

A Review on Demand-side Tools in Electricity Market

R Sharifi¹

S.H Fathi²

V Vahidinasab³

reza.sharifi@aut.ac.ir

fathi@aut.ac.ir

v_vahidinasab@sbu.ac.ir

¹ Ph.D Student in Electrical Engineering Department, Amirkabir University of Technology, Tehran, Iran

² Professor in Electrical Engineering Department, Amirkabir University of Technology, Tehran, Iran

³ Assistant Professor in Electrical Engineering Department, Shahid Beheshti University, Tehran, Iran

Abstract

With the advent of restructuring in the electricity markets, the Supply-side quickly adapted to the new environment but the story in demand side was different. Demand side dealt with electric energy as a commodity available to the necessary extent. This caused the Supply-side to realize that the demand side would admit to purchase electric energy at any price and this resulted in the advent of bidding strategies in the Supply-Side, known as “hockey-stick bidding”. The most important result was transfer of the demand side assets to the Supply-side. After a while, the demand side was noticed self-sloppy condition, therefore looked for tools to deal with these threats. This subject is examined by this paper.

Keywords: Demand side, Supply-side, Demand side management (DSM), Bidding Strategy, Purchase Allocation

I. Introduction

Until a few decades ago, the government was responsible for the management and control of the electric power system and rarely was it owned by the private sector. This structure did not contain suitable efficiency due to exclusivity and thus ensuring that producers benefit; therefore, there remained no way but to eradicate this exclusive structure. Competitive electricity market requires the competitiveness of this market at the level of generation, transmission and distribution. In general, the electricity power industry, after privatization, was split into two parts:

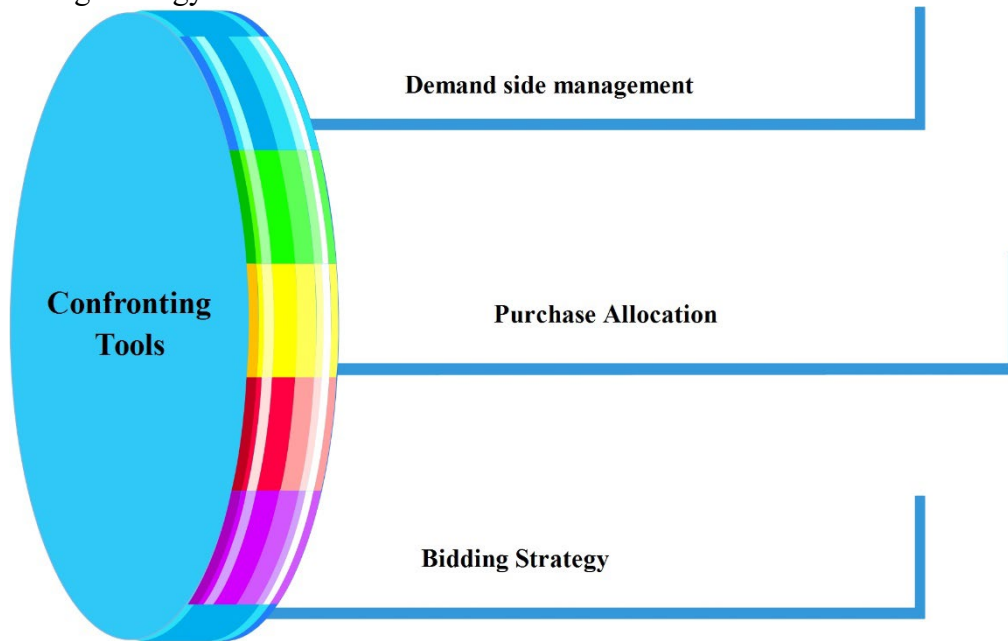
- The Wholesale Sector
- The Retail Sector

The wholesale sector is the generation companies, which generate electric energy in high volume and transmit it to load centers through high-voltage transmission lines. In the next step, retailer companies, on behalf of the Demand-side and, occasionally, Big Consumers, purchase their required energy independently [1]-[8].

In the Deregulated Electricity Market, until recently, in the wholesale sector, only the generation companies would seek to compete with each other to sell their electric energy to customers with the objective of increasing profit, yet the Demand-side had no function in this area. It means that the Demand-side dealt with electric energy as a commodity available to the extent necessary, which indicates its inflexibility. Overall, the Demand-side had not been adapted to the new environment. This incompatibility of the Demand-side caused the increasing greed in generation companies and soon it was realized that the Demand-side would yield to any price to purchase electric energy, resulting in the advent of bidding strategies in the Supply-Side, known as “hockey-stick bidding” [9].

36 Thus, the prodigious asset transfers of the Demand-side towards Supply-Side may be viewed as
37 the most important impact of restructuring until recently [10]. The primary reasons for this
38 incompatibility in demand-side were due to the lack of sufficient knowledge and confronting tools
39 to participate effectively in the electricity markets. Having gradually identified this issue, the
40 Demand-side looked for some confronting tools in order to avoid being placed in this situation.
41 In this paper, after reviewing the solutions and confronting tools that have been proposed so far to
42 reduce or avoid this issue, these tools are classified into three different categories as shown in
43 Figure (1):

- 44 • Demand Side Management (DSM) Programs
- 45 • Purchase Allocation
- 46 • Bidding Strategy



47
48 **Fig.1: Confronting tools of Demand-side**
49 After awareness from its lethargy in the initial years of restructuring and the ensuing problems,
50 demand-side managed to use these three tools to tackle the imposed problems. Using the DSM
51 programs, demand-side managed to amend load profiles as required to increase its profits, reduce
52 the risk of buying from a single producer by diversifying its sources, and create an optimal
53 bidding strategy to achieve higher profits.
54 This paper reviews and evaluates these tools that give the demand-side a leverage against supply-
55 side, and carefully examines the work that has been carried out in this field in order to identify
56 the challenges ahead and provide a clear image and framework for future studies.

57 **II. Demand Side Management**

58 As mentioned, the Demand-side, which realized the avarice of the Supply-Side, sought a solution
59 in order to escape from this situation. One of the first strategies of the Demand-side was to adjust
60 its consumption levels according to the price levels, leading to the advent of an extensive
61 discussion, called the Demand Side Management (DSM), in the electricity markets. In most cases,
62 the concept of DSM implies a Supply/Demand-side relationship that presents mutually beneficial
63 results.

64 The implementation of DSM plans contains numerous profits for a great number of beneficiaries
 65 in the deregulated distribution system; therefore, this expansion and all-encompassing profitability
 66 of such plans causes this option to constantly be considered as one of the substantial research cases
 67 and many actors who are somehow involved in the Demand-side want to investigate different
 68 aspects of these plans on their profit and loss.

69 One of the first papers in the field of DSM is reference [11]. In this article, a framework is provided
 70 for the responsibility of a simple consumer to Spot Prices. In reference [12], some aspects of the
 71 electricity market, from the perspective of the Demand-side and tools needed by the consumers
 72 and retailers to more actively and effectively participate in electricity markets, are introduced and
 73 discussed. According to this reference, if consumers have the tools of forecasting prices and also
 74 energy storage, they can alter their consumption pattern and transfer their consumption from times
 75 of high energy prices to other times. Therefore, in this reference, a decision-making framework,
 76 suitable for consumers and significant in terms of the Demand-side, is presented.

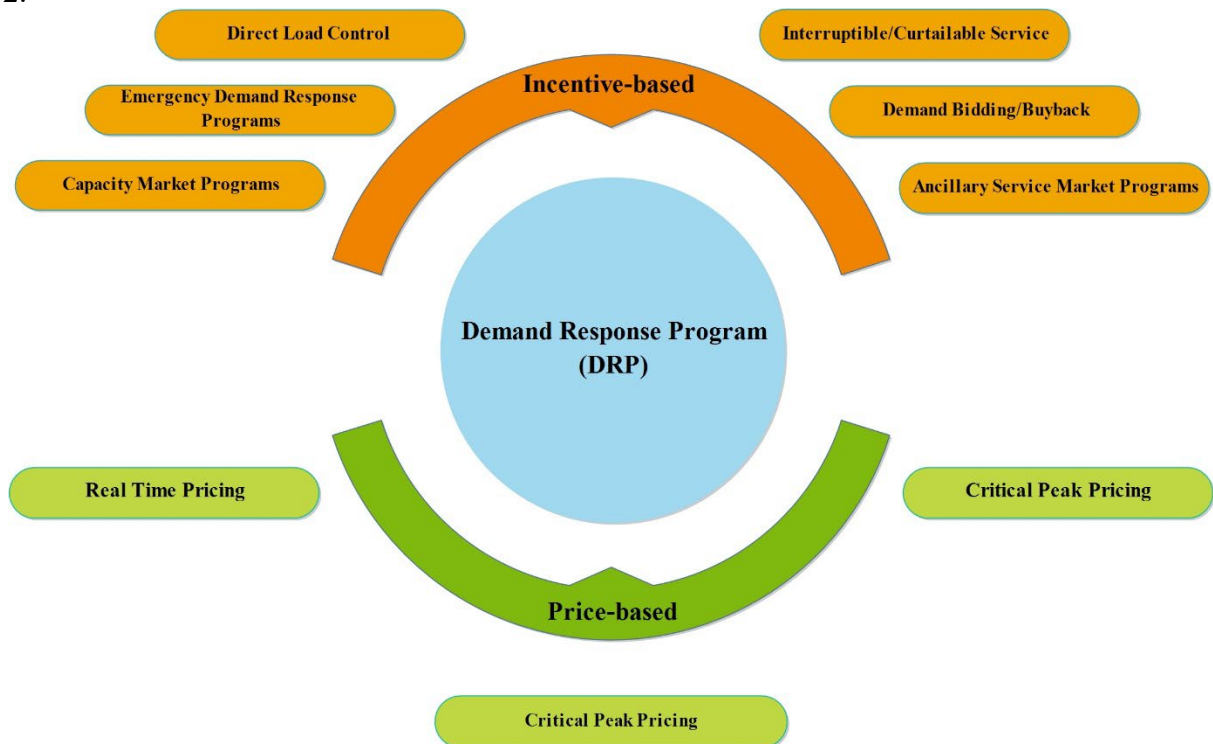
77 In order for consumers to be able to use the benefits of cheap electric energy at times of low energy
 78 price, there must be an interaction between consumers and retailer. In reference [13], a general
 79 model of interaction is proposed between sellers and consumers in the electricity market.

80 DSM programs have been divided into techniques [14]:

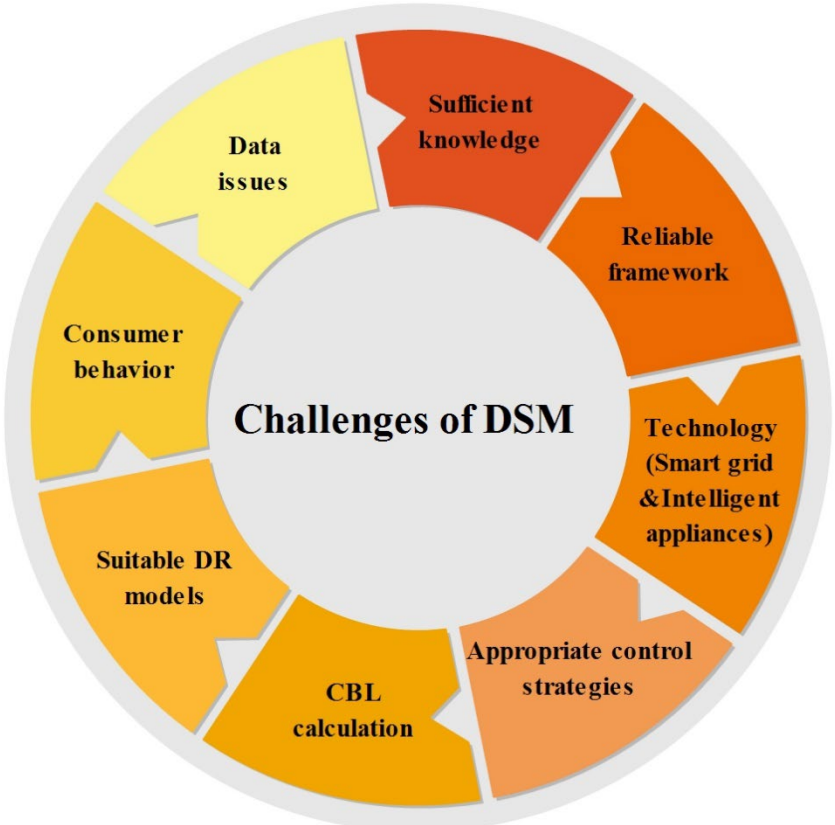
81 (1) energy efficiency improvement programs: reduce the amount of energy required. For instance,
 82 double glazed windows, insulation, sealing, installation of light dimmers to control the power
 83 consumption, solar water heating systems, etc. [15].

84 (2) Demand Response (DR) Program: an optional temporary adjustment of consumption as a
 85 reaction to a price signal or reliability conditions [16]. In [17], it has been shown that increasing
 86 the capability of demand-side to react to electricity prices decreases the total costs. also, it can
 87 alleviate the rate volatility of prices during peak times.

88 DR programs are divided into two main categories and several subcategories, which are shown in
 89 Figure 2.



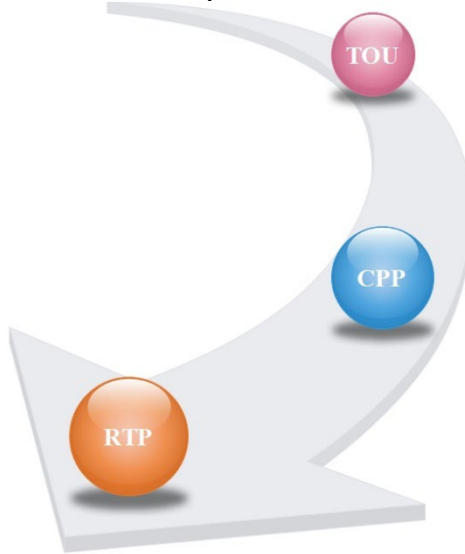
91 Fig.2: Categories of Demand Response Programs [18]
 92 In reference [19], the benefits and challenges of DSM plans are discussed in the context of
 93 England’s Electric System. In reference [20], it is demonstrated that although DSM programs have
 94 myriad benefits, they contain challenges as well which must be overcome. Of the most significant
 95 challenges pointed out in this reference is the creation of appropriate control strategies and reliable
 96 framework in such a way as to optimally utilize the generated sources of DSM plans.
 97 Consequently, the biggest problem in the implementation of DSM plans is to establish
 98 communication between Supply-Side and Demand-side. With the advent of the Smart Grid, this
 99 problem is slightly solved. Smart Grids are known as a controlled electric network, which can
 100 transmit electric energy from the producer to the consumers in a clever way [21].
 101 Reference [22] have also examined the obstacles and challenges ahead of implementation of DSM
 102 programs, and has reported the most important challenges in this regard to be as follows: (1)
 103 Consumer Behavior: the uncertainty in regard to how consumers react to these programs. (2) Data
 104 issue: the lack of adequate data because of the lack of experience in this field and the novelty of
 105 these programs. (3) Customer Baseline (CBL) Calculation: CBL calculation is one of the most
 106 important step for assessing the success of DR programs. CBL is the pattern of consumption to be
 107 expected in the absence of DR programs, and its accurate calculation is a major achievement in
 108 the implementation of DR programs. In [23], it has been shown that inaccurate calculation of CBL
 109 will lead to lower customer participation and the mechanism of this effect has been explained.
 110 Some of the most important challenges in the implementation of DSM programs are illustrated in
 111 Figure 3.



112 Fig.3: Most important challenges facing the DSM programs [20]-[22]
 113 Among the methods available in price-based DR programs, real-time pricing (RTP) is particularly
 114 popular among market economists [24]. In references [25], benefits of implementing RTP plan in
 115

116 an electricity market are introduced. In reference [26], By using simple simulations with real
117 parameters, it has been demonstrated that the amount of profit gained from the implementation of
118 RTP is considerable, even at times when the demand response is low compared with electricity
119 price changes.

120 Figure 4 shows the consumer risk/ reward in different price-based DR programs. As shown at the
121 left, with TOU rates offering the lowest risk compared to a RTP but also the lowest reward [27].



Risk / Reward

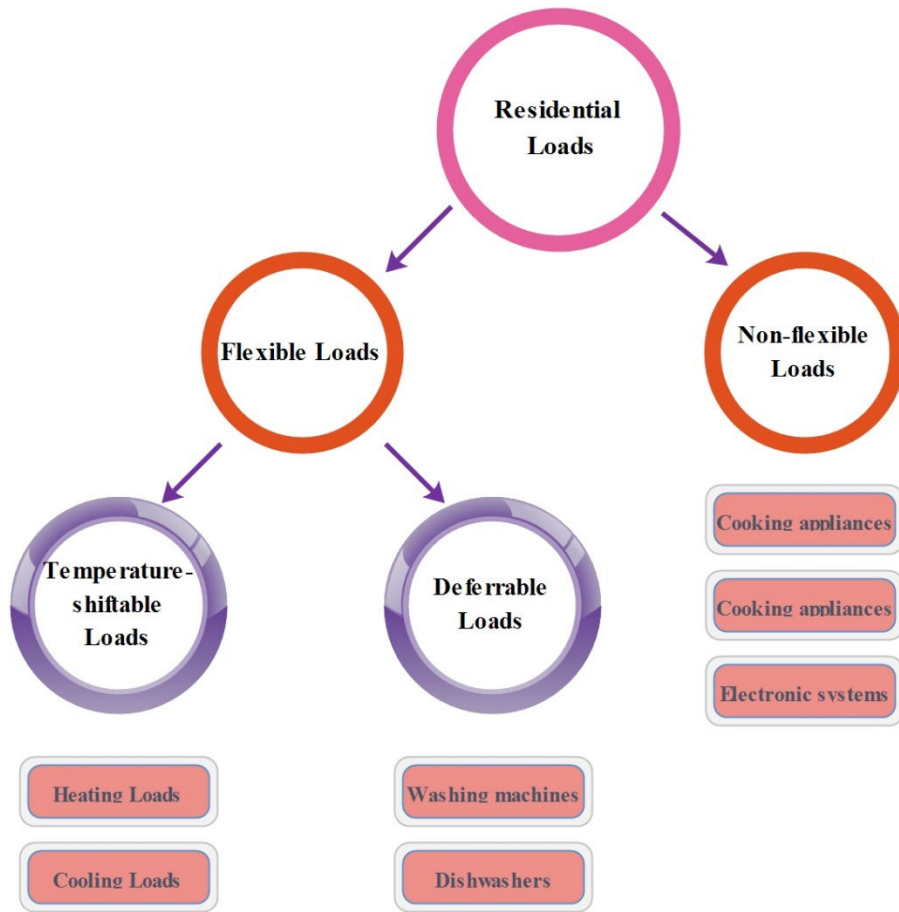
Fig.4: Consumer risk / reward in different electricity pricing methods

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123
124 Meanwhile, the growing tendency toward the use of renewable energy sources has led to problem
125 such as uncertainty in power source [28]. The renewable resources have lower reliability and
126 controllability than conventional ones, which make the networks containing such resources more
127 complex and more difficult to operate. These problems can be tackled by several methods, such as
128 predicting a suitable reserve in the conventional power plants to support renewable resources,
129 providing connections to nearby alternative grids, and implementation and use of DSM programs.
130 In [29], it has been shown that the use of DSM methods is by far the most efficient and cost-
131 effective approach among mentioned solutions. In [30], after examining the uncertainties in the
132 wind sources as well in demand, a robust optimization approach has been employed to develop a
133 new framework for handling both types of uncertainty and their portrayal over uncertainty sets.
134 Although DSM programs can effectively result in the reduction of electricity generation prices and
135 bill of customers, still, for networks with several retailers and consumers, each of whom thinking
136 about maximizing their own profit, this is an open and unresolved issue. In reference [31], this
137 issue has been evaluated and, by offering a method based on the Game Theory between retailers
138 and consumers, attempted to maximize the profit of each of these actors.
139 In reference [32], a Bi-level Stochastic Programming between retailer and consumers has been
140 presented. At Upper Level, the price-taker retailer makes decisions based on purchasing energy
141 from the market and then selling it to the customer with the purpose of increasing its profit. In this
142 reference, the retailers consider three methods of RTP, TOU and Flat Rate in order to sell energy
143 to the customers. And at Lower Level, the customers alter their consumption pattern according to
144 the offered prices with the purpose of reducing energy purchase prices. The consequent results
145 indicate the priority of RTP to the alternate methods.

146 There are also other important issues in regard to DSM programs that mostly pertain to industrial
147 and commercial sectors. Implementation of DSM programs in the industrial sector eliminates the
148 need for expensive energy storage, and given the size of demand of this sector, they can be of great
149 use for reducing the price of electricity. In [33], the applications of DR programs in the industrial
150 sector have been thoroughly studied.

151 The biggest consumer of electric power is the Residential Sector; however, due to its numerous
152 complexities, there are far fewer works in regard to applications of DSM programs in the
153 residential sector than for industrial and commercial sectors. In [34], the challenges ahead of
154 implementation of DR programs in the residential sector has been discussed.

155 In [35], the role of DR programs in the residential sector as envisioned in new markets have been
156 investigated. As shown in Figure 5, in the residential sector, demand loads are divided into two
157 categories of flexible loads and non-flexible loads. Non-flexible loads consist of for example
158 lighting loads, which are bound to happen at certain hours and cannot be shifted, but flexible loads
159 can be pushed from one hour to another.



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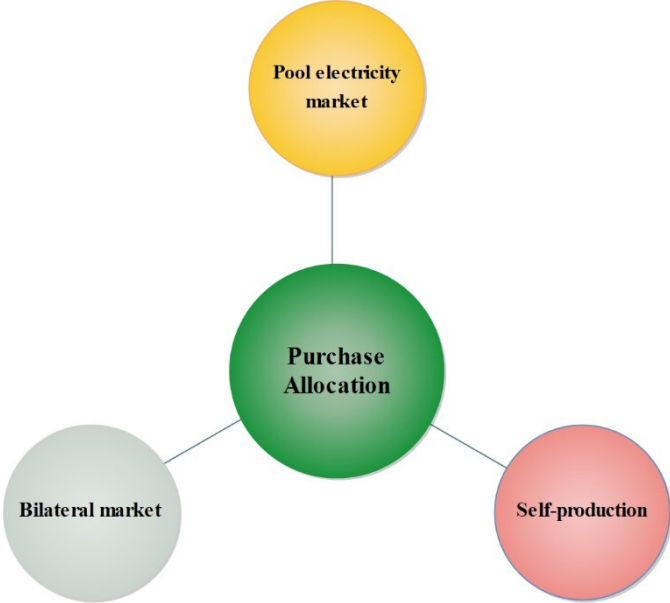
161 Fig.5: electricity loads in the residential sector [36]

162 One of the challenges facing the DSM program and especially RTP program in the residential
163 sector is how to create a mechanism in which flexible loads be responsive to changes in power
164 prices of different hours, and although great strides have been made in the provision of equipment
165 and facilities required for such mechanisms, the actual use of these mechanisms is still at an early
166 stage. Authors of [37] have provided a new thermostat design that can respond to price signals,
167 and can be used to make the energy intensive appliances such as heating and cooling systems

168 responsive. In [38], the benefits of a RTP program in the residential sector in the presence of such
169 price-responsive appliances have been discussed, and the manner in which consumption profile
170 shifts to adapt to new prices and minimize the electricity bill have been demonstrated.
171 Meanwhile, the advent and development of new electrical loads with high energy storage potential
172 such as plug-in electric vehicles have led to new opportunities for the development of DSM
173 programs for the residential sector [39]-[45].
174 One of the most essential problems in the Residential Sector is the presence of some customers
175 who are not sensitive about the price changes [46]. In reference [47], the issue of how flexibility
176 of electricity demand is effective on determining electricity price in the market has been discussed.
177 Moreover, various responses of different consumers to electricity price changes have been
178 modeled.
179 In addition to DSM discussion, the Demand-side, in order to further reduce electric power purchase
180 prices, expanded its aggressive mode and another new discussion named “Purchase Allocation”
181 was shaped. In this discussion, retailers and big consumers seek to resolve the problem of how to
182 procure their needs from various sources of electric energy supply in order to increase their profit
183 and decrease risk. This issue is addressed in the following sections of the paper.

184 III. Purchase Allocation

185 As shown in Figure 6, the retailer can supply its needs from various sources including bilateral
186 markets, self-productions and pool electricity market [48].



187 Fig.6: Classification of Source Purchase Allocation

188 The retailer must decide either to use these sources or not, and determine the share of each of these
189 sources. In consequence, the evaluation of ways of supplying electricity required by retailers from
190 source basket is one of the most substantial measures which must be conducted by a retailer in the
191 competitive market [49]. Performing bilateral contracts reduces the fluctuation risk of pool
192 electricity and if consumers have their self-productions as well, this risk will contain a far greater
193 reduction. Thus, consumers encounter an exchange between bilateral markets, pool and their self-
194 productions. Since prices have numerous uncertainties in different markets based on different
195

196 conditions, the purchase allocation of each of these markets is an important problem and one of
197 the most substantial difficulties faced by retailers and big consumers.

198 Since some of the most essential factors in the pool system based market, such as the power
199 demand and price, are ambiguous and uncertain, a stochastic programming problem is faced with.
200 In reference [50], the amount of energy purchase allocation of a big consumer from each electric
201 energy supply has been estimated, while the consumer has its own generating source as well.

202 Reference [51] has addressed the problem of optimal purchase for electricity markets and pricing
203 method for the intended demand. In this reference, price fluctuations have been considered in the
204 problem of purchase allocation and the nature of Successive changes has been proposed by
205 stochastic models.

206 In [52], a two-stage problem concerning the optimal size of electricity purchase from bilateral
207 markets and pool electricity market with the objective of minimizing the risk and cost of purchase
208 has been examined. The results of the solution method proposed in this article has shown partial
209 success in achieving this objective.

210 Authors of [53] have developed a hybrid approach for optimal purchase of electricity from all
211 available sources based on binary imperialist competitive algorithm (BICA) and binary particle
212 swarm optimization (BPSO). According to the reported results, this method has a good efficiency
213 in the optimal allocation of purchases. In [54], the mathematical models and mixed-integer
214 stochastic programming have been used to develop a bidding strategy for a retailer purchasing
215 electricity from several sources. In [55], a stochastic model for the purchase of electricity from
216 several sources has been developed. The model provided in this article also reflects the effect of
217 DR program and energy storage systems on the purchase price reduction. In [56], a two-stage
218 decision-making model for purchase from reserve market has been developed, and it has been
219 demonstrated that this model can reduce the cost of purchase from this market.

220 In a competitive electric market, a retailer encounters two major issues. On the one hand, electric
221 energy must be supplied with a variable price from the wholesale market or bilateral contracts
222 (which usually consist of a rate higher than the average price). On the other hand, it faces
223 consumers who have a vague amount of demand and may also have the capability to change their
224 retailer in case of dissatisfaction from the offered prices. In reference [57], this problem has been
225 evaluated and, by providing a suitable stochastic framework, decisions have been adopted on
226 electric energy buying and selling method so as to both maximize the resultant profit and lead to
227 consumer satisfaction as well.

228 In reference [58], a decision-making framework is proposed for a retailer in an average-term based
229 on a Bi-level Stochastic Programming. These decisions include determining electricity sales price
230 to consumers according to TOU and also determining a plan to allocate purchase from various
231 markets to supply their demand with the objective of risk reduction. In this reference, consumer
232 response to the prices of retailers and also competition of retailers have been considered. In
233 reference [59], a method has been introduced based on Stochastic Programming to optimally solve
234 the problem of electricity purchase for a big consumer in the electricity market. Supply sources
235 include bilateral contracts, self-productions and electricity market based on pool system.

236 Reference [60] provides a Bi-Level Programming to solve the problem of purchase allocation. The
237 price-taker retailer makes decisions with the purpose of maximizing its profit based on the method
238 of the company in Futures markets and Day-Ahead Markets and also the pricing method to
239 consumers. In this model, numerous uncertain variables have been considered such as Day-Ahead
240 Market prices, consumer demand and prices of other retailer competitors. Here, consumer response
241 to retail price and competition among retailers both have been taken into account in the proposed

242 model. In reference [61], contractual policies relevant to energy purchase of an industrial consumer
243 under the electricity market are investigated. In reference [62], industrial consumer strategies for
244 electric energy purchase in the electricity market are examined.

245 One other subject, which appeared in the field of Demand-side, was the problem of pricing
246 strategies. In this problem, price-maker retailers and occasionally big consumers seek to extract
247 their Bidding Curves in markets based on pool system with the purpose of enhancing their profit,
248 dealing with the greed of production companies and manipulating market prices to their advantage
249 with the help of bidding strategies.

250 This subject is addressed in the next section of the article.

251 **IV. Bidding Strategy**

252 As was seen, in the markets based on pool system, similar to the supply side, the Demand-side
253 also introduces its proposed prices to the pool. According to the microeconomic theory, the best
254 bidding method for each participator in the market with complete competition, is bidding based on
255 marginal costs. However, the presence of some participators, who are capable of affecting market
256 prices, has usually led the electricity markets not to be the type of markets with complete
257 competition. Normally, the price offered by these participators is more than the competitive level
258 or marginal costs. This behavior, the so-called “bidding strategy”, is caused by the power market
259 of this type of participators [63].

260 In the economics texts, the power market is viewed as one of the market parameters, effective on
261 the commodity price in the market and often for making a profit more than the conditions of perfect
262 competition. Consequently, from this angle, we can immediately deduce the conclusion that the
263 power market is not limited to the producer power alone, but in some conditions, in the Demand-
264 side, some retailers have the power market [64].

265 It must be noted that the power market is a natural phenomenon based on the rational behavior of
266 market participants, since it is assumed that the market participants are constantly expanding their
267 benefits. Nevertheless, the main point is that every market must have a specific model according
268 to different conditions and, as a result, every market is a designer and creator. It is the duty of the
269 designer to provide the necessary steps in order to prevent creation of this phenomenon. Thus, the
270 need for assessment of removing such cases in deregulated distribution system and price control
271 seems an essential matter [64]. However, despite all these considerations, electricity markets in
272 the whole world still contain some degrees of this power market. In reference [65], a set of
273 indicators is presented for the measurement of the power market.

274 In general, participants in the market are divided into two categories based on the power market:

- 275 • Price Maker
- 276 • Price Taker

277 The first category refers to the participants who affect market prices, namely have the power
278 market, whereas the second category has no effect on the prices. Thus, in fact, the bidding method
279 of a price-taker participant in the market is a Bidding Problem yet this very problem is a bidding
280 strategy for a price-maker participant [66].

281 The number of articles presented in the field of bidding strategies in the supply side are numerous
282 and are not comparable with the Demand-side. However, the rate of expansion of papers in this
283 context in the Demand-side, especially in the last few years, indicates the increased interest of
284 researchers in this subject.

285 According to the economic logic of markets, the suitable economic price at which social welfare
286 is maximum is equal to the Market Clearing Price of electric energy wholesale. In this price, social

287 welfare is the highest. Accurate bidding for the retailers is performed based on costs, customers
288 and competitors. Whenever each of these variables changes, the best price might also change.
289 Therefore, to adopt optimal bidding strategy, it is necessary that the retailer uses an efficient
290 method for bidding in the wholesale market based on different factors. For this purpose, the retailer
291 must understand different bidding methods, their traits, advantages and disadvantages. Therefore,
292 it is necessary to conduct comprehensive researches in this regard [64]. In this context, the number
293 of performed studies is very few.

294 In reference [67], a framework is introduced for the comprehensive assessment of possible
295 scenarios to implement the bidding mechanism of the Demand-side in the electricity market and
296 evaluate the impact of bidding of the Demand-side in the total production costs, ultimate price and
297 allocated merits between producers and consumers. In reference [68], it has been demonstrated
298 how the bidding of the Demand-side can prevent price jumps in electricity markets. Furthermore,
299 in reference [69], the effects of bidding in markets based on pool system have been evaluated and
300 it has revealed that in case the production programming is based on minimizing the production
301 costs in everyday horizon, then the bidding of the Demand-side can lead to unexpected price jump
302 in the market.

303 Overall, there are two general methods for the development of bidding strategies:

- 304 • Game Theory Based Methods
- 305 • Forecasting and Estimation Based Methods

306 So far, various methods have been presented based on the Game Theory, the most common of
307 which include [70]:

- 308 • Bertrand Equilibrium(BE)
- 309 • Cournot Equilibrium(CE)
- 310 • Supply Function Equilibrium(SFE)
- 311 • Stackleberg Equilibrium(SE)
- 312 • ConjectorVariation (CV) and Conjector SFE Equilibrium

313 Each of these methods is employed in different competitive levels in the market and is of utmost
314 significance in the evaluation of markets in which the power market exists.

315 In a complex and severely competitive market, forecasting and assessing demand seems difficult.
316 Retailers can attempt bidding as much as possible according to different methods, after conducting
317 a proper prediction of load, price and or grid to participate in the market. Surely, this bidding
318 depends on numerous factors such as the required load, system conditions, climate conditions,
319 forecasted price, rate of acceptable risk for retailers and the like.

320 Retailers must have the opportunity and will power to adopt the most optimal bidding strategy in
321 the competitive market. To obtain this goal, after modeling the competitors and choosing the
322 bidding strategy, the retailer should have a simple, fast and accurate software in order to be able
323 to compete in the distribution market and perform the bidding according to conditions, limitations
324 and objectives, using the chosen method. To do this, retailers should transform their bidding
325 strategies with the help of mathematical algorithms into simple and efficient software's, which
326 requires research in this context and use of the experiences of Software experts [64].

327 In reference [71], a method is proposed for all participators in pool-based electricity markets to
328 construct their bidding strategies. In this reference, it is assumed that both producers and
329 purchasers offer a linear supply/demand function to the market operator. The market operator
330 performs market mechanism with the aim of maximizing the public welfare. Every producer and
331 purchaser chooses coefficients for their supply/demand function whose objective is the expansion

332 of their profit. These coefficients depend on predictions which are considered in relation to other
333 competitors.

334 In reference [72], a stochastic linear programming model has been proposed to make piecewise-
335 linear bidding curves to offer to the Nord Pool market. In this model, a price maker retailer is
336 introduced which has the duty of supplying electric power for a number of consumers. Moreover,
337 it is assumed that consumers are sensitive to price fluctuations. The purpose of the proposed model
338 is to minimize energy purchase prices from the day-ahead electricity market and the balancing
339 market.

340 In reference [73], consumers are classified into two groups of Price-Based and Must-Serve in
341 relation to price and, in continuance, the optimal bidding functions of each is deduced.

342 In reference [74], a model of electricity purchaser in Norway has been provided, which performs
343 bidding in the day-ahead market. The purchasers must arrange their purchase for an indecisive
344 demand. Any kind of difference between purchase and demand must be compensated for in the
345 secondary market after the day-ahead market. In this reference, a Cournot Equilibrium has been
346 considered and assumed that the purchaser has perfect knowledge of generator production
347 function; of course, this model is suitable for today's structures of pool-based electricity markets.

348 In reference [75], a method is proposed for the extraction of bidding strategies in the day-ahead
349 market for big consumers who supply their demand from the day-ahead market and adjustment
350 market. In this reference, a method has been used for the derivation of bidding curves based on
351 Information Gap Decision Theory (IGDT).

352 In reference [76], an algorithm is presented based on Monte Carlo to solve the coalition problem
353 of consumers equipped with the demand response plan. This coalition must determine the bidding
354 method in the day-ahead market in which they encounter uncertainties such as prices offered by
355 producers.

356 In reference [77], a method is presented to determine optimal bidding strategy for a retailer, which
357 provides electricity for its consumers. The purpose of this strategy is to reduce energy purchase
358 prices.

359 In reference [78], a Dynamic Programming method is proposed in order to make bidding curves
360 for the Demand-side with the aim of enhancing consumer profit and increasing market efficiency
361 for New Zealand. In reference [79], a Stochastic Complementarity Model is suggested to describe
362 the strategic behavior of a big consumer, the obtained results of which make the bidding curves.

363 In reference [80], a bidding strategy formulation of an electric utility in view of the risk is offered.
364 This utility includes the retail sector which is equipped with the demand response plan. The retail
365 sector is responsible for supplying the demanded electric power. The profit of this utility is
366 obtained by attending the day-ahead market and also selling electric energy to customers through
367 the retail sector. In this paper, IGDT theory has been applied to obtain robust scheduling method
368 against undesirable deviations from market prices. The consequent results refer to desirable effects
369 of the presented strategy and also higher profit by considering the demand response plan.

370 In [81], a similar work has been carried out for an industrial consumer equipped with cogeneration
371 facilities, and the obtained results have also confirmed the good performance of the proposed
372 method. In [82], a bidding strategy for the Demand-side in the presence of a smart grid has been
373 provided. In this strategy, which has been developed for a day-ahead market, consumers form a
374 consumption profile to maximize their profit depending on the hourly electricity prices and submit
375 it to the retailer one day before the date of consumption. The retailer then sums the submitted load
376 profiles to determine the Demand-side price curves. In [83], a model for optimal purchase by a
377 retailer from pool market have been developed using the bidding strategy and purchase allocation.

378 The presented method is based on a robust optimization approach, and its results provide the
379 retailer with sufficient data to obtain an optimum bidding strategy.
380 As can be seen, in recent years, several articles have attempted a combination of methods to
381 challenge the excessive demands of supply-side in electricity markets, and this is a direction that
382 researchers are expected to follow in the coming years.

383 V. Conclusion

384 As observed, with the advent of deregulated electricity markets, when the Demand-side stretched
385 and bended in compliance with this new environment, it was the supply side that ruled the market
386 and by offering the bidding strategies, the Demand-side asset was captured. This process continued
387 until recently when the Demand-side also sensed and sought a solution.

388 In the context of electricity markets based on the electricity pool, the main problem is the lower
389 flexibility of Demand-side compared to the supply side. Since most of generation companies can
390 change their rate of production, with less consequences, in order for affecting the prices, yet the
391 Demand-side has less flexibility in consumption reduction for the construction of bidding curves.
392 As was mentioned, one of the suitable strategies for the expansion of flexibility is to utilize DSM
393 programs. It is suggested that researchers surge their studies in the context of optimization
394 strategies towards the investigation and derivation of bidding curves by implementing DSM
395 discussion; i.e., consider a retailer whose some customers have enthusiasm to participate in DSM
396 programs. The response of customers leads to expansion of flexibility of retailer more than before.
397 In fact, the retailer becomes equipped and can be effective on the price, in favor of his benefit, by
398 considering suitable bidding strategies. In this context, a few works have been done, yet they are
399 not considerable and require more attempts.

400 On the other hand, retailers and big consumers can, for the reduction of their risk, cater their needs
401 from different sources of electricity such as bilateral markets, self-productions and electricity pool.
402 Using each of these sources has its own cons and pros which requires comprehensive studies in
403 this field.

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