

Peer-to-Peer, Community Self-Consumption, and Transactive Energy: A Systematic Literature Review of Local Energy Market Models

Timothy Capper^{a,1,*}, Anna Gorbacheva^{b,1}, Mustafa A. Mustafa^{c,d}, Mohamed Bahloul^e, Jan Marc Schwidtal^f, Ruzanna Chitchyan^g, Merlinda Andoni^h, Valentin Robu^{i,j}, Mehdi Montakhabi^k, Ian J. Scott^l, Christina Francis^m, Tanaka Mbavariraⁿ, Juan Manuel Espana^o, Lynne Kiesling^p

^a*Tyndall Centre for Climate Change Research, School of Engineering, The University of Manchester, Oxford Road, Manchester, M13 9PL, United Kingdom*

^b*Energy Institute, University College London, 14 Upper Woburn Place, London, WC1H 0NN, United Kingdom*

^c*Department of Computer Science, The University of Manchester, Oxford Road, Manchester, M13 9PL, United Kingdom*

^d*imec-COSIC, KU Leuven, Kasteelpark Arenberg 10, bus 2452, Leuven-Heverlee, B-3001, Belgium*

^e*International Energy Research Centre, Tyndall National Institute, Cork, Ireland*

^f*Department of Industrial Engineering, University of Padua, Via Giovanni Gradengio 6/a, Padova (PD), 35131, Italy*

^g*Department of Computer Science, University of Bristol, Bristol, BS8 1TH, United Kingdom*

^h*Smart Systems Group, School of Engineering and Physical Sciences, Heriot-Watt University, Edinburgh, EH14 4AS, United Kingdom*

ⁱ*CWI, National Research Institute for Mathematics and Computer Science, Amsterdam, 1098XG, Netherlands*

^j*Algorithmics Group, EEMCS, Delft University of Technology (TU Delft), 2628 XE Delft, Netherlands*

^k*imec-SMIT, Vrije Universiteit Brussel, Pleinlaan 9, Brussels, 1050, Belgium*

^l*NOVA Information Management School (NOVA IMS), Universidade Nova de Lisboa, Campus de Campolide, Lisbon, 1070-312, Portugal*

^m*School of Engineering, University of Edinburgh, Edinburgh, EH9 3FB, United Kingdom*

ⁿ*Institute for Innovation and Technology Management, Lucerne University of Applied Sciences & Arts, Horw, 6048, Switzerland*

^o*Universidad EIA, Vda. El Penasco, Enviado, Antioquia, Colombia*

^p*University of Colorado-Denver, Denver, United States*

Abstract

Peer-to-peer, community or collective self-consumption, and transactive energy markets offer new models for trading energy locally. Over the past five years, there has been significant growth in the amount of academic literature examining how these local energy markets might function. This systematic literature review of 139 peer-reviewed journal articles examines the market designs used in these energy trading models. A modified version of the Business Ecosystem Architecture Modelling framework is used to extract market model information from the literature, and to identify differences and similarities between the models. This paper examines how peer-to-peer, community self-consumption and transactive energy markets are described in current literature. It explores the similarities and differences between these markets in terms of participation, governance structure, topology, and design. This paper systematises peer-to-peer, community self-consumption and transactive energy market designs, identifying six archetypes. Finally, it identifies five evidence gaps which require future research before these markets could be widely adopted. These evidence gaps are the lack of: consideration of physical constraints; a holistic approach to market design and operation; consideration about how these market designs will scale; consideration of information security; and, consideration of market participant privacy.

Word count: 11,320

Keywords: peer-to-peer, community self-consumption, transactive energy, market model, electricity trading, energy trading, smart grid, local energy market, prosumer

*Corresponding author

Email address: timothy.capper@manchester.ac.uk (Timothy Capper)

¹TC and AG have contributed equally to this work.

Nomenclature

CSC	Community or collective self-consumption
DER	Distributed energy resource
DSO	Distribution system operator
EV	Electric vehicle
LEM	Local energy market
P2P	Peer-to-peer
PV	Photovoltaic
TE	Transactive energy
TEAM	The Business Ecosystem Architecture Modelling framework

1. Introduction

Fundamental changes are transforming energy markets globally. Distributed energy resources (DERs), such as photovoltaic (PV) and wind generators, and storage devices are being installed at ever increasing rates [1]. DERs can help to reduce emissions and meet the carbon reduction targets many countries have committed to under the Paris Agreement [2]. However, the intermittent nature of most renewable energy sources creates challenges for network and system operators. Keeping energy supply and demand in balance poses a greater challenge with lower proportions of dispatchable generation. Simultaneously, demand is likely to increase due to the electrification of heating and transportation [3]. Existing energy markets are limited in their ability to respond to these new challenges [4]. To avoid high grid reinforcement costs, and to respond to the changes in load behaviour and volume, new market and balancing mechanisms are needed.

Local energy markets (LEMs) have emerged as a leading approach to foster the integration of more DERs into the electricity system [4]. The purpose of LEMs is to incentivise small energy consumers, producers and prosumers to exchange energy with one another in a competitive market, and to balance energy supply and demand locally [5]. In this literature review, we provide a systematisation of knowledge of the market design and transaction aspects of LEMs. We aim to help researchers in this area understand the types of LEMs being researched and the nuances of the different market types.

Three distinct types of LEM have emerged. Firstly, peer-to-peer (P2P) markets allow direct trading of energy without an intermediary. They aim to provide energy users with an incentive to actively engage in energy markets [6]. Secondly, community or collective self-consumption (CSC) is when co-located energy prosumers trade their surplus energy in a market arrangement [7–9]. The term CSC originates from a regulatory context that focuses on the empowerment of energy users [7]. Its definition is a collection of the participants’ activities, rather than the organisational market structure [8]. Finally, transactive energy (TE) markets balance supply and demand in electricity systems via decentralised coordination [10]. The aim of TE markets is to manage decentralised resources in an autonomous way using price signals to provide system stability [11]. While the three market types share common features, they have distinct characteristics in terms of size, operational scale and the main trading purpose. In the current literature, these LEM types are used interchangeably, with a lack of consensus on their meaning and the differences between the market types.

Several recent review articles analyse LEMs. Khorasany et al. [12] review market designs for local energy trading, focusing on scalability, overheads, and how they address grid constraints. Tushar et al. [13] review P2P electricity trading techniques, providing an overview of their key features and the benefits they bring to the grid and prosumers. Their focus is on market clearing mechanisms. Similarly, Jin et al. [14] classify and organise the literature on market designs and clearing methods, with a focus on local flexibility markets. Tsaousoglou et al. [15] review LEMs focusing on four key

38 attributes of the market: scope, modelling assumptions, objectives, and mechanisms. Sousa et al.
39 [16] review consumer-centric electricity markets, integrating the behaviour of all market participants,
40 not only prosumers. Zhou et al. [17] review P2P market designs, as well as trading platforms,
41 physical and ICT infrastructure, social science perspectives and policy implications. Soto et al.
42 [18] analyse trading platforms, blockchain, game theory, simulations, optimisation methods and
43 algorithms used in P2P markets. Aggarwal et al. [19] focus on optimisation models used in P2P
44 markets, providing a comprehensive taxonomy. Andoni et al. [20] provide a systematic review of how
45 blockchain technology is used in the energy sector. Similarly, Siano et al. [21] explore the application
46 of distributed ledger technology in TE markets, experimenting with different consensus mechanisms.
47 Kirli et al. [22] review the application of smart contracts in energy systems.

48 These review articles make a valuable contribution to the current state-of-the-art. However, the
49 systematisation of knowledge of the market design and transaction aspects of LEMs presented in
50 this paper gives an insight into the different applications of these markets. It outlines the underlying
51 operating conditions needed for these markets to function successfully. By identifying the key
52 evidence gaps in the field of LEMs, we help researchers direct their efforts to provide the evidence
53 policy makers, regulators and companies will need to design and adopt these markets. The terms
54 P2P, CSC and TE are ill-defined. The results in this paper are broken down by each of the three
55 market types to reveal overlaps and differences between them. This systematic literature review
56 makes four important contributions:

- 57 (1) It examines the types of markets described as either P2P, CSC or TE in the academic literature.
58 This review analyses the similarities, differences and overlaps between these three types of
59 market.
- 60 (2) It develops six archetypal market designs based on the market types found in the literature,
61 which are presented alongside the main price formation mechanisms used.
- 62 (3) It presents detailed information about the value proposition, the size of participants, scale and
63 operating conditions of the markets, broken down by the market type.
- 64 (4) It details five significant evidence gaps found in the literature. These are the lack of: consider-
65 ation of physical constraints; a holistic approach to market design and operation; consideration
66 about how these market designs will scale; consideration of information security; and, consid-
67 eration of participant privacy.

68 The remainder of this paper is structured as follows. Section 2 presents the methodology
69 used for the systematic literature review, including the literature search, decision on paper inclu-
70 sion/exclusion, data extraction and analysis. Section 3 presents the results of the analysis and
71 a discussion of the results. Section 4 details the research gaps found during the review. Finally,
72 Section 5 provides concluding remarks. Appendix A contains additional supporting results data.
73 Appendix B contains the code book for the data extraction table used in this analysis.

74 2. Methodology

75 This literature review followed a systematic process for paper selection and data extraction. This
76 section details the process used to search for relevant literature, make decisions on which literature
77 to include in, or exclude from the review, and to extract and analyse data consistently from each
78 piece of literature.

79 2.1. Literature search

80 To identify a relevant set of literature we conducted a systematic search using the Scopus and
81 Web of Science databases. The search term was (“peer to peer” OR “peer-to-peer” OR P2P) OR
82 (“self consumption” OR “self-consumption” OR CSC) OR (transactive OR TE) AND electricity.
83 The paper title, abstract and keywords fields were searched in Scopus. The topic field was searched
84 in Web of Science, which includes title, abstract, author keywords, and keywords plus. The results

85 were filtered to only include peer-reviewed journal articles. Both databases were searched on 25
86 March 2020. Scopus returned 759 results and Web of Science returned 587 results. A total of 892
87 journal articles were returned by the search after the removal of 454 duplicate search results.

88 The choice of search term was based on the fact that P2P, CSC and TE are ill-defined terms.
89 By minimising the search terms to variations of P2P, CSC and TE, plus 'electricity', we aimed
90 to find the widest possible range of literature which the authors define as concerning one of these
91 markets. Search terms in Scopus and Web of Science must appear in the results for it to be included.
92 Therefore, adding additional terms would exclude results, rather than widen the search.

93 The only filter applied to the search results was to limit them to peer-reviewed journal articles.
94 No limits were placed on the year of publication, country of study or other factors.

95 2.2. Inclusion criteria

96 We first reviewed the title and abstract of each paper against the inclusion criteria listed below.
97 The title and abstract review was completed by one person. Papers were kept in the review at the
98 title and abstract review stage if the reviewer was in doubt. During the title and abstract review,
99 675 paper were removed, leaving 217 papers in the full text review.

100 **Inclusion criteria:**

- 102 • The paper is written in English.
- 103 • The paper concerns electricity markets.
- 104 • The author defines the subject of the paper as P2P, CSC or TE uses of electricity – there are
105 no universally agreed upon definitions for P2P, CSC or TE; therefore papers were included
106 based on whether the author defined their paper as concerning one of these topics.
- 107 • The paper analyses one or more entities which transact, or a market.
- 108 • The paper has been published in a peer-reviewed journal.

109 Following the title and abstract review, we reviewed the full text of the remaining papers. The
110 same inclusion criteria were used for the title and abstract review and the full text review. The full
111 text of each paper was reviewed by one person. Where that person had a doubt about one of the
112 criteria, a second reviewer checked it. There were 72 papers removed during the full text review,
113 leaving 145 papers for data extraction. During the data extraction process a further six papers were
114 removed, leaving a total of 139 papers in the review.

115 **Number of papers included in the review:**

- 117 • Total results: 892 (Scopus 759, Web of Science 587, duplicates 454)
- 118 • Remaining papers after title and abstract review: 217 (675 removed)
- 119 • Remaining papers after full text review: 145 (72 removed)
- 120 • Papers included in review: 139 (6 removed during data extraction)

121 2.3. Data extraction

122 Data was consistently extracted from each paper included in the review using a data extraction
123 table. The data extraction table was designed for this study, but is based on *The Business Ecosystem*
124 *Architecture Modelling* (TEAM) framework [23]. The TEAM framework is designed to analyse a
125 group of businesses that do not have a central coordinator controlling them, but rely on common ICT
126 infrastructure. The businesses in the ecosystem must cooperate on things such as communication
127 protocols, but compete with each other on price. This mixture of cooperation and competition is
128 described as a competition game.

129 This leaderless coopetition game is very analogous to LEMs. There is not necessarily a central
130 coordinator directing the market, each individual may act in the market as they see fit. However,
131 for the market to function, all individuals must agree on common means of communicating bids,
132 creating contracts and proving that the contracted energy has been supplied and demanded. The
133 market participants also compete with each other in the purchase and sale of energy or other market
134 commodities. The TEAM framework therefore provides a good basis for analysing P2P markets and
135 other LEMs.

136 The TEAM framework examines three broad aspects of a market: the needs of the customers
137 and participants of the market; the distribution of costs, risks and benefits within the market; and
138 the data sharing requirements within the market. The holistic analysis of the market provided
139 by the TEAM framework looks not just at the main businesses, but also at the rule makers and
140 complimenting businesses in the market. This makes it appropriate for examining energy markets
141 where regulators, wire operators and system operators must be considered alongside the energy
142 traders.

143 The TEAM framework was adapted by the authors of this study to make it more specific to the
144 P2P, CSC and TE markets this study is analysing. The amendments to the TEAM framework for
145 this study include:

- 146 • Additional data about whether the author defines the market in the paper as a P2P, CSC or
147 TE market, and how the author defines those terms.
- 148 • Additional data about modelling assumptions used in the paper, including whether there is
149 uncertainty about future events, and whether physical constraints are considered.
- 150 • Additional data about the market participants.
- 151 • Additional information about the market, such as the length of the settlement period and the
152 length of the model run.
- 153 • Additional information about the size of the market and the resources available to market
154 participants.
- 155 • Consolidation of information about cash flows and risks.
- 156 • Removal of information about ICT and technology requirements.

157 A complete list of the data extracted for each paper can be found in [Appendix B](#). Details about
158 how to access the completed data extraction table for this study can be found in [Section 6](#).

159 Data extraction was undertaken by one researcher per paper. The unit of analysis for data
160 extraction was a market, i.e. all data was extracted for each market presented in a paper.

161 Following data extraction, the data was checked for validity and completeness. Each data field
162 was checked by one reviewer to ensure data had been extracted consistently for each paper. In-
163 consistencies found during the review were addressed by the researcher who originally did the data
164 extraction for that paper.

165 **3. Results and Analysis**

166 The results of the literature review identify six archetypal P2P, CSC and TE market designs
167 ([Section 3.2](#)). These archetypal market designs are backed up by a more detailed analysis of specific
168 aspects of the markets, including the price formation mechanism ([Section 3.3](#)), the market value
169 proposition ([Section 3.4](#)), and the market participants and the resources available to them ([Sec-
170 tion 3.5](#)). This section begins with a summary of the types of papers discovered in the literature
171 search, and a discussion of the defining characteristics of P2P, CSC and TE markets ([Section 3.1](#)).

172 Of the 139 papers included in this analysis, 77 modelled a P2P market, 61 modelled a TE market,
173 but only 6 modelled a CSC market. The very small sample size of CSC markets in the results limits
174 the extent to which conclusions about CSC markets can be drawn. Results for CSC markets are still
175 presented, but caution is required when generalising these. Note that five papers present multiple

176 markets. Therefore, the number of markets modelled is more than the number of papers included
177 in the review.

178 Only two of the 139 papers in the review are case studies of pilot projects [24, 25]. Of the
179 remaining 137 papers, 135 were mathematical models of markets and 2 were surveys. Although
180 some of the mathematical models used real data, such as from loads, generations (e.g. [26–29]) or
181 grid models (see Section 3.6.3), the mathematical models tend to focus on particular aspects of a
182 market, rather than creating a model which could be directly implemented. This means that not all
183 papers present information on all market elements covered in this analysis. Therefore, some sections
184 of analysis do not include all 139 papers, where some of the papers did not include the information
185 for that particular analysis.

186 3.1. Defining characteristics of P2P, CSC and TE markets

187 The terms P2P, CSC and TE are ill-defined and are used to describe a diverse range of markets.
188 This section examines how the terms P2P, CSC and TE are used by categorising the markets in the
189 reviewed literature. This analysis only includes papers that provide a definition of P2P, CSC or TE,
190 or give a statement on the purpose of the market. Of the 139 papers in the review, 70 were included
191 in this analysis. Table 1 presents references for each characteristic of the respective market type.

192 Only papers in the review concerning P2P markets explicitly discuss the size of the market
193 participants. These range from small participants, e.g. residential energy consumers and pro-
194 sumers [25, 28, 30, 31], to larger ones such as buildings and microgrids [32, 33]. Market participant
195 size is discussed further in Section 3.6.2.

196 P2P markets tend to be more decentralised than CSC markets. In CSC markets, participants
197 are typically closely geographically located [34]. Participants in P2P markets can trade energy
198 with each other directly [6, 26, 32, 35–42], or through centralised third parties [26, 27, 43]. CSC
199 markets are generally operated in a more collaborative manner, for example using a non-profit
200 centralised manager [44]. None of the papers considering TE markets gives information on the
201 market governance.

202 P2P and CSC markets tend to operate at small scales, e.g. within distribution networks, whereas
203 TE markets operate at all scales. Whilst there are examples of small TE markets [45–48], there are
204 also examples of TE markets which trade over entire electricity networks [49–51]. P2P and CSC
205 markets often aim to incentivise the use of local generation [25, 26, 31, 34, 52–54] or other local
206 resources [26, 38, 55, 56, 56].

207 TE markets focus more on providing grid services than P2P and CSC markets. Papers presenting
208 TE markets frequently aim to create a secure and efficient energy supply [57, 58]. They do this by
209 focusing on the balance of energy supply and demand [45, 46, 49–51, 59–63], and the integration of
210 flexible loads or storage devices [58, 63–69].

211 TE markets more frequently consider technical complications and operating conditions [76, 79], or
212 reliability and demand constraints [47, 78]. They also provide demand-side response [47, 68, 69, 76].
213 There are some examples of P2P markets providing flexibility [24, 56, 75] and stability services to the
214 network [33, 80]. There are fewer examples of CSC markets providing grid services. One example
215 which was found involved a community manager coordinating prosumers to provide peak shaving
216 services by minimising the maximum imported energy [44].

217 Papers considering P2P and TE markets tend to put more emphasis on specifying the market
218 structure and design than papers focusing on CSC markets. The concept of P2P energy trading is
219 based on a competitive market structure [52] where users engage in bilateral negotiation [40, 42, 82–
220 84], making use of contracts for the settlements [31, 85]. In TE markets, engagement is generally
221 through bidding [45, 79], price negotiations [68, 94] or auction based market clearing mechanisms [46,
222 48, 94]. TE markets can be operated as an extension of [81, 86] or replacement to [65] wholesale
223 markets. TE markets can also operate as a sub-system of existing markets [67]. TE systems are set
224 up in a market-based environment [48, 59, 62, 64, 69, 78, 81] aligning participants' interests with
225 those of the wider energy system [50] by using economic incentives [48, 49, 57, 59, 63, 78, 81, 86].
226 The use of locational marginal pricing [61, 67, 87] and the response to price signals [46, 66, 87, 88]
227 can optimise load behaviour. More details on markets structure and price formation can be found
228 in Sections 3.2 and 3.3, respectively.

Table 1: Defining characteristics of P2P, CSC and TE markets

Category	Characteristics	P2P	TE	CSC
Participation	Small-scale participants	[25, 28, 30, 31]	-	-
	Participants from various scales	[32, 33]	-	-
	Participants located in one community	-	-	[34]
Governance	Energy trading without intermediary	[6, 26, 32, 35–42]	-	-
	Energy trading with intermediary	[26, 27, 43]	-	[44]
Locality & typology	Local energy generation	[25, 26, 31, 52–54]	[58, 63–67]	[34]
	Local energy consumption	[38, 55, 56]	-	[26]
	Close geographical proximity	[26, 55, 70–74]	[45–48]	-
	Virtual trading of energy and different layers of the grid	[40, 70]	-	-
	Operating across various grid layers	-	[49–51]	-
Market services	Demand-side response	[24, 56, 75]	[47, 68, 69, 76]	-
	Supply/demand balancing	-	[45, 46, 49–51, 59–63]	[44, 77]
	Response to grid constraints	-	[47, 76, 78, 79]	-
	Grid stability and system efficiency	[33, 80]	[57, 58]	-
Market design	Competitive market structure	[52]	[48, 59, 62, 64, 69, 78, 81]	-
	Bilateral market transactions	[40, 42, 82–84]	-	-
	Contracts	[31, 85]	-	-
	Price signals and economic incentives		[46, 48, 49, 57, 59, 63, 66, 78, 81, 86–88]	-
Market transactions	Maximise total welfare	[71, 89]	-	-
	Set own trading preferences	[85, 89, 90]	[50]	-
	Trading of surplus energy	[26, 74, 75, 80, 89, 91–93]	-	[26, 44]

229 While all three market types share characteristics, the analysis of the definitions shows that they
230 each have a particular focus. P2P markets incentivise individuals to participate in energy markets.
231 CSC markets create energy communities which act for the benefit of the group. TE markets optimise
232 resources, providing services to the electricity system.

233 3.2. Market design

234 Six archetypal market designs have been identified in the papers: futures market, real time
235 market, mixed decentralised/centralised market, mixed futures/real time market, multi-layer market,
236 and settlement after the fact. The market design is the manner in which the price formation
237 mechanisms are strung together to form a complete market (see Section 3.3 for more detail on
238 individual price formation mechanisms). Figure 1 shows flowcharts for each of the archetypal market
239 designs. In some cases, such as a futures market (Figure 1a), a single price formation mechanism is
240 used. Whereas in other market designs, such as a mixed decentralised/centralised market (Figure 1c),
241 several different price formation mechanisms are used in succession over different time periods. In
242 this section, each of the market designs found in the reviewed literature is described, along with an
243 analysis of how each is typically used. Figure 2 shows the number of papers that use each type of

244 market design and price formation mechanism. Table A.5 in Appendix A shows the price formation
245 mechanism and market design used in each paper. Of the 139 papers included in the review, 55
246 provided sufficient information to be included in the market design analysis.

247 *Futures market:* In a futures market, all trading happens before the settlement period. During
248 the settlement period, market participants attempt to stick as closely to their traded positions as
249 possible. Any energy imbalances resulting from a deviation from the traded position are dealt
250 with during settlement. Single auction, double auction and bilateral negotiation price formation
251 mechanisms are all found paired with futures markets. Futures markets are the most common
252 market design found in the reviewed literature. They are also the most similar to the way many
253 existing electricity markets work, e.g. in Great Britain [95]. Figure 1a shows an archetypal flowchart
254 for a futures market.

255 *Real time market:* In real time markets, there is no trading ahead of the settlement period.
256 All trading is done during the settlement period. This allows market participants to update their
257 position in the market throughout the settlement period based on their actual supply and demand
258 for energy. Therefore, all market participants should theoretically come out of the settlement period
259 with a balanced position. However, there are reasons why market participants may not have a
260 balanced position, for example, if total supply and demand in the market are not matched. Most
261 papers reviewed assume the markets are linked to larger traditional electricity systems which act
262 as an infinite bus and are able to absorb any excess supply and demand. Else the papers assume
263 there is sufficient flexible energy generation or load that price signals in the market are sufficient to
264 balance supply and demand for energy. This allows all market participants to balance their position
265 during every settlement period. Single auctions, double auctions and bilateral negotiations are all
266 found in real time markets in the reviewed literature. Figure 1b shows an archetypal flowchart for
267 a real time market.

268 *Mixed decentralised/centralised market:* In a mixed decentralised/centralised market, there is a
269 period of bilateral negotiation, where market participants attempt to clear the market as far as
270 possible without intervention from a market operator. The bilateral negotiation is followed by a
271 centralised auction run by a market operator to clear the remainder of the market. The centralised
272 auction may simply be within the P2P/CSC/TE market, or the market operator might trade with
273 a larger traditional market in order to further clarify the P2P/CSC/TE market. Both single and
274 double auctions are used for the centralised part of the market in the reviewed literature. Figure 1c
275 shows an archetypal flowchart for a mixed decentralised/centralised market.

276 *Mixed futures/real time market:* In a mixed futures/real time market, there is some trading ahead
277 of the settlement period based on predicted supply and demand for energy. There is then further
278 trading during the settlement period, at which time market participants can correct their position
279 in the market due to any forecasting errors. Mixed futures/real time markets are found with both
280 single and double auctions in the papers reviewed. Figure 1d shows an archetypal flowchart for a
281 mixed futures/real time market.

282 *Multi-layer market:* Multi-layer markets are settled at multiple levels. For example, there may
283 be multiple markets at the bottom level which are cleared internally. An aggregator within each of
284 these markets then participates in a higher level market to clear excess supply or demand in the
285 lower level markets. Multi-layer markets are found with both single and double auctions in the
286 papers reviewed. Figure 1e shows an archetypal flowchart for a multi-layer market.

287 *Settled after the fact:* In a small number of cases, there was no trading before the end of the
288 settlement period. In these markets, participants are paid or charged for energy they supplied or
289 demanded after the settlement period. These markets use a system-determined price formation
290 mechanism, energy is bought or sold at a fixed price. Market participants can purchase or sell as
291 much energy as they require at these fixed prices. Therefore, no trading to determine an equilibrium
292 price and volume is done ahead of the settlement period. Figure 1f shows an archetypal flowchart
293 for a market settled after the fact.

294 3.3. Price formation mechanism

295 Price formation is the mechanism by which market prices are discovered. Exchange takes place
296 within the context of a market institution, the rules that specify which messages (e.g. buyer bids,

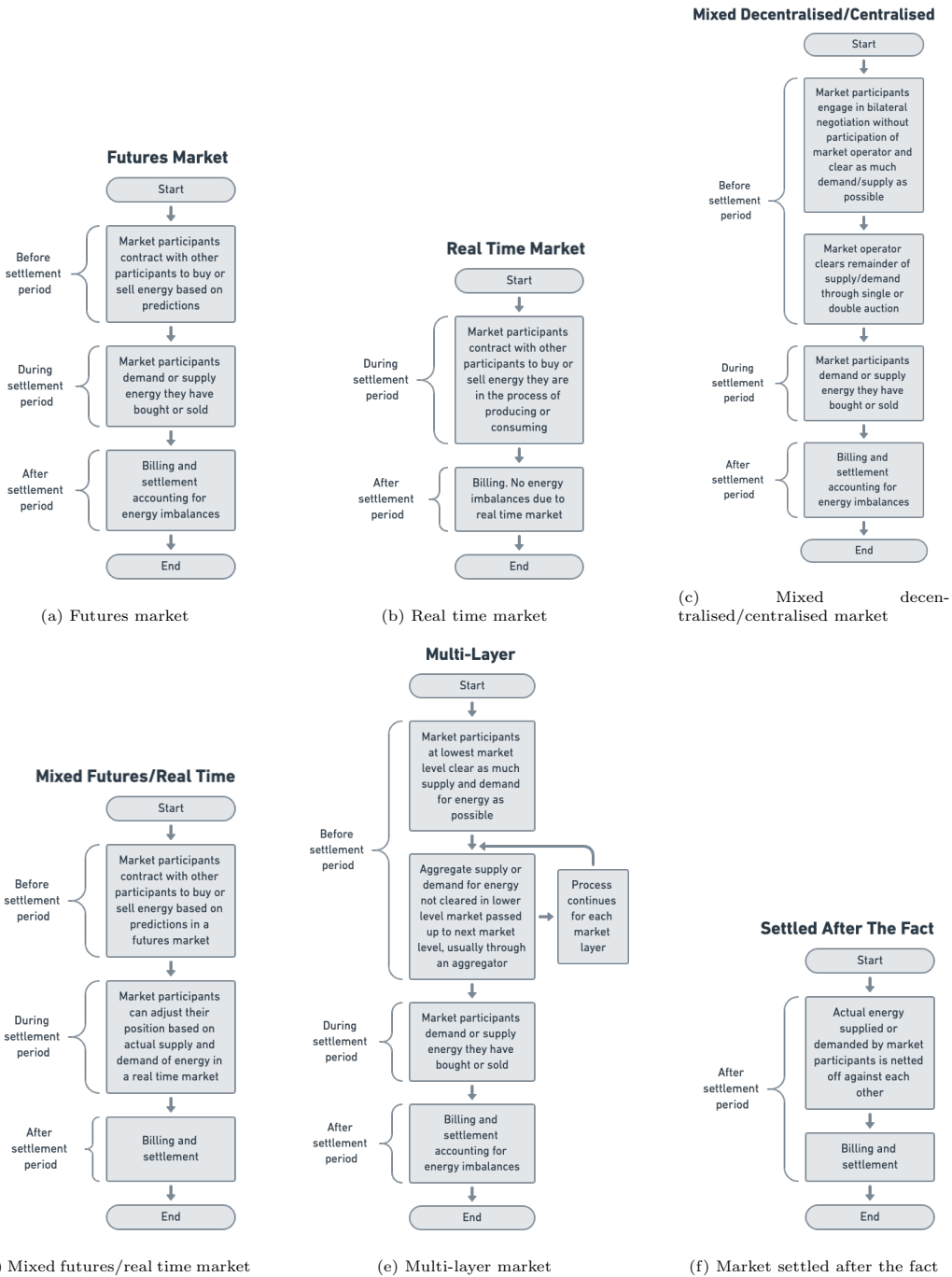


Figure 1: Market design flowcharts

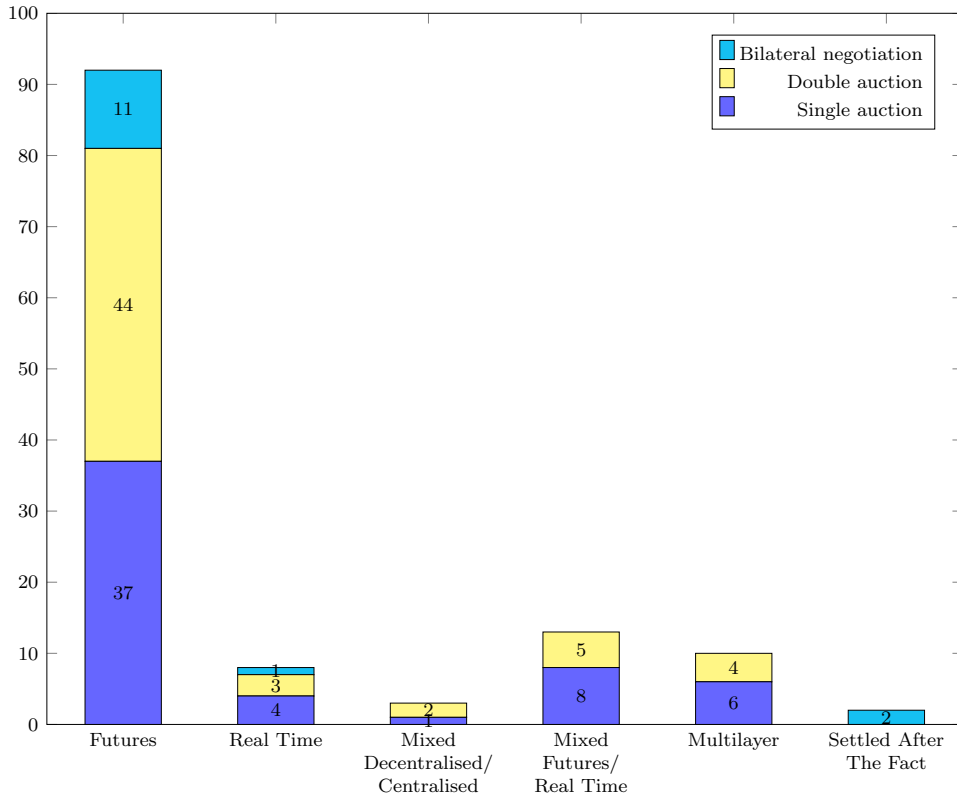


Figure 2: Number of markets using each market design and price formation mechanism

297 seller asks) are permitted, which agents are allowed to communicate messages, and how agents
 298 transact. Market institutions thus define price formation processes. Of the 139 papers included in the
 299 review, 53 provided sufficient information to be included in the price formation mechanism analysis.
 300 In the papers reviewed for this survey, five main categories of price formation mechanism were
 301 employed and tested: single auction, double auction, system-determined mechanisms, negotiation-
 302 based mechanisms, and equilibrium-based mechanisms.

303 *Single auction:* In a single auction, only agents on one side of the market communicate messages.
 304 This market institution is more common in settings where one side of the market is a single agent.
 305 In procurement auctions, for example, a single buyer solicits offers from suppliers.

306 The single auctions used in the reviewed papers (15% of markets reviewed) generally involve
 307 consumers submitting bids which are then cleared by a market operator. The market operator role
 308 can be performed by an aggregator, local energy operator and even distribution system operator
 309 (DSO), amongst others. Examples of single auctions include consumers in a community bidding to
 310 acquire units of excess renewable energy available at a given time (an ascending, one-side auction,
 311 with varying supply) [81], and demand response units bidding to offer flexibility or energy reduction
 312 services at a particular time (which is a reverse auction, up to the limit required by the system
 313 operator) [96]. Figure 3a shows a flowchart for a typical single auction price formation mechanism.

314 *Double auction:* The double auction is a common market institution in P2P, CSC and TE energy
 315 systems. Twenty-five percent of the 139 papers reviewed used some form of a double auction. It
 316 has been used and tested both theoretically and empirically since the original GridWise Olympic
 317 Peninsula TE project [97]. The double auction is the largest and probably the most well understood
 318 category of price formation mechanisms in the reviewed papers, being widely used in both wholesale
 319 energy markets and financial markets. While the double auction has many forms, its defining feature

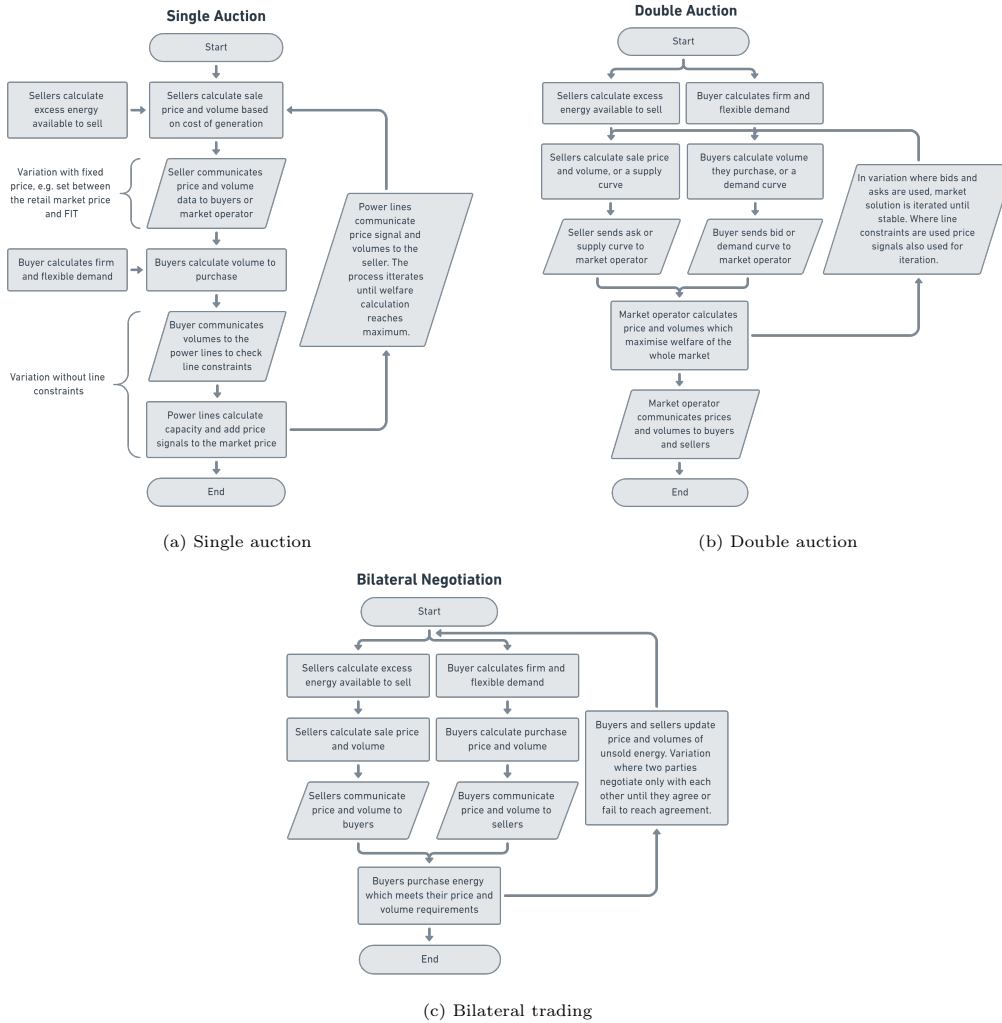


Figure 3: Price formation mechanism flowcharts

320 is the ability of both buyers and sellers to send messages. Buyer bids communicate willingness to pay
 321 that reflect underlying utility and preferences. Seller asks communicate willingness to accept that
 322 reflect underlying costs. When the double auction is repeated (as is usually the case in electricity
 323 market applications), it yields highly efficient outcomes through an information-rich environment
 324 that enables considerable learning among market agents [98]. The institutions used in the literature
 325 include several subcategories, with the two most common being a double clock auction and a con-
 326 tinuous double auction. A double clock auction is cleared at specific time points or regular intervals,
 327 usually in real time but also for day-ahead forward markets [88, 99]. In a continuous double auction,
 328 the market is cleared continuously, such as in stock markets that use order books to keep track of
 329 standing bids and offers [41, 100]. Figure 3b shows a flowchart for a typical double auction price
 330 formation mechanism.

331 *System-determined mechanisms:* Market institutions and price formation vary by industry and
 332 context. The requirement for real-time physical coordination and balance in electric systems has led
 333 to price formation in some projects that relies on system-determined mechanisms (23% of papers
 334 reviewed). This category encompasses all mechanisms that do not rely on market bids and offers,

335 and are instead set by a platform operator, based on a pre-agreed or pre-set mechanism or formula.
336 The “system operator” setting the prices is broadly defined and varies from paper to paper – it
337 could potentially be the community energy aggregator, local retailer, or DSO. Common types of
338 mechanisms mentioned include:

- 339 • Uniform or fixed prices, up to a limit or per unit.
- 340 • Pricing such as fixed feed-in tariffs on the generation side, or time-of-use prices on the demand
341 side.
- 342 • Mechanisms where the price set for local renewable energy is set at some fixed ratio (e.g.
343 mid-point or average between peak import and export prices).
- 344 • Mechanisms that use a function of demand or some other signal (e.g. quadratic on demand).
- 345 • Mechanism where the community aggregator uses an established technique from cooperative
346 game theory (e.g. Shapley value) to redistribute benefits in the local TE scheme participants.

347 *Negotiation-based mechanisms:* The auction institutions described above typically involve a cen-
348 tralised market platform in which buyers and sellers participate. A more decentralised approach
349 that resembles bilateral search uses negotiation-based mechanisms. Negotiation-based P2P trans-
350 actions are often automated with specialised, AI-enabled software, such as negotiating autonomous
351 agents. Unlike single and double auctions, which are a more structured method of price formation,
352 negotiation prices depend on the local one-to-one (or sometimes one-to-many) offers being made and
353 accepted. However, they have the potential to allow truly decentralised P2P energy transactions.
354 Eleven percent of the papers reviewed used a form of negotiation-based price formation. Figure 3c
355 shows a flow chart for a typical bilateral negotiation price formation mechanism.

356 *Equilibrium-based mechanisms:* Equilibrium-based mechanisms include those mechanisms where
357 price is formed based on bids/offers from the agents (usually prosumers, but could also be suppliers,
358 flexibility providers, etc.), but price is formed as a derived equilibrium of the interaction, using a
359 game-theoretic solution concept to construct the equilibrium. Several papers explore how an iterated
360 exchange of bids results in convergence to a price equilibrium. The game-theoretic equilibrium
361 concepts employed include Nash equilibrium (most frequent), but also Cournot, Stackelberg, or other
362 competitive market equilibrium. Eight percent of the papers reviewed used a form of equilibrium-
363 based price formation.

364 *Not specified or not explicitly mentioned:* A sizeable number of the reviewed papers (18%) do
365 not include a description of how the price is formed, mostly because price is not a key element of
366 the paper. Several papers are completely unrelated to prices (they are about forecasting, low-level
367 control etc.) Another insightful reason is that several P2P and TE exchange mechanisms (especially
368 in the context of local communities) are “relationship based”, not price based. For example, in some
369 local community energy projects, exchanging excess energy is done on a reciprocal basis, not on
370 price, or the excess is redistributed by a local aggregator or operator based on some fairness criteria,
371 not monetary payment.

372 3.4. Market value proposition

373 The value proposition of the market is the benefit which the market brings to its participants
374 through the trading of a commodity. In this section, we analyse the commodities traded in the
375 markets, and the value brought by these trades to the participants. The benefits of the market are
376 described as the needs of the market participants in the following sections.

377 3.4.1. Market commodity

378 Of the 139 papers included in the review, 130 provided information on the commodity traded in
379 the market. Electrical energy was traded in all the markets reviewed which provided that information
380 (130 of 130 papers). In most cases, electrical energy was sold by generators to consumers (102 of
381 130 papers). In other cases, the market paid for flexibility, either alongside a market for the sale
382 of energy (11 of 130 papers) [56, 62, 63, 90, 101–107], or in a flexibility only market (10 of 130

papers) [47, 49, 69, 76, 77, 79, 108–111]. Finally, some markets traded ancillary services such as reactive power, either alongside energy (five of 130 papers) [50, 51, 112–114], or as a standalone ancillary services market (two of 130 papers) [61, 115].

Although electrical energy was always traded in the markets reviewed, it was sometimes combined with other forms of energy. Combined heat and power markets are found in five of 130 papers [91, 116–119]. One presented a combined power and gas market [120], and one paper presented a combined power, heat and gas market [121]. It should be noted that the search term used in this study contained ‘electricity’, so pure heat or gas markets are excluded.

Almost all P2P markets only trade electrical energy. This could be due to the fact that P2P markets typically focus on providing services to prosumers, who demand or supply electrical energy. The majority of TE markets trade flexibility alongside electrical energy. This could be due to the fact that TE markets provide services to the electricity system, which needs flexibility to keep supply and demand for energy in balance. Three of the five CSC markets only traded electrical energy, while two also traded flexibility.

3.4.2. Benefits to market participants

Of the 139 papers reviewed, 128 provided information on the benefits of participating in the market. These benefits are primarily financial, e.g. profits from the sale of energy [40, 74, 120, 122, 123] or minimising the price paid for energy [84, 86, 93, 124]. Many markets also had secondary objectives, e.g. ensuring power line thermal limits are not exceeded [39, 41, 43, 62, 84, 104, 115, 125, 126]. Figure 4 breaks down the primary and secondary market benefits by number of papers. Table A.6 in Appendix A provides references for the primary and secondary benefits (needs) of the market participants, broken down by commodity (see Section 3.4.1 for more details on market commodities). Figure 4 and Table A.6 differentiate between the following terms closely-related to financial benefits: total welfare (also known as economic surplus), profit, cost and electricity cost. We use the term total welfare if a market provides the end users, e.g. prosumers, with higher profits or lower costs, depending on their role in the market (seller or buyer). If a market only provides one financial benefit to the market participants then we use the specific term instead of total welfare. We use the term electricity cost if the market aims to reduce the electricity cost, which is beneficial to all grid users, not only the market participants.

Energy buyers and sellers both benefit in P2P, CSC and TE markets. Buyers benefit by purchasing energy at below the retail market rate. Sellers benefit by selling energy at above the feed-in tariff rate, if one exists, or by selling energy at all if not [28, 59]. The distribution of the benefits between the buyer and seller depends on the market price (see Section 3.3 for more detail on market prices). Many papers do not explicitly compare the P2P/CSC/TE market price to retail market and feed-in tariff prices. Therefore, it is often not possible to quantify the benefit of the P2P/CSC/TE market over the traditional market.

For some sellers in P2P, CSC and TE markets, there may be no other means of selling their excess energy. P2P, CSC and TE markets are also less rigid than traditional markets about the types of generation which are permissible. Feed-in tariff schemes have limitations on the type and

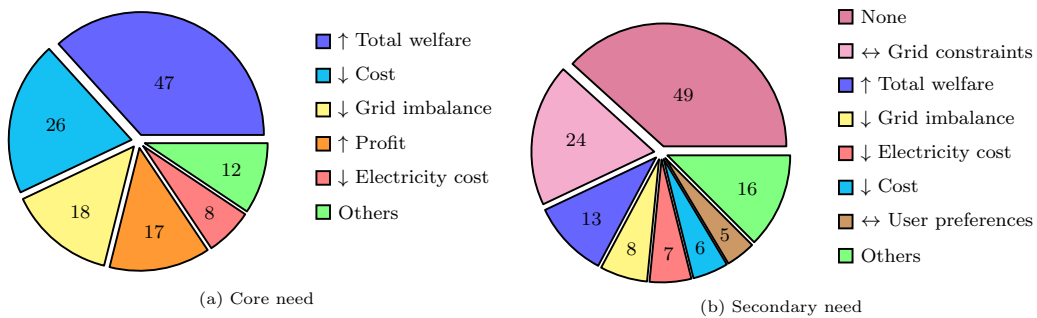


Figure 4: Needs of market participants (↑ Increase; ↓ Reduce; ↔ Respect)

422 size of generation which is allowed [127]. Typically, storage is not compensated under feed-in tariff
423 schemes.

424 Although many papers state that the P2P/CSC/TE market price is lower than the retail market
425 price, they neglect non-energy costs which are included in the retail market price [26, 35, 52, 128].
426 These include balancing costs ² and network costs ³. It is likely that P2P, CSC and TE markets will
427 be subject to some level of balancing and network costs [129, 130]. However, they may be lower than
428 in traditional markets. For example, CSC markets aim to use electricity locally. Therefore, they
429 may not be subject to the same level of network costs and geographic balancing costs. However,
430 these costs are still likely to reduce the value of these markets for their participants when compared
431 to the models presented in the current literature.

432 Some markets also provided a service to the grid, such as energy balancing ⁴. These services are
433 normally compensated through time-of-use pricing. For example, a flexible load can be compensated
434 for shifting in time by the fact that they buy energy at a lower price. Or, a storage device can be
435 compensated by purchasing energy at a low price and selling it at a high price (arbitrage). These
436 devices are providing a service beyond simply selling energy. They are making adjustments to the
437 supply and demand for energy at short notice.

438 Unlike in P2P, CSC and TE markets, traditional energy systems procure these balancing services
439 in a separate market to energy. In liberalised electricity markets, balancing services are often pro-
440 cured by a different entity to energy (system operator and energy supplier respectively). Balancing
441 services are normally valued more highly than energy in traditional markets to reflect the fact that
442 the changes to supply and demand are being made at short notice (typically less than an hour). It
443 is therefore possible that by only paying balancing services at arbitrage rates in P2P/CSC/TE mar-
444 kets, they are being under-compensated when compared to their value added to the system. Their
445 compensation will be lower than the market price for energy in P2P/CSC/TE markets, compared
446 to above the market price for energy in traditional markets.

447 In traditional electricity markets, there are normally minimum bid sizes for balancing markets.
448 The types of resources which can participate in balancing in P2P/CSC/TE markets are often too
449 small to provide those services in traditional markets. The fact they can be compensated for bal-
450 ancing services at all in P2P, CSC and TE markets is additional value to those participants.

451 One reason these flexible resources are not fully compensated for their true service is that most
452 P2P, CSC and TE markets in the papers reviewed are not subject to imbalance charges. Either the
453 papers assume that market participants can perfectly predict their supply and demand for energy
454 and always balance their position in the futures market, or the papers do not consider cash out
455 at all. If the papers considered imbalance charges, flexible resources may be valued more highly
456 because their price would be compared to the cash out price, rather than the energy price.

457 The majority of the articles reviewed either only provide information about the benefits of partic-
458 ipating in P2P, CSC or TE markets, or provide limited information about the costs of participating.
459 In addition, a predominant assumption in the papers reviewed is that the market participants al-
460 ready possess the necessary assets (e.g. storage, PV, etc.) to generate and trade electricity. The
461 value proposition of these markets then takes as a benchmark the benefits one can obtain from
462 using these assets in the traditional market and derives the benefits obtained by participating in the
463 P2P/CSC/TE market.

464 What then becomes even more interesting is to find out the value proposition vis-à-vis cost
465 involved in participating in P2P/CSC/TE electricity markets considering the capital investments in
466 assets. Although important, this analysis is out of the scope of this paper as the TEAM framework
467 does not facilitate the collection of sufficient data to perform this analysis.

²Balancing costs are charged to electricity market participants by the system operator. They are used to recover the costs of the system operator and are charged in proportion to market participants' energy imbalances.

³Network costs are charged to market participants by the distribution and transmission network operator to cover the capital and operating costs of the electricity network.

⁴Energy balancing involves shifting supply or demand for energy between settlement periods to keep the overall supply and demand for energy in balance.

468 *3.5. Market participants*

469 In the following section, we take a detailed look at the participants involved in the markets. We
 470 look at the types of participants, taking a frequentist approach, and analyse the assets participants
 471 contribute to the market.

472 *3.5.1. Types of market participants*

473 Market designs and operating conditions can be distinguished based on the participants involved
 474 in the market. We differentiate between seven different types of market participants: pure generators,
 475 pure consumers, prosumers, aggregators, retailers, central market operators and grid operators.
 476 Figure 5 shows the types of market participants, split by type of market. Some papers are represented
 477 multiple times if more than one market was discussed. Of the 139 papers included in this review,
 478 136 papers contained the correct information to be included in this analysis. Detailed references for
 479 the types of market participants considered by each paper can be found in Table A.7 in Appendix
 480 A. A description of each participant can be found in the code book in Appendix B.

481 Around 94% of P2P markets have prosumers, followed by 55% which have pure consumers, 46%
 482 have central market operators and 29% have grid operators. Other market participants represented
 483 in P2P markets include aggregators and retailers, with pure generators being the least frequently
 484 represented. This distribution of participants highlights the focus of P2P markets on individual
 485 energy end-users and the goal to offer them a platform to trade energy. However, the inclusion of
 486 other participants such as retailers, grid operators and aggregators shows the diversity P2P markets
 487 and the different ways they integrate into existing energy markets.

488 In TE markets, grid operators and prosumers play the most significant role. Both are represented
 489 in 64% of papers. They are closely followed by pure consumers, in 62% of markets. Fifty-five percent
 490 of papers include a central market operator. Around half of all papers include pure generators and
 491 aggregators. Retailers were the least frequent market participant, appearing in 23% of markets. TE
 492 markets have a more even distribution of market participant types than P2P markets. This supports
 493 the defining characteristic of TE markets (Section 3.1) that they can operate at various levels of the
 494 grid with a diverse range of participants.

495 Over 83% of CSC markets are centred around energy prosumers. A central market operator
 496 existed in 67% of cases. Half of the papers considered pure consumers. Retailers, pure generators
 497 and grid operators were the least prominent market players in CSC markets. None included an
 498 aggregator. This highlights the centralised nature of CSC markets. It should be stressed that only
 499 a small sample size of CSC markets have been analysed.

500 The dominant participants in all three types of market are prosumers, pure consumers and market
 501 operators. TE markets put a stronger focus on grid operators, pure generators and aggregators than
 502 P2P markets. This supports the findings in Section 3.1 that TE markets are more focused on
 503 providing grid services than incentivising individuals to trade amongst each other. Furthermore, TE
 504 is a concept that focuses on supporting the electricity grid, explaining a more equal distribution of
 505 different market participants. This is supported by the characteristics identified in Section 3.1 where

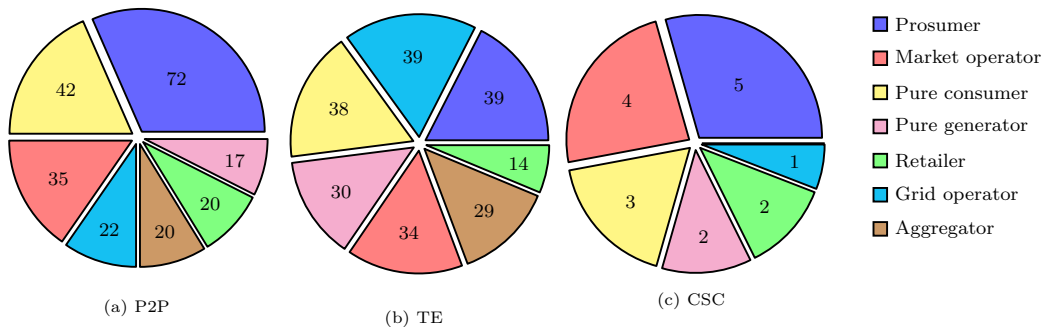


Figure 5: Types of market participants

506 locality plays a rather small role in TE markets compared to P2P markets. An important observation
 507 to make is that the diversity of participants in a market is important for pooling resources to create
 508 diversity of load and generation profiles. However, that diversity might also increase complexity
 509 when operating the market, as a wider range of market behaviours have to be taken into account.

510 3.5.2. Assets of market participants

511 Assets participating in the market were classified as either controllable or non-controllable. Con-
 512 trollable assets are energy generators or loads that can be dispatched on demand. Controllable loads
 513 can either be shifted, curtailed or completely disconnected depending on their specific properties.
 514 These assets can provide power balance or voltage control services. Energy storage systems are
 515 considered to be controllable assets. They can either generate or absorb power from the electricity
 516 grid. Non-controllable assets are generation units that cannot be dispatched or are intermittent in
 517 nature, and loads that are not shiftable or shapeable. Of the 139 papers included in the review, 123
 518 contained the correct data to be included in the analysis of market participants' assets.

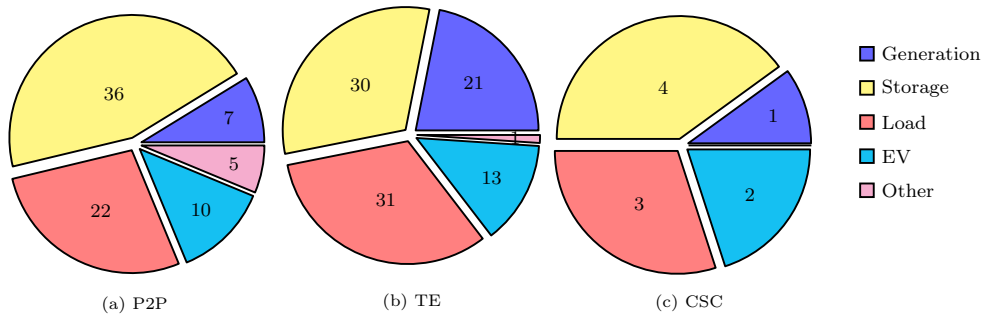


Figure 6: Types of controllable market assets

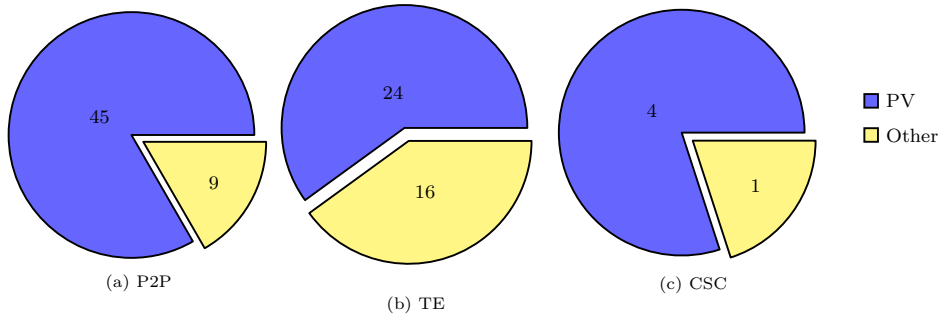


Figure 7: Types of non-controllable market assets

519 Assets participating in markets directly and indirectly (e.g. through a home energy manager)
 520 were considered in this analysis. Figure 6 shows the frequency of controllable asset types, split by
 521 market type. Nearly 80% of all markets include controllable assets. Storage devices and dispatchable
 522 loads played a major role in all types of market. In most markets, small scale residential energy
 523 storage systems were used, with a few exceptions. For example, in the cases where community or
 524 utility size storage systems [53, 128] or thermal storage units [67, 117, 118] were considered.

525 All three market types integrated controllable load in their designs. In P2P and CSC markets,
 526 controllable loads were usually shiftable appliances [33, 101, 102, 124, 131], air conditioners [90, 111,
 527 124] or heat pumps [33]. In TE markets, shiftable appliances were also a key source of flexibility [59,
 528 68, 103, 109, 119]. Heat pumps were frequently used as the main source of load control [49, 59, 68,
 529 88, 99, 116, 117]. TE markets put a stronger focus on dispatchable generation, including combined

530 heat and power [67, 116–118] or traditional fuel-based generators [49, 57, 119]. In a few cases, P2P
 531 markets made use of diesel generators [42, 132, 133]. All three models considered electric vehicles
 532 (EV) in their markets, although not as frequently as other controllable assets. An overview of the
 533 references that used controllable assets can be found in Table A.8 in Appendix A.

534 There is a clear difference between the non-controllable assets found in P2P and CSC markets
 535 when compared to TE markets. Figure 7 shows the types of non-controllable generation units found
 536 in the literature, grouped as either PV generators or other distributed generators. P2P markets
 537 mainly include PV generators. When size is explicitly mentioned, most markets refer to small-scale
 538 rooftop PV systems. In a few cases, multiple generation units have been considered, mostly PV
 539 paired with wind generation [56, 114, 121, 134]. By contrast, TE markets more frequently include
 540 other types of distributed generation. In these cases, wind energy is dominant [61, 105, 113, 114,
 541 120]. In CSC markets, most non-controllable generation units were PV installations, with one
 542 exception [77].

543 3.6. Market scale

544 The scale of a market is key to understanding its operating conditions. This section first looks
 545 at the size of the markets in terms of the number of nodes or participants involved. Secondly, it
 546 investigates the scale of the participants in each market.

547 3.6.1. Participation in markets

548 This section focuses on analysing the size and scale of the markets in terms of the number of
 549 participants involved. Where multiple markets have been tested, the one with the highest number
 550 of participants was included in this analysis. An overview of the number of papers and size of
 551 the markets is given in Figure 8. Instead of specifying the number and type of participants, some
 552 papers referred to nodes which is usually the number of agents or buses a market is optimised for,
 553 e.g. [81, 113, 134]. Where the number of participants was not given, the number of nodes was used in
 554 the analysis instead. Of the 139 papers in this review, 117 provided information about the number
 555 of market participants and are included in this analysis.

556 Most papers present small energy markets with 1-10 participants, followed by markets with 11-50
 557 participants. These two group sizes make up more than half of all papers. Sixteen papers present
 558 markets with 51-100 participants, 13 papers involve 101-500 participants, 5 papers involve 501-1000
 559 participants and 6 papers look at more than 1000 participants. A detailed overview of the number
 560 of participants considered in each paper can be found in Table 2.

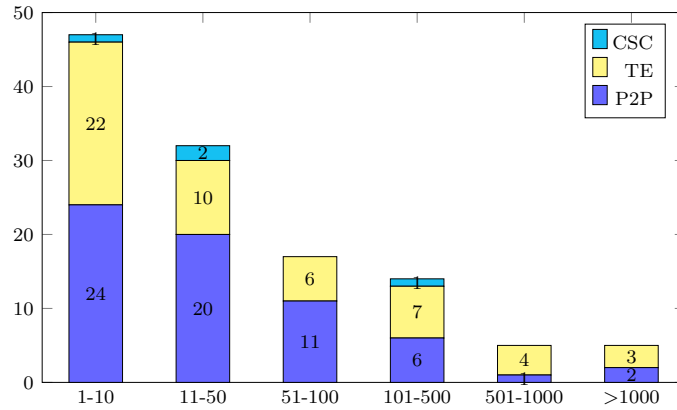


Figure 8: Number of nodes/participants in the market

561 Most authors built their markets using small participation numbers to demonstrate the function-
 562 ality of their market mechanisms. While this can help to evaluate the performance of a market, it
 563 only provides limited insights into the real-life applicability and scalability of such markets. Markets

564 with larger numbers of participants usually focus on scheduling of devices, such as EVs or ther-
 565 mostatically controlled loads [60, 79, 109, 123], rather than individual households optimising load
 566 profiles.

567 For all papers with more than 500 participants, the test duration varied between a few hours
 568 and a maximum of one day, with one exception where the test duration was two months [135].
 569 Although the models look at larger scale adoption, they are not tested for resiliency or diversity
 570 of load. However, where fewer participants have been included in the market, longer simulation
 571 durations have been tested [35, 81, 136]. More research is required into markets operating at larger
 572 scales, with a couple of hundred participants or more.

Table 2: Number of market participants

Participation	P2P	TE	CSC
1-10 participants	[6, 25–29, 36, 39, 52, 53, 56, 71, 72, 74, 91, 124, 128, 134, 137–142]	[45, 46, 57, 64, 65, 68, 81, 86, 99, 103, 104, 108, 112, 115, 117, 143–149]	[26]
11-50 participants	[24, 30–32, 35, 37, 40–42, 54, 84, 102, 106, 122, 131, 132, 150–153]	[37, 51, 59, 87, 94, 107, 110, 118, 120, 125]	[44, 102]
51-100 participants	[55, 73, 85, 93, 96, 100, 121, 126, 154–156]	[47, 62, 113, 136, 157, 158]	-
101-500 participants	[70, 75, 90, 92, 101, 159]	[63, 76, 105, 119, 160–162]	[34]
501-1000 participants	[111]	[50, 60, 69, 88]	-
>1000 participants	[123, 135]	[78, 79, 109]	-

573 3.6.2. Size of market participants

574 A second important characteristic is the scale of participants in the market. The scale here
 575 refers to the size of the market participants. We divide participants into small-scale, building-
 576 scale, microgrid/community-scale or grid-scale. In cases where multiple scales of participants were
 577 present, the scale was selected according to the key targeted group of the market. Small-scale market
 578 participants are predominantly residential/individual energy users. In markets with building-scale
 579 participants, multiple buildings trade with each other. They can be either larger residential or com-
 580 mercial/industrial buildings. Community or microgrid-scale markets do not focus on the individual
 581 energy users in the market, but rather operate as a community. Grid-scale market participants are
 582 directly linked and provide benefits to the distribution or transmission network. Identifying the scale
 583 of market participants helps us to understand the main trading purpose of a market, by means of
 584 who the market was designed for, and its ability to scale in the future. Of the 139 papers included in
 585 the review, 131 provided information on the size of the market participants and have been included
 586 in this analysis. An overview of the scale of market participants can be seen in Figure 9. Table 3
 587 provides the associated references.

588 Most papers focus on developing markets for small-scale participants. In the case of P2P markets,

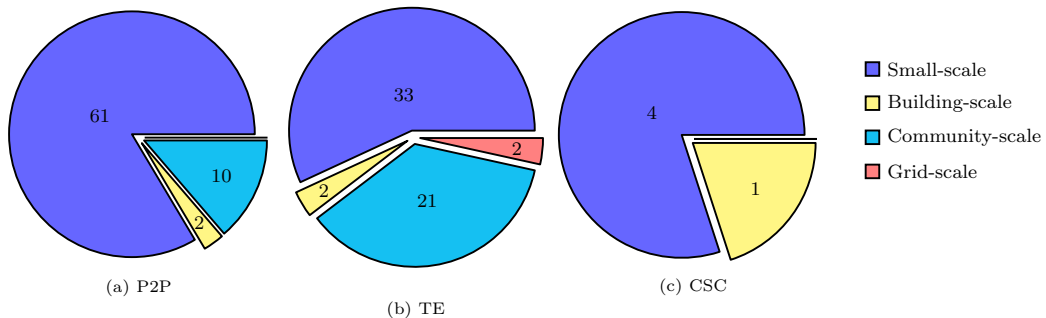


Figure 9: Scale of market participants

Table 3: Scale of market participants

Participant-scale	P2P	TE	CSC
Small-scale	[6, 24, 25, 27–31, 33, 35–37, 43, 52, 54–56, 70–75, 80, 82–85, 91–93, 96, 100–102, 106, 111, 121–123, 126, 131, 133–135, 137, 139–142, 150–156, 159, 163, 164]	[37, 45, 47, 48, 50, 51, 59, 60, 62, 63, 68, 69, 78, 79, 81, 87, 88, 99, 103, 108, 109, 113, 115, 116, 119, 120, 128, 136, 147, 148, 157, 158, 161]	[34, 44, 77, 102]
Building-scale	[26, 124]	[67, 149]	[26]
Microgrid/ Community-scale	[32, 39–42, 53, 90, 114, 132, 138]	[46, 49, 57, 64, 76, 86, 94, 105, 107, 110, 112, 114, 117, 118, 125, 143–146, 160, 162]	-
Grid-scale	-	[61, 104]	-

589 nearly all papers focus on small-scale residential energy users, or in some cases EVs [54, 73, 151]. A
590 few papers have considered trading at community-scale. These markets usually include transactions
591 between microgrids [32, 39, 132], within virtual power plants [40] or with industrial energy users [42,
592 90, 138]. Examples of building-scale trading includes trading between campus-buildings [26] or build-
593 ings in clusters [124]. Similarly, papers proposing CSC markets mainly consider small-scale energy
594 users in their analysis [34, 44, 77, 102]. The scale of users in TE markets is more diverse, although the
595 key target group are still small-scale users. Building-scale TE models consider commercial buildings,
596 such as schools and offices, or manufacturing plants [67, 149]. Most microgrid/community-scale pa-
597 pers with TE markets focus on trading between microgrids [57, 118, 143, 145]. However, two papers
598 focus on trading between aggregators [76, 125], and one conducts trading through a virtual power
599 plant [110]. The grid-scale markets operate at higher grid levels and are targeted specifically at the
600 transmission or distribution grids [61, 145]. Although small-scale participants are dominant in TE
601 markets, those papers included proportionally more grid-scale markets than papers examining P2P
602 or CSC markets. This shows that TE markets operate across various scales, from small scale to grid
603 scale applications.

604 An analysis comparing the number of market participants and the market scale to the price
605 formation mechanism and market design was conducted to examine the relationship between mar-
606 ket size and complexity. No correlation was found between the market design or price formation
607 mechanism and the market scale or number of participants. Only a small number of papers model
608 markets with a large number of participants (five models contained more than 1000 participants),
609 and most papers modelled small scale markets. Therefore, it is possible that the reviewed literature
610 would not identify issues relating to scaling complexity of the market designs and price formation
611 mechanisms. Section 4.3 provides further discussion of the scalability research gap.

612 3.6.3. Types of grid model

613 Due to the link between LEMs and low/medium voltage networks, many papers have been
614 devoted to analysing grid integration constraints. Forty-eight of the 139 papers reviewed used a grid
615 model to test the effect of their market on the power network. Along with voltage range operation
616 limits [126], other constraints have been highlighted, including but not limited to, phase imbalance,
617 power peaks, upstream generation, transmission capacity, and line congestion [33, 52, 56, 128, 134].
618 It is worth noting that besides grid constraint, power losses have an essential impact on the physical
619 implementation of the commercial transaction too [84, 140]. A detailed analysis of the technical
620 aspect of power losses and network constraints integration to the transaction design has been assessed
621 by Dudjak et al. [165].

622 Different grid models have been used in the models presented, including IEEE and CIGRE test
623 feeders, simulation case test feeders, and in some cases, real test feeders. Table A.9 in Appendix
624 A provides references for each paper that considers grid models, including the grid model used and
625 the type of analysis performed. The relatively small number of papers using each grid model and
626 performing each type of analysis limits the bench-marking which can be done between the different
627 analyses.

628 *3.7. Market operation*

629 In the following section, we discuss the type of data shared between participants and the user
630 preferences considered (Section 3.7.1). We then provide insights into the settlement period and gate
631 closure times used in the markets (Section 3.7.2).

632 *3.7.1. Data sharing and user preferences*

633 In order to persuade end-users to actively engage and participate in LEMs, markets should treat
634 participants fairly and provide them with means of informed decision-making. Therefore, one crucial
635 aspect of the markets is the data/information shared amongst participants. Of the 139 papers in
636 the review, 113 provided information about data sharing and user preferences.

637 In cases when the trade is between one or two large buyers (e.g. grid operators [87] or ag-
638 gregators [76]) and many smaller sellers (e.g. prosumers or consumers), the buyers usually share
639 information about the volume of the commodity they wish to purchase and potentially price infor-
640 mation. Based on this information, the sellers can then form their bids and participate in the market.
641 The sellers' bids usually contain at least information about the volume of commodity available for
642 the announced price [60, 69], the price for which the requested commodity can be provided [64] or
643 both [50, 51, 88, 110, 112]. This is the usual data flow in TE markets, where aggregators sit between
644 prosumers and the central market operator, whose role in many cases is played by the grid operators
645 themselves [76, 87]. Table 4 provides a summary of the types of information shared in different
646 markets.

647 In all market types, electricity price and volume information for a specific trading period are
648 the main types of data shared by prosumers, either with the other prosumers if the market is
649 fully decentralised [52, 72, 94, 99, 132, 141, 161], or with a central market operator that clears
650 the market [6, 32, 51, 66, 80, 88, 112, 155, 157]. Therefore, the vast majority of markets use only
651 these two data items to determine the market output. Supply and demand curves are the main
652 data items shared by participants in markets where the bidding takes place for several trading
653 periods [36, 37, 62, 68, 106, 149], for example in day-ahead markets. In a few markets, prosumers
654 only share electricity price [33, 64, 67, 133] or volume [24, 28, 60, 85, 121, 139]. This is due to the
655 fact that the markets have buyers (e.g. grid operator in TE models or prosumers in P2P models)
656 who announce only price or volume information. Hence the prosumers who sell only need to submit
657 volume or price information. These types of markets offer limited flexibility as prosumers can only
658 express their trading preferences via one parameter – price or volume.

659 *3.7.2. Settlement period & gate closure*

660 The settlement period of an electricity market is the period of time over which a market partici-
661 pant must balance their supply and demand of energy. Gate closure is the length of time before the
662 settlement period when the wholesale market closes. Of the 139 papers in the review, 110 provided
663 information about the settlement period and gate closure in the market. Together, the settlement
664 period and gate closure length determine how far in advance a market participant must predict their
665 supply and demand for energy, and over what period they must make that prediction. In traditional
666 electricity markets, settlement periods are typically around 30 minutes [95], but can be as short as
667 5 minutes [166]. Gate closure is around one hour prior to the start of the settlement period [95].

668 The papers included in the review had settlement periods ranging from 15 seconds to 1 day. Gate
669 closure ranged from zero, i.e. a real time market, to one day. For very short settlement periods, there
670 is a strong correlation between the settlement period length and gate closure. Only one paper [27]
671 had a settlement period of less than one minute (15 seconds) and that was also the only paper to
672 model a gate closure of less than one minute (20 seconds).

673 As the settlement period increases, there is less correlation between settlement period and gate
674 closure. The two papers which model three minute settlement periods both use one hour gate
675 closures [147, 155]. The gate closure of papers modelling a five minute settlement period ranges
676 from five minutes [65, 154] to one day, e.g. [77, 106, 109, 124, 138]. As the settlement period grows
677 longer, there is less use of short gate closures. At a settlement period of 15 minutes, the smallest gate
678 closure is 15 minutes [75, 141], and they go up to one day [59, 100, 123, 153]. This trend continues

Table 4: Data shared in markets

Data type	Recipient	Market type & references			
		P2P	TE	CSC	Combined
Price	Prosumer	[133]	[67]	-	-
	Central market operator	[33]	[64]	-	-
Volume	Prosumer	[28, 43, 70, 85, 93, 121, 139]	-	-	-
	Consumer	[24, 138, 163]	-	-	-
	Retailer	-	[60, 69]	-	-
Price & volume	Prosumer	[25, 35, 39, 41, 42, 52, 72, 73, 75, 82, 91, 99, 100, 122, 132, 134, 135, 137, 141, 151, 159]	[47, 94, 117, 143, 144, 147, 161]	[77]	-
	Central market operator	[6, 29, 30, 32, 35, 71, 80, 84, 99, 101, 137, 150, 152, 155]	[46, 48, 50, 51, 61, 66, 78, 81, 88, 104, 110, 112, 113, 145, 157]	-	[102, 114, 119]
Demand & supply curve	Prosumer	[36, 54, 90, 154]	-	-	-
	Central market operator	[27, 31, 53, 55, 89, 92, 96, 101, 106, 123, 131, 142]	[45, 57, 59, 62, 63, 68, 76, 79, 86, 87, 103, 107, 108, 115, 116, 125, 148, 149, 158, 160]	[34]	[37]
Controllable loads	Prosumer	[124]	[162]	-	-
Flexibility available	Central market operator	[106, 123, 142]	[62, 87, 108]	-	-
Battery SoC	Central market operator	[53, 92, 142]	-	-	-
Distribution line distance	Central market operator	[31]	[112]	-	-
Discomfort level	Central market operator	-	[59]	-	-
Eagerness factor	Central market operator	[35, 96]	-	-	-
Willingness to pay/accept	Prosumer	[40]	-	-	-

679 with 30 minutes [74] and one hour [42, 144] settlement periods, where the shortest gate closure is
680 the same as the length of the settlement period, and the longest is one day [92, 106, 134, 143].

681 4. Research gaps and future research directions

682 The results in the previous sections have highlighted the key differences and similarities of P2P,
683 CSC and TE markets and also LEMs as a whole, showing how the concepts are currently addressed
684 and described in the literature. The analysis has also shown that there are substantial gaps in the
685 current academic literature that need to be addressed for P2P, CSC and TE markets to operate at
686 scale. This section highlights five key research gaps that require further analysis.

687 4.1. Consideration of physical constraints

688 LEMs incentivise energy transactions between participants connected to the medium/low voltage
689 distribution networks. This creates bidirectional power flows in systems designed for unidirectional
690 power flows. It is therefore important to consider physical grid constraints when clearing LEMs. Only
691 about one-fifth of the analysed markets incorporate a comprehensive market mechanism that takes
692 into account physical grid constraints [45, 109, 113, 125] (see Table A.6). The rest of the analysed
693 markets either focus on the virtual market layer where transactions among market participants are
694 agreed, or only examine a single type of grid constraint such as congestion [79]. Further research is
695 needed to design market mechanisms that can incorporate the full range of grid constraints. This
696 could be achieved by grid operators feeding the market with various parameters which would indicate
697 the grid status. The market would have to have mechanisms in place to translate these parameters
698 to concrete desired actions with regards to the physical grid (e.g. reduce/increase supply at a specific
699 grid access point). Once this is in place, the market clearance phase could take this into account when
700 matching market participants. Transactions that would further violate the grid constraints could be
701 vetoed while the ones that would have a positive effect on the grid could be prioritised. Bundling
702 the grid constraints with pricing mechanisms and user preferences would potentially result in more
703 complete markets that take into account the physical infrastructure as well as user preferences.

704 In addition, a key aspect of successfully managing the physical constraints of the grid infrastruc-
705 ture is a close integration of LEMs with the current power system, as well as their integration and
706 coordination with the traditional energy markets such as wholesale, retail and balancing markets.
707 Some work has already been done in this direction (see for example [15, 167, 168]). Furthermore,
708 apart from their integration, quantifying the effect of these local energy markets on the traditional
709 markets is something that needs in-depth investigation.

710 4.2. Lack of holistic approach to market operation

711 Although there is a rich literature on different P2P, CSC and TE markets, existing solutions
712 focus mainly on the market clearance phase, including bid/offer submission, market price deter-
713 mination and market participant matching/transaction selection. Other crucial phases, such as
714 bid/offer creation incorporating user preferences, strategic bidding, billing/settlements and dispute
715 resolution [169], have been largely neglected.

716 The bid/offer creation phase should be able to capture (i) the diverse available resources of
717 the users, (ii) the predicted user supply and demand, (iii) users' preferences in terms of level of
718 comfort and available flexibility (e.g. deviations in battery levels, room temperature), and (iv) users'
719 preferences in terms of market participation (e.g. favouring community over profit, trading with
720 preferred peers). Existing approaches either take into account only user resources and completely
721 ignore user preferences or consider only the user preferences in terms of their comfort level within
722 their household [44, 96].

723 Strategic bidding is another phase that has seen little attention. User bids and offers can be
724 devised based on the available resources and user preferences. However, determining the best time,
725 volume and price needs external information about the market and possibly information about the
726 other users' intentions. As shown in Table 4, only limited information is shared between mar-
727 ket participants in the current models, mainly focusing on the price and volume of electricity re-
728 quested/offered.

729 Billing and settlements is the phase proceeding market clearance [170]. Once the transaction
730 details such as prices and volumes have been set, the next phase is to sort out the payments amongst
731 the market participants. In contrast to the retail market, where users have contractual obligations
732 with only one entity, their supplier, in P2P, CSC and TE markets, users can potentially trade with
733 every other market participant. Most markets have the market clearing phase before the settlement
734 period. Volumes to be traded, prices and transaction parties are determined in advance. Markets
735 assume that the volumes agreed in advance will be delivered during the trading period. In practice,
736 this might not be the case due to errors in the predictions.

737 Another important phase that has been largely ignored by the literature is dispute resolution
738 [171]. In any market that involves transactions between participants, there must be mechanisms
739 in place to deal with any disagreements.

740 4.3. Scalability and replicability

741 Few studies have tested their market proposal on large numbers of participants [41, 85, 87,
742 101, 123, 159–161]. The majority of markets operate within fixed environments and set boundary
743 conditions such as the type of stakeholders involved or the governance models applied. However,
744 to enable successful uptake of P2P, CSC and TE markets in the future, market designs need to be
745 able to respond to the dynamic nature of real-life applications. Dynamic parameters from within
746 the market, as well as dynamic environmental conditions will impact the performance of a market.

747 To enable the uptake of LEMs, market designs need to satisfy two key criteria, namely market
748 scalability and replicability. Our analysis has shown we have to differentiate between two types of
749 scalability. Firstly, markets need to be able to react to increasing numbers of participants. Our
750 analysis has not found any correlation between market size and complexity. However, Section 3.2
751 has shown that most market designs and settlement mechanisms have been tested using low numbers
752 of participants to provide an initial proof of concept. Secondly, markets need to be able to react to
753 changing market conditions over time, such as the type of assets in the market. More research on
754 the performance of markets with a high number of participants and changing market participation
755 over time is required.

756 The concept of replicability has barely been touched upon in the papers analysed. Replicability
757 can also be assessed from two perspectives. Firstly, a particular market design could be replicated in
758 different contexts and locations. This could include being exposed to various internal and external
759 parameters. These might include different types of participants, assets, requirements and electricity
760 grid typologies. Secondly, replicability also refers to the different regulatory contexts in which
761 markets must operate. This is especially the case when replicating a pilot project in a different
762 region or country with divergent policy and regulatory landscapes or norms and values.

763 4.4. Information security

764 P2P, CSC and TE markets rely on vast volumes of data. These data are either exchanged directly
765 among the market participants in fully decentralised models, or indirectly via central market opera-
766 tors in centralised models. The source of these data could range from small sensors on distribution
767 lines and prosumers' assets (e.g. remote terminal units, smart meters, home energy management sys-
768 tems) to large equipment (e.g. substations) and other market participants (e.g. suppliers, network
769 operators, aggregators, etc.). As the market outcome heavily depends on these data, the reliability,
770 authenticity and trustworthiness of these data are of paramount importance [172].

771 4.5. Prosumer privacy

772 The bids and offers submitted by market participants contain data about their energy use which
773 may be classed as personal data [173]. The reviewed papers do not consider the risks of loss of this
774 personal data either during transfer or from a market operator.

775 **5. Conclusion**

776 LEMs have seen increased interest in the academic literature as they are regarded as an ap-
777 propriate tool to respond to some of the challenges energy markets are currently facing. They can
778 incentivise the integration and uptake of renewable energy which is urgently needed to meet global
779 carbon reduction targets. P2P, CSC and TE markets are some of the most common LEM concepts.
780 However, these terms are currently used interchangeably and lack a clear definition, which can lead
781 to misconceptions amongst the scientific community and result in slower development. Through
782 the systematisation of knowledge of recent studies, we create an overview of the current state-of-art
783 research with regards to the market design and transaction aspects of LEMs. We contribute to a
784 transparent and clear representation of the underlying concepts and assumptions of LEMs. The
785 results of this review highlight the main differences and similarities between P2P, CSC and TE
786 markets and disclose key evidence gaps that require further research for LEMs to be successfully
787 implemented in the future.

788 To analyse the current academic literature in a structured manner, we adapted the TEAM
789 framework [23], which is used to analyse businesses that must both compete and cooperate in order
790 to make a market function (Section 2.3). A total of 139 peer-reviewed papers have been assessed
791 considering the strategy, technology and value of each proposed market. The framework was further
792 extended to gather data about the assumptions made in the markets, and the participants involved.

793 Our analysis of the defining characteristics of P2P, CSC and TE markets shows that P2P and
794 CSC markets mainly focus on providing a financial incentive to market participants. TE markets
795 have a stronger focus on providing grid-related services. Compared to the P2P and TE markets,
796 CSC markets are poorly represented in the literature. CSC markets focus on the community and
797 locality aspects of energy markets and follow a rather centralised governance structure (Section 3.1).

798 We have identified six archetypal designs used in P2P, CSC and TE markets. They mainly vary
799 with regards to their degree of centralisation and the number and types of price formation mecha-
800 nisms needed to settle the market (Section 3.2). The assessment of the price formation mechanisms
801 showed that there are three key archetypal mechanisms predominately used across the literature;
802 single and double auctions and bilateral negotiations (Section 3.3).

803 We assessed the value proposition of the markets. The most common commodity traded in
804 P2P energy markets is electrical energy. TE markets more frequently trade flexibility. This can be
805 referred back to the fact that P2P markets are more focused on providing services to the market
806 participants, while TE markets have a stronger focus on providing services to the grid (Section 3.4.1).
807 Most markets provide benefits to the participants, compensating them for their services by increasing
808 the total welfare in the market or reducing the costs of the participants. However, most papers do
809 not consider installation costs, which limits their applicability in real contexts (Section 3.4.2).

810 We evaluated the types of market participants involved and provided an overview of the assets
811 in the markets (Sections 3.5.1 and 3.5.2). While P2P markets mainly focus on small-scale individual
812 energy users, TE markets have a more diverse range of market participants across different scales.
813 All market types showed strong dependence on energy storage capacity. The assessment of the
814 number of market participants showed that most market mechanisms modelled are tested with only
815 a small number of participants. They are mainly case studies as a proof-of-concept of the proposed
816 market mechanism. This limits their replicability for real-life implementation, especially for markets
817 with a couple of hundred participants or more (Section 3.6.1).

818 While both P2P and CSC markets mainly focus on small scale energy users, TE markets have a
819 more diverse scale of operation. This supports the finding that TE markets operate across various
820 scales of the energy system. An assessment of the types of grid models and constraints highlighted
821 that only P2P and TE markets focus on the operation of the grid and the typology of the infrastruc-
822 ture (Section 3.6.3).

823 We concluded the paper by providing an overview of the key research gaps identified during
824 the review. These research gaps are the lack of: consideration of physical constraints; a holistic
825 approach to market design and operation; consideration about how these market designs will scale;
826 consideration of information security; and, consideration of market participant privacy.

827 The vast majority of papers in this review (137 of 139) were simulations or surveys and typically

828 focused on a specific aspect of the market. Pilot projects, by contrast, must take a holistic approach
829 to market design because they are actually implemented, albeit often with deviations from regula-
830 tions. Well studied pilot projects with thorough and publicly available results are an essential next
831 step in testing the feasibility of LEMs.

832 6. Data Availability

833 The completed data extraction table [174] which formed the basis of the analysis presented in
834 this paper is available at <https://doi.org/10.48420/16930768>.

835 Acknowledgements

836 This publication is part of the work of the Global Observatory on Peer-to-Peer, Community Self-
837 Consumption and Transactive Energy Models (GO-P2P), a task of the User-Centred Energy Systems
838 Technology Collaboration Programme (Users TCP), run under the auspices of the International
839 Energy Agency (IEA). GO-P2P benefits from the support of Australia, Belgium, Ireland, Italy, the
840 Netherlands, Switzerland, the United Kingdom and the United States.

841 The author TC received financial support from the EPSRC (grant number EP/L016141/1)
842 through the Power Networks Centre for Doctoral Training.

843 The author AG received funding from the EPSRC Centre for Doctoral Training in Energy De-
844 mand (LoLo), grant numbers EP/L01517X/1 and EP/H009612/1.

845 The author MAM received funding from the EPSRC through the project EnnCore EP/T026995/1,
846 Flemish Government through the FWO SBO project SNIPPET S007619, and The University of
847 Manchester through DKO Fellowship.

848 The author MB would like to acknowledge the support of the Department of the Environment,
849 Climate and Communications (DECC) for the GO-P2P project.

850 The author RC received funding from the EPSRC and InnovateUK co-funded EnergyREV project
851 (grant number EP/S031863/1).

852 References

- 853 [1] International Renewable Energy Agency, Global Energy Transformation: A Roadmap
854 to 2050 (2019 Edition), 2019. URL: [https://www.irena.org/publications/2019/Apr/
855 Global-energy-transformation-A-roadmap-to-2050-2019Edition](https://www.irena.org/publications/2019/Apr/Global-energy-transformation-A-roadmap-to-2050-2019Edition).
- 856 [2] United Nations Framework Convention on Climate Change, Paris agreement, 2015. URL:
857 https://unfccc.int/sites/default/files/english_paris_agreement.pdf.
- 858 [3] International Energy Agency, Net zero by 2050: A roadmap for the global energy sector, 2021.
859 URL: <https://www.iea.org/reports/net-zero-by-2050>.
- 860 [4] F. Hvelplund, Renewable energy and the need for local energy markets, Energy 31 (2006)
861 2293–2302. doi:10.1016/j.energy.2006.01.016.
- 862 [5] E. Mengelkamp, B. Notheisen, C. Beer, D. Dauer, C. Weinhardt, A blockchain-based smart
863 grid: towards sustainable local energy markets, Computer Science - Research and Development
864 33 (2017) 207–214. doi:10.1007/s00450-017-0360-9.
- 865 [6] C. Zhang, J. Wu, Y. Zhou, M. Cheng, C. Long, Peer-to-peer energy trading in a microgrid,
866 Applied Energy 220 (2018). doi:10.1016/j.apenergy.2018.03.010.
- 867 [7] European Parliament, Directive (EU) 2018/2001 of the european parliament and of the council
868 of 11 december 2018 on the promotion of the use of energy from renewable sources
869 (recast), 2018. URL: [https://eur-lex.europa.eu/legal-content/EN/TXT/?uri=uriserv:OJ.L_.
870 2018.328.01.0082.01.ENG](https://eur-lex.europa.eu/legal-content/EN/TXT/?uri=uriserv:OJ.L_.2018.328.01.0082.01.ENG).

- 871 [8] F. Dorian, T. Andreas, R. Josh, d. Stanislas, A. Gubina, Collective self-consumption and
872 energy communities: Overview of emerging regulatory approaches in europe, 2019. URL: <https://www.compile-project.eu/>.
873
- 874 [9] D. Frieden, A. Tuerk, C. Neumann, J. R. S. D’herbemont, J. Roberts, R. Eu,
875 M. Furlan, L. Herenčić, B. Pavlin, R. Marouço, N. Primo, A. R. Antunes,
876 B. Rónai, Collective self-consumption and energy communities: Trends and chal-
877 lenges in the transposition of the eu framework, 2020. URL: [https://www.rescoop.
878 eu/uploads/rescoop/downloads/Collective-self-consumption-and-energy-communities.
879 -Trends-and-challenges-in-the-transposition-of-the-EU-framework.pdf](https://www.rescoop.eu/uploads/rescoop/downloads/Collective-self-consumption-and-energy-communities.-Trends-and-challenges-in-the-transposition-of-the-EU-framework.pdf).
- 880 [10] S. Chen, C. C. Liu, From demand response to transactive energy: state of the art, *Journal of*
881 *Modern Power Systems and Clean Energy* 5 (2017) 10–19. doi:10.1007/s40565-016-0256-x.
- 882 [11] The GridWise Architecture Council, GridWise Transactive Energy Framework Version 1.0,
883 2015. URL: https://www.gridwiseac.org/pdfs/te_framework_report_pnnl-22946.pdf.
- 884 [12] M. Khorasany, Y. Mishra, G. Ledwich, Market framework for local energy trading : A re-
885 view of potential designs and market clearing approaches, *IET Generation, Transmission &*
886 *Distribution* (2018). doi:10.1049/iet-gtd.2018.5309.
- 887 [13] W. Tushar, T. K. Saha, C. Yuen, D. Smith, H. V. Poor, Peer-to-peer trading in electricity
888 networks: An overview, *IEEE Transactions on Smart Grid* 11 (2020) 3185–3200. doi:10.1109/
889 TSG.2020.2969657.
- 890 [14] X. Jin, Q. Wu, H. Jia, Local flexibility markets: Literature review on concepts, models and
891 clearing methods, 2020. doi:10.1016/j.apenergy.2019.114387.
- 892 [15] G. Tsaousoglou, J. S. Giraldo, N. G. Paterakis, Market mechanisms for local electricity mar-
893 kets: A review of models, solution concepts and algorithmic techniques, *Renewable and*
894 *Sustainable Energy Reviews* 156 (2022) 111890. doi:10.1016/j.rser.2021.111890.
- 895 [16] T. Sousa, T. Soares, P. Pinson, F. Moret, T. Baroche, E. Sorin, Peer-to-peer and community-
896 based markets: A comprehensive review, 2019. doi:10.1016/j.rser.2019.01.036.
- 897 [17] Y. Zhou, J. Wu, C. Long, W. Ming, State-of-the-art analysis and perspectives for peer-to-peer
898 energy trading, 2020. doi:10.1016/j.eng.2020.06.002.
- 899 [18] E. A. Soto, L. B. Bosman, E. Wollega, W. D. Leon-Salas, Peer-to-peer energy trading: A
900 review of the literature, *Applied Energy* 283 (2021). doi:10.1016/j.apenergy.2020.116268.
- 901 [19] S. Aggarwal, N. Kumar, S. Tanwar, M. Alazab, A survey on energy trading in the smart grid:
902 Taxonomy, research challenges and solutions, *IEEE Access* 9 (2021). doi:10.1109/ACCESS.
903 2021.3104354.
- 904 [20] M. Andoni, V. Robu, D. Flynn, S. Abram, D. Geach, D. Jenkins, P. McCallum, A. Peacock,
905 Blockchain technology in the energy sector: A systematic review of challenges and opportuni-
906 ties, *Renewable and Sustainable Energy Reviews* 100 (2019). doi:10.1016/j.rser.2018.10.
907 014.
- 908 [21] P. Siano, G. D. Marco, A. Rolan, V. Loia, A survey and evaluation of the potentials of
909 distributed ledger technology for peer-to-peer transactive energy exchanges in local energy
910 markets, *IEEE Systems Journal* 13 (2019). doi:10.1109/JSYST.2019.2903172.
- 911 [22] D. Kirli, B. Couraud, V. Robu, M. Salgado-Bravo, S. Norbu, M. Andoni, I. Antonopoulos,
912 M. Negrete-Pincetic, D. Flynn, A. Kiprakis, Smart contracts in energy systems: A systematic
913 review of fundamental approaches and implementations, *Renewable and Sustainable Energy*
914 *Reviews* 158 (2022) 112013. doi:10.1016/j.rser.2021.112013.

- 915 [23] R. Wieringa, W. Engelsman, J. Gordijn, D. Ionita, A business ecosystem architecture modeling
916 framework, volume 1, Institute of Electrical and Electronics Engineers Inc., 2019, pp. 147–156.
917 doi:[10.1109/CBI.2019.00024](https://doi.org/10.1109/CBI.2019.00024).
- 918 [24] L. P. Klein, A. Krivoglazova, L. Matos, J. Landeck, M. de Azevedo, A novel peer-to-peer
919 energy sharing business model for the portuguese energy market, *Energies* 13 (2019). doi:[10.
920 3390/en13010125](https://doi.org/10.3390/en13010125).
- 921 [25] E. Mengelkamp, J. Gärttner, K. Rock, S. Kessler, L. Orsini, C. Weinhardt, Designing microgrid
922 energy markets, *Applied Energy* 210 (2018). doi:[10.1016/j.apenergy.2017.06.054](https://doi.org/10.1016/j.apenergy.2017.06.054).
- 923 [26] D. L. Rodrigues, X. Ye, X. Xia, B. Zhu, Battery energy storage sizing optimisation for different
924 ownership structures in a peer-to-peer energy sharing community, *Applied Energy* 262 (2020).
925 doi:[10.1016/j.apenergy.2020.114498](https://doi.org/10.1016/j.apenergy.2020.114498).
- 926 [27] M. Yebiyo, R. Mercado, A. Gillich, I. Chaer, A. Day, A. Paurine, Novel economic modelling of
927 a peer-to-peer electricity market with the inclusion of distributed energy storage—the possible
928 case of a more robust and better electricity grid, *The Electricity Journal* 33 (2020). doi:[10.
929 1016/j.tej.2020.106709](https://doi.org/10.1016/j.tej.2020.106709).
- 930 [28] W. Tushar, T. K. Saha, C. Yuen, M. I. Azim, T. Morstyn, H. V. Poor, D. Niyato, R. Bean,
931 A coalition formation game framework for peer-to-peer energy trading, *Applied Energy* 261
932 (2020). doi:[10.1016/j.apenergy.2019.114436](https://doi.org/10.1016/j.apenergy.2019.114436).
- 933 [29] J. M. Zepter, A. Lüth, P. C. del Granado, R. Egging, Prosumer integration in wholesale elec-
934 tricity markets: Synergies of peer-to-peer trade and residential storage, *Energy and Buildings*
935 184 (2019). doi:[10.1016/j.enbuild.2018.12.003](https://doi.org/10.1016/j.enbuild.2018.12.003).
- 936 [30] W. Tushar, T. K. Saha, C. Yuen, T. Morstyn, Nahid-Al-Masood, H. V. Poor, R. Bean, Grid
937 influenced peer-to-peer energy trading, *IEEE Transactions on Smart Grid* 11 (2020). doi:[10.
938 1109/TSG.2019.2937981](https://doi.org/10.1109/TSG.2019.2937981).
- 939 [31] B.-C. Neagu, O. Ivanov, G. Grigoras, M. Gavrilas, A new vision on the prosumers energy
940 surplus trading considering smart peer-to-peer contracts, *Mathematics* 8 (2020). doi:[10.3390/
941 math8020235](https://doi.org/10.3390/math8020235).
- 942 [32] N. Wang, W. Xu, Z. Xu, W. Shao, Peer-to-peer energy trading among microgrids with multi-
943 dimensional willingness, *Energies* 11 (2018). doi:[10.3390/en11123312](https://doi.org/10.3390/en11123312).
- 944 [33] W. Liu, D. Qi, F. Wen, Intraday residential demand response scheme based on peer-to-peer
945 energy trading, *IEEE Transactions on Industrial Informatics* 16 (2020) 1823–1835. doi:[10.
946 1109/TII.2019.2929498](https://doi.org/10.1109/TII.2019.2929498).
- 947 [34] V. Heinisch, M. Odenberger, L. Göransson, F. Johnsson, Organizing prosumers into electricity
948 trading communities: Costs to attain electricity transfer limitations and self-sufficiency goals,
949 *International Journal of Energy Research* 43 (2019) 7021–7039. doi:[10.1002/er.4720](https://doi.org/10.1002/er.4720).
- 950 [35] W. Amin, Q. Huang, M. Afzal, A. A. Khan, K. Umer, S. A. Ahmed, A converging non-
951 cooperative & cooperative game theory approach for stabilizing peer-to-peer electricity trading,
952 *Electric Power Systems Research* 183 (2020). doi:[10.1016/j.epsr.2020.106278](https://doi.org/10.1016/j.epsr.2020.106278).
- 953 [36] Y. Wang, X. Wu, Y. Li, R. Yan, Y. Tan, X. Qiao, Y. Cao, Autonomous energy community
954 based on energy contract, *IET Generation, Transmission & Distribution* 14 (2020). doi:[10.
955 1049/iet-gtd.2019.1223](https://doi.org/10.1049/iet-gtd.2019.1223).
- 956 [37] M. Khorasany, Y. Mishra, G. Ledwich, Hybrid trading scheme for peer-to-peer energy trading
957 in transactive energy markets, *IET Generation, Transmission & Distribution* 14 (2020). doi:[10.
958 1049/iet-gtd.2019.1233](https://doi.org/10.1049/iet-gtd.2019.1233).

- 959 [38] D. Brown, S. Hall, M. E. Davis, Prosumers in the post subsidy era: an exploration of new
960 prosumer business models in the UK, *Energy Policy* 135 (2019). doi:[10.1016/j.enpol.2019.](https://doi.org/10.1016/j.enpol.2019.110984)
961 [110984](https://doi.org/10.1016/j.enpol.2019.110984).
- 962 [39] J. Zhang, C. Hu, C. Zheng, T. Rui, W. Shen, B. Wang, Distributed peer-to-peer electricity
963 trading considering network loss in a distribution system, *Energies* 12 (2019). doi:[10.3390/](https://doi.org/10.3390/en12224318)
964 [en12224318](https://doi.org/10.3390/en12224318).
- 965 [40] Lyu, Xu, Wang, Fu, Xu, A two-layer interactive mechanism for peer-to-peer energy trading
966 among virtual power plants, *Energies* 12 (2019). doi:[10.3390/en12193628](https://doi.org/10.3390/en12193628).
- 967 [41] M. Troncia, M. Galici, M. Mureddu, E. Ghiani, F. Pilo, Distributed ledger technologies
968 for peer-to-peer local markets in distribution networks, *Energies* 12 (2019). doi:[10.3390/](https://doi.org/10.3390/en12173249)
969 [en12173249](https://doi.org/10.3390/en12173249).
- 970 [42] E. Sorin, L. Bobo, P. Pinson, Consensus-based approach to peer-to-peer electricity markets
971 with product differentiation, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/](https://doi.org/10.1109/TPWRS.2018.2872880)
972 [TPWRS.2018.2872880](https://doi.org/10.1109/TPWRS.2018.2872880).
- 973 [43] W. Hou, L. Guo, Z. Ning, Local electricity storage for blockchain-based energy trading in
974 industrial internet of things, *IEEE Transactions on Industrial Informatics* 15 (2019). doi:[10.](https://doi.org/10.1109/TII.2019.2900401)
975 [1109/TII.2019.2900401](https://doi.org/10.1109/TII.2019.2900401).
- 976 [44] F. Moret, P. Pinson, Energy collectives: A community and fairness based approach to future
977 electricity markets, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/TPWRS.](https://doi.org/10.1109/TPWRS.2018.2808961)
978 [2018.2808961](https://doi.org/10.1109/TPWRS.2018.2808961).
- 979 [45] H. Nezamabadi, V. Vahidinasab, Microgrids bidding strategy in a transactive energy market,
980 *Scientia Iranica* (2019). doi:[10.24200/sci.2019.54148.3616](https://doi.org/10.24200/sci.2019.54148.3616).
- 981 [46] G. Prinsloo, A. Mammoli, R. Dobson, Customer domain supply and load coordination: A
982 case for smart villages and transactive control in rural off-grid microgrids, *Energy* 135 (2017).
983 doi:[10.1016/j.energy.2017.06.106](https://doi.org/10.1016/j.energy.2017.06.106).
- 984 [47] T. Morstyn, A. Teytelboym, M. D. McCulloch, Designing decentralized markets for distribu-
985 tion system flexibility, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/TPWRS.](https://doi.org/10.1109/TPWRS.2018.2886244)
986 [2018.2886244](https://doi.org/10.1109/TPWRS.2018.2886244).
- 987 [48] T. Chen, W. Su, Indirect customer-to-customer energy trading with reinforcement learning,
988 *IEEE Transactions on Smart Grid* 10 (2019). doi:[10.1109/TSG.2018.2857449](https://doi.org/10.1109/TSG.2018.2857449).
- 989 [49] N. Good, E. A. M. Ceseña, C. Heltorp, P. Mancarella, A transactive energy modelling and
990 assessment framework for demand response business cases in smart distributed multi-energy
991 systems, *Energy* 184 (2019). doi:[10.1016/j.energy.2018.02.089](https://doi.org/10.1016/j.energy.2018.02.089).
- 992 [50] J. P. Palacios, M. E. Samper, A. Vargas, Dynamic transactive energy scheme for smart dis-
993 tribution networks in a latin american context, *IET Generation, Transmission & Distribution*
994 13 (2019). doi:[10.1049/iet-gtd.2018.5272](https://doi.org/10.1049/iet-gtd.2018.5272).
- 995 [51] H. T. Nguyen, S. Battula, R. R. Takkala, Z. Wang, L. Tesfatsion, An integrated transmission
996 and distribution test system for evaluation of transactive energy designs, *Applied Energy* 240
997 (2019). doi:[10.1016/j.apenergy.2019.01.178](https://doi.org/10.1016/j.apenergy.2019.01.178).
- 998 [52] M. Khorasany, Y. Mishra, G. Ledwich, A decentralized bilateral energy trading system
999 for peer-to-peer electricity markets, *IEEE Transactions on Industrial Electronics* 67 (2020).
1000 doi:[10.1109/TIE.2019.2931229](https://doi.org/10.1109/TIE.2019.2931229).
- 1001 [53] A. Lüth, J. M. Zepter, P. C. del Granado, R. Egging, Local electricity market designs for
1002 peer-to-peer trading: The role of battery flexibility, *Applied Energy* 229 (2018). doi:[10.1016/](https://doi.org/10.1016/j.apenergy.2018.08.004)
1003 [j.apenergy.2018.08.004](https://doi.org/10.1016/j.apenergy.2018.08.004).

- 1004 [54] X. Yang, G. Wang, H. He, J. Lu, Y. Zhang, Automated demand response framework in
1005 elns: Decentralized scheduling and smart contract, *IEEE Transactions on Systems, Man, and*
1006 *Cybernetics: Systems* 50 (2020). doi:[10.1109/TSMC.2019.2903485](https://doi.org/10.1109/TSMC.2019.2903485).
- 1007 [55] A. Basnet, J. Zhong, Integrating gas energy storage system in a peer-to-peer community
1008 energy market for enhanced operation, *International Journal of Electrical Power & Energy*
1009 *Systems* 118 (2020). doi:[10.1016/j.ijepes.2019.105789](https://doi.org/10.1016/j.ijepes.2019.105789).
- 1010 [56] H. L. Cadre, P. Jacquot, C. Wan, C. Alasseur, Peer-to-peer electricity market analysis: From
1011 variational to generalized nash equilibrium, *European Journal of Operational Research* 282
1012 (2020). doi:[10.1016/j.ejor.2019.09.035](https://doi.org/10.1016/j.ejor.2019.09.035).
- 1013 [57] M. Marzband, M. H. Fouladfar, M. F. Akorede, G. Lightbody, E. Pouresmaeil, Framework for
1014 smart transactive energy in home-microgrids considering coalition formation and demand side
1015 management, *Sustainable Cities and Society* 40 (2018). doi:[10.1016/j.scs.2018.04.010](https://doi.org/10.1016/j.scs.2018.04.010).
- 1016 [58] H. Hao, C. D. Corbin, K. Kalsi, R. G. Pratt, Transactive control of commercial buildings for
1017 demand response, *IEEE Transactions on Power Systems* 32 (2017). doi:[10.1109/TPWRS.2016.](https://doi.org/10.1109/TPWRS.2016.2559485)
1018 [2559485](https://doi.org/10.1109/TPWRS.2016.2559485).
- 1019 [59] M. S. H. Nizami, M. J. Hossain, E. Fernandez, Multiagent-based transactive energy manage-
1020 ment systems for residential buildings with distributed energy resources, *IEEE Transactions*
1021 *on Industrial Informatics* 16 (2020). doi:[10.1109/TII.2019.2932109](https://doi.org/10.1109/TII.2019.2932109).
- 1022 [60] X. Tan, A. Leon-Garcia, Y. Wu, D. H. Tsang, Posted-price retailing of transactive energy: An
1023 optimal online mechanism without prediction, *IEEE Journal on Selected Areas in Communi-*
1024 *cations* 38 (2020) 5–16. doi:[10.1109/JSAC.2019.2951930](https://doi.org/10.1109/JSAC.2019.2951930).
- 1025 [61] M. Rayati, S. A. Goghari, Z. N. Gheidari, A. M. Ranjbar, An optimal and decentralized
1026 transactive energy system for electrical grids with high penetration of renewable energy sources,
1027 *International Journal of Electrical Power and Energy Systems* 113 (2019) 850–860. doi:[10.](https://doi.org/10.1016/j.ijepes.2019.06.017)
1028 [1016/j.ijepes.2019.06.017](https://doi.org/10.1016/j.ijepes.2019.06.017).
- 1029 [62] J. Hu, G. Yang, C. Ziras, K. Kok, Aggregator operation in the balancing market through
1030 network-constrained transactive energy, *IEEE Transactions on Power Systems* 34 (2019) 4071–
1031 4080. doi:[10.1109/TPWRS.2018.2874255](https://doi.org/10.1109/TPWRS.2018.2874255).
- 1032 [63] Z. Liu, Q. Wu, K. Ma, M. Shahidehpour, Y. Xue, S. Huang, Two-stage optimal scheduling of
1033 electric vehicle charging based on transactive control, *IEEE Transactions on Smart Grid* 10
1034 (2019). doi:[10.1109/TSG.2018.2815593](https://doi.org/10.1109/TSG.2018.2815593).
- 1035 [64] W. Liu, J. Zhan, C. Y. Chung, A novel transactive energy control mechanism for collaborative
1036 networked microgrids, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/TPWRS.](https://doi.org/10.1109/TPWRS.2018.2881251)
1037 [2018.2881251](https://doi.org/10.1109/TPWRS.2018.2881251).
- 1038 [65] S. Moazeni, B. Defourny, Optimal control of energy storage under random operation permis-
1039 sions, *IIESE Transactions* 50 (2018). doi:[10.1080/24725854.2017.1401756](https://doi.org/10.1080/24725854.2017.1401756).
- 1040 [66] R. Ghorani, M. Fotuhi-Firuzabad, M. Moeini-Aghtaie, Optimal bidding strategy of transactive
1041 agents in local energy markets, *IEEE Transactions on Smart Grid* 10 (2019). doi:[10.1109/](https://doi.org/10.1109/TSG.2018.2878024)
1042 [TSG.2018.2878024](https://doi.org/10.1109/TSG.2018.2878024).
- 1043 [67] Y. Chen, M. Hu, Swarm intelligence-based distributed stochastic model predictive control
1044 for transactive operation of networked building clusters, *Energy and Buildings* 198 (2019).
1045 doi:[10.1016/j.enbuild.2019.06.010](https://doi.org/10.1016/j.enbuild.2019.06.010).
- 1046 [68] M. Nizami, M. Hossain, B. R. Amin, E. Fernandez, A residential energy management sys-
1047 tem with bi-level optimization-based bidding strategy for day-ahead bi-directional electricity
1048 trading, *Applied Energy* 261 (2020). doi:[10.1016/j.apenergy.2019.114322](https://doi.org/10.1016/j.apenergy.2019.114322).

- 1049 [69] J. Lian, H. Ren, Y. Sun, D. J. Hammerstrom, Performance evaluation for transactive energy
1050 systems using double-auction market, *IEEE Transactions on Power Systems* 34 (2019) 4128–
1051 4137. doi:[10.1109/TPWRS.2018.2875919](https://doi.org/10.1109/TPWRS.2018.2875919).
- 1052 [70] U. J. Hahnel, M. Herberz, A. Pena-Bello, D. Parra, T. Brosch, Becoming prosumer: Revealing
1053 trading preferences and decision-making strategies in peer-to-peer energy communities, *Energy*
1054 *Policy* 137 (2020). doi:[10.1016/j.enpol.2019.111098](https://doi.org/10.1016/j.enpol.2019.111098).
- 1055 [71] S. Kuruseelan, C. Vaithilingam, Peer-to-Peer Energy Trading of a Community Connected with
1056 an AC and DC Microgrid, *Energies* 12 (2019). doi:[10.3390/en12193709](https://doi.org/10.3390/en12193709).
- 1057 [72] A. Paudel, K. Chaudhari, C. Long, H. B. Gooi, Peer-to-peer energy trading in a prosumer-
1058 based community microgrid: A game-theoretic model, *IEEE Transactions on Industrial Elec-*
1059 *tronics* 66 (2019). doi:[10.1109/TIE.2018.2874578](https://doi.org/10.1109/TIE.2018.2874578).
- 1060 [73] J. Kang, R. Yu, X. Huang, S. Maharjan, Y. Zhang, E. Hossain, Enabling localized peer-to-peer
1061 electricity trading among plug-in hybrid electric vehicles using consortium blockchains, *IEEE*
1062 *Transactions on Industrial Informatics* 13 (2017). doi:[10.1109/TII.2017.2709784](https://doi.org/10.1109/TII.2017.2709784).
- 1063 [74] S. Zhou, Z. Hu, W. Gu, M. Jiang, X.-P. Zhang, Artificial intelligence based smart energy
1064 community management: A reinforcement learning approach, *CSEE Journal of Power and*
1065 *Energy Systems* (2019). doi:[10.17775/CSEEJPES.2018.00840](https://doi.org/10.17775/CSEEJPES.2018.00840).
- 1066 [75] K. Chen, J. Lin, Y. Song, Trading strategy optimization for a prosumer in continuous dou-
1067 ble auction-based peer-to-peer market: A prediction-integration model, *Applied Energy* 242
1068 (2019). doi:[10.1016/j.apenergy.2019.03.094](https://doi.org/10.1016/j.apenergy.2019.03.094).
- 1069 [76] A. Masood, J. Hu, A. Xin, A. R. Sayed, G. Yang, Transactive energy for aggregated electric
1070 vehicles to reduce system peak load considering network constraints, *IEEE Access* 8 (2020)
1071 31519–31529. doi:[10.1109/ACCESS.2020.2973284](https://doi.org/10.1109/ACCESS.2020.2973284).
- 1072 [77] A. Amato, B. D. Martino, M. Scialdone, S. Venticinque, Distributed architecture for agents-
1073 based energy negotiation in solar powered micro-grids, *Concurrency and Computation: Prac-*
1074 *tice and Experience* 28 (2016). doi:[10.1002/cpe.3757](https://doi.org/10.1002/cpe.3757).
- 1075 [78] M. Mukherjee, L. Marinovici, T. Hardy, J. Hansen, Framework for large-scale implementation
1076 of wholesale-retail transactive control mechanism, *International Journal of Electrical Power &*
1077 *Energy Systems* 115 (2020). doi:[10.1016/j.ijepes.2019.105464](https://doi.org/10.1016/j.ijepes.2019.105464).
- 1078 [79] M. S. Nazir, I. A. Hiskens, A dynamical systems approach to modeling and analy-
1079 sis of transactive energy coordination, *IEEE Transactions on Power Systems* 34 (2019).
1080 doi:[10.1109/TPWRS.2018.2834913](https://doi.org/10.1109/TPWRS.2018.2834913).
- 1081 [80] J. An, M. Lee, S. Yeom, T. Hong, Determining the peer-to-peer electricity trading price and
1082 strategy for energy prosumers and consumers within a microgrid, *Applied Energy* 261 (2020).
1083 doi:[10.1016/j.apenergy.2019.114335](https://doi.org/10.1016/j.apenergy.2019.114335).
- 1084 [81] Z. Li, M. Shahidepour, A. Alabdulwahab, Y. Al-Turki, Valuation of distributed energy
1085 resources in active distribution networks, *The Electricity Journal* 32 (2019). doi:[10.1016/j.](https://doi.org/10.1016/j.tej.2019.03.001)
1086 [tej.2019.03.001](https://doi.org/10.1016/j.tej.2019.03.001).
- 1087 [82] C. Etukudor, B. Couraud, V. Robu, W. G. Früh, D. Flynn, C. Okereke, Automated negotiation
1088 for peer-to-peer electricity trading in local energy markets, *Energies* 13 (2020). doi:[10.3390/](https://doi.org/10.3390/en13040920)
1089 [en13040920](https://doi.org/10.3390/en13040920).
- 1090 [83] A. Leelasantitham, A Business Model Guideline of Electricity Utility Systems Based on
1091 Blockchain Technology in Thailand: A Case Study of Consumers, Prosumers and SMEs,
1092 *Wireless Personal Communications* 115 (2020). doi:[10.1007/s11277-020-07202-8](https://doi.org/10.1007/s11277-020-07202-8).

- 1093 [84] T. Baroche, P. Pinson, R. L. G. Latimier, H. B. Ahmed, Exogenous cost allocation in peer-
1094 to-peer electricity markets, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/
1095 TPWRS.2019.2896654](https://doi.org/10.1109/TPWRS.2019.2896654).
- 1096 [85] S. Chakraborty, T. Baarslag, M. Kaisers, Automated peer-to-peer negotiation for energy
1097 contract settlements in residential cooperatives, *Applied Energy* 259 (2020). doi:[10.1016/j.
1098 apenergy.2019.114173](https://doi.org/10.1016/j.apenergy.2019.114173).
- 1099 [86] Y. K. Renani, M. Ehsan, M. Shahidehpour, Optimal transactive market operations with
1100 distribution system operators, *IEEE Transactions on Smart Grid* 9 (2018). doi:[10.1109/TSG.
1101 2017.2718546](https://doi.org/10.1109/TSG.2017.2718546).
- 1102 [87] M. N. Faqiry, L. Wang, H. Wu, HEMS-enabled transactive flexibility in real-time operation
1103 of three-phase unbalanced distribution systems, *Journal of Modern Power Systems and Clean
1104 Energy* 7 (2019). doi:[10.1007/s40565-019-0553-2](https://doi.org/10.1007/s40565-019-0553-2).
- 1105 [88] S. Behboodi, D. P. Chassin, N. Djilali, C. Crawford, Transactive control of fast-acting demand
1106 response based on thermostatic loads in real-time retail electricity markets, *Applied Energy*
1107 210 (2018). doi:[10.1016/j.apenergy.2017.07.058](https://doi.org/10.1016/j.apenergy.2017.07.058).
- 1108 [89] Y. Yu, Y. Guo, W. Min, F. Zeng, Trusted transactions in micro-grid based on blockchain,
1109 *Energies* 12 (2019). doi:[10.3390/en12101952](https://doi.org/10.3390/en12101952).
- 1110 [90] Y. Li, W. Yang, P. He, C. Chen, X. Wang, Design and management of a distributed hybrid
1111 energy system through smart contract and blockchain, *Applied Energy* 248 (2019). doi:[10.
1112 1016/j.apenergy.2019.04.132](https://doi.org/10.1016/j.apenergy.2019.04.132).
- 1113 [91] R. Jing, M. N. Xie, F. X. Wang, L. X. Chen, Fair P2P energy trading between residential
1114 and commercial multi-energy systems enabling integrated demand-side management, *Applied
1115 Energy* 262 (2020). doi:[10.1016/j.apenergy.2020.114551](https://doi.org/10.1016/j.apenergy.2020.114551).
- 1116 [92] S. Nguyen, W. Peng, P. Sokolowski, D. Alahakoon, X. Yu, Optimizing rooftop photovoltaic
1117 distributed generation with battery storage for peer-to-peer energy trading, *Applied Energy*
1118 228 (2018). doi:[10.1016/j.apenergy.2018.07.042](https://doi.org/10.1016/j.apenergy.2018.07.042).
- 1119 [93] C. Long, J. Wu, Y. Zhou, N. Jenkins, Peer-to-peer energy sharing through a two-
1120 stage aggregated battery control in a community microgrid, *Applied Energy* 226 (2018).
1121 doi:[10.1016/j.apenergy.2018.05.097](https://doi.org/10.1016/j.apenergy.2018.05.097).
- 1122 [94] T. Pinto, R. Faia, M. A. F. Ghazvini, J. Soares, J. M. Corchado, Z. Vale, Decision support
1123 for small players negotiations under a transactive energy framework, *IEEE Transactions on
1124 Power Systems* 34 (2019). doi:[10.1109/TPWRS.2018.2861325](https://doi.org/10.1109/TPWRS.2018.2861325).
- 1125 [95] Elexon, Balancing & settlement code, 2020. URL: [https://www.elexon.co.uk/bsc-and-codes/
1126 balancing-settlement-code/](https://www.elexon.co.uk/bsc-and-codes/balancing-settlement-code/).
- 1127 [96] T. Morstyn, M. D. McCulloch, Multiclass energy management for peer-to-peer energy trading
1128 driven by prosumer preferences, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.
1129 1109/TPWRS.2018.2834472](https://doi.org/10.1109/TPWRS.2018.2834472).
- 1130 [97] D. J. Hammerstrom, R. Ambrosio, T. A. Carlon, J. G. DeSteese, G. R. Horst, R. Kajfasz,
1131 L. L. Kiesling, P. Michie, R. G. Pratt, M. Yao, J. Brous, D. P. Chassin, R. T. Guttromson,
1132 S. Katipamula, N. T. Le, T. V. Oliver, S. E. Thompson, Pacific northwest gridwise testbed
1133 demonstration projects; part i. olympic peninsula project, 2008. doi:[10.2172/926113](https://doi.org/10.2172/926113).
- 1134 [98] D. Easley, J. Ledyard, Theories of price formation and exchange in double oral auction markets,
1135 1993.

- 1136 [99] W. El-Baz, P. Tzscheutschler, U. Wagner, Integration of energy markets in microgrids: A
 1137 double-sided auction with device-oriented bidding strategies, *Applied Energy* 241 (2019).
 1138 doi:[10.1016/j.apenergy.2019.02.049](https://doi.org/10.1016/j.apenergy.2019.02.049).
- 1139 [100] Z. Wang, X. Yu, Y. Mu, H. Jia, A distributed peer-to-peer energy transaction method for
 1140 diversified prosumers in urban community microgrid system, *Applied Energy* 260 (2020).
 1141 doi:[10.1016/j.apenergy.2019.114327](https://doi.org/10.1016/j.apenergy.2019.114327).
- 1142 [101] Z. Zhang, R. Li, F. Li, A novel peer-to-peer local electricity market for joint trading of
 1143 energy and uncertainty, *IEEE Transactions on Smart Grid* 11 (2020). doi:[10.1109/TSG.2019.](https://doi.org/10.1109/TSG.2019.2933574)
 1144 [2933574](https://doi.org/10.1109/TSG.2019.2933574).
- 1145 [102] S. Zhou, F. Zou, Z. Wu, W. Gu, Q. Hong, C. Booth, A smart community energy management
 1146 scheme considering user dominated demand side response and P2P trading, *International Jour-*
 1147 *nal of Electrical Power & Energy Systems* 114 (2020). doi:[10.1016/j.ijepes.2019.105378](https://doi.org/10.1016/j.ijepes.2019.105378).
- 1148 [103] K. A. Melendez, V. Subramanian, T. K. Das, C. Kwon, Empowering end-use consumers of
 1149 electricity to aggregate for demand-side participation, *Applied Energy* 248 (2019). doi:[10.](https://doi.org/10.1016/j.apenergy.2019.04.092)
 1150 [1016/j.apenergy.2019.04.092](https://doi.org/10.1016/j.apenergy.2019.04.092).
- 1151 [104] C. Liu, J. Zhou, Y. Pan, Z. Li, Y. Wang, D. Xu, Q. Ding, Z. Luo, M. Shahidehpour, Multi-
 1152 period market operation of transmission-distribution systems based on heterogeneous decom-
 1153 position and coordination, *Energies* 12 (2019). doi:[10.3390/en12163126](https://doi.org/10.3390/en12163126).
- 1154 [105] G. Mohy-ud-din, K. M. Muttaqi, D. Sutanto, Transactive energy-based planning framework
 1155 for VPPs in a co-optimised day-ahead and real-time energy market with ancillary services,
 1156 *IET Generation, Transmission & Distribution* 13 (2019). doi:[10.1049/iet-gtd.2018.5831](https://doi.org/10.1049/iet-gtd.2018.5831).
- 1157 [106] M. R. Alam, M. St-Hilaire, T. Kunz, Peer-to-peer energy trading among smart homes, *Applied*
 1158 *Energy* 238 (2019). doi:[10.1016/j.apenergy.2019.01.091](https://doi.org/10.1016/j.apenergy.2019.01.091).
- 1159 [107] H. S. V. S. K. Nunna, D. Srinivasan, Multiagent-based transactive energy framework for
 1160 distribution systems with smart microgrids, *IEEE Transactions on Industrial Informatics* 13
 1161 (2017). doi:[10.1109/TII.2017.2679808](https://doi.org/10.1109/TII.2017.2679808).
- 1162 [108] M. Babar, J. Grela, A. Ożadowicz, P. Nguyen, Z. Hanzelka, I. Kamphuis, Energy flexometer:
 1163 Transactive energy-based internet of things technology, *Energies* 11 (2018). doi:[10.3390/](https://doi.org/10.3390/en11030568)
 1164 [en11030568](https://doi.org/10.3390/en11030568).
- 1165 [109] P. H. Divshali, B. Choi, H. Liang, L. Söder, Transactive Demand Side Management Programs
 1166 in Smart Grids with High Penetration of EVs, *Energies* 10 (2017). doi:[10.3390/en10101640](https://doi.org/10.3390/en10101640).
- 1167 [110] J. Qiu, K. Meng, Y. Zheng, Z. Y. Dong, Optimal scheduling of distributed energy resources
 1168 as a virtual power plant in a transactive energy framework, *IET Generation, Transmission &*
 1169 *Distribution* 11 (2017) 3417–3427. doi:[10.1049/iet-gtd.2017.0268](https://doi.org/10.1049/iet-gtd.2017.0268).
- 1170 [111] I. Lopez-Rodriguez, M. Hernandez-Tejera, Infrastructure based on supernodes and software
 1171 agents for the implementation of energy markets in demand-response programs, *Applied*
 1172 *Energy* 158 (2015). doi:[10.1016/j.apenergy.2015.08.039](https://doi.org/10.1016/j.apenergy.2015.08.039).
- 1173 [112] Y. Wang, Z. Huang, M. Shahidehpour, L. L. Lai, Z. Wang, Q. Zhu, Reconfigurable distribution
 1174 network for managing transactive energy in a multi-microgrid system, *IEEE Transactions on*
 1175 *Smart Grid* 11 (2020). doi:[10.1109/TSG.2019.2935565](https://doi.org/10.1109/TSG.2019.2935565).
- 1176 [113] M. N. Faqiry, L. Edmonds, H. Wu, A. Pahwa, Distribution locational marginal price-based
 1177 transactive day-ahead market with variable renewable generation, *Applied Energy* 259 (2020).
 1178 doi:[10.1016/j.apenergy.2019.114103](https://doi.org/10.1016/j.apenergy.2019.114103).

- 1179 [114] K. Moslehi, A. B. R. Kumar, Autonomous Resilient Grids in an IoT Landscape Vision for
1180 a Nested Transactive Grid, *IEEE Transactions on Power Systems* 34 (2019). doi:[10.1109/
1181 TPWRS.2018.2810134](https://doi.org/10.1109/TPWRS.2018.2810134).
- 1182 [115] M. L. D. Silvestre, P. Gallo, M. G. Ippolito, R. Musca, E. R. Sanseverino, Q. T. T. Tran,
1183 G. Zizzo, Ancillary services in the energy blockchain for microgrids, *IEEE Transactions on
1184 Industry Applications* 55 (2019). doi:[10.1109/TIA.2019.2909496](https://doi.org/10.1109/TIA.2019.2909496).
- 1185 [116] D. Wang, Q. Hu, H. Jia, K. Hou, W. Du, N. Chen, X. Wang, M. Fan, Integrated demand
1186 response in district electricity-heating network considering double auction retail energy market
1187 based on demand-side energy stations, *Applied Energy* 248 (2019). doi:[10.1016/j.apenergy.
1188 2019.04.050](https://doi.org/10.1016/j.apenergy.2019.04.050).
- 1189 [117] M. Marzband, F. Azarinejadian, M. Savaghebi, E. Pouresmaeil, J. M. Guerrero, G. Light-
1190 body, Smart transactive energy framework in grid-connected multiple home microgrids under
1191 independent and coalition operations, *Renewable Energy* 126 (2018). doi:[10.1016/j.renene.
1192 2018.03.021](https://doi.org/10.1016/j.renene.2018.03.021).
- 1193 [118] Y. Chen, M. Hu, Balancing collective and individual interests in transactive energy manage-
1194 ment of interconnected micro-grid clusters, *Energy* 109 (2016). doi:[10.1016/j.energy.2016.
1195 05.052](https://doi.org/10.1016/j.energy.2016.05.052).
- 1196 [119] P. Siano, D. Sarno, L. Straccia, A. T. Marrazzo, A novel method for evaluating the impact
1197 of residential demand response in a real time distribution energy market, *Journal of Ambient
1198 Intelligence and Humanized Computing* 7 (2016). doi:[10.1007/s12652-015-0339-y](https://doi.org/10.1007/s12652-015-0339-y).
- 1199 [120] J. Qiu, J. Zhao, H. Yang, Z. Y. Dong, Optimal scheduling for prosumers in coupled transactive
1200 power and gas systems, *IEEE Transactions on Power Systems* 33 (2018). doi:[10.1109/TPWRS.
1201 2017.2715983](https://doi.org/10.1109/TPWRS.2017.2715983).
- 1202 [121] H. Zhang, Y. Li, D. W. Gao, J. Zhou, Distributed optimal energy management for energy
1203 internet, *IEEE Transactions on Industrial Informatics* 13 (2017). doi:[10.1109/TII.2017.
1204 2714199](https://doi.org/10.1109/TII.2017.2714199).
- 1205 [122] K. Saxena, A. R. Abhyankar, Agent based bilateral transactive market for emerging distri-
1206 bution system considering imbalances, *Sustainable Energy, Grids and Networks* 18 (2019).
1207 doi:[10.1016/j.segan.2019.100203](https://doi.org/10.1016/j.segan.2019.100203).
- 1208 [123] J. Wu, J. Hu, X. Ai, Z. Zhang, H. Hu, Multi-time scale energy management of electric
1209 vehicle model-based prosumers by using virtual battery model, *Applied Energy* 251 (2019).
1210 doi:[10.1016/j.apenergy.2019.113312](https://doi.org/10.1016/j.apenergy.2019.113312).
- 1211 [124] S. Cui, Y.-W. Wang, J.-W. Xiao, Peer-to-peer energy sharing among smart energy buildings
1212 by distributed transaction, *IEEE Transactions on Smart Grid* 10 (2019). doi:[10.1109/TSG.
1213 2019.2906059](https://doi.org/10.1109/TSG.2019.2906059).
- 1214 [125] M. Faqiry, L. Edmonds, H. Zhang, A. Khodaei, H. Wu, Transactive-market-based operation of
1215 distributed electrical energy storage with grid constraints, *Energies* 10 (2017). doi:[10.3390/
1216 en10111891](https://doi.org/10.3390/en10111891).
- 1217 [126] J. Guerrero, A. C. Chapman, G. Verbic, Decentralized P2P Energy Trading Under Network
1218 Constraints in a Low-Voltage Network, *IEEE Transactions on Smart Grid* 10 (2019). doi:[10.
1219 1109/TSG.2018.2878445](https://doi.org/10.1109/TSG.2018.2878445).
- 1220 [127] Ofgem, Smart export guarantee: Guidance for generators, 2019. URL: [https://www.ofgem.
1221 gov.uk/system/files/docs/2020/02/seg_generator_guidance_-_final_for_publication.pdf](https://www.ofgem.gov.uk/system/files/docs/2020/02/seg_generator_guidance_-_final_for_publication.pdf).
- 1222 [128] C. Feng, Z. Li, M. Shahidehpour, F. Wen, Q. Li, Stackelberg game based transactive pricing
1223 for optimal demand response in power distribution systems, *International Journal of Electrical
1224 Power & Energy Systems* 118 (2020). doi:[10.1016/j.ijepes.2019.105764](https://doi.org/10.1016/j.ijepes.2019.105764).

- 1225 [129] X. Wen, D. Abbas, B. Francois, Modeling of photovoltaic power uncertainties for impact
1226 analysis on generation scheduling and cost of an urban micro grid, *Mathematics and Computers*
1227 *in Simulation* 183 (2021) 116–128. doi:[10.1016/j.matcom.2020.02.023](https://doi.org/10.1016/j.matcom.2020.02.023).
- 1228 [130] B. Zhou, Y. Meng, W. Huang, H. Wang, L. Deng, S. Huang, J. Wei, Multi-energy net load
1229 forecasting for integrated local energy systems with heterogeneous prosumers, *International*
1230 *Journal of Electrical Power and Energy Systems* 126 (2021). doi:[10.1016/j.ijepes.2020.](https://doi.org/10.1016/j.ijepes.2020.106542)
1231 [106542](https://doi.org/10.1016/j.ijepes.2020.106542).
- 1232 [131] S. Noor, W. Yang, M. Guo, K. H. van Dam, X. Wang, Energy demand side management
1233 within micro-grid networks enhanced by blockchain, *Applied Energy* 228 (2018). doi:[10.](https://doi.org/10.1016/j.apenergy.2018.07.012)
1234 [1016/j.apenergy.2018.07.012](https://doi.org/10.1016/j.apenergy.2018.07.012).
- 1235 [132] T. Morstyn, A. Teytelboym, M. D. McCulloch, Bilateral contract networks for peer-to-peer
1236 energy trading, *IEEE Transactions on Smart Grid* 10 (2019). doi:[10.1109/TSG.2017.2786668](https://doi.org/10.1109/TSG.2017.2786668).
- 1237 [133] N. K. Meena, J. Yang, E. Zacharis, Optimisation framework for the design and operation of
1238 open-market urban and remote community microgrids, *Applied Energy* 252 (2019). doi:[10.](https://doi.org/10.1016/j.apenergy.2019.113399)
1239 [1016/j.apenergy.2019.113399](https://doi.org/10.1016/j.apenergy.2019.113399).
- 1240 [134] E. Reihani, P. Siano, M. Genova, A new method for peer-to-peer energy exchange in distri-
1241 bution grids, *Energies* 13 (2020). doi:[10.3390/en13040799](https://doi.org/10.3390/en13040799).
- 1242 [135] L. Park, S. Lee, H. Chang, A sustainable home energy prosumer-chain methodology with
1243 energy tags over the blockchain, *Sustainability* 10 (2018). doi:[10.3390/su10030658](https://doi.org/10.3390/su10030658).
- 1244 [136] E. A. M. Ceseña, N. Good, A. L. Syri, P. Mancarella, Techno-economic and business case
1245 assessment of multi-energy microgrids with co-optimization of energy, reserve and reliability
1246 services, *Applied Energy* 210 (2018). doi:[10.1016/j.apenergy.2017.08.131](https://doi.org/10.1016/j.apenergy.2017.08.131).
- 1247 [137] K. Anoh, S. Maharjan, A. Ikpehai, Y. Zhang, B. Adebisi, Energy peer-to-peer trading in
1248 virtual microgrids in smart grids: A game-theoretic approach, *IEEE Transactions on Smart*
1249 *Grid* 11 (2020). doi:[10.1109/TSG.2019.2934830](https://doi.org/10.1109/TSG.2019.2934830).
- 1250 [138] C. Dang, J. Zhang, C.-P. Kwong, L. Li, Demand side load management for big industrial
1251 energy users under blockchain-based peer-to-peer electricity market, *IEEE Transactions on*
1252 *Smart Grid* 10 (2019). doi:[10.1109/TSG.2019.2904629](https://doi.org/10.1109/TSG.2019.2904629).
- 1253 [139] U. Cali, O. Cakir, Energy policy instruments for distributed ledger technology empowered
1254 peer-to-peer local energy markets, *IEEE Access* 7 (2019). doi:[10.1109/ACCESS.2019.2923906](https://doi.org/10.1109/ACCESS.2019.2923906).
- 1255 [140] M. L. D. Silvestre, P. Gallo, M. G. Ippolito, E. R. Sanseverino, G. Zizzo, A technical approach
1256 to the energy blockchain in microgrids, *IEEE Transactions on Industrial Informatics* 14 (2018).
1257 doi:[10.1109/TII.2018.2806357](https://doi.org/10.1109/TII.2018.2806357).
- 1258 [141] W. Tushar, T. K. Saha, C. Yuen, P. Liddell, R. Bean, H. V. Poor, Peer-to-peer energy
1259 trading with sustainable user participation: A game theoretic approach, *IEEE Access* 6 (2018).
1260 doi:[10.1109/ACCESS.2018.2875405](https://doi.org/10.1109/ACCESS.2018.2875405).
- 1261 [142] M. R. Alam, M. St-Hilaire, T. Kunz, An optimal P2P energy trading model for smart homes
1262 in the smart grid, *Energy Efficiency* 10 (2017). doi:[10.1007/s12053-017-9532-5](https://doi.org/10.1007/s12053-017-9532-5).
- 1263 [143] Y. Liu, H. B. Gooi, Y. Li, H. Xin, J. Ye, A secure distributed transactive energy management
1264 scheme for multiple interconnected microgrids considering misbehaviors, *IEEE Transactions*
1265 *on Smart Grid* 10 (2019). doi:[10.1109/TSG.2019.2895229](https://doi.org/10.1109/TSG.2019.2895229).
- 1266 [144] S. A. Janko, N. G. Johnson, Scalable multi-agent microgrid negotiations for a transactive
1267 energy market, *Applied Energy* 229 (2018). doi:[10.1016/j.apenergy.2018.08.026](https://doi.org/10.1016/j.apenergy.2018.08.026).

- 1268 [145] Y. Liu, K. Zuo, X. A. Liu, J. Liu, J. M. Kennedy, Dynamic pricing for decentralized energy
1269 trading in micro-grids, *Applied Energy* 228 (2018). doi:[10.1016/j.apenergy.2018.06.124](https://doi.org/10.1016/j.apenergy.2018.06.124).
- 1270 [146] T. Pinto, M. F. Ghazvini, J. Soares, R. Faia, J. Corchado, R. Castro, Z. Vale, Decision
1271 support for negotiations among microgrids using a multiagent architecture, *Energies* 11 (2018).
1272 doi:[10.3390/en11102526](https://doi.org/10.3390/en11102526).
- 1273 [147] G. Prinsloo, R. Dobson, A. Mammoli, Synthesis of an intelligent rural village microgrid
1274 control strategy based on smartgrid multi-agent modelling and transactive energy management
1275 principles, *Energy* 147 (2018). doi:[10.1016/j.energy.2018.01.056](https://doi.org/10.1016/j.energy.2018.01.056).
- 1276 [148] M. Akter, M. Mahmud, A. Oo, A hierarchical transactive energy management system for
1277 energy sharing in residential microgrids, *Energies* 10 (2017). doi:[10.3390/en10122098](https://doi.org/10.3390/en10122098).
- 1278 [149] W. Qi, B. Shen, H. Zhang, Z.-J. M. Shen, Sharing demand-side energy resources - a conceptual
1279 design, *Energy* 135 (2017). doi:[10.1016/j.energy.2017.06.144](https://doi.org/10.1016/j.energy.2017.06.144).
- 1280 [150] H. Huang, S. Nie, J. Lin, Y. Wang, J. Dong, Optimization of Peer-to-Peer Power Trading
1281 in a Microgrid with Distributed PV and Battery Energy Storage Systems, *Sustainability* 12
1282 (2020). doi:[10.3390/su12030923](https://doi.org/10.3390/su12030923).
- 1283 [151] H. Liu, Y. Zhang, S. Zheng, Y. Li, Electric Vehicle Power Trading Mechanism Based on
1284 Blockchain and Smart Contract in V2G Network, *IEEE Access* 7 (2019). doi:[10.1109/ACCESS.
2019.2951057](https://doi.org/10.1109/ACCESS.2019.2951057).
- 1286 [152] Y. Zhou, J. Wu, C. Long, Evaluation of peer-to-peer energy sharing mechanisms based on
1287 a multiagent simulation framework, *Applied Energy* 222 (2018). doi:[10.1016/j.apenergy.
2018.02.089](https://doi.org/10.1016/j.apenergy.2018.02.089).
- 1289 [153] A. Ghosh, V. Aggarwal, H. Wan, Strategic prosumers: How to set the prices in a tiered
1290 market?, *IEEE Transactions on Industrial Informatics* 15 (2019). doi:[10.1109/TII.2018.
2889301](https://doi.org/10.1109/TII.2018.2889301).
- 1292 [154] B. Hayes, S. Thakur, J. Breslin, Co-simulation of electricity distribution networks and peer to
1293 peer energy trading platforms, *International Journal of Electrical Power & Energy Systems*
1294 115 (2020). doi:[10.1016/j.ijepes.2019.105419](https://doi.org/10.1016/j.ijepes.2019.105419).
- 1295 [155] H. Zhang, H. Zhang, L. Song, Y. Li, Z. Han, H. V. Poor, Peer-to-Peer Energy Trading in DC
1296 Packetized Power Microgrids, *IEEE Journal on Selected Areas in Communications* 38 (2020).
1297 doi:[10.1109/JSAC.2019.2951991](https://doi.org/10.1109/JSAC.2019.2951991).
- 1298 [156] M. Khorasany, Y. Mishra, B. Babaki, G. Ledwich, Enhancing scalability of peer-to-peer energy
1299 markets using adaptive segmentation method, *Journal of Modern Power Systems and Clean
1300 Energy* 7 (2019). doi:[10.1007/s40565-019-0510-0](https://doi.org/10.1007/s40565-019-0510-0).
- 1301 [157] J. Lin, M. Pipattanasomporn, S. Rahman, Comparative analysis of auction mechanisms and
1302 bidding strategies for P2P solar transactive energy markets, *Applied Energy* 255 (2019).
1303 doi:[10.1016/j.apenergy.2019.113687](https://doi.org/10.1016/j.apenergy.2019.113687).
- 1304 [158] M. H. Y. Moghaddam, A. Leon-Garcia, A fog-based internet of energy architecture for trans-
1305 active energy management systems, *IEEE Internet of Things Journal* 5 (2018). doi:[10.1109/
JIOT.2018.2805899](https://doi.org/10.1109/JIOT.2018.2805899).
- 1307 [159] F. Luo, Z. Y. Dong, G. Liang, J. Murata, Z. Xu, A distributed electricity trading system in
1308 active distribution networks based on multi-agent coalition and blockchain, *IEEE Transactions
1309 on Power Systems* 34 (2019). doi:[10.1109/TPWRS.2018.2876612](https://doi.org/10.1109/TPWRS.2018.2876612).
- 1310 [160] Z. Liu, L. Wang, L. Ma, A transactive energy framework for coordinated energy management
1311 of networked microgrids with distributionally robust optimization, *IEEE Transactions on
1312 Power Systems* 35 (2020). doi:[10.1109/TPWRS.2019.2933180](https://doi.org/10.1109/TPWRS.2019.2933180).

- 1313 [161] J. Li, C. Zhang, Z. Xu, J. Wang, J. Zhao, Y.-J. A. Zhang, Distributed transactive energy
1314 trading framework in distribution networks, *IEEE Transactions on Power Systems* 33 (2018).
1315 doi:[10.1109/TPWRS.2018.2854649](https://doi.org/10.1109/TPWRS.2018.2854649).
- 1316 [162] F. Lezama, J. Soares, P. Hernandez-Leal, M. Kaisers, T. Pinto, Z. Vale, Local energy markets:
1317 Paving the path toward fully transactive energy systems, *IEEE Transactions on Power Systems*
1318 34 (2019). doi:[10.1109/TPWRS.2018.2833959](https://doi.org/10.1109/TPWRS.2018.2833959).
- 1319 [163] O. Samuel, A. Almogren, A. Javaid, M. Zuair, I. Ullah, N. Javaid, Leveraging Blockchain Tech-
1320 nology for Secure Energy Trading and Least-Cost Evaluation of Decentralized Contributions
1321 to Electrification in Sub-Saharan Africa, *Entropy* 22 (2020). doi:[10.3390/e22020226](https://doi.org/10.3390/e22020226).
- 1322 [164] K. Inayat, S. O. Hwang, Load balancing in decentralized smart grid trade system using
1323 blockchain, *Journal of Intelligent & Fuzzy Systems* 35 (2018). doi:[10.3233/JIFS-169832](https://doi.org/10.3233/JIFS-169832).
- 1324 [165] V. Dudjak, D. Neves, T. Alskaf, S. Khadem, A. Pena-Bello, P. Saggese, B. Bowler, M. Andoni,
1325 M. Bertolini, Y. Zhou, B. Lormeteau, M. A. Mustafa, Y. Wang, C. Francis, F. Zobiri, D. Parra,
1326 A. Papaemmanouil, Impact of local energy markets integration in power systems layer: A
1327 comprehensive review, *Applied Energy* 301 (2021) 117434. URL: [https://doi.org/10.1016/j.](https://doi.org/10.1016/j.apenergy.2021.117434)
1328 [apenergy.2021.117434](https://doi.org/10.1016/j.apenergy.2021.117434). doi:[10.1016/j.apenergy.2021.117434](https://doi.org/10.1016/j.apenergy.2021.117434).
- 1329 [166] Australian Energy Market Operator, Five-minute settlement: High level
1330 design, 2017. URL: [https://www.aemc.gov.au/sites/default/files/content/](https://www.aemc.gov.au/sites/default/files/content/b862be5a-4460-4b72-a90b-8f73117f301c/5MS-HLD-Final-4-Sep.pdf)
1331 [b862be5a-4460-4b72-a90b-8f73117f301c/5MS-HLD-Final-4-Sep.pdf](https://www.aemc.gov.au/sites/default/files/content/b862be5a-4460-4b72-a90b-8f73117f301c/5MS-HLD-Final-4-Sep.pdf).
- 1332 [167] H. Gerard, E. I. R. Puente, D. Six, Coordination between transmission and distribution system
1333 operators in the electricity sector: A conceptual framework, *Utilities Policy* 50 (2018) 40–48.
1334 doi:[10.1016/j.jup.2017.09.011](https://doi.org/10.1016/j.jup.2017.09.011).
- 1335 [168] J. Iria, P. Scott, A. Attarha, Network-constrained bidding optimization strategy for aggrega-
1336 tors of prosumers, *Energy* 207 (2020) 118266. doi:[10.1016/j.energy.2020.118266](https://doi.org/10.1016/j.energy.2020.118266).
- 1337 [169] A. Abidin, R. Callaerts, G. Deconinck, V. D. G. Shenja, A. Madhusudan, M. Montakhabi,
1338 M. A. Mustafa, S. Nikova, D. Orlando, J. Schroers, S. Vanhove, F. Zobiri, Poster: Snippet –
1339 secure and privacy-friendly peer-to-peer electricity trading, Internet Society, 2020.
- 1340 [170] R. Thandi, M. A. Mustafa, Privacy-enhancing settlements protocol in peer-to-peer energy
1341 trading markets, *IEEE*, 2021. doi:[10.1109/ISGT.2014.6816376](https://doi.org/10.1109/ISGT.2014.6816376).
- 1342 [171] M. A. Mustafa, S. Cleemput, A. Abidin, A local electricity trading market: Security analysis,
1343 *IEEE*, 2016. doi:[10.1109/ISGTEurope.2016.7856269](https://doi.org/10.1109/ISGTEurope.2016.7856269).
- 1344 [172] G. Kalogridis, M. Sooriyabandara, Z. Fan, M. A. Mustafa, Toward unified security and privacy
1345 protection for smart meter networks, *IEEE Systems Journal* 8 (2014). doi:[10.1109/JSYST.](https://doi.org/10.1109/JSYST.2013.2260940)
1346 [2013.2260940](https://doi.org/10.1109/JSYST.2013.2260940).
- 1347 [173] E. L. Quinn, Privacy and the new energy infrastructure, *SSRN Electronic Journal* (2009).
1348 doi:[10.2139/ssrn.1370731](https://doi.org/10.2139/ssrn.1370731).
- 1349 [174] T. Capper, A. Gorbacheva, J. M. Schwidtal, M. A. Mustafa, M. Andoni, R. Chitchyan,
1350 V. Robu, M. Montakhabi, P. Piccini, M. Bahloul, T. Mbavarira, L. Kiesling, I. J. Scott,
1351 C. Francis, J. M. Espana, M. Troncia, Peer-to-peer, self-consumption and transactive energy
1352 literature review data extraction table, 2022. doi:[10.48420/16930768](https://doi.org/10.48420/16930768).
- 1353 [175] S. Wang, A. F. Taha, J. Wang, K. Kvaternik, A. Hahn, Energy crowdsourcing and peer-to-peer
1354 energy trading in blockchain-enabled smart grids, *IEEE Transactions on Systems, Man, and*
1355 *Cybernetics: Systems* 49 (2019) 1612–1623. doi:[10.1109/TSMC.2019.2916565](https://doi.org/10.1109/TSMC.2019.2916565).

1356 **Appendix A. Additional Data**

1357 This appendix contains tables of supporting data and references. Each table is referenced in the
 1358 relevant part of the results section, and is briefly introduced here as well.

1359 Table A.5 provides references for the market design and price formation mechanisms. The papers
 1360 are grouped based on market design, price formation mechanism and market type (P2P, CSC or
 1361 TE). Discussion about market design is provided in Section 3.2 and discussion about price formation
 1362 mechanism is provided in Section 3.3.

1363 Table A.6 provides references based on the different market participant needs and the market
 1364 commodity, broken down by market types (P2P, CSC or TE). The market commodity is discussed
 1365 further in Section 3.4.1 and the needs of the market participants are discussed in Section 3.4.2.

1366 Table A.7 provides references for the types of market participants, split by market type (P2P,
 1367 CSC or TE). Further discussion of market participants can be found in Section 3.5.1.

1368 Table A.8 provides references for the different types of assets of market participants split by
 1369 market type (P2P, CSC or TE). Further discussion about the assets of market participants can be
 1370 found in Section 3.5.2.

1371 Table A.9 provides references for each type of grid model used, split by market type (P2P or
 1372 TE) and what the grid was used to model (constraints, power loss or other). Further information
 1373 about the grid models used in the reviewed literature is available in Section 3.6.3.

Table A.5: Price formation mechanism and market design

Price FM	Market design						Type
	F	RT	Mixed C/D	Mixed F/RT	Multilayer	S.A.T.F	
Single auc- tion	[6, 27, 29, 31, 43, 52, 56, 84, 89, 92, 96, 106, 111, 121, 134, 135, 138, 142, 163, 164]	[133]	[53]	[123, 175]	[159]	-	P2P
	[49, 50, 57, 62, 65, 66, 79, 81, 86, 87, 105, 112, 149, 158, 162]	[60, 61, 119]	-	[45, 51, 59, 63, 67, 120]	[47, 104, 145, 148, 160]	-	TE
	[26, 34]	-	-	-	-	-	CSC
Double auction	[21, 25, 28, 30, 32, 33, 36, 37, 40, 41, 55, 72, 74, 75, 90, 100, 101, 126, 128, 131, 150- 155]	[54, 73]	[35, 124]	[132]	[114]	-	P2P
	[46, 64, 69, 76, 94, 103, 108-110, 115, 116, 118, 125, 143, 146, 147, 157]	[88]	-	[48, 68, 99, 107]	[78, 117]	-	TE
	[44]	-	-	-	[102]	-	CSC
Bilateral negotia- tion	[42, 82, 85, 122, 137, 156, 161]	[71]	-	-	-	[24, 139]	P2P
	[39, 94, 144]	-	-	-	-	-	TE
	[77]	-	-	-	-	-	CSC

* FM – Formation Mechanism; F – Futures; RT – Real Time; C – Centralised; D – Decentralised; S.A.T.F. – Settled After the Fact

Table A.6: Needs of participants addressed by P2P, CSC and TE markets

Core need	Secondary need	Commodity	P2P	TE	CSC
↑ Total welfare	None	Electricity	[6, 27, 28, 52, 56, 70, 89, 135, 139, 151, 153, 154, 175]	[60, 118]	-
↑ Total welfare	None	Flexibility	-	[108]	-
↑ Total welfare	↔ Grid constraints	Electricity	[37, 39, 134, 155]	[37, 69, 125, 146, 161]	-
↑ Total welfare	↔ Grid constraints	Flexibility	-	[50, 59, 112, 113]	-
↑ Total welfare	↓ Electricity cost	Electricity	[24, 72, 82]	-	-
↑ Total welfare	↓ Electricity cost	Flexibility	[102]	-	[102]
↑ Total welfare	↓ Grid imbalance	Electricity	[36, 54, 100]	[117, 145]	-
↑ Total welfare	↔ User preferences	Electricity	[42]	-	-
↑ Total welfare	↔ User preferences	Flexibility	[85]	-	-
↑ Total welfare	↓ Consumption	Electricity	[150]	-	-
↑ Total welfare	↓ Electricity loss	Electricity	[31]	-	-
↑ Total welfare	↓ CO2 emissions	Electricity	[137]	-	-
↑ Total welfare	↑ RES use	Electricity	[32]	-	-
↑ Total welfare	Fair cost distribution	Electricity	[106]	-	-
↑ Total welfare	↑ Self-consumption	Electricity	[55]	-	-
↑ Profit	None	Electricity	[26, 35, 80, 122]	[48, 66, 94, 120]	[26]
↑ Profit	None	Flexibility	[123]	[65]	-
↑ Profit	↔ Grid constraints	Electricity	[40, 126]	-	-
↑ Profit	↔ Grid constraints	Flexibility	-	[62]	-
↑ Profit	↑ RES use	Electricity	[74]	[116]	-
↑ Profit	↓ Grid imbalance	Electricity	-	[110]	-
↓ Cost	None	Electricity	[71, 83, 91, 92, 138, 141, 156, 159]	[67, 148, 158, 162]	-
↓ Cost	None	Flexibility	-	[78, 109]	-
↓ Cost	↔ Grid constraints	Electricity	[43]	[64, 104]	-
↓ Cost	↔ User preferences	Electricity	[96]	-	-
↓ Cost	↔ User preferences	Flexibility	-	[63, 68]	-
↓ Cost	↓ Grid imbalance	Flexibility	[90]	[103]	-
↓ Cost	↑ Total welfare	Electricity	[30]	-	-
↓ Cost	↓ Electricity cost	Electricity	-	[143]	-
↓ Cost	↑ Self-consumption	Electricity	-	-	[34]
↓ Cost	↑ Return on investment	Electricity	[133]	-	-
↓ Electricity cost	None	Electricity	[124]	[144]	-
↓ Electricity cost	↑ Total welfare	Electricity	[93]	-	-
↓ Electricity cost	↑ Total welfare	Flexibility	[128]	[86]	-
↓ Electricity cost	↔ Grid constraints	Electricity	[84]	-	-
↓ Electricity cost	↓ Cost	Flexibility	[53]	-	-
↓ Electricity cost	Fair cost distribution	Flexibility	[142]	-	-
↓ Grid imbalance	None	Electricity	[164]	[147]	-
↓ Grid imbalance	None	Flexibility	-	[46, 149]	-
↓ Grid imbalance	↑ Total welfare	Electricity	[73, 121]	[45]	-
↓ Grid imbalance	↑ Total welfare	Flexibility	-	[47, 49]	-
↓ Grid imbalance	↓ Electricity cost	Electricity	-	[160]	-
↓ Grid imbalance	↓ Cost	Electricity	[131]	-	-
↓ Grid imbalance	↓ Cost	Flexibility	[29]	[88]	-
↓ Grid imbalance	↔ Grid constraints	Flexibility	[41]	[79]	-
↓ Grid imbalance	↑ Profit	Electricity	[75]	-	-
↓ Grid imbalance	↑ Profit	Flexibility	-	[105]	-
↓ Grid imbalance	↓ Grid dependence	Flexibility	-	[107]	-
↔ Grid constraints	↑ Total welfare	Electricity	[132]	[61]	-
↔ Grid constraints	↓ Cost	Flexibility	-	[87]	-
↑ Flexible demand use	↑ Total welfare	Flexibility	[33, 101]	-	-
↑ Self-consumption	None	Flexibility	-	-	[77]
↑ Self-consumption	↓ Cost	Flexibility	-	[99]	-
↓ Grid dependence	↑ Self-consumption	Electricity	[163]	-	-
↓ Peak load	↔ Grid constraints	Flexibility	-	[76]	-
↑ Ancillary services	↔ Grid constraints	Electricity	-	[115]	-
↔ User preferences	None	Electricity	-	-	[44]
↑ DER use	↑ Profit	Electricity	-	[57]	-

Legend: ↑ Increase; ↓ Reduce; ↔ Respect

Table A.7: Market participants

Participant type	P2P	TE	CSC
Pure generators			
Entities which only generate energy	[32, 41–43, 74, 83, 89, 101, 114, 121–123, 132, 133, 137, 138, 141]	[45, 46, 50, 51, 57, 61, 64, 66, 67, 86, 88, 94, 103–105, 107, 108, 110, 113, 114, 116–120, 125, 136, 145–147]	[44, 77]
Pure consumers			
Entities which only consume energy	[21, 24, 25, 29, 31–33, 35, 36, 41–43, 53, 56, 70, 71, 74, 75, 80, 82, 83, 89, 92, 93, 101, 102, 111, 114, 121, 122, 124, 126, 131, 133, 134, 137–139, 150, 152, 163, 164]	[21, 45, 46, 48, 49, 59–63, 66, 69, 86–88, 94, 103–105, 107–109, 113, 114, 116, 117, 119, 120, 125, 136, 144–148, 157, 160, 162]	[44, 77, 102]
Prosumers			
Entities which consume and generate energy	[6, 21, 24–33, 35–37, 39–43, 52–56, 70–75, 80, 82–85, 90–93, 96, 100, 102, 106, 111, 114, 121–124, 126, 128, 132–135, 137, 139–142, 150–156, 159, 163, 164, 175]	[21, 37, 45, 47, 48, 50, 51, 57, 59, 62, 65, 67, 68, 78, 81, 86–88, 99, 104, 105, 107, 112, 114, 115, 117, 120, 125, 136, 144, 145, 147–149, 157, 158, 160–162]	[26, 34, 44, 77, 102]
Aggregator			
Entity that act on behalf of a group of smaller market participants	[21, 33, 36, 39–42, 73, 74, 85, 89, 93, 111, 114, 123, 124, 128, 132, 139, 151]	[21, 47, 49–51, 62, 63, 68, 76, 78, 79, 87, 94, 104, 105, 107, 108, 114, 116, 119, 120, 144–149, 160, 162]	-
Retailer			
Entity that connects to other large markets	[24, 26, 35, 36, 42, 52, 53, 55, 72, 80, 85, 101, 114, 124, 128, 131, 139, 152, 153, 159]	[48–51, 57, 60, 94, 104, 105, 112, 114, 146, 160, 162]	[26, 44]
Central market operator			
Single agent which runs the market or the platform	[26, 27, 30–33, 35, 37, 41, 43, 53, 55, 56, 72, 73, 80, 83, 92, 96, 101, 102, 106, 111, 114, 123, 138, 140, 142, 150–152, 155, 159, 163, 175]	[37, 45, 46, 48, 50, 51, 57, 59, 61, 65–68, 76, 78, 81, 86, 88, 99, 105, 107, 113, 114, 116, 119, 125, 145, 146, 148, 149, 157, 158, 160, 162]	[26, 34, 44, 102]
Grid operator			
Entity that operates the electricity network and interacts with the market	[21, 32, 37, 41, 71, 72, 83–85, 93, 100–102, 111, 114, 123, 131, 133, 141, 151, 152, 175]	[21, 37, 45, 47, 49–51, 58, 59, 61, 62, 64, 65, 67, 69, 76, 78, 79, 81, 86, 87, 94, 99, 103, 104, 110, 112–115, 118, 119, 136, 145–147, 158, 160, 162]	[102]

Table A.8: Controllable and non-controllable assets of P2P, CSC and TE markets

Type of control	Type of assets	P2P	TE	CSC
Controllable assets	Generation Storage Load	-	[45, 49, 57, 117, 118, 145]	-
	Storage Load EV	[91, 102]	[50, 59, 68, 79, 107]	[102]
	Generation Storage	[114, 133]	[67, 110, 114, 125, 143]	-
	Storage Load	[21, 29, 33, 39, 43, 90, 106, 121, 128, 131]	[21, 87, 99, 104, 105, 108, 113, 120, 148, 158]	[77]
	Load EV	[101, 152]	[103, 109]	-
	Generation Load	[132]	[78, 88, 116, 119]	[44]
	Storage EV	[54, 135]	[47]	[34]
	Generation	[42, 141, 153]	[61, 64, 66, 86, 94, 112]	-
	Storage	[26–28, 53, 55, 72, 74, 82, 85, 92, 93, 96, 126, 134, 150, 155, 159, 163]	[115, 144, 147]	[26]
	Load	[6, 36, 52, 111, 124, 138, 142]	[46, 51, 69, 160]	-
	EV	[73, 151]	[60, 62, 63, 76, 149]	-
	Other	[40, 41, 83, 137, 175]	[136]	-
Non-controllable assets	PV Other	[29, 56, 114, 121, 134, 159]	[57, 61, 64, 81, 104, 113, 114, 117]	[77]
	PV	[6, 21, 24–28, 30, 31, 33, 35, 36, 53, 55, 71, 72, 80, 82, 85, 90–92, 96, 100–102, 106, 123, 126, 128, 132, 133, 135, 139, 150, 152–154, 163]	[21, 47, 50, 59, 67, 68, 88, 108, 115, 118, 144, 147, 148, 157, 158, 161]	[26, 44, 102]
	Other	[43, 52, 74]	[45, 60, 94, 105, 120, 125, 143, 149]	-

Table A.9: Types of grid model

Grid model	P2P			TE		
	Grid Constraints	Power Loss	Other	Grid Constraints	Power Loss	Other
IEEE 13 bus	[52, 75, 111]	-	-	[50, 76, 92, 125]	[50, 76, 125]	-
IEEE 14 bus	[56]	-	[35]	-	-	-
IEEE 24 bus	-	-	-	[105]	[105]	-
IEEE 30 bus	[33]	-	-	[61]	[61]	-
IEEE 33 bus	[128]	-	-	[112, 160]	[112, 160]	-
IEEE 37 bus	-	-	-	[104, 107, 109, 161]	[109, 161]	-
IEEE 39 bus	[84]	[84]	-	-	-	-
IEEE 55 bus*	[96, 132, 154]	[96, 154]	-	[47]	[47]	-
IEEE 69 bus	-	-	-	[87, 113]	[87, 113]	-
IEEE 118 bus	-	-	-	[105]	[105]	-
IEEE 123 bus	[28, 33, 128]	-	-	[64, 76, 160, 161]	[64, 76, 160, 161]	-
ISO 5-bus**	-	-	-	[51]	[51]	-
CIGRE 6 bus***	[6]	-	-	-	-	-
CIGRE 15 bus*	[41]	-	-	-	-	-
SCE 56 bus**	[175]	-	-	-	-	-
WECC 240 node***	-	-	-	[78]	[78]	-
PJM 5 bus	-	-	-	[103, 104]	[103]	-
Real Network	[126, 140]	[126, 140]	[31]	[62]	-	[162]
Simulation Case	[42, 134]	[42, 134]	-	[81, 86, 104, 115, 120]	[115, 120]	[104, 110, 119, 144]

*:European Low Voltage Test Feeder, ** ISO 5-bus transmission test system, ***CIGRE Benchmark LV Microgrid network, *CIGRE 15bus European benchmark,**Southern California Edison (SCE) 56-bus test feeder,***CAISO- 240 node WECC

1374 Appendix B. Data Extraction Table Code Book

1375 This study developed a data extraction table which was used to consistently extract data from
1376 each paper in the review. The data extraction table is based on *The Business Ecosystem Architecture*
1377 *Modelling* (TEAM) framework [23]. For more details on the data extraction process see Section 2.3.
1378 Details about how to access the full data extraction table are available in Section 6. Table B.10
1379 contains the code book for the data extraction table. The code book contains a list of all data
1380 extraction fields, the type of data required and a description of the data required.

Data Ex- traction Field	Data Type	Description
Research question	Free text	Why was this paper written (i.e. what question is this paper addressing)?
Future work	Free text	What is noted as still to be researched/addressed as continuation/building on this work?
Category of definition: P2P or TE or CSC	Choice of: P2P, TE, CSC	Please choose the category which best fits the paper given the definitions.
Definitions	Free text	How does the paper define the respective P2P / CSC / TE market? (Please copy/paste the definition verbatim from the text)

Assumptions	Forecast uncertainty	Boolean: yes/no	Does the agent know what his/her supply and demand will be for the trading period (where agent can be household, or a market if trade is between markets, or microgrids, etc.).
	Rationality	Boolean: yes/no	Are the agents expected to be rational (e.g. act in accordance with a utility function, know/calculate precisely what their benefits are, etc.)? Note, models which are based on empirical data may not require agent rationality.
	Perfect information	Boolean: yes/no	Do the agents know and share with each other all information about the market? (e.g, how much energy is generated, traded, who the agents are, etc.)
	Transaction charges	Boolean: yes/no	The financial charges to be paid by the agents to undertake each transactions.
	Supplier of last resort	Boolean: yes/no	Is the market grid-connected and so can the agents fall back to the grid if the supply from peers is short/used up?
	Type of tariffs	Choice: static, dynamic, time of use	Which kind of tariff does the supplier (of last resort) apply to the market? E.g. static, dynamic, time of use, or something else?
	Grid constraints	Boolean: yes/no	Does the model account for grid constraints?
	Power losses	Boolean: yes/no	Does the model account for power losses?
	Type of grid model	Free text	Does the model use a specific model of grid, e.g. IEEE-33 bus grid?
	Origin of data	Free text	Where does load and generation data come from?
Market Participants	Pure generators	Boolean: yes/no	Does the modelled market include entities which only generate energy?
	Pure consumers	Boolean: yes/no	Does the modelled market include entities which only consume energy?
	Prosumers	Boolean: yes/no	Does the modelled market include entities which consume and generate energy?
	Aggregator	Boolean: yes/no	Does the modelled market include an entity which acts on behalf of a group of smaller market participants?
	Retailer	Boolean: yes/no	Does the modelled market include an entity which connects to another large market?
	Central market operator	Boolean: yes/no	Does the modelled market include a single agent which runs either the market or the platform, e.g. this could be an entity which is only a market operator, it could be a function carried out by an aggregator or DSO, or it could be a transaction server. However it does not include many entities sharing this task in a decentralised manner.
	Grid operator	Boolean: yes/no	Does the modelled market include a grid operator that interacts with the market?
Strategic Layer	Customers	Free text	Agents being supplied with one of the commodities through the market.
	Internal competitors	Free text	Agents who participate in the market for one of the commodities being traded and engage in competitive behaviour.
	External competitors	Free text	Agents outside the market competing with the market for one of the commodities being traded in the market.
	Enablers	Free text	Entities who do not directly participate in the market but supply essential products or services to make the market work, e.g. blockchain miner, or ICT provider.

	Rule makers, associations	Free text	Entities who do not directly participate in the market but set market rules or constraints (e.g. thermal constraints).
	Core needs	Free text	Need in terms of main trade purpose.
	Secondary needs	Free text	Need in terms of (optional) secondary trade purpose.
	Commodity / attribute being traded	Free text	Commodity or attribute traded in the market (e.g. electricity, flexibility, reactive power, active power, renewable energy, battery capacity, etc.)
	Price formation mechanism	Free text	The system by which market prices are determined, e.g. single auction, double auction, merit ordering.
	Time scale	Free text	The time between the market being cleared and the product being delivered, e.g. 1 day, 1 hour, 15 minute.
	Settlement period	Free text	The duration of time over which the energy can be delivered.
	Test duration	Free text	The length of the experiment or simulation.
	Market size	Free text	The number nodes in the market.
	Controllable assets	Free text	Any equipment, generation, demand or storage, which can be controlled. e.g. batteries, appliances which can participate in demand response, CHP plants.
	Non-controllable assets	Free text	Any equipment, generation or demand, which cannot be controlled. e.g. solar panels, non-controllable loads.
	Coordination paradigms	Choice: individual optimisation, central optimisation, multiple optimisation	If there is a market optimisation taking place, does it take place on the individual agent level or is the market optimised centrally for the whole community?
	Strategic behaviour	Boolean: yes/no	Do agents adjust their strategy based on speculation or the expected behaviour of other agents?
	Switching costs	Boolean: not specified/specified	What costs are incurred by agents who want to switch into or out of the market?
	Value transfers	Free text	Movement of the commodity that has been purchased in the market.
Value Layer	Commercial transactions	Free text	All financial flows, including payments to e.g. blockchain miners, network operators, aggregators. Describe the flow of money between parties.
	Transaction dependencies	Free text	Which financial / commercial factors affect contract creation and which factors might prevent a contract being fulfilled. To whom do they apply and how?
	Settlement	Free text	How are different energy contracts settled.
	Fraud	Boolean: yes/no	Do market participants act against the market rules?
	Other market risks	Boolean: yes/no	Are there any other factors which might adversely affect the market, e.g. data loss, hardware failure, etc?
	Specific the other market risk	Free text	Describe the other market risk.

	Distribution of benefits, costs or risks	Free text	Any information in the paper about how benefits, costs or risks arising from the respective market participation/operation are distributed between participants.
Technology Layer	Semantics	Free text	What information is shared?
	Ontologies	Free text	Who is that information shared with?
	Privacy	Free text	Do agents specify any privacy preferences with regard to data sharing?
	Choreography	Free text	The order in which market functions occur.
	Physical dependencies	Free text	Are there any physical market constraints, e.g. thermal line limits, state of charge of batteries? To whom do they apply and how?
	Country link	Free text	Is the paper about a specific country?

Table B.10: Data extraction table code book