22]
Energy data management	M	H	H	H	H	H	Reliable, secure and privacy-assured data communication channels and storage options Authenticity of algorithms developed for data analysis	Enervalis, Jouliette, Energy Bazaar [22]

H High impact

M Medium impact

L Low impact

637 9. Conclusion and Future Research Directions

638 The paper analyzes six directions of blockchain utilization for facilitating the real-
639 ization of the envisaged, intelligent, autonomous EI grids. Energy sustainability through
640 heterogeneity, improved trust, security and privacy, ultimate reliability and stability,
641 decentralized scalability, advanced big data management and grid intelligence are identified
642 as these directives while challenges and the future research for maximizing the
643 benefits of the envisaged architecture are elaborated with respect to each identified key
644 direction.

645 The paper presents a clear insight of the successful integration of blockchain tech-
646 nologies with AI and ML based predictive analysis in future electricity networks, to
647 reshape the energy industry to fully-autonomous, self-resilient grids. xAI, federated ML
648 and DL would enable trusted and transparent grid intelligence in future smart grids.
649 AV era would be realizable through such intelligent smart grids. Beyond 5G communi-
650 cation architecture would facilitate the integration of edge intelligence for distributed
651 processing thereby, offer scalable, efficient and responsive EI grids.

652 Blockchain offers great flexibility in providing security solutions through the in-
653 herent features. However, challenges have been identified in the areas of AI and ML
654 integration, which would require alternatives beyond blockchain.

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664 Abbreviations

665 The following abbreviations are used in this manuscript:

5G	Fifth Generation
6G	Sixth Generation
AI	Artificial Intelligence
AV	Autonomous Vehicles
BC	Blockchain
CHP	Combined Heat and Power
DDoS	Distributed Denial of Service
DER	Distributed Energy Resources
DG	Distributed Generation
DL	Deep Learning
DLT	Distributed Ledger Technology
DNP	Distributed Network Protocol
DR	Demand Response
DSI	Demand Side Integration
DSO	Distribution System Operator
EI	Energy Internet
ESS	Energy Storage Systems
EMS	Energy Management System
666 EV	Electric Vehicles
FDI	False Data Injection
FL	Federated Learning
ICT	information and Communication Technology
IoT	Internet of Things
IP	Internet Protocol
IPFS	Inter Planetary File System
6LowPAN	IPv6 over Low power wireless Personal Area Network
ML	Machine Learning
PKI	Public Key Infrastructure
PV	Photovoltaic
P2P	Peer-to-Peer
RES	Renewable Energy Sources
TCP	Transmission Control Protocol
VPP	Virtual Power Plants
V2G	Vehicle-to-Grid
V2V	Vehicle-to-Vehicle
xAI	Explainable AI
ZSM	Zero-touch network Service Management

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