

Wind Power Producers' Costs And Associated Market Regulations: The Source of Wind Power Producers' Market Power

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Abstract

In this paper, I build a two-stage-multiple-hour model to analyze wind power producers' (WPPs) ability to manipulate price (ATMP) and market-power strategies in a sequentially structured electricity market. By exploring WPPs' cost structures and the dynamics that prices respond to wind-energy generation, the analyses demonstrate that WPPs can have significant ATMPs even though their marginal fuel costs are zero. Actually, the bidding rule regulating wind energy, which is different from the bidding rule regulating other technologies, provide WPPs a high flexibility to exercise their market power. The bidding rule, which allows WPPs separately determine their hourly generation, provide WPPs a particular strategy of utilizing wind-energy fluctuation and conventional generators' ramp constraints. My empirical simulation, which is based on data from Texas in 2012, demonstrates that WPPs already have ability to manipulate price in more than 900 hours in 2012. In some hours, they can inflate price by around 25%.

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2 **1. Introduction**

3 The increase of wind-energy penetration brings concerns about wind power producers'
4 (WPPs) ability of to manipulate the price (ATMP). In Spain, annual wind-energy genera-
5 tion has already supplied over 20% of demands. In some other countries, the market share
6 of wind energy in some hours can exceed 50%. In addition to clarifying traditional concerns
7 about market power, studying the market-power issue of wind energy is critical in answer-
8 ing another important question in wind-energy market design: whether and when WPPs
9 should be allowed to aggregately make bids in a electricity market. Actually, in order to
10 control the growing wind-energy forecast error associated with increasing wind-energy pen-
11 etration, researchers are discussing the business models that allow WPPs to aggregately
12 bid into electricity markets[3, 16]. If I can demonstrate that WPPs do not have ATMP
13 even if they collude in some hours, they should be allowed to aggregately bid in order to

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14 reduce forecast errors in those hours. Therefore, it is necessary to systematically analyze
15 whether and when WPPs have unignorable ATMPs.

16 In this paper, I build a two-stage-multiple-hour model to examine WPPs ATMPs and
17 strategies to exercise their market power. The two stages, which include the day-ahead
18 (DA) stage and the real-time (RT) stage, simulate the sequential structure of electricity
19 systems [17, 16]. I consider multiple hours in the model because I would like to examine
20 WPPs' ATMPs and strategies when the WPPs and their GenCo competitors are regulated
21 by different rules and have different bidding processes. In addition to theoretical analyses,
22 I use data from the Electricity Reliable Commission of Texas (ERCOT) market in 2012 to
23 measure WPPs ATMP when they aggregately bid.

24 This paper contributes a systematical analyzing framework to study the market-power
25 issue of wind energy in literature. By using this analyzing framework, I provide conditions
26 determining when a WPP has significant ATMPs even though their marginal fuel costs
27 are zero. Because my framework considers multiple hours, I can examine how net-demand
28 fluctuations and GenCos' ramp constraints impact WPPs' ATMPs and optimal strategies.
29 To my best knowledge, the interaction between net-demand fluctuation with market-power
30 strategies has not been systematically studied. Furthermore, by considering both the
31 DA and RT markets, the framework built in this study can be also used to compare a
32 WPP's ATMPs when it participates into different sequential markets. The empirical case
33 demonstrates that this framework can be used in a real electricity market to monitor
34 market-power issue of wind energy.

35 In fact, market power is the core issue for electricity-market regulation and has been
36 deeply examined in a system mainly supplied by conventional electricity-generation compa-
37 nies(GenCos) [23, 24, 4, 10]. The electricity crisis in California demonstrated that market
38 regulations, such as bidding rules and dispatch protocols, play essential rules to determine
39 whether market players have mechanism to exercise their market power[6, 5, 11, 15, 22].

40 The California electricity crisis also inspired researchers to explore factors that determine
41 GenCos' ATMPs and willingness to exercise market power[21, 13]. [13] explained why a
42 GenCo can have ATMP in a electricity market supplied by multiple GenCos and devel-
43 oped a framework to measure GenCo's ATMP and willingness to exercise its market power.
44 While the penetration of wind energy have been growing, researchers begin to focus on the
45 interaction between wind-energy integration and market power [20, 12, 19, 2]. However,
46 most these studies either focus on GenCos' market-power strategy when they own WPPs.

47 However, the models used to analyze GenCo's ATMP cannot be directly used to ex-
48 amine WPP's ATMP because WPPs and GenCos face to different grid-access and bidding
49 rules when they exercise their market power.

50 GenCos and WPPs are regulated by different grid-access rules because they have differ-
51 ent physical natures. Conventional technologies for electricity generation are controllable.
52 Therefore, GenCos are required to provide supply curves in the DA market when they
53 access to a power-grid system. In contrast, wind energy is intermittent. Therefore, the
54 process that a WPP accesses into a power-grid system depends on how this power-grid
55 system deals with the wind-energy uncertain cost. If the WPP is required to pay its own
56 uncertain costs, it will be defined as a capacity resource (CR) and required to submit its
57 hourly generation commitment in the DA market. If the WPP's generation is less than
58 its DA commitment in an hour, it must purchase electricity from the RT market of that
59 hour to compensate for its insufficient generation. If consumers or other market players
60 are required to pay the costs associated with wind-energy uncertainty, the WPP is defined
61 as a non-capacity resource (NCR) and can participate into the RT market directly. In the
62 DA market, the SO reserves a market share for potential wind energy in each hour.

63 GenCos and WPPs are regulated by different bidding rules because they have different
64 cost structures. Each GenCo is allowed to provide one supply curve per day because its
65 fuel cost usually keeps the same in one day. In contrast, WPPs are forced to bid at at zero

66 costs but allowed to separately determine hourly generations because wind energy has zero
67 fuel costs and brings the whole system an uncertain cost that varies in different hours.

68 The above differences of market rules result in two consequences. First, WPPs do not
69 necessarily have similar marginal fuel costs with their competitors. If a WPP is a CR, its
70 marginal cost (MC) in the DA market is the marginal expected payment in the RT market
71 rather than the marginal fuel cost. If a WPP is a NCR, it has a zero MC but competes
72 with fringe GenCos in the RT market. It is necessary to explore factors that determine a
73 WPP's competitors as well as ATMPs. Second, WPPs' strategies to exercise market power
74 can be different from GenCos' while WPPs are required to determine hourly generation
75 levels rather than provide one supply curve.

76 Thus, it is necessary to develop a research framework to examine WPP's ATMP and
77 strategies to exercise their market power. However, WPPs' market power does not attract
78 enough attentions because of wind energy's zero marginal fuel cost, which causes that
79 policy makers usually assume wind-integration can always decrease price and ignore the
80 potential market power issue of WPPs[14]. Only a few papers discuss WPPs' market-power
81 strategies when they are regulated by the same rules with GenCos but do not discuss the
82 impacts of special grid-access and bidding rules regulating WPPs[1, 18].

83 The remainder of the paper is organized as follows: the theoretical power market model
84 is described in Section 2; in Section 3, I build a framework to analyze WPPs' ATMPs and
85 develop index to measure WPPs' ATMPs; then, Section 5 includes discussion about how the
86 special market regulations impact the WPPs' ATMP and strategies; Section 6 summarizes
87 the impact of WPPs' market-power strategies on the fluctuation of wind energy; Section ??
88 includes the discussion about the scenario when WPPs are NRCs; the empirical study
89 based on ERCOT 2012 data is included in to Section 8; lastly, in Section 9, I draw final
90 conclusions.

91 **2. Three-generator market model for theoretical analysis**

92 *2.1. Overview the basic structure of a electricity market in the U.S.*

93 My market model include two stages. The first stage is called the day-ahead (DA)
94 market, which occurs in one day ahead of the operation time. The second stage is called
95 the real-time (RT) market, which occurs one hour ahead of the operation time. In the DA
96 market, every GenCo submits its willingness-to-supply curve for the whole day to the SO.
97 Simultaneously, consumers provide their aggregated demand level for each hour. According
98 to the demand and supply curves, the SO integrally determines the hourly generation plans
99 for the next day by solving an aggregated daily cost-minimization problem. If the demand
100 or supply sides would like to change their contract made in the DA market, they can
101 trade again in the RT market. In the RT market, the SO separately solves the market
102 equilibriums hour by hour. In the DA market, the whole market know the the distribution
103 function of wind energy in each hour. In the RT market, the wind-energy forecast is quite
104 accurate. Thus in this research, I assume that the exact available wind energy is revealed
105 in the RT market.

106 I first consider a simplified three-generator-two-hour model in which two GenCos and
107 a WPP compete for supplying demands in two hours. In this model, a SO integrally
108 calculates the market equilibrium of two hours in the DA market and separately calculates
109 market equilibrium for each hour in the RT market. In this research, I use the superscript
110 a (r) to represent factors in the DA (RT) market. In the first part of this research, I mainly
111 focus on the situation in which WPPs are defined as CR. The situation in which WPPs
112 are not CRs will be compared in later sections.

113 I do not consider the effects of demand uncertainties in this model because those effects
114 have limited relationships with a WPP's market power. I also do not include the effects
115 of transmission losses and limits because they do not essentially affect the conclusions.
116 Because it is illegal and difficult for a WPP to collude with other market players, I in this

117 paper examine the scenario that a strategic WPP competes with other GenCos.

118 *2.2. The three-generator-two-hour (TGTH) model*

In the TGTH model, the three generators include a coal-fired generator G_c , a gas-fired generator G_g , and a WPP w . For a GenCo, the cost function of G_i ($i \in \{c, g\}$) is

$$c_i(q_i) = \alpha_i q_i + \frac{\beta_i}{2} q_i^2. \quad (1)$$

119 I assume GenCos are price takers so that their bidding curves are their marginal costs. I
120 further assume that G_c has a limited maximum ramp rate r so that the difference between
121 G_c 's generation in two neighboring hours cannot exceed r .

122 I assume that the demands are inelastic and use L_j to denote the total demand in hour
123 j . For hour j , there will be W_j MWhs wind energy available. In the DA market, the
124 WPP knows the distribution of W_j , which is a truncated normal distribution between 0
125 and installed wind-energy capacity. The wind distribution in hour j has the mean $E[W_j]$
126 and standard deviation σ_j . To simulate the market structure when the WPP is defined as
127 a CR, I assume that the SO requires the WPP to submit its hourly DA commitment q_{wj}^a
128 separately with zero cost. In contrast, a GenCo is required to provide one marginal-cost
129 curve (MC) for both hours.

130 If the WPP commits to produce q_{wj}^a in hour j in the DA market, I call $L_j - q_{wj}^a$ as the
131 net load in hour j . I first examine the situation that the WPP's commitments result in a
132 ramp up of the net load, which indicates that $L_1 - q_{w1}^a$ is less than $L_2 - q_{w2}^a$. I refer to this
133 situation as that the net demand is ramping up. The analysis of the situation that the net
134 demand is ramping down is symmetric.

In the DA market, the SO will integrally determine the generation plan for both hours according to GenCos' MCs and the WPP's commitments. Then, the market equilibrium

$\{q_{ij}^{a*}\}$ ($i \in \{c, g\}$) is solved from the following cost minimization problem.

$$\begin{aligned} \min_{q_{cj}^a, q_{gj}^a} \quad & \sum_{j=1}^2 c_{cj}(q_{cj}^a) + c_{gj}(q_{gj}^a) \\ \text{s.t.} \quad & q_{cj}^a + q_{gj}^a \geq L_j - q_{wj}^a \quad j = 1, 2 \\ & -r \leq q_{c1}^a - q_{c2}^a \leq r. \end{aligned} \quad (2)$$

In the RT market of hour j , the WPP must procure electricity from GenCos if its generation is less than its DA commitment. The equilibrium of the RT market for hour j is solved by the SO from the following problems.

$$\min_{q_{cj}^r, q_{gj}^r} c_{cj}(q_{cj}^{a*} + q_{cj}^r) + c_{gj}(q_{gj}^{a*} + q_{gj}^r). \quad (3)$$

In hour 2, the optimization problem (3) must satisfy G_c 's ramp constrain.

$$|(q_{c1}^a + q_{c1}^r) - (q_{c2}^a + q_{c2}^r)| \leq r. \quad (4)$$

135 In the DA market, if G_c 's ramp constraint is binding, prices in both hours are affected.
 136 However, if G_c 's ramp constraint is binding in the RT market, only the equilibrium in hour
 137 2 is affected.

138 **3. Measure the WPP's ability to manipulate price (ATMP) by strategically** 139 **reducing commitment levels**

140 *3.1. Price response to the WPP's commitment*

A WPP can manipulate the market price by strategically reducing its commitment levels in the DA market. In the Appendix, I calculate the market equilibriums in the DA and RT markets. When GenCos compete rather than collude with the WPP, market prices are still functions of the WPP's commitments. According to Eq.(2) , a WPP's commitment pair (q_{w1}^a, q_{w2}^a) will lead to corresponding market equilibriums, which include the DA price

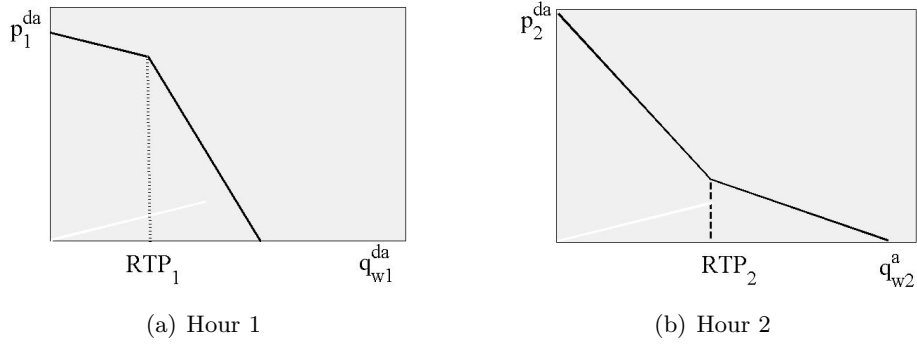


Figure 1: The WPP's residual inverse demand (RID) curves in the three-generator case p_j^a of hour j . Therefore, the price p_j^a is a function of the pair (q_{w1}^a, q_{w2}^a) . In my TGTH model, the prices are

$$p_1^a = \begin{cases} \phi_1' - \frac{\beta_g \beta_c}{\beta_g + \beta_c} q_{w1}^a, & \text{if } q_{w1}^a \leq RTP_1; \\ \phi_1 + \frac{\beta_g^2}{2(\beta_c + \beta_g)} q_{w2}^a - \frac{\beta_g(2\beta_c + \beta_g)}{2(\beta_c + \beta_g)} q_{w1}^a & \text{if } q_{w1}^a \geq RTP_1. \end{cases} \quad (5)$$

and

$$p_2^a = \begin{cases} \phi_2' - \frac{\beta_g \beta_c}{\beta_g + \beta_c} q_{w2}^a, & \text{if } q_{w2}^a \geq RTP_2; \\ \phi_2 + \frac{\beta_g^2}{2(\beta_c + \beta_g)} q_{w1}^a - \frac{\beta_g(2\beta_c + \beta_g)}{2(\beta_c + \beta_g)} q_{w2}^a & \text{if } q_{w2}^a \leq RTP_2. \end{cases} \quad (6)$$

Here RTP_1 and RTP_2 are the tipping points that determine whether G_c 's generation is limited by its own ramp rate. The tipping points are

$$RTP_1 = [L_1 - L_2 + q_{w2}^a + \frac{\beta_c + \beta_g}{\beta_g} r]_+ \quad (7)$$

and

$$RTP_2 = [L_2 - L_1 + q_{w1}^a - \frac{\beta_c + \beta_g}{\beta_g} r]_+. \quad (8)$$

141 I call p_j^a , which is a function of (q_{w1}^a, q_{w2}^a) , the WPP's residual inverse demand curve
 142 in hour j and refer to it as RID_j . I conceptually show the residual inverse demand (RID)
 143 curves in Fig. 1

144 The two figures demonstrate that price will increase when the WPP reduces its commit-
 145 ment even if GenCos compete rather than collude with the WPP. In fact, the price increases

146 because GenCos' marginal costs increase when they generate more to compensate for the
 147 WPP's commitment reduction.

148 RID curves have a piece-wise characteristic, which reflects that GenCos' ramp rates can
 149 change the relationship between market price and the WPP's commitment. For example,
 150 in Fig. 1(b), the piece on the left of the RTP_2 has a steep slope, but the right piece has a
 151 flat slope. The slopes are different because G_c 's ramp constraint is tight once $q_{w2}^a < RTP_2$,
 152 which leads the price to be more sensitive to WPP's commitment change. According to
 153 Eq. (6), one less unit commitment from the WPP inflates the price by $\frac{\beta_g \beta_c}{\beta_g + \beta_c} q_{wj}^a + \frac{\beta_g^2}{2(\beta_c + \beta_g)}$
 154 when $q_{w2}^a < RTP_2$. In contrast, one less unit commitment from the WPP only inflates the
 155 price by $\frac{\beta_g \beta_c}{\beta_g + \beta_c} q_{wj}^a$ when $q_{w2}^a \geq RTP_2$.

156 The effects of GenCos' ramp rates on the price's sensitivity to the WPP's commitment
 157 varies in different hours. In both two hours, price increase while the WPP reduces its
 158 commitment. However, the price increase slows down in hour 1 but speeds up in hour 2.
 159 This is because a large WPP's commitment in hour 1 will tighten GenCos' ramp constraints.
 160 In contrast, in hour 2 when net demand is high, a low WPP's commitment can tighten
 161 GenCos' ramp constraints tight. Therefore, I have the following theorem.

162 **Theorem 3.1.** *In an hour with a high net demand, the market price becomes more sensitive*
 163 *to WPPs' commitments while WPPs reduce their commitments. In an hour with a low net*
 164 *demand, the market price becomes less sensitive to WPPs' commitments while WPPs reduce*
 165 *their commitments.*

166 3.2. Index to measure the WPP's ATMP

167 I measure the WPP's ATMP by using the slope of RID_j between \hat{q}_{wj}^a , the WPP's
 168 commitment as a price taker, and q_{wj}^{a*} , the WPP's commitment as a market power. I use
 169 the slope between these two points because a rational WPP's commitment will not exceed
 170 its price-taker commitment level or be lower than its market-power level. Then, I define
 171 the following index to measure the WPP's ATMP.

Definition I define the inverse elasticity of the RID in period j as

$$\eta_j^a = \frac{p_j^a(\hat{q}_{wj}^a) - p_j^a(q_{wj}^{a*})}{\hat{q}_{wj}^a} / \frac{\hat{q}_{wj}^a - p_j^a(q_{wj}^{a*})}{\hat{q}_{wj}^a}. \quad (9)$$

172 The WPP has a high ATMP when η_j is large

173 I would like to particularly emphasize that, in contrast with GenCos' ATMPs, a WPP's
 174 ATMP is contingent to how the WPP optimizes its own profit. A WPP can separately
 175 determine its market-power commitments by maximizing its total expected profit of each
 176 hour. Or, the WPP can integrally determine several hours' market-power commitments by
 177 maximizing its total expected profits of these hours. how many hours the WPP integrally
 178 maximize its profits determines this WPP's market-power commitment level q_{wj}^{a*} in hour
 179 j . Therefore, the value of η_j^a also depends on how the WPP selects its market-power com-
 180 mitment. Thus, **I use η_{jN}^a to represent the WPP's ATMP when WPP determine**
 181 **its market-power commitment by integrally maximizing N hours' profit.**

182 **4. Factors determine when the WPP has significant ATMP**

183 *4.1. The WPP's marginal commitment cost*

184 When defined as RC, the WPP faces a marginal commitment cost (MCC) in each hour
 185 that reflects the expected penalty when this WPP commits one more unit of generation
 186 in the DA market. I use mcc_j to represent the WPP's MCC in hour j . Because a WPP's
 187 ATMP in hour j is the average slope of RID_j between q_{wj}^{a*} and \hat{q}_{wj}^a , which are determined by
 188 mcc_j , the WPP's MCC has decisive impacts on the sam WPP's ATMP (Fig. 2). However,
 189 the WPP's MCC varies by hour. Thus, the WPP's ATMP in each hour of the same day
 190 can differ significantly.

191 The WPP's MCC impacts the ATMP by determining the WPP's competitors in each
 192 hour. η_j is the average slope of RID_j between q_{wj}^{a*} and \hat{q}_{wj}^a . Therefore, when a WPP
 193 reduces its commitment from \hat{q}_{wj}^a to q_{wj}^{a*} , the WPP competes with GenCos whose marginal
 194 costs are in between $mcc_j(q_{wj}^{a*})$ and $mcc_j(\hat{q}_{wj}^a)$. And the price inflation reflects the change of

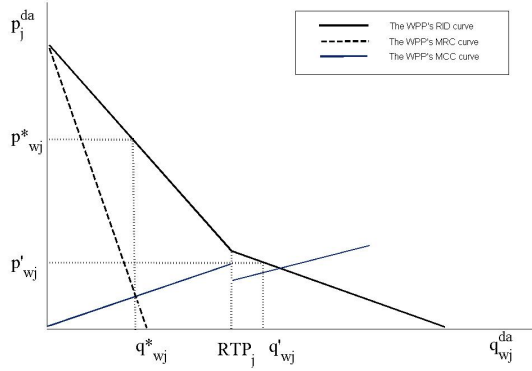


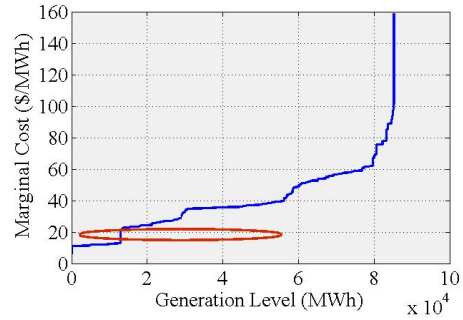
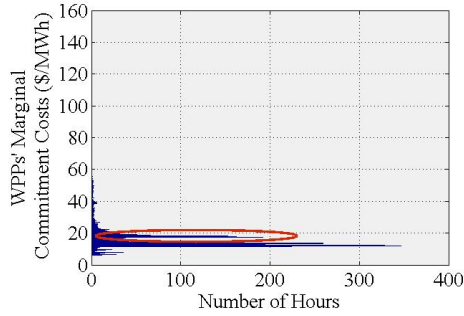
Figure 2: The WPP's MCC and ATMP in hour 2

195 these GenCos' marginal costs. Thus, The WPP has a high ATMP in hour j when GenCos,
 196 whose marginal costs is in between $mcc_j(q_{wj}^{a*})$ and $mcc_j(\hat{q}_{wj}^a)$, has steep-slope supply curves.
 197 Then, I have the following theorem.

198 **Theorem 4.1.** *A WPP has a high ATMP in hour j if GenCos whose supply curves are*
 199 *in between $mcc_j(q_{wj}^{a*})$ and $mcc_j(\hat{q}_{wj}^a)$ have steep slopes.*

200 Again, I would like to emphasize, the WPP's ATMP varies in different hours because
 201 the WPP's MCC varies in different hours.

202 In fact, in contrast with GenCos, which mostly have high ATMPs when demand is quite
 203 high, WPPs can have high ATMPs even if the demand is moderate. For example, WPPs can
 204 have high ATMPs in some hours when demands are moderate but marginal GenCos change
 205 from coal-fired GenCos to gas-fired GenCos. I calculate the average of WPPs' hourly price-
 206 taker MCC $mcc_j(q_{wj}^{a*})$ and market-power MCC $mcc_j(\hat{q}_{wj}^a)$ in ERCOT 2012 and plot the
 207 histogram in Fig. 3(a). In Fig. 3(b), I plot the aggregated marginal-cost curve of GenCos
 208 in the ERCOT in 2012. By comparing the two figures, I suggest that WPPs can have high
 209 ATMPs in hours when their average MCCs are around 20 \$/MWh. In these hours, the
 210 supply curve, which reflect the marginal costs of WPPs' GenCos, sharply increases. In
 211 contrast, GenCos whose marginal costs are around 20 \$/MWh usually have very limited
 212 ATMPs because they can only provide one supply curve in each day. If they strategically
 213 inflate cost curves for the hour when they are fringe generators, they will not be dispatched



(a) Histogram of WPPs' Hourly MCCs in ER- (b) Aggregated GenCos' Marginal Cost Curves
COT 2012 in ERCOT 2012

214 in other hours when the demand is low. When the demand is relatively high, their bidding
 215 curves have little influence on the price. **Therefore, WPPs have more opportunity to**
 216 **exercise market power because they are allowed determine hourly generation**
 217 **levels.**

218 Actually, because the WPPs's MCCs varies by hours, they are allowed to separately
 219 determine their hourly generation levels rather than provide daily supply curve. A WPP's
 220 MCC is determined by wind-energy distribution and MCs of GenCos who can provide
 221 electricity market in the RT market. Additionally, a WPP's MCC in an hour can be
 222 impacted by its own generations and demands in neighboring hour's because of the effects
 223 of GenCos' ramp constraints. In the Appendix, I detailed discuss the MCC of the WPP
 224 in my TGTH model and summarize the analyses in the following theorem.

225 **Theorem 4.2.** *A WPP's MCC is a increase function of its DA commitment level in*
 226 *the same hour. In a high net-demand hour, the WPP's MCC is sensitive to its own DA*
 227 *commitment in the same hour if*

- 228 • *its competitors have limited ramp rates,*
- 229 • *or this WPP make a high commitment in a low-net-load neighboring hours when*
 230 *GenCos ramp constraints are binding.*

231 *4.2. GenCos' ramp rates and market conditions in neighboring hours.*

232 WPPs' ATMP in a high-net-demand hour can be significantly impacted by GenCos
 233 ramp rates and market conditions in neighboring low-net-demand hours. In my TGTH

234 model, the WPP's ATMP in hour 2 is determined by the location of RTP_2 once RTP_2
 235 falls in between q_{w2}^{a*} and \hat{q}_{w2}^a . According to Eq.(8), RTP_2 is a function of G_c 's ramp rate
 236 r , demand in hour 1 L_1 , and the WPP's commitment in hour 1 q_{w1}^a . Furthermore, these
 237 factors also impact q_{w2}^{a*} by determining the WPP's MCC in hour 2. Therefore, these three
 238 factors impact the WPP's ATMP.

239 GenCos's ramp rates impacts a WPP's ATMP by determining its RID and MCC curves.
 240 In fact, a WPP has a high ATMP when GenCos ramp rates are small. For example, in my
 241 TGTH model, G_c 's ramp rate r impacts the WPP's ATMP in hour 2 by determining the
 242 value of RTP_2 and q_{w2}^{a*} . According to Eq.(8), RTP_2 is a decrease function of G_c 's ramp rate
 243 r . Therefore, a decrease of r will increase the value of RTP_2 . In contrast to the effect on
 244 RTP_2 , the dynamic that r impacts q_{w2}^{a*} is complicated. Actually, a decrease of r raises both
 245 the WPP's RID and MCC curves. If the effect on the RID curve is stronger than that on
 246 the MCC curve, r 's decrease raises q_{w2}^{a*} . Otherwise, r 's decrease reduces q_{w2}^{a*} . However, in
 247 the Appendix, I demonstrate that the distance between q_{w2}^{a*} and RTP_j is always increasing
 248 while r is decreasing no matter how r impacts q_{w2}^{a*} . Consequently, r 's decrease raises the
 249 value of η_2^a according to Eq.(9). The analyses for hour 1 is symmetric.

250 Because RTP_j impacts η_j and is determined by market conditions in neighboring hours,
 251 the WPP's ATMPs in hour j are impacted by these conditions including demand and the
 252 WPP's commitment level. For example, η_2 is impacted by L_1 and q_{w1}^a . Given the same
 253 L_2 , a larger $L_1 - q_{w1}^a$ helps the WPP gain higher ATMP.

254 Actually, a decrease of L_1 raises the WPP's AEMP by rising the steep-slope piece of
 255 RIDC in hour 2. Consequently, both RTP_2 and q_{w2}^{a*} increase while L_1 is declining. Because
 256 RTP_2 increases more than q_{w2}^{a*} does, η_j^a increases with L_1 .

257 Simultaneously, a increase of q_{w1}^a exacerbates the WPP's AEMP. Similar to the effect of
 258 increasing L_2 , an increase of q_{w1}^a also raise the steep-slope piece of the RID curve in hour 2.
 259 Additionally, q_{w1}^a 's increase also influences the WPP's MCC curve in hour 2. When q_{w1}^a is

260 increasing, mcc_2 becomes less and less sensitive to q_{w2}^a . The aggregated effect of these two
 261 dynamics results in increases of RTP_2 and q_{w2}^{a*} . However, η_j^a still increases because RTP_2
 262 increases more than q_{w2}^{a*} does.

263 In the appendix, I mathematically demonstrate the above dynamics and have the fol-
 264 lowing theorem.

265 **Theorem 4.3.** *Once the WPP's strategic commitment reduction causes GenCos genera-*
 266 *tions to be limited by their ramp rates, the WPP has a high ATMP in the hours with high*
 267 *net demand if*

- 268 • *GenCos' ramp rates are small,*
- 269 • *the net-demand ramp is large,*
- 270 • *demands are low in hours with low net demands,*
- 271 • *or the WPP itself makes high generation commitments in hours with low net demands.*

272 **5. WPP's strategy of utilizing fluctuations of net demands and GenCos' ramp** 273 **rates**

274 Compared with GenCos, WPPs have a particular strategy to manipulate the market
 275 price by utilizing fluctuations of net demands. By adopting this particular strategy, WPPs
 276 can inflate price higher but produce more when net-demand fluctuations are significant than
 277 when net-demand fluctuations are moderate. WPPs have this particular strategy because
 278 WPPs are allowed separately submit hourly commitment. In this section, I analyze this
 279 particular strategy and its impacts on the forecasted wind-energy fluctuations.

280 In my TGTH model, the WPP can gain a higher profit by integrally maximizing profits
 281 of the two neighboring hours when G_c 's ramp rate can be tightened because of the fluctu-
 282 ation of the net demands. In the above analyses, I demonstrate that the WPP's AEMP
 283 in hour 2 can be impacted by the WPP's own commitment in hour 1 q_{w1}^a . Therefore, if
 284 WPPs integrally determines its optimal commitments in the two hours, the WPP can gain
 285 market-power rents by utilizing G_c 's ramp rate to enhance its own AEMP in hour 2. By
 286 utilizing G_c 's ramp rate, the WPP can generate more in hour 1 to enhance the WPP's

287 ATMP in hour 2. Consequently, the WPP in hour 2 can inflate price to a higher level but
 288 reduce less commitment level than when the WPP separately maximize its hourly profit.

In fact, if the WPP in the TGTH model integrally determines its commitments in the two hours, the WPP's profit maximization problem is,

$$\begin{aligned} \max_{q_{wj}^a, q_{wj}^r} \quad & p_1^a q_{w1}^a + p_2^a q_{w2}^a + \tau(w_1 + w_2) - E[p_1^r(q_{w1}^a - q_{w1}^r) + p_2^r(q_{w2}^a - q_{w2}^r)] \\ \text{s.t.} \quad & q_{wj}^r \leq W_j. \end{aligned} \quad (10)$$

Here, τ is the subsidy for the WPP's per-unit generation. Therefore, the WPP's market-power commitments in the two hours are solved from

$$\tau + p_1^a + \frac{\partial p_1^a}{\partial q_{w1}^a} q_{w1}^a + \frac{\partial p_2^a}{\partial q_{w1}^a} q_{w2}^a \mathbf{1}(q_{w2}^a < RTP_2) = mcc_1, \quad (11)$$

$$\tau + p_2^a + \frac{\partial p_2^a}{\partial q_{w2}^a} q_{w2}^a + \frac{\partial p_1^a}{\partial q_{w2}^a} q_{w1}^a \mathbf{1}(q_{w1}^a > RTP_1) = mcc_2, \quad (12)$$

289 From these two conditions, I can solve for f_{21} , the WPP's best response function of hour 2
 290 that reflect how q_{w2}^{a*} respond to q_{w1}^a . Similarly, I can solve for f_{12} , the WPP's best response
 291 function of hour 1 to its own commitment in hour 2.

292 In fact, the WPP's marginal benefit curve in each hour includes two parts. One part is
 293 the marginal benefit from manipulating price in the current hour. The other is the marginal
 294 benefit from manipulating the neighboring hour's price. For example, on the left-hand side
 295 of Eq. (11), which is the WPP's marginal benefit curve in hour 1, $\tau + p_1^a + \frac{\partial p_1^a}{\partial q_{w1}^a} q_{w1}^a$ is
 296 the marginal benefit from using q_{w1}^a to manipulate p_2^a . There is an additional term $\frac{p_2^a}{q_{w1}^a}$
 297 that reflects the WPP's marginal benefit from using q_{w1}^a to manipulate p_2^a . Because $\frac{p_2^a}{q_{w1}^a}$ is
 298 positive according to Eq.(6), the WPP has incentive to commit more in hour 1 than when
 299 the WPP separately determines its hourly optimal strategy.

300 The economical explanation of the above dynamic is that the WPP has a incentive
 301 to raise q_{w1}^a in exchange for a high profit in hour 2 if the WPP integrally determine its
 302 strategies in the two hours. A increase of q_{w1}^a has three effects: inflating the DA price in

hour 2, enlarging the WPP's ATMP in hour 2, and decreasing the WPP's MCC in hour
2. All three effects incentivize the WPP to generate more but keep the price in hour 2 at
a high level. Therefore, I have the following theorem.

Theorem 5.1. *If GenCos' ramp rates limit their generations in two hours while WPPs exercise their market power, WPPs have incentives to make more generation commitment in low-net-demand hours when they integrally determine strategies in these two hours than when they separately determine the hourly strategies. WPPs make more commitment in low-net-demand hours to in exchange for a higher ATMP and lower MCC in high-net-demand hours. I call this effects the **ramping rate's rebound(R3) effect**.*

I would like to emphasize that R3 effect can occur both in the scenarios when net load is ramping up and the scenario when net load is ramping down. When the net load is ramping up, the WPP will make high commitment in low-net-demand hour in exchange for higher profit in the following hours. In contrast, when the net load is ramping down, the WPP will make high commitment in low-net-demand hour in exchange for high profit in the previous hours.

6. WPP's strategic behavior and wind-energy fluctuation

While net-demand fluctuations determines WPPs' strategies explained in the last section, WPPs' strategies also impact net-demand fluctuations. Once wind-energy penetration is significant, wind-energy fluctuations unignorably affect the extent of net-demand fluctuations. However, the extends of wind-energy fluctuation are determined by strategies adopted by WPPs for DA-commitment making. Therefore, net-demand fluctuations will be sensitive to strategies adopted by WPPs. In particular, net-demand fluctuations are different when WPPs separately determine commitments of each hour from when WPPs integrally determine commitments of several hours.

For example, wind-energy fluctuation, as well as net-demand fluctuation, is different when the WPP in the TGTH model choose different strategies. In Theorem 5.1, I demonstrate that R3 effect causes the WPP to make a high generation commitment in hour 1

330 when the WPP integrally determine strategies in two hours than when the WPP separately
 331 determines its hour strategy. The R3 effect also impacts the WPP's commitment in hour
 332 2 and the wind-energy fluctuation. I summarize the impacts in the following theorem.

Theorem 6.1. *If the WPP integrally determine its strategies in two hours rather than separately determine hourly strategies, net-demand-energy fluctuation is aggravated if*

$$\frac{\partial q_{w2}^{a*}}{\partial p_2^a} \frac{\partial p_2^a}{\partial q_{w1}^a} + \frac{\partial q_{w2}^{a*}}{\partial m_{cc2}} \frac{\partial m_{cc2}}{\partial q_{w1}^a} < 1. \quad (13)$$

333 *Otherwise, the net-demand-energy fluctuation keeps the same or is moderated.*

334 The proof of the above theorem is also the economical explanation of the condition
 335 Eq. (13). The first term of the left-hand side of Eq. (13) is the increase of q_{w2}^{a*} that respond
 336 to p_2^a 's change of caused by one-unit more q_{w1}^a . The second term is the increase of q_{w2}^{a*}
 337 that respond to m_{cc2} 's change caused by one-unit more q_{w1}^a . If the integrated effects of
 338 increasing q_{w1}^a by one-more unit causes q_{w2}^{a*} to increase by less than one unit, the wind-
 339 energy fluctuation will be aggravated because the value of $q_{w1}^a - q_{w2}^{a*}$ enlarges.

340 In fact, the q_{w2}^{a*} 's sensitivity to q_{w1}^a depends on competitors' supply curves and the joint
 341 distribution of wind energy in the two hours according to Eq. (5) and Eq. (B.1). In fact,
 342 q_{w1}^a 's growth stimulates q_{w2}^{a*} to increase more if the ratio β_g/β_c has a larger value. **There-**
 343 **fore, if the slow-ramping GenCos have much lower costs than fast-ramping Gen-**
 344 **Cos, R3 effect can shrink the wind-energy fluctuations.** However, if $E[W_1 - W_2]$
 345 is large, increasing q_{w1}^a have small effect of decreasing m_{cc_j} . Consequently, q_{w2}^{a*} 's growth
 346 is small when R3 occurs. **Therefore, R3 can enlarge the wind-energy fluctuations**
 347 **when the wind-energy forecasts has already significantly fluctuated.**

348 7. WPPs' market power when they are NRC

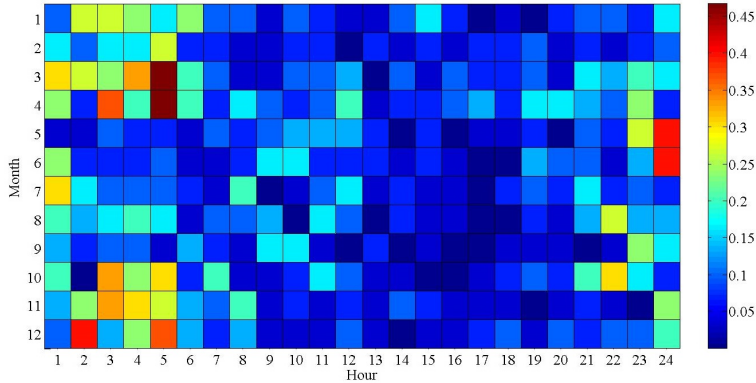
349 If the WPP is defined as NRC, their ATMPs can be analyzed by the same framework
 350 explained above. However, there are two essential differences. First, the WPP's ATMP is
 351 determined by demands and wind-energy forecast rather than its MC. In fact, the WPP

352 can participate in the RT market and has zero MCs, and the SO will reserve market shares
353 for it in the DA market. The low-cost GenCos will be dispatched in the DA market to
354 balance the net load, which is the forecast wind energy subtracted from the demand. The
355 fringe GenCos will be used in the RT market if the WPP's generation is less than the
356 reserved market share. Therefore, the fringe GenCos' MCs determine the slopes of WPP's
357 RIPs and marginal-benefit (MB) curves. Because WPP's marginal cost is zero, the RIPs
358 determines the value of η_j . Actually, all characteristics of RIPs, which include the slope
359 of each piece and the location of tipping point, are determined by demands, wind-energy
360 forecasts, fringe GenCos's ramp rates and MCs. In addition, the demand and wind-energy
361 forecast determine which GenCos are fringe in an hour. Therefore, the demand level and
362 wind-energy forecasts determine the WPP's ATMP.

363 In addition, the SO separately calculate RT market equilibrium for each hour in the
364 RT market in contrast with integrally calculate market equilibrium for all hours in the DA
365 market. Therefore, The R3 effect will occur only when the net load in the RT market is
366 ramping up.

367 **8. WPPs' market power in the ERCOT market in 2012**

368 In order to examine WPPs' ATMPs in a real electricity market by using my analyzing
369 framework presented in this paper, I calculated the ATMPS of the WPPs in the ERCOT
370 market in 2012 if the WPPs are aggregately bid in the RT market. I assume WPPs
371 separately optimal their hourly profits. In the DA market, the SO will determine the
372 generation plan for the next 24 hours that starts from 12 am the next day and reserve
373 market shares for WPPs according to hourly wind-energy forecast. I assume GenCos are
374 price takers and their marginal generation costs are their heat rates times fuel costs. In
375 the RT market, the SO separately solves the market equilibriums hour by hour. In this
376 study, I ignore the effects of demand forecast error and assume hourly demand is inelastic
377 and known in the DA market.



378 Figure 3: The average probability of the WPPs have ability to manipulate the prices
 The data of hourly demands and wind-energy forecasts are provided by ERCOT[9].
 379 The data include wind-energy hourly generation, the DA wind energy forecast, and 20%
 380 quantiles of forecast errors. The fuel-price data is from the Energy Information Agency
 381 (EIA)[7]. The technological features of the generators are from the Emissions & Generation
 382 Resource Integrated Database (eGRID), issued by the United States Environmental Pro-
 383 tection Agency (EPA)[8]. The eGRID database provides the heat rates (MMBtu/MWh)
 384 and the maximum generation capacities of 235 generators in the ERCOT
 385 The calculation results demonstrate that WPPs already have had significant ATMPs in
 386 some hours even at the 2012’s penetration levels in ERCOT, which is around 9% of the total
 387 electricity generation. In 2012, there are more than 900 hours in which the WPPs have
 388 ATMPs that are greater than zero. In 93 hours, 1 MWh’s decrease of WPP’s generation can
 389 result in nearly 9\$/MWh, which is around 25% of the DA price level. In order to examine
 390 when WPPs have high potential to have ATMPs, I also calculate the monthly probability
 391 of each hour in which the WPPs have ATMPs. The results are summarized in Figure 3. I
 392 observed that hours that WPPs have ATMPs concentrate in some particular period such
 393 as late night and early morning. The WPPs have market power in these periods because
 394 GenCos have limited ramp rates and wind-energy fluctuations are significant.

395 9. Conclusions

396 In this paper, I build a two-stage-multiple-hour model to analyze wind power pro-
397 ducers' (WPPs) ability to manipulate price and market-power strategies in a sequentially
398 structured electricity market. By separately examining two scenarios when WPPs partic-
399 ipate in the day-ahead and real-time markets, I clarify the cost structure of WPPs and
400 explore which factors determine WPPs' ATMPs. I examine when WPPs have significant
401 ATMPs. The analyses demonstrate that WPPs can have significant ATMPs even though
402 their marginal fuel costs are zero. Furthermore, the current bidding regulation that allows
403 WPPs to separately determine their hourly generations provide WPPs more flexibility to
404 exercise their market power. Because of this regulation, WPPs can gain high ATMPs
405 in peak-demand hours by adjusting their generation in low-demand hours. Furthermore,
406 WPPs can utilize conventional generators' ramp constraints to exercise their market power
407 so that they can inflate prices higher and produce more electricity than suppliers who are
408 only allowed to provide one supply curve per day in the day-ahead market. My empiri-
409 cal simulation based on data from Texas in 2012 demonstrates that WPPs already have
410 ATMPs in over 900 hours.

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468 10. Appendix

469 Appendix A. Market equilibrium of the TGTH model

470 In DA, the generators submit their bid curves to the SO. Because I focus on the market
 471 power of wind, I assume convention generators are truthful in their bids and submits their
 472 marginal cost curves. If the WPP is defined as CR, its DA commitment level in hour j is
 473 q_{wj}^a .

By solving (2), the day-ahead dispatch is:

$$q_{c1}^a = \begin{cases} \frac{\alpha_g - \alpha_c}{\beta_c} + \frac{\beta_g}{\beta_g + \beta_c} (L_1 - q_{w1}^a - \frac{\alpha_g - \alpha_c}{\beta_c}), \\ \text{if } |q_{c1}^a - q_{c2}^a| < r; \\ \frac{\beta_g(L_1 - q_{w1}^a + L_2 - q_{w2}^a) + 2(\alpha_g - \alpha_c)}{2(\beta_c + \beta_g)} \mp \frac{r}{2}, \\ \text{if } q_{c1}^a - q_{c2}^a = \mp r. \end{cases} \quad (\text{A.1})$$

$$q_{c2}^a = \begin{cases} \frac{\alpha_g - \alpha_c}{\beta_c} + \frac{\beta_g}{\beta_g + \beta_c} (L_2 - q_{w2}^a - \frac{\alpha_g - \alpha_c}{\beta_c}), \\ \text{if } |q_{c1}^a - q_{c2}^a| < r; \\ \frac{\beta_g(L_1 - q_{w1}^a + L_2 - q_{w2}^a) + 2(\alpha_g - \alpha_c)}{2(\beta_c + \beta_g)} \pm \frac{r}{2}, \\ \text{if } q_{c1}^a - q_{c2}^a = \mp r. \end{cases} \quad (\text{A.2})$$

$$q_{g1}^a = \begin{cases} \frac{\beta_c}{\beta_g + \beta_c} (L_1 - q_{w1}^a - \frac{\alpha_g - \alpha_c}{\beta_c}), \\ \text{if } |q_{c1}^a - q_{c2}^a| < r; \\ \frac{(2\beta_c + \beta_g)(L_1 - q_{w1}^a) - \beta_g(L_2 - q_{w2}^a) - 2(\alpha_g - \alpha_c)}{2(\beta_c + \beta_g)} \pm \frac{r}{2}, \\ \text{if } q_{c1}^a - q_{c2}^a = \mp r. \end{cases} \quad (\text{A.3})$$

$$q_{g2}^a = \begin{cases} \frac{\beta_c}{\beta_g + \beta_c}(L_2 - q_{w2}^a - \frac{\alpha_g - \alpha_c}{\beta_c}), \\ \text{if } |q_{c1}^a - q_{c2}^a| < r; \\ \frac{(2\beta_c + \beta_g)(L_2 - q_{w2}^a) - \beta_g(L_1 - q_{w1}^a) - 2(\alpha_g - \alpha_c)}{2(\beta_c + \beta_g)} \mp \frac{r}{2}, \\ \text{if } q_{c1}^a - q_{c2}^a = \mp r, \end{cases} \quad (\text{A.4})$$

474 If the WPP is NCR, the SO will determine the DA market equilibrium according to expected
475 wind-energy level $E[W_j]$. Therefore, I can get the day-ahead dispatch by replacing q_{wj}^a by
476 $E[W_j]$ in Eq. (A.1) Eq. (A.4). According to the dispatch, I can get the market-clearing
477 price as Eq. (5) and Eq. (6)

Given WPP's generation level W_j in hour j , the optimal RT dispatch strategy for period
1 is

$$q_{c1}^r = \frac{\beta_g}{\beta_g + \beta_c}(q_{w1}^a - W_1) + (\frac{1}{\beta_c} - \frac{\beta_g}{\beta_c(\beta_g + \beta_c)})(\alpha_g + \alpha_c) \\ + (\frac{\beta_g}{\beta_c} - \frac{\beta_g^2}{\beta_c(\beta_g + \beta_c)})q_{g1}^a + (\frac{\beta_g}{\beta_g + \beta_c} - 1)q_{c1}^a, \quad (\text{A.5})$$

$$q_{g1}^r = \frac{\beta_c}{\beta_g + \beta_c}(q_{w1}^a - W_1) - \frac{1}{\beta_g + \beta_c}(\alpha_g + \alpha_c) \\ - \frac{\beta_g}{\beta_g + \beta_c}q_{g1}^a + \frac{\beta_c}{\beta_g + \beta_c}q_{c1}^a. \quad (\text{A.6})$$

Given WPP's generation level w_2 , the RT dispatch strategy for period 2 is

$$q_{c2}^r = \begin{cases} q_{c1}^r \pm r, \text{ if (4) is binding;} \\ \frac{\beta_g}{\beta_g + \beta_c}(q_{w2}^a - W_2) + (\frac{1}{\beta_c} - \frac{\beta_g}{\beta_c(\beta_g + \beta_c)})(\alpha_g + \alpha_c) + \\ (\frac{\beta_g}{\beta_c} - \frac{\beta_g^2}{\beta_c(\beta_g + \beta_c)})q_{g2}^a + (\frac{\beta_g}{\beta_g + \beta_c} - 1)q_{c2}^a, \\ \text{if (4) is not binding.} \end{cases} \quad (\text{A.7})$$

$$q_{g2}^r = \begin{cases} q_{w2}^a - W_2 - (q_{c2}^r \pm r), \\ \text{if (4) is binding;} \\ \frac{\beta_c}{\beta_g + \beta_c}(q_{w2}^a - W_2) - \frac{1}{\beta_g + \beta_c}(\alpha_g + \alpha_c) - \\ \frac{\beta_g}{\beta_g + \beta_c}q_{g2}^a + \frac{\beta_c}{\beta_g + \beta_c}q_{c2}^a, \\ \text{if (4) is not binding.} \end{cases} \quad (\text{A.8})$$

478 I argue that a strategic WPP, which is defined as a RC, will not strategically hold back
479 its generation capacity and generating electricity up to $\min\{W_j, q_j^a\}$ MWhs. If the WPP's
480 generation q_{wj}^r is less than $\min\{W_j, q_j^a\}$ MWhs in hour j , its net benefit is $-p_j^r(q_j^a - q_j^r)$
481 instead of $\min\{0, -p_j^r(q_j^a - W_j)\}$. Consequently, the WPP's net profit decreases. Therefore,
482 I have the following corollary.

483 **Corollary Appendix A.1.** *The WPP's optimal strategy in the RT market is to adopt*
484 *the price-taker strategy, therefore it has no ability to affect the RT price.*

485 Appendix B. WPP's marginal commitment costs (MCC)

486 Parallel with the RID curve, the WPP's MCC curve also affects the WPP's ability to
487 manipulate the price. In contrast with the RID curve that reflects the price sensitivity
488 to the WPP's commitment, the MCC curve reflects the WPP's marginal-cost sensitivity
489 to its own commitment. The MCC curve associated with the RID curve determines the
490 WPP's market-power commitment q_{wj}^{a*} and price-taker commitment $q_{wj}^{\prime a}$, which together
491 determine the inverse elasticity of the WPP's RID curve. Furthermore, the WPP's MCC
492 also determines the WPP's willingness to exercise its market power. If the MCC quickly
493 increases as the WPP's commitment grows, the WPP has high incentive to exercise its
494 market power for not just inflating price but also preventing high marginal commitment
495 costs. In the rest of this section, I analyze the WPP's MCC curve in hour 2, in which
496 the net demand is high. The conclusions can be symmetrically generalized to get the
497 characteristics of the MCC in hour 1, in which the net demand is low.

Because of the effects of G_c ' ramp rate, the WPP's MCC curve in my three-generator case is a discontinuous function. I conceptually show the WPP's MCC curves of the two hours in Fig. 1. The RID's tipping point also splits the MCC curve in the same hour. The WPP's MCC curve of hour 2 in the DA market is a piece-wise function described in the following equation.

$$mcc_2 = \begin{cases} \int_0^{q_{w_2}^a} \chi(w_2) f(w_2) dw_2 + q_{w_2}^a \{ Prob(W_1 \geq q_{w_1}^a \cup (W_1 < q_{w_1}^a \cap W_2 < W_1 - \frac{\beta_g + \beta_c}{\beta_g} r)) \beta_g \\ + Prob(W_1 < q_{w_1}^a \cap W_2 \geq W_1 - \frac{\beta_g + \beta_c}{\beta_g} r) \frac{\beta_c}{\beta_g + \beta_c} \beta_g \}, \text{ if } q_{w_2}^a < RTP_2; \\ \int_0^{q_{w_2}^a} \chi'(w_2) f(w_2) dw_2 + q_{w_2}^a \{ Prob(W_2 < q_{w_2}^a \cap W_2 < W_1 - \frac{\beta_g + \beta_c}{\beta_g} r) \beta_g \\ + Prob(W_2 \geq q_{w_2}^a \cup (W_2 < q_{w_2}^a \cap W_2 \geq W_1 - \frac{\beta_g + \beta_c}{\beta_g} r)) \frac{\beta_c}{\beta_g + \beta_c} \beta_g \}, \text{ if } q_{w_2}^a \geq RTP_2; \end{cases} \quad (B.1)$$

498 Here, given w_2 , χ and χ' are functions of demands and conventional GenCos' bidding curves.
 499 The WPP's MCC curve's discontinuity reselects the heterogeneity of the MCC's sensitivity
 500 to its own commitment, as in Eq. B.1. When the WPP's commitment $q_{w_j}^a < RTP_j$,
 501 the WPP's MCC is more sensitive to its own commitment than in the scenario when
 502 $q_{w_j}^a \geq RTP_j$.

503 The MCC's sensitivity is heterogenous with respect to its own commitment because of
 504 two reasons. First, the RT-price sensitivity with respect to the WPP' DA commitment
 505 is different by whether G_c 's ramp constrain is binding in the RT market. Second, as
 506 shown in Fig., the probability of that G_c 's ramp constrain is binding in the RT market
 507 discontinuously jumps to a significantly high level if the WPP increase its commitment
 508 $q_{w_2}^a$ from just lower than RTP_2 to just higher than RTP_2 . Therefore, the WPP's MCC
 509 curve discontinuously drops to a low level at RTP_2 and has flatter slope when $q_{w_2}^a > RTP_2$
 510 because the MCC is the expected RT-price. (Because the fact the $q_{w_2}^a < RTP_2$ will
 511 essentially expand the probability of the situation that G_c 's ramp constrain is binding,
 512 under which situation the RT price p_r is more sensitive to $q_{w_2}^a$, the WPP's MCC $E[p_r^2]$ is
 513 more sensitive to $q_{w_2}^a$ when $q_{w_2}^a < RTP_2$ than when $q_{w_2}^a \geq RTP_2$. Consequently, the MCC

514 curve has steeper slope in the segment of $q_{wj}^a < RTP_j$ than in the segment of $q_{wj}^a \geq RTP_j$.)

515 By summarizing above analyse, I have the following theorem.

516 **Theorem Appendix B.1.** *If a GenCo's ramp constrain is becoming binding because a*
517 *WPP reduces its DA commitment, the WPP's MCC is a discontinuous increase function*
518 *of its own commitment. The MCC is more sensitive to the WPP's DA commitment when*
519 *the GenCo's ramp constrain is binding than when the constrain is not binding.*

520 *Proof.* First, the probability of the situation that G_c ' ramp constrain is binding in the RT
521 market changes with the WPP's DA commitment q_{w2}^a discontinuously at the point RTP_2 .
522 As shown in Eq.(B.1), G_c ' ramp rate has much more opportunity to limit its generation
523 in the RT market if $q_{w2}^a < RTP_2$ than in the scenario if $q_{wj}^a \geq RTP_j$. I would like to
524 emphasize that the fact that G_c 's ramp constrain is binding in the DA market does not
525 necessarily indicate that the same constrain is binding in the RT market. For example, if
526 the WPP's generation in the hour 1 is sufficiently small in RT market, G_c can generate
527 more than its commitment in hour 1 such that $q_{w1}^r > q_{w2}^a$. Consequently, G_c 's generation
528 capacity in hour 2 is $q_{w1}^r + r$ in the RT market instead of $q_{w2}^a + r$ and the ramp constrain
529 can be no binding. I in Fig. compare the probability of ramp-constrain binding given
530 the WPP's different DA commitment levels. If $q_{w2}^a < RTP_j$, G_c 's generation will not be
531 constrained only if the WPP's generation capacities in both two hours are less than the
532 commitment levels and $W_2 - W_1$ is sufficiently large. In contrast, if $q_{w2}^a \geq RTP_2$, G_c 's
533 generation will not be constrained once the WPP's generation capacity W_2 in hour 2 is
534 large than its commitment or $W_2 - W_1$ is relatively large when $W_j < q_{wj}^a$. Therefore, G_c
535 are more likely to be constrained by its own ramp rate in the RT market when $q_{w2}^a < RTP_2$
536 than when $q_{w2}^a \geq RTP_2$.

537 The second reason (that causes the WPP's heterogenous MCC sensitivities to its DA
538 commitment) is that the RT-price sensitivity to the WPP' DA commitment depends on
539 whether G_c 's ramp constrain is binding. The RT price, which determines the WPP's MCC,
540 is more sensitive to the WPP's DA commitment if G_c 's ramp constrain is binding in the RT
541 market than if the ramp constrain is not binding. For example, the WPP's MCC in hour
542 2 $E[p_2^r]$, which is the expected price that the WPP needs to pay for purchasing electricity
543 in the RT market, is more sensitive to the WPP's DA commitment q_{w2}^a when G_c ' ramp
544 constrain is binding than when the constrain is no binding. Actually, one unit more DA
545 commitment from the WPP will inflate the RT price by β_g when G_c 's ramp constrain is
546 binding rather than by $\frac{\beta_c}{\beta_g + \beta_c} \beta_g$ when the ramp constrain is not binding. \square

547 According to Eq. B.1, the WPP's MCC in hour 2 is also determined by G_c 's ramp rate
548 r and the WPP's commitments in both two hours, and the extend of difference between
549 β_g and β_c . The ramp rate r and the WPP's commitments in two hours affect mcc_2 by

550 impacting the the probability of the situation that G_c ' ramp constrain is binding in the
 551 RT market. The probability is high if r is small. q_{w1}^a can affect the probability only if
 552 $q_{w2}^a < RTP_2$. In this scenario, the smaller the q_{w1}^a , the higher the provability that G_1 's
 553 ramp constrain is binding in the RT market. In contrast, q_{w2}^a can affect the probability
 554 only if $q_{w2}^a \geq RTP_2$. In this scenario, the smaller the q_{w2}^a , the lower the provability that
 555 G_1 's ramp constrain is binding in the RT market. I summarize the analyses in the following
 556 theorem.

557 Appendix C. Proofs of Theorems

558 Proof of Theorem 4.3

Proof. When $q_{w2}^{a*} < RTP_2$, the WPP's market-power-profit-maximization commitment level q_{w2}^{a*} is solved from

$$p_j^a(q_{w2}^{a*})q_{w2}^{a*} + p_2^a(q_{w2}^{a*}) = mcc_2.$$

Therefore, q_{w2}^{a*} can be represented by $q_{w2}^{a*}(p_2^a, p_2^a, mcc_2)$. In particular, when $q_{w2}^{a*} < RTP_2$, GenCos' ramp rates affect p_2^a by determining the market price p_2^a when the WPP commit to provide zero MWhs in hour j . Therefore, the change of q_{w2}^{a*} caused by a change of r can be expressed as

$$\frac{dq_{w2}^{a*}}{dr} = \frac{dq_{w2}^{a*}}{dp_2^a} \frac{dp_2^a}{d\phi_2} \frac{d\phi_2}{dr} + \frac{dq_{w2}^{a*}}{dmcc_2} \frac{dmcc_2}{dr} \Big|_{q_{w2}^{a*} < RTP_2}. \quad (C.1)$$

Symmetrically, \hat{q}_{w2}^a is solved from

$$p_2^a(q_{w2}^a) = E[p_2^a].$$

Therefore, \hat{q}_{w2}^a can be represented by $\hat{q}_{w2}^a(p_2^a, mcc_2)$ when $\hat{q}_{w2}^a > RTP_2$. Then, the change of \hat{q}_{w2}^a caused by a change of r can be expressed as

$$\frac{d\hat{q}_{w2}^a}{dr} = \frac{d\hat{q}_{w2}^a}{dmcc_2} \frac{dmcc_2}{dr} \Big|_{\hat{q}_{w2}^a > RTP_2}. \quad (C.2)$$

Similarly, the change of RTP_2 caused by a change of r can be expressed as

$$\frac{dRTP_2}{dr} = \frac{dRTP_2}{d\phi_2} \frac{d\phi_2}{dr}. \quad (C.3)$$

559 Then, if $\frac{d(RPT_j - q_{wj}^{a*})}{dr}$ is greater than $\frac{d(\hat{q}_{w2}^a - RTP_j)}{dr}$ when G_c 's ramp rate decreases by
 560 dr , the ramp-rate decrease enlarges the vale of η as well as enhances the WPP's ATMP.
 561 Because r 's change only affect RIP_2 when $q_{w2}^a < RTP_2$, the change of r impacts q_{wj}^{a*}
 562 by increasing both RIP_2 and mcc_2 curve. The increases of RIP_2 and mcc_2 curve have
 563 opposite effects on q_{wj}^{a*}). In contrast, the change of r impacts \hat{q}_{w2}^a 's only by steeping mcc_2 .
 564 Furthermore, the piece of mcc_2 in between $q_{w2}^a \in [0RTP_2]$ is steeper than the piece of
 565 $q_{w2}^a > RTP_2$. Therefore, $\frac{d(RPT_j - q_{wj}^{a*})}{dr}$ is always greater than $\frac{d(\hat{q}_{w2}^a - RTP_j)}{dr}$. Consequently,
 566 the ramp-rate decrease enlarges the vale of η as well as enhances the WPP's ATMP.

567 Following similar processes, I can demonstrate that L_1 's decrease and q_{w2}^a 's increase
 568 enlarge the vale of η as well as enhances the WPP's ATMP. □