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BEFORE THE PUBLIC SERVICE COMMISSION OF UTAH

In the Matter of the Application of Rocky
Mountain Power to Establish Export Credits
for Customer Generated Electricity

Docket No. 17-035-61 Phase 2

REVISED AFFIRMATIVE TESTIMONY OF MICHAEL MILLIGAN, PH.D.

ON BEHALF OF

VOTE SOLAR

~~March 3~~ May 8, 2020

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1 **I. INTRODUCTION**

2 **Q. Please state your name and business address.**

3 A. My name is Michael Milligan. My business address is 9584 W 89th Avenue,
4 Westminster, Colorado 80021.

5 **Q. On whose behalf are you submitting this revised direct testimony?**

6 A. I am submitting this revised testimony on behalf of Vote Solar.

7 **Q. By whom are you employed and in what capacity?**

8 A. I am principle consultant with Milligan Grid Solutions, Inc., an independent power
9 system consulting firm.

10 **Q. Please summarize your education and professional experience.**

11 A. I have a Ph.D. in Economics from the University of Colorado and a B.A. from
12 Albion College in Mathematics. My experience includes working in the power system
13 industry for about seven years. I then was Principal Researcher at the National Renewable
14 Energy Laboratory (“NREL”) for 25 years, where I authored/co-authored more than 225
15 technical reports, journal articles, and book chapters. I served on multiple technical
16 committees at the Western Electricity Coordinating Council (“WECC”) and the North
17 American Electric Reliability Corporation (“NERC”), which is the official reliability
18 regulator in the U.S., and I was a charter member of the IEEE Wind and Solar Coordinating
19 Committee. For many years, I served on the International Energy Agency Task 25 – Large-
20 scale Wind Integration – research team where I led multiple international research papers

21 on integrating wind into the power system. As an independent consultant, my clients have
22 included NERC, the Electric Power Research Institute, the Southwest Power Pool,
23 GridLab, and multiple trade and educational/research organizations. Exhibit 1-MM
24 provides a statement of my qualifications and experience.

25 **Q. Have you previously testified before the Utah Public Service Commission**
26 **(“PSC” or “Commission”)?**

27 A. No.

28 **II. PURPOSE OF TESTIMONY**

29 **Q. What is the purpose of your testimony in this proceeding?**

30 A. The purpose of my testimony is to quantify several aspects of the value of exported
31 customer generation (“CG”) on the Rocky Mountain Power (“RMP”) system to support
32 the overall valuation provided in the testimony of Vote Solar witness, Dr. Carolyn Berry.
33 Specifically, I will address three categories of value: (i) Avoided Energy Costs;
34 (ii) Avoided Generation Capacity Costs; and (iii) Avoided Emissions volume.

35 Currently, over 99% of customers with CG in RMP’s Utah service territory have
36 distributed generation (“DG”) solar.¹ As a result, I focus my analysis of the value of CG
37 exports on the characteristics of DG solar. It is my understanding that the compensation
38 mechanism for exported CG that will be approved in this case will take effect in 2021.

39 Accordingly, all of my results are provided [in](#) 2021 dollars and cover the 20-year period

¹ Vote Solar, [Revised Affirmative Testimony of Briana Kobor, Section IV, Sachu Constantine, lines 175–77.](#)

40 2021-2040. I focus on this time period because it provides an accurate representation of
41 the value of CG commensurate with the minimum expected lifetime of DG solar.

42 The *first* value that I quantify in my testimony is the value of avoided energy costs
43 associated with CG exports. When solar energy is generated at the customer's location and
44 exported to the grid, that energy flows to the nearest load sink and offsets the amount of
45 energy that needs to be provided by RMP. The reduction in electricity demand is matched
46 by either a reduction in electricity produced by RMP or in an additional unit of energy that
47 RMP can sell to a neighboring utility. My analysis examines the specific attributes of
48 exported CG in RMP's Utah service territory to derive a value for avoided energy costs
49 inclusive of avoided line losses.

50 The *second* value that I quantify is the avoided generation capacity cost. To the extent that
51 CG exports are produced during times of system peak, the required capacity that RMP
52 needs to acquire to serve its demand will decrease. In my testimony, I calculate the value
53 of this avoided capacity need that results from CG exports.

54 *Finally*, I calculate avoided carbon dioxide emissions ("CO₂") that result from CG exports
55 by examining the emissions profile of RMP's fossil-fuel powered fleet and how those
56 resources may change over time.

57 The results of each of my analyses were provided to Dr. Berry who then provided the total
58 assessment of the value of CG.

59 **III. SUMMARY OF RECOMMENDATIONS**

60 **Q. Please provide a brief summary of your recommendations.**

61 A. I recommend that the avoided cost results from my testimony be adopted in the
62 valuation of CG exports to inform just and reasonable compensation that captures the value
63 of this resource. I find that CG exports provide benefits to RMP in the form of avoided
64 energy costs, avoided line losses, avoided generation capacity costs, and avoided carbon
65 emissions.

66 As described in full detail in my testimony, I recommend the following values be
67 incorporated into the value of CG analysis:

- 68 1. An avoided energy value of ~~3.65~~3.55 c/kWh;
- 69 2. An avoided line loss value of 0.31 c/kWh;
- 70 3. An avoided generation capacity value of ~~1.60~~1.48 c/kWh; and
- 71 4. Avoided carbon emissions based on my annual projections that average to
72 ~~229,097~~232 tons/year.

73 My lack of comments on RMP's affirmative testimony filed on February 3, 2020 should
74 not be interpreted as acquiescence or agreement with RMP. I reserve the right to express
75 additional opinions, to amend or supplement the opinions in this testimony, or to provide
76 additional rationale for these opinions as additional documents are produced, and new facts
77 are introduced during discovery and trial. I also reserve the right to express additional
78 opinions in response to any opinions or testimony offered by other parties to this
79 proceeding.

80 IV. AVOIDED ENERGY COSTS

81 Q. What drives energy costs for a utility such as RMP?

82 A. The cost of energy is driven primarily by fuel costs and maintenance costs. These
83 costs are variable costs – they depend directly on the volume of electricity produced.
84 Demand for electricity varies significantly from hour to hour, day to day, and even year to
85 year. To ensure demand is met at all times, utilities, including RMP, utilize complex
86 software and telecommunications systems so that the electric energy can be supplied
87 reliably and at minimum cost. The computer software evaluates the suite of available
88 resources – including their energy cost, physical characteristics, and capabilities – and
89 develops a generation schedule for each resource that takes all of these items into account.
90 This is generally referred to as a “least-cost dispatch” (also called “economic dispatch”
91 because the solution is economic—the lowest-cost solution to meet demand) and is
92 standard operating practice in the power system industry.

93 Figure 1 shows how demand generally varies throughout the day, with a representation of
94 weekly demand on the lower side of the diagram.² At low levels of demand, the cost-
95 minimization algorithm chooses the resource with the lowest marginal cost,³ followed by
96 the resource with the next-lowest marginal cost if needed. The process generally occurs at
97 intervals of 5-minutes up to one hour throughout the day; PacifiCorp is a member of the
98 Western Energy Imbalance Market, which carries out this dispatch process every five

² See Michael Milligan, Erik Ela, Jeff Hein, Thomas Schneider, Gregory Brinkman, and Paul Denholm, *Renewable Electricity Futures Study Volume 4: Exploration Bulk Electric Power Systems: Operations and Transmission Planning*, National Renewable Energy Laboratory, v. 4, p. 25-1, June 2012, <https://www.nrel.gov/docs/fy12osti/52409-4.pdf>.

³ The resource with the lowest marginal cost is the cost of producing one additional MWh or GWh.

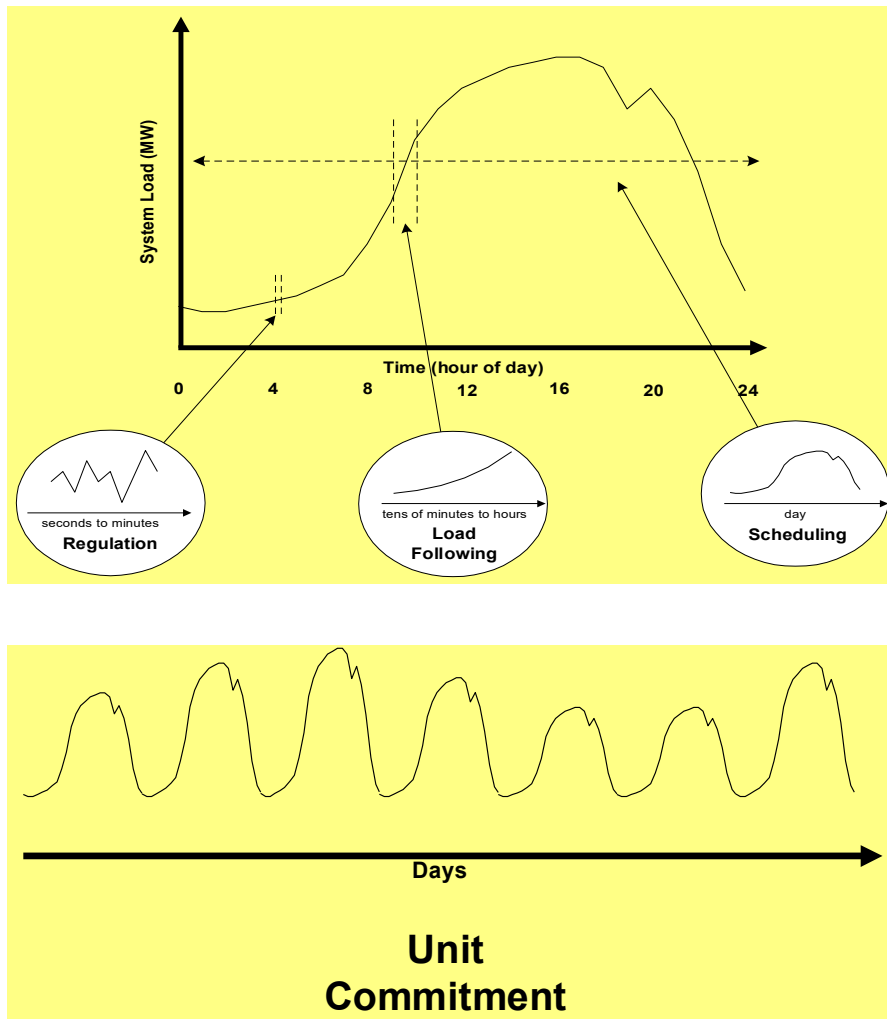
99 minutes.⁴ At the time of peak demand, if all low-marginal-cost resources are already being
100 dispatched, relatively expensive resources will need to be deployed. The economic dispatch
101 model will, however, minimize total production cost by avoiding the use of these expensive
102 resources whenever possible.⁵ Minimizing production cost is also a general requirement
103 imposed on utilities by state regulatory commissions.

104 The “least-cost” nature of the dispatch process guarantees that resources with zero or near-
105 zero marginal cost are chosen before a more expensive resource is chosen. Examples of
106 these low-cost resources include hydro power, wind power, and solar power. Although
107 there are exceptional times when these resources may be curtailed, they are utilized
108 whenever system balance and reliability can be maintained. Viewing Figure 1 in the
109 context of economic dispatch, if a new inexpensive resource becomes available, it will
110 displace the resource at the “top” of the stack as the displaced resource is the highest cost
111 resource.⁶

⁴ Western Energy Imbalance Market, *About*, 2020, <https://www.westerneim.com/Pages/About/default.aspx>.

⁵ For a short description of common power system models, see Erin Boyd, *Power Sector Modeling 101*, U.S. Department of Energy – Office of Energy Policy and Systems Analysis, https://www.energy.gov/sites/prod/files/2016/02/f30/EP_SA_Power_Sector_Modeling_FINAL_021816_0.pdf.

⁶ Michael Milligan, Erik Ela, Bri-Mathias Hodge, Brendan Kirby, Debra Lew, Charlton Clark, Jennifer DeCesaro, and Kevin Lynn, *Cost-Causation and Integration Cost Analysis for Variable Generation*, National Renewable Energy Laboratory, June 2011, <https://www.nrel.gov/docs/fy11osti/51860.pdf>; *see also, e.g.*, U.S. Energy Information Administration, *Electric generator dispatch depends on system demand and the relative cost of operation* (Aug. 17, 2012), <https://www.eia.gov/todayinenergy/detail.php?id=7590>.



112

113

Figure 1. Sample daily and weekly electricity demand

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The discussion above has particular relevance for the energy value of DG solar. Solar

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power, generated at a customer location on the distribution network, will reduce demand,

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which then reduces the need for electricity from the most expensive generator on the

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system. Thus, the value of the solar-generated energy is the cost of the energy from the

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most expensive resource. This concept is illustrated by a hypothetical supply curve from

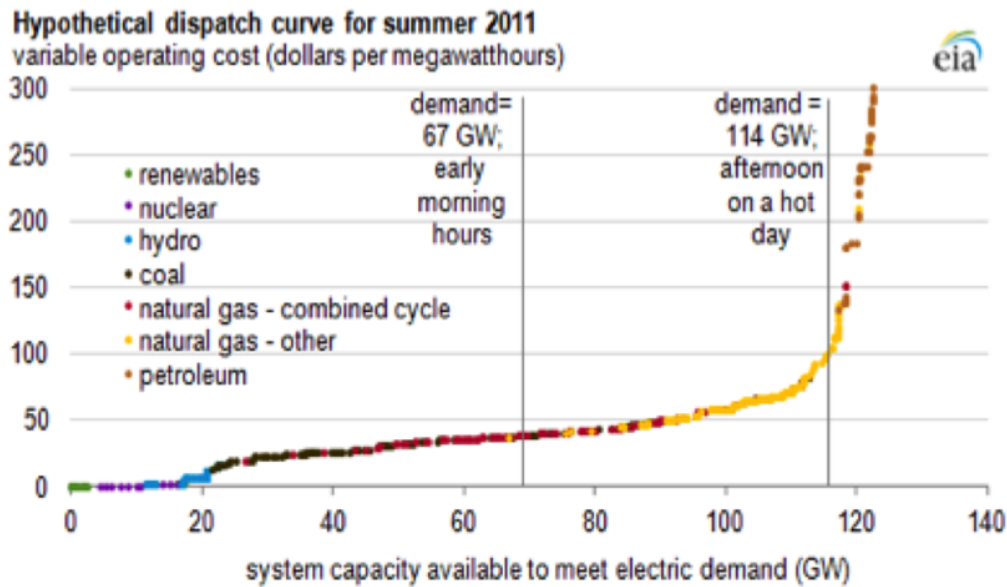
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the U.S. Energy Information Administration in Figure 2. The supply curve shows that

120

resources with low or zero marginal cost, such as solar energy, are dispatched first. As

121 demand increases, higher-cost resources are deployed. Conversely, as demand falls, the
122 operating cost falls as expensive units are taken off dispatch. Solar energy, whether
123 connected to the transmission system or to the distribution system, will displace relatively
124 high-cost resources on the supply curve.



125

126 **Figure 2. Example demand curve shows that renewables are dispatched first because of**
127 **their low marginal cost⁷**

128 The economic dispatch as described above does not consider constraints that may occur on
129 the grid that prevent the least-cost dispatch from being realized. For example, the least-cost
130 resource might be unavailable because it is behind a network constraint, which means there
131 is insufficient transmission or distribution capacity to deliver the energy to the load center.
132 In such situations, an alternative, more expensive resource may be needed to avoid the
133 congested path, which can increase the cost of energy.

⁷ U.S. Energy Information Administration, *Electric generator dispatch depends on system demand and the relative cost of operation* (Aug. 17, 2012), <https://www.eia.gov/todayinenergy/detail.php?id=7590>.

134 During extreme peak periods, the utility may need to deploy its most expensive resource
135 (highest marginal cost resource) to meet demand. In some cases, it may be necessary to
136 import electricity from a neighboring system during such a peak period. The value of
137 energy in such a transaction is determined by a combination of the marginal cost of each
138 of the two neighboring systems, along with potential network constraints. When a
139 transaction is carried out, the price paid is a direct indicator of the value to both the buyer
140 and seller.

141 In the U.S. Western Interconnection,⁸ there are several large substations that are often used
142 as electricity delivery points for market transactions between utilities. Utilities have well-
143 developed trading frameworks in place at these market hubs that facilitate common market
144 products.⁹ Many utilities that trade in these hubs have a desire to forecast future market
145 fundamentals—including detailed price forecasts for each trading hub covering time
146 periods from the present to the future—so that they can plan how best to invest in new
147 facilities to meet future demand. The trading price at any hub at a given time will depend
148 upon the utilities’ positions on their supply curve for the hour in question, the internal
149 resources that are available and at what price, and what bids from other resource-owners
150 are available, along with other information such as projected transmission congestion.

151 RMP participates in some of these markets, and it has developed a comprehensive price
152 forecast for its key trading hubs, hourly by hub, through 2040. This price forecast is useful

⁸ The Western Interconnection is described on the web site of the Western Electricity Coordinating Council (“WECC”). *The Western Interconnection*, Western Electricity Coordinating Council, <https://www.wecc.org/epubs/StateOfTheInterconnection/Pages/The-Western-Interconnection.aspx>.

⁹ See, e.g., S&P Global, *Methodology and specifications guide, North American Electricity*, p. 8, Feb. 2020, https://www.spglobal.com/platts/plattscontent/_assets/_files/en/our-methodology/methodology-specifications/na_power_method.pdf.

153 for evaluating avoided energy costs associated with exported CG in RMP’s Utah service
154 territory.

155 **Q. How can CG result in avoided energy costs?**

156 A. For every MWh of CG energy that is produced, the utility reduces its energy
157 delivery requirement to its end-use customers by one MWh. Because power plants are
158 operated according to cost-minimization principles, a low-cost or no-cost resource, such as
159 CG, will cause a reduction in power and energy production for the utility. Since the power
160 system must be in balance at all times according to rules set out by the NERC,¹⁰ if the
161 aggregate increase in CG is 100 MW over an hour, that energy will appear as a reduction
162 in demand on the distribution system and a commensurate amount of generation will be
163 turned down for the hour. Therefore, a power system operator for a utility like RMP must
164 decide which unit(s) to reduce so that there is no excess generation on the grid. In
165 ~~practicality~~[practice](#), these decisions will be made by sophisticated software tools that are
166 well equipped to handle fluctuations in demand (whether resulting from CG exports or
167 other customer activities).

168 Alternatively, if there is an offer from another utility to purchase the excess energy made
169 available by CG exports, then another resource in a different utility control area must
170 reduce its power for the hour so that it can import the extra power and energy. In this

¹⁰ NERC balancing standards are described in *Standard BAL-001-2 – Real Power Balancing Control Performance*, North American Electric Reliability Corporation, <https://www.nerc.com/pa/Stand/Reliability%20Standards/BAL-001-2.pdf>. NERC’s “Balancing and Frequency Control” technical document describes other aspects of the balancing standard including the equation for Area Control Error (“ACE”) that shows the relationship between the level of generation, demand, frequency, imports, and exports. NERC Resources Subcommittee, *Balancing and Frequency Control*, North American Electric Reliability Corporation, Jan. 26, 2011, <https://www.nerc.com/docs/oc/rs/NERC%20Balancing%20and%20Frequency%20Control%20040520111.pdf>.

171 scenario, RMP must therefore reduce its dispatch stack or find a buyer for the excess
172 energy. In the first case, the value of displaced energy depends on which unit(s) are reduced
173 and their marginal cost. If 100 MWh is displaced from a combined cycle gas unit with cost
174 of \$50/MWh, then the energy value of the CG is 100 MWh x \$50/MWh = \$5,000. In the
175 second case, the avoided energy value of CG is determined in the market when the power
176 is sold.

177 **Q. How did you determine the appropriate value of avoided energy costs**
178 **associated with exported CG?**

179 A. I based avoided energy costs associated with exported CG on the market price for
180 energy in RMP’s Utah service territory. I based assumptions regarding market pricing on
181 PacifiCorp’s Official Forward Price Curve (“OFPC”).¹¹ The OFPC is updated quarterly
182 and represents PacifiCorp’s “official quarterly outlook.”¹² It is developed based on market
183 forwards and fundamentals derived from a WECC-wide market model.¹³ At the time of
184 writing, the most recently available OFPC was developed September 30, 2019 and
185 provided to Vote Solar under the confidentiality agreement in this Proceeding.^{14,15}

¹¹ PacifiCorp is the parent company of RMP.

¹² *2019 Integrated Resources Plan*, PacifiCorp, Volume 1, p. 180, October 18, 2019, https://www.pacificorp.com/content/dam/pcorp/documents/en/pacificorp/energy/integrated-resource-plan/2019_IRP_Volume_I.pdf.

¹³ *Id.*

¹⁴ Exhibit 2-MM, 1909 OFPC (CY2018_2041)_2019 10 10 (min 0.01)###_GLOBAL_csv, RMP’s Responses to Vote Solar 9th Set Data Requests – Attach 9.2-1 (Feb. 6, 2020).

¹⁵ The OFPC does not include any assumptions for CO₂ pricing. *See supra* n.12 at 180. As described in more detail in Section VI, I developed a separate analysis of avoided emissions volume to support the avoided carbon costs discussed in the testimony of Dr. Berry.

186 **Q. Why is the market price a reasonable basis for the determination of avoided**
187 **energy costs?**

188 A. Market prices provide a measure of value, and the OFPC prices represent
189 PacifiCorp's official assessment of the state of the power system—and the value of
190 energy—at each hour and location from 2020-2040. At electricity trading hubs, any
191 participating utility can offer electricity for sale or bid to purchase electricity, and as its
192 needs change throughout the day, month, and year, a given utility may participate as a
193 buyer during some periods, and a seller at other periods. There may be times at which the
194 utility chooses not to participate.¹⁶

195 The decision of whether to buy or sell at the trading hub is primarily driven by economics
196 and public service commission oversight that requires cost-effective operation of the power
197 system, subject to reliability constraints. For example, in RMP's case, it may find that there
198 are times during which there is electricity for sale at a price that is lower than what it could
199 generate itself with its incumbent fleet of resources. At other times, RMP might find that
200 it has excess generation (energy) that it could sell in the market, obtaining a higher price
201 than the cost of generating that level of energy.

202 The other utilities that trade at these hubs face the same economic decision. Transactions
203 are therefore characterized when a buyer, who has determined that its cost to purchase is
204 less than its cost to generate (or purchase) from internal resources, finds it economical to
205 buy power. At the same time, the seller has determined that it can generate energy and sell

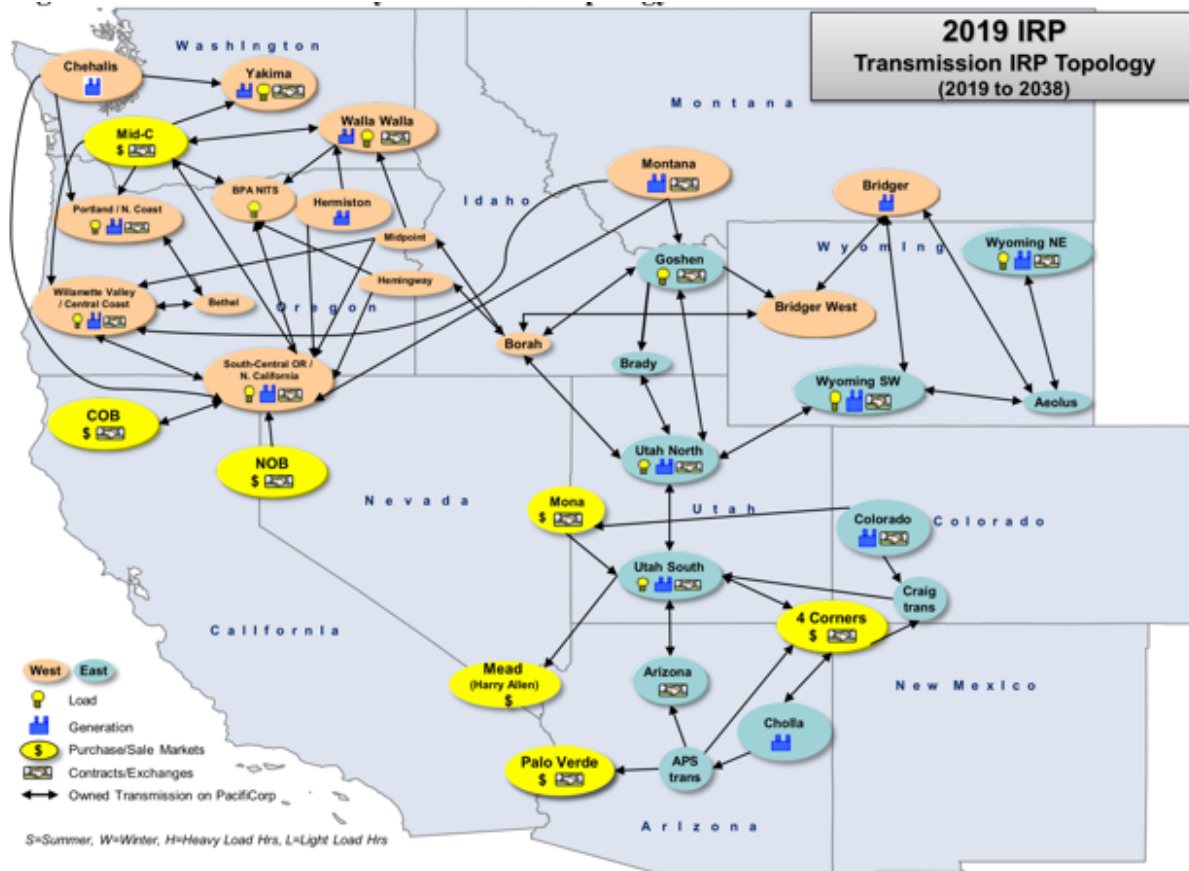
¹⁶ For an in-depth discussion of electricity markets, see Daniel S. Kirschen and Goran Strbac, *Fundamentals on Power System Economics* (2d ed. 2018).

206 that energy at a cost that exceeds its marginal cost. The decisions of both buyers and sellers
207 are driven by the relationship between their internal marginal cost and market prices.
208 Specifically, sellers find that their marginal cost is less than the market price, and buyers
209 find that their marginal cost exceeds the market price. Prices at electricity trading hubs
210 represent the characteristics of operational cost for electricity that intersect when a
211 transaction is executed. If RMP were to find that it must commit an additional generating
212 unit to meet demand, but instead realizes that it can buy energy at a lower cost, the value
213 of these additional MWhs is determined by the combined marginal cost curves of the buyer
214 and seller.

215 The market prices that comprise the OFPC represent RMP's best effort to value electricity
216 in the trading region. As described below, the OFPC accurately reflects relative scarcity
217 accompanying high levels of demand. During these periods, the value of electricity is high
218 because relatively high-cost resources must be dispatched to help meet demand.
219 Neighboring systems may themselves be in similar situation, which is captured in the price
220 forecasts. During off-peak times, the OFPC prices are lower, reflecting the relative lack of
221 value of energy during those times. An economically rational utility will always compare
222 its own internal cost of energy against the price it could pay in the market. If its own internal
223 cost is high at the same time that a neighboring system is selling at a low price, then the
224 economically rational utility will enter into a transaction that allows it to purchase energy
225 at a price less than its own internal cost. In ~~practicality~~[practice](#), utilities may not always be
226 able to act in an economically rational way as operational or system constraints may impact
227 their ability to trade in the market. Thus, market prices provide a conservative estimate of
228 the value of CG exports.

229 **Q. Please describe the market hub data included in the OFPC as it relates to**
230 **RMP’s service territory.**

231 A. The OFPC data that PacifiCorp developed represents hourly electricity prices at
232 several relevant trading hubs, from 2020-2040. Because RMP is in the PacifiCorp East
233 Balancing Area (“PACE”), several of these trading hubs that are connected to the
234 PacifiCorp West Balancing Area (“PACW”) may not be directly accessible to RMP. This
235 can be seen on a map that was taken from the 2019 Integrated Resource Plan (“IRP”) of
236 PacifiCorp, shown below in Figure 3.



237

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Figure 3. Transmission Map with Trading Hubs from the 2019 PacifiCorp IRP¹⁷

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The map shows the bifurcation of the PacifiCorp system into two control areas (PACW and PACE)—otherwise known as balancing areas (“BA”). This bifurcation makes it difficult to move energy between the two BAs when there is insufficient transmission through the path at Borah. For the purposes of my analysis, I assumed that it is never possible to move energy from PACW to PACE, or conversely from PACE to PACW. This represents a conservative assumption because PacifiCorp can often move energy between control areas.

¹⁷ 2019 Integrated Resources Plan, PacifiCorp, Volume II, p. 279, Oct. 18, 2019, https://www.pacifiCorp.com/content/dam/pcorp/documents/en/pacifiCorp/energy/integrated-resource-plan/2019_IRP_Volume_I.pdf.

246 Focusing on PACE, the map shows that Utah South has direct connections to trading hubs
247 at Mead, Mona, and Four Corners. Depending on transmission loading and economics,
248 Utah South may also have access to Palo Verde, and Utah North may, at times, have access
249 to transmission through Borah, which could then allow access to other trading hubs.
250 However, I assume that Utah South will primarily connect with its direct market hubs and
251 flows between Utah North and South are generally unimpeded.

252 **A. ANALYSIS OF OFPC**

253 **Q. How did you analyze the OFPC data?**

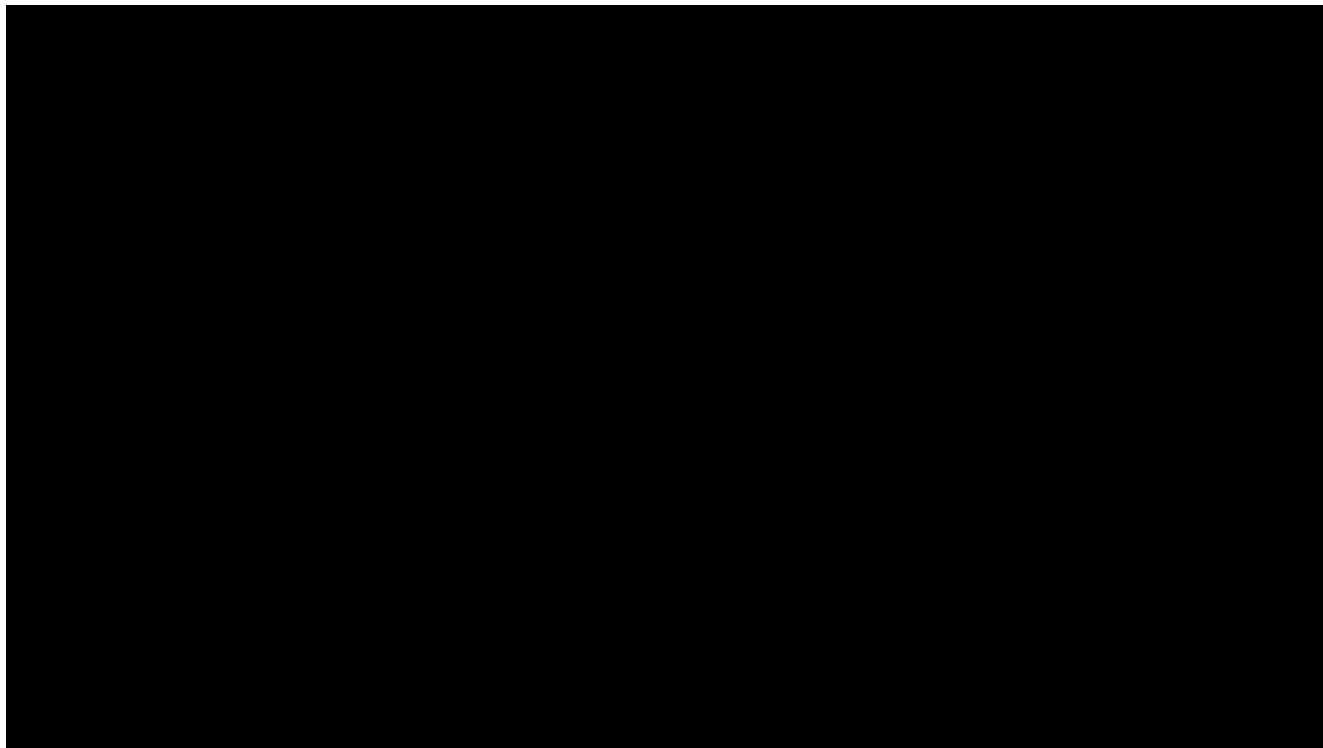
254 A. For my analysis I used the OFPC data for the Mead, Mona, and Four Corners
255 trading hubs. My analysis is divided into several sections:

- 256 1. Analysis of the OFPC data from the three trading hubs;
- 257 2. Combining the prices from all three hubs into a representative single market
258 price assessment;
- 259 3. Calculating the value of solar, using the Vote Solar Load Research Study
260 (“Vote Solar LRS”) data; and
- 261 4. Further analysis of the pricing and solar-value results as described below.

262 **Q. What do market prices look like at the three trading hubs based on the OFPC?**

263 A. In order to get a general sense of pricing at the three hubs that are directly connected
264 to RMP, I examined the OFPC data from 2021. This year was chosen because it is not far
265 out in the future, and as such, is more likely representative of relative prices than periods

266 that are far in the future, where forecasting errors are likely higher. The first step I took
267 was to compare prices from the three hubs by first calculating the average, maximum, and
268 standard deviation for each trading hub. The results appear in Figure 4. The graph indicates
269 that the maximum price, average prices, and standard deviations from Four Corners and
270 Mona are similar. Therefore, the two prices have similar variability properties and overall
271 range. Mead has a lower maximum price, a higher average, and lower standard deviation.
272 On balance, Mead has somewhat lower volatility, is higher on average, but has less range
273 than the other two hubs.



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Figure 4. Statistics for Three Trading Hubs, 2021¹⁸

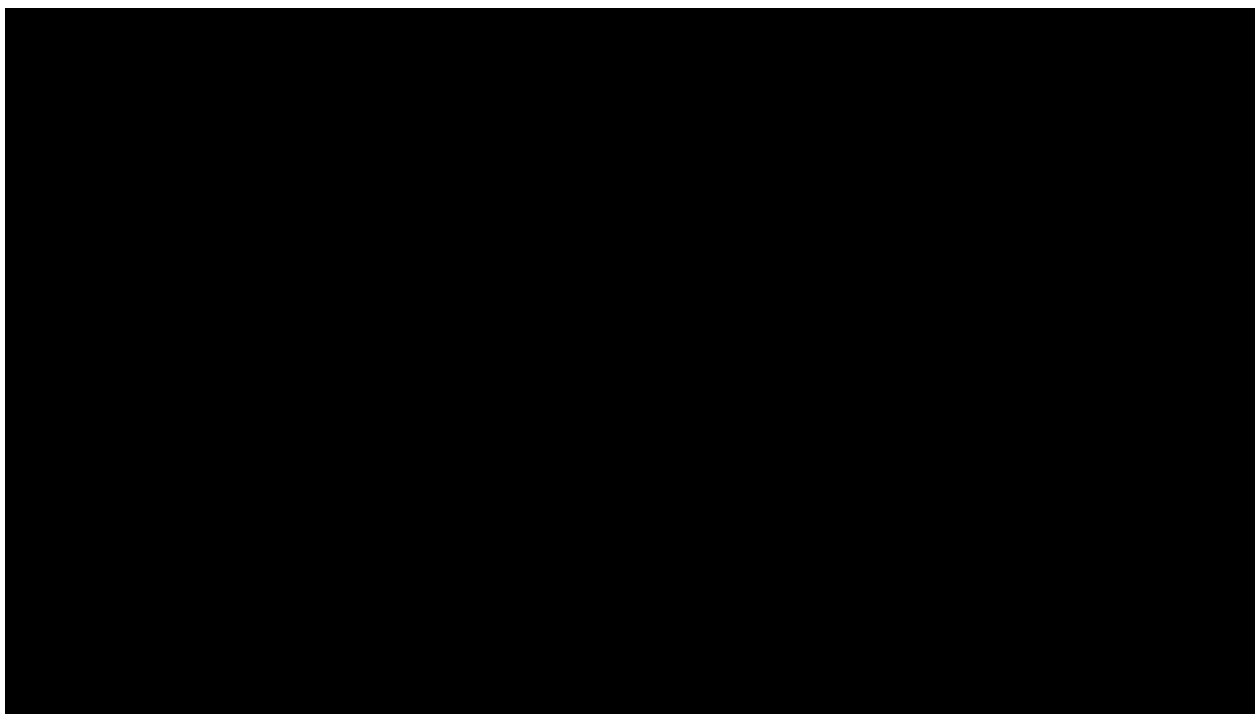
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Next, I examined average price on a monthly basis for each of the trading hubs. A graphical comparison as shown in Figure 5 demonstrates that there are two key patterns. The first is

¹⁸ See *supra* n.14.

278 illustrated by the common movement of Four Corners’ prices and Mona prices, which
279 follow a very similar pattern for the 12-month period. Mead’s prices are different;
280 significantly higher in winter months and somewhat lower in the peak months of July-
281 September.



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Figure 5. Average Monthly Prices¹⁹

284 I additionally calculated the average price for each hour of each month. This results in a
285 block of 12x24 price averages and shows the average price by hour and month for each
286 hub. The graphs appear in Appendix Fig. 1.²⁰ Each panel of the chart shows one of the
287 trading hubs, and each cell of the table shows the average price at the relevant location for
288 the month and hour that are indicated by the row label and the column header, respectively.
289 To help visualize the relatively large quantity of data in each of the three tables, a “heat

¹⁹ *Id.*

²⁰ *Id.*

290 map” colorization was applied so that the highest prices are shown with a red background,
291 moderate prices with a yellow background, and low prices with a blue background. Price
292 levels that lie between these ranges are given appropriate color mixes.

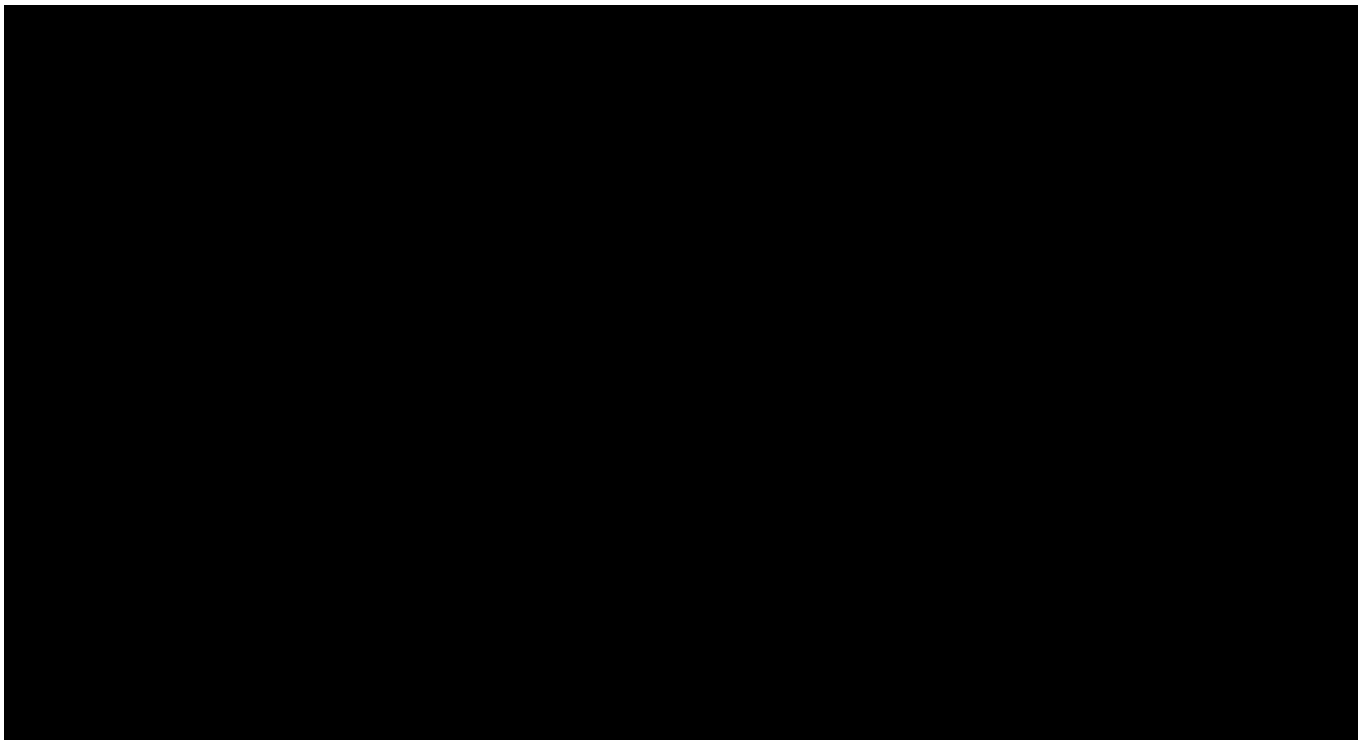
293 Although the absolute prices levels differ—especially those of Mead—from each other,
294 the overall price patterns are similar. Prices tend to peak in the late afternoon/early evening
295 of July, August, and September. In time periods surrounding these periods, prices are
296 moderate-to-high and are colored in some combination of orange and yellow. Moderately
297 low prices appear in early morning hours, especially in July, August, and September, with
298 more moderate prices appearing in the early morning hours of February, March, and
299 occasionally January.

300 **Q. How did you use this information?**

301 A. Because these three trading hubs are directly accessible to RMP without wheeling
302 energy across other systems, I averaged prices across the three hubs. The average price
303 table appears in Appendix Fig. 2. I then calculated the 12x24 price block to facilitate
304 comparison with Appendix Fig. 1.²¹ The combined prices exhibit similar properties
305 throughout the year, as can be seen by comparing Appendix Fig. 1 with Appendix Fig. 2.
306 Although the characteristics of Mead prices are somewhat different than the other two, the
307 overall shape of the 12x24 average prices are similar, and the average of the three hubs
308 does not show significant changes from the individual hubs.

²¹ *Id.*

309 To illustrate the wide range of prices, I developed a price duration curve as shown in Figure
310 6. Although it is difficult to ascertain precise point-values from the graph, it indicates that
311 the price exceeded \$ [REDACTED]/MWh approximately [REDACTED] times in the 8,760 hours in the year. The
312 price of \$ [REDACTED]/MWh was exceeded [REDACTED] times. From this information, I determined that the
313 overall price level is low—below about \$ [REDACTED]/MWh about half of the time, and higher than
314 \$ [REDACTED]/MWh about [REDACTED]% of the time. Very high prices—those above [REDACTED]/MWh—occurred
315 just over [REDACTED]% of the time.



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Figure 6. Price Duration Curve for 3-hub Average Price²²

²² *Id.*

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B. CALCULATING THE ENERGY VALUE OF CG EXPORTS

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Q. How did you use the OFPC to calculate the avoided energy costs associated with CG exports?

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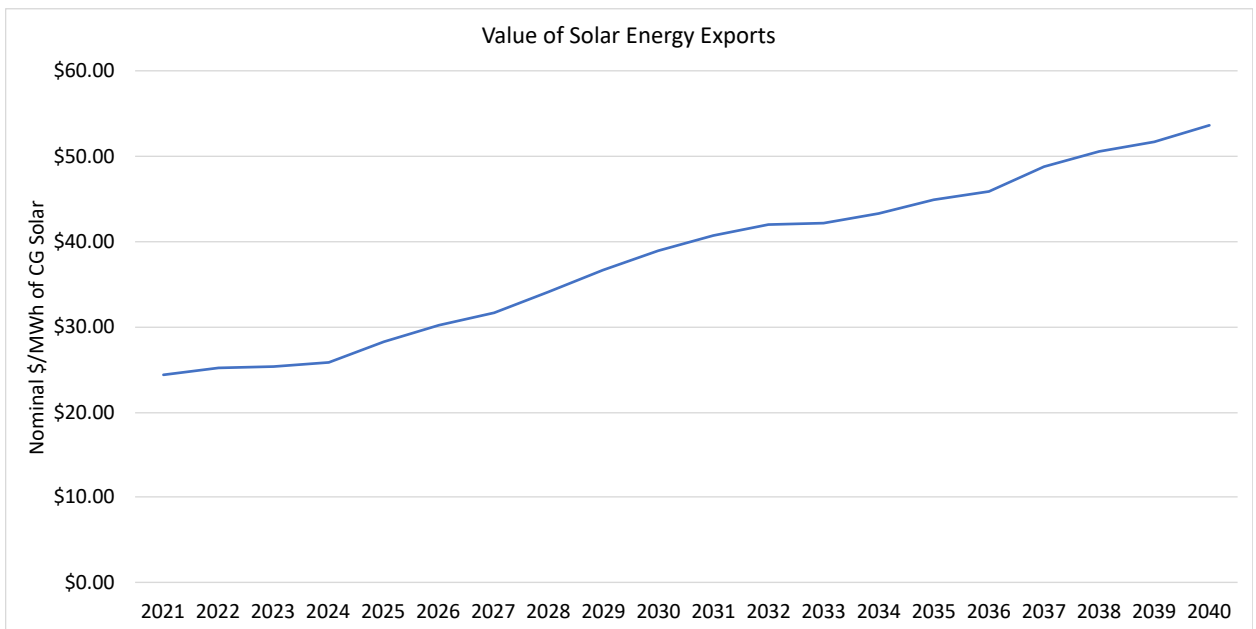
A. To evaluate the avoided energy costs associated with CG exports, I applied the 3-hub average hourly OFPC price for 2021-2040 to the shape of CG exports. I obtained the CG export shape from the Vote Solar LRS, as shown in Exhibit 1-~~AJL~~AJL-REVISED. Figure 7 below provides the resulting average avoided energy costs in nominal dollars for the duration of the study period.

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Figure 7. Value of Solar Energy

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The avoided energy costs are a direct function of the price levels at the time that CG is exporting power. There is some overlap of CG exports with high prices, but that overlap is offset somewhat because, even though solar generation is high during peak periods of the summer, it produces power at a lower rate later in the day during the highest-price periods.

329

330

331

332 This is shown in Appendix Fig. 3,²³ where the two temperature maps are stacked for ease
333 of comparison. The value of solar is highest just after the high-price periods during the
334 middle of the summer days.

335 **Q. What did you conclude regarding avoided energy costs associated with CG**
336 **exports?**

337 A. To produce an estimate of avoided energy costs, I applied an inflation rate of 2.28%
338 and a discount rate of 6.92% consistent with PacifiCorp's assumptions in its 2019 IRP.²⁴
339 The levelized value for the years 2021-2040 is ~~3.65~~3.55 cents/kWh in 2021 dollars.

340 **Q. Does your value of ~~3.65~~3.55 cents/kWh include avoided line loss costs?**

341 A. No. My value of ~~3.65~~3.55 cents/kWh represents the avoided energy costs associated
342 with CG exports at the location of the CG solar. For this amount of energy to be delivered
343 to the distribution system, an additional amount of energy would be required to be
344 generated and delivered to the customer's location. To capture the full avoided costs of
345 energy, it is necessary to include avoided lines losses as well. Using data provided Mr. Curt
346 Volkmann, I applied an energy loss percentage of 8.62% to calculate the additional avoided
347 energy loss value of 0.31 cents/kWh.

²³ *Id.*

²⁴ *Supra* n.12 at 179.

348 **V. AVOIDED GENERATION CAPACITY COSTS**

349 **Q. What drives generation capacity costs for a utility?**

350 A. Utility capacity costs are driven by the need to meet peak demand and satisfy
351 reserve requirements. Utilities need to have sufficient capacity to meet peak demand with
352 a planning reserve margin that protects against unforeseen operational problems or
353 inaccurate peak demand forecasts. Capacity is measured by the expected availability of the
354 resources on the system, after considering power plant performance characteristics that
355 include capacity ratings and availability, which can be impacted by unexpected (forced)
356 outages. In resource planning processes like [PacifiCorp's that of PacifiCorp](#) the utility
357 identifies future expected capacity needs and plans for additional resources as necessary.

358 **Q. How can CG exports allow a utility to avoid generation capacity costs?**

359 A. CG exports can allow a utility to avoid generation capacity costs if the exports
360 provide capacity support for the utility. Generally, this means that the exports are available
361 during peak periods. The extent to which CG can reduce capacity cost must be analyzed
362 with appropriate calculations, as discussed in detail below.

363 **Q. What information is required to evaluate the avoided generation capacity costs
364 associated with CG?**

365 A. There are two primary components required to evaluate the avoided capacity costs:
366 (1) an estimate of the capacity value of the CG resource, measured in MW and (2) a cost
367 of capacity on a \$/kW or \$/MW basis. I will address each in turn.

368 **A. RESOURCE CAPACITY VALUE**

369 **Q. What is a resource capacity value?**

370 A. The resource capacity value is a measure of the contribution of a resource to
371 planning reserves. The basic algorithms to estimate resource capacity value have existed
372 for many years, pre-dating the adoption of renewable resources. Every thermal plant has a
373 rated capacity expressed in MW and also has a non-zero forced outage rate (“FOR”). This
374 rate is a function of plant size and type, and generally older plants have higher FORs. The
375 capacity value of any resource measures how it contributes to the overall planning reserve
376 requirement, which is often expressed as a “loss-of-load” probability or mathematical
377 expectation (“LOLP” or “LOLE,” respectively). The most robust calculation of capacity
378 value is based upon these LOLP models, and it is common to have a resource adequacy
379 target of expected loss of load 1 day in every 10 years. The capacity contribution metric
380 recommended by the North American Electric Reliability Corporation (“NERC”) is the
381 effective load carrying capability (“ELCC”) metric or a similar variant that is based on
382 loss-of-load probability or related metric.²⁵ The ELCC metric calculates the additional
383 demand that can be served by the resource in question, returning long-term reliability to its
384 level prior to adding the new resource.

385 Depending on the system characteristics, a thermal plant with nameplate capacity of 100
386 MW and FOR of 0.10 will generally have an ELCC of 90-92% of its rated capacity. Older

²⁵ North American Electric Reliability Corporation, *Methods to Model and Calculate Capacity Contributions of Variable Generation for Resource Adequacy Planning*, Mar. 2011, <https://www.nerc.com/comm/PC/Integration%20of%20Variable%20Generation%20Task%20Force%20IVGT/Sub%20Teams/Probabilistic%20Techniques/IVGTF1-2.pdf>.

387 plants may have FORs ranging up to 0.15 or even 0.20, with corresponding ELCCs of
388 approximately 80-85% of rated capacity. Variable generation such as wind and solar have
389 lower capacity contributions because of the variable nature of the wind or solar fuel.

390 **Q. Please describe the ELCC Method.**

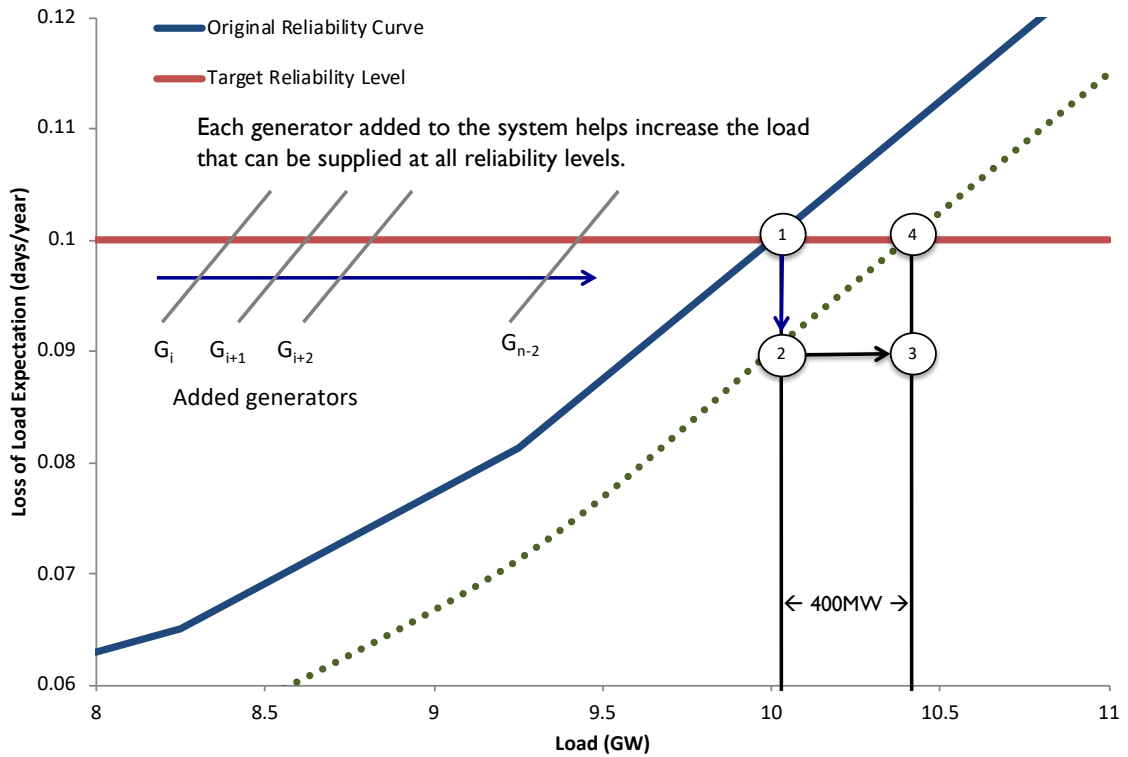
391 A. The ELCC method determines how much additional demand can be served by a
392 new resource, holding reliability constant. Reliability is defined as a metric such as LOLP,
393 LOLE, or expected unserved energy (“EUE,” where “expected” refers to a probabilistic
394 expected value), or similar metric. Much of the fundamental work on this was done by Roy
395 Billinton and Ronald Allen.²⁶

396 To use the ELCC method, a reliability target is chosen—a loss of load expectation of 0.1
397 day/year is common—and the base system is evaluated and adjusted so that it achieves the
398 target. The new resource, such as solar generation, is added to the resource mix and the
399 LOLE is recalculated. The new LOLE will have fallen because of the new resource. Then,
400 demand is incremented until the LOLE increases to its original target. The amount of
401 increased load that can be served while holding reliability constant is the ELCC of the new
402 resource. Figure 8 is an adaptation of a graphic from the NERC 2011 report that illustrates
403 the concept.²⁷ At point one, the system target of 0.1 day/year is achieved. A new resource
404 is added, which shifts the reliability curve down and to the right. The new reliability level

²⁶ Roy Billinton and Ronald N. Allan, *Reliability Evaluation of Power Systems* (2d. ed. 1996).

²⁷ North American Electric Reliability Corporation, *Methods to Model and Calculate Capacity Contributions of Variable Generation for Resource Adequacy Planning*, Mar. 2011, <https://www.nerc.com/comm/PC/Integration%20of%20Variable%20Generation%20Task%20Force%20IVGT/Sub%20Teams/Probabilistic%20Techniques/IVGTF1-2.pdf>.

405 is depicted by point 2, and it is approximately 0.09 days/year. This is more reliable than
 406 the target, and because reliability is expensive, we gradually increase demand until the
 407 target reliability level is reached again, traveling through point 3 and arriving at point 4.



Milligan and Ibanez presentation to WECC, June 2015

408

409

Figure 8. Graphical depiction of ELCC

410

Q. Is ELCC Method always the best approach?

411

A. Even though the ELCC is the preferred method, it is a relatively complex
 412 calculation, not particularly transparent, and requires significant data and computing
 413 capability. Because of these factors, there are several methods that simplify or approximate
 414 ELCC. Some of these simplified methods have been benchmarked against ELCC so their
 415 ability to approximate ELCC is well understood.

416 **Q. What information did you review to determine the best method to employ in**
417 **this case?**

418 A. In its 2019 IRP, PacifiCorp included a discussion and analysis of resource capacity
419 value. In preparation for my testimony, I reviewed the method employed by PacifiCorp as
420 well as the underlying data and analyses, which were provided in discovery.

421 **Q. What method did PacifiCorp employ to determine the resource capacity value**
422 **in its IRP?**

423 A. According to the IRP, PacifiCorp utilized a LOLP model to calculate hourly LOLP
424 value for its system by constructing a sample year with energy-not-served data (an output
425 from an LOLP model), from June-September of 2030, and October-May from 2036. RMP
426 then used these hourly LOLP values as weights, applied to the solar generation, to calculate
427 the capacity value.²⁸

428 **Q. Is PacifiCorp's IRP method reasonable for adoption in this case?**

429 A. No. The method used in the PacifiCorp IRP suffers from two related deficiencies:
430 (1) it is based upon a method that has been shown to be less accurate than other simplified
431 approximations to ELCC and (2) it is based solely on hourly LOLP values from two half-
432 years that are unlikely to represent periods of long-term risk, which is what LOLP methods
433 are intended to do.

²⁸ See *supra* n.12 at 397-405. Although wind and solar energy differ in some of their qualities, they both have variable and uncertain output, therefore it is reasonable to apply the results of the 1997 study to [solar](#).

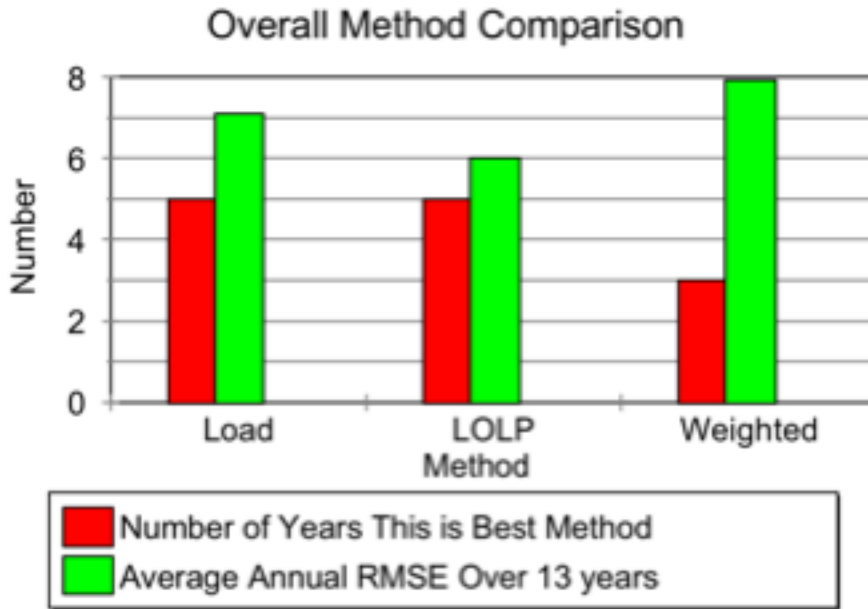
434 **Q. Please describe how the PacifiCorp method compares to other approximation**
435 **methods.**

436 A. The method used by PacifiCorp has not been sufficiently analyzed, nor has it been
437 validated as an accurate way to estimate ELCC. Interestingly, PacifiCorp’s analysis refers
438 to a paper that I co-authored in 1997. In the 1997 paper, my co-author and I compared 3
439 simplified ELCC approximation methods to the rigorous ELCC calculation itself: (1) top
440 load hours; (2) LOLP indicators; and (3) LOLP-weighted (the same method as used by
441 PacifiCorp). Figure 9 is reproduced from that paper.²⁹ The three methods were compared
442 to a full ELCC calculation for wind energy in two ways.³⁰ The *first* was a simple counting
443 metric: how many years, out of the 13 years studied did each method out-perform the other
444 methods in replicating ELCC? This comparison is shown in the red bars in Figure 8. A
445 higher score indicates better performance. The *second* comparison was based on the root
446 mean square error (“RMSE”) statistic,³¹ which was calculated for the 13-year period. This
447 comparison is shown in the green bars in Figure 8. For RMSE, a higher score is associated
448 with poorer performance.

²⁹ Michael Milligan and Brian Parsons, *A Comparison and Case Study of Capacity Credit Algorithms for Intermittent Generators*, <https://www.nrel.gov/docs/legosti/fy97/22591.pdf>.

³⁰ *Id.*

³¹ RMSE is a common metric that is used to calculate statistical errors. The calculation involves calculating the sum of the squares of the errors, and then taking the mean value of that sum.



449

450 **Figure 9. Comparison of capacity value methods, Milligan and Parsons, 1997**

451 The “weighted” method that PacifiCorp used to evaluate the capacity contribution of
 452 renewables is shown to be the worst-performing metric among the ones studied in my 1997
 453 paper. It outperformed the other methods in only 3 of the 13 years studied, and it had the
 454 highest RMSE of all the methods.

455 **Q. What additional elements should be considered in the selection of a method to**
 456 **approximate ELCC?**

457 A. There are other considerations beyond the accuracy of a method to match ELCC in
 458 a long-term planning study related to data availability. Because both demand and solar
 459 generation are clearly influenced by the weather, a single year of data will not represent
 460 the long-term performance of the resource. As described by NERC,³² multiple years of

³² North American Electric Reliability Corporation, *Methods to Model and Calculate Capacity Contributions of Variable Generation for Resource Adequacy Planning*, Mar. 2011,

461 demand and renewables data should be used to ascertain the long-term level of resource
462 adequacy that wind and solar energy can provide. Section 2.3 of the NERC report focuses
463 on inter-annual variability and shows how the capacity value is sensitive to annual
464 variations.

465 With a limited data set such as that used in the PacifiCorp IRP, a false sense of security
466 may be found in a precise calculation that is based on hourly LOLP values that are likely
467 to be quite different in other years. Generally, utilities are well aware of periods of time
468 when there are risks of insufficient resources, and methods such as the “Load” method in
469 my 1997 paper are the preferred approach for capturing long-term capacity value when
470 data is limited, as it is in this case. It is extremely unlikely that RMP’s hourly LOLP results
471 will ever be replicated in a different year. In the absence of multiple years of renewable
472 data, a more robust estimate of capacity value is necessary for long-term planning
473 documents, such as an IRP.

474 **Q. Do you recommend that RMP’s IRP method be used in the determination of**
475 **the resource capacity value associated with CG in this case?**

476 A. No. I don’t recommend the RMP method for any long-term planning use because
477 of the concerns described above: (1) it has not been carefully and successfully compared
478 to ELCC and long-term capacity value and (2) there are better methods.

479 **Q. What method do you recommend be adopted for determination of the resource**
480 **capacity value?**

481 A. Ideally, a full ELCC analysis would be conducted, taking into account multiple
482 years of time-coincident demand, because solar, wind, and hydro data are each driven by
483 weather, which rarely if ever repeats itself. I discussed this issue in a recent presentation,³³
484 and it is discussed in the NERC document.³⁴ Because PacifiCorp’s data does not allow for
485 the ideal calculation, the next-best approach must be adopted.

486 I recommend that resource capacity value be developed based on the “Load” method from
487 my 1997 paper.³⁵ This method is often called the “capacity factor” method and is similar
488 to a method utilized for many years by the PJM Regional Transmission Organization.³⁶
489 The capacity factor method utilizes a range of hours based upon the annual peak demand
490 to determine the performance of algorithms that use the top 1%, 2%, through to the top
491 30% of load hours. In my judgment, based upon decades of researching and publishing on
492 the subject, a capacity factor method that examines 10% of the top load hours will provide
493 an adequate approximation of ELCC. In situations where multiple years of demand and

³³ Michael Milligan, *Building the Power Grid of the Future: Resource Adequacy Issues*, Presentation to Minnesota Clean Energy Advocates, Oct. 25, 2019, <http://www.milligangridolutions.com/MCEA%20Symposium%202019.pdf>.

³⁴ North American Electric Reliability Corporation, *Methods to Model and Calculate Capacity Contributions of Variable Generation for Resource Adequacy Planning*, Mar. 2011, <https://www.nerc.com/comm/PC/Integration%20of%20Variable%20Generation%20Task%20Force%20IVGT/Sub%20Teams/Probabilistic%20Techniques/IVGTF1-2.pdf>.

³⁵ Michael Milligan and Brian Parsons, *A Comparison and Case Study of Capacity Credit Algorithms for Intermittent Generators*, National Renewable Energy Laboratory, Mar. 1997, <https://www.nrel.gov/docs/legosti/fy97/22591.pdf>.

³⁶ M. Milligan and K. Porter, *Determining the Capacity Value of Wind: An Updated Survey of Methods and Implementation*, National Renewable Energy Laboratory, p. 12–14, June 2008, <https://www.nrel.gov/docs/fy08osti/43433.pdf>.

494 production data are not available, such as this, the capacity factor method will provide a
495 more robust analysis for resource capacity value than PacifiCorp's LOLP-based method.

496 **Q. Can you elaborate why PacifiCorp's LOLP-based method is less robust?**

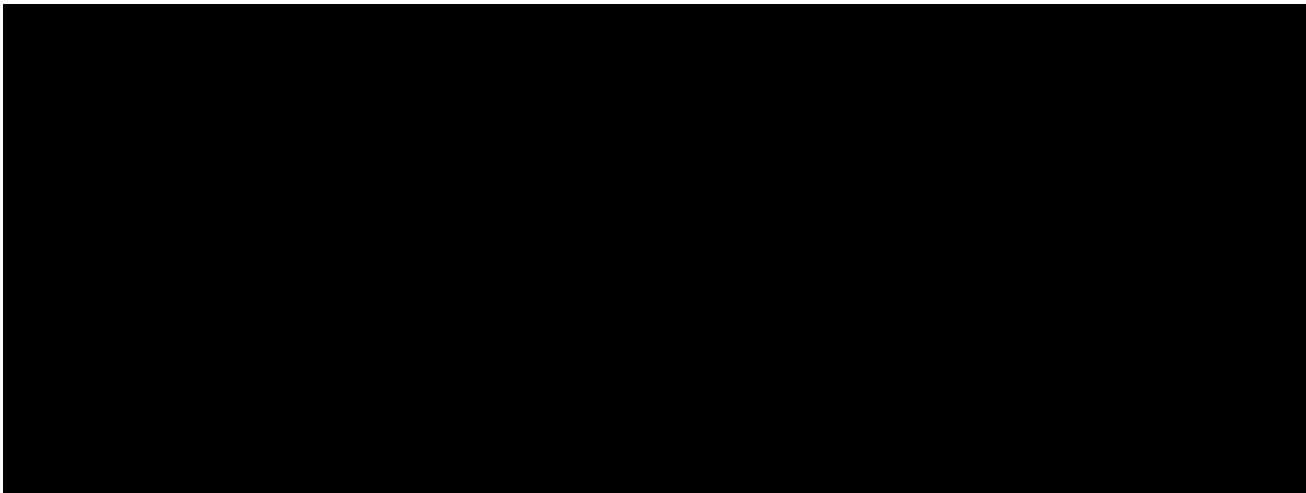
497 A. LOLP-based methods are dependent upon finding periods of relatively high risk.
498 The focus is the risk of having insufficient generation. Over a multiple-year period, weather
499 can vary substantially. Because weather is an important driver for demand patterns along
500 with renewable energy generation patterns, a period of LOLP risk in one year does not
501 necessarily translate to LOLP risk at the same time(s) in another year. Therefore,
502 employing a precise set of equations to a specific year of data may not capture the long-
503 term risk patterns. The risk of LOLP has generally been shown to occur during high-
504 demand periods, although other factors can contribute, such as maintenance schedules and
505 off-system imports and exports.³⁷

506 To illustrate, I used a 10-year subset of the load data provided by RMP in discovery³⁸ to
507 show periods of potential LOLP risk throughout the 10-year period starting with 2021. I
508 then calculated the MW demand of the top 10% of load-hours, or 876 hours of the year. I
509 tabulated the periods of time that demand is in the top 10% of load hours. This captures
510 potential time periods during which RMP could potentially be at LOLP risk. A graphical
511 depiction of the monthly results appears in Figure 10.

³⁷ North American Electric Reliability Corporation, *Methods to Model and Calculate Capacity Contributions of Variable Generation for Resource Adequacy Planning*, Mar. 2011, <https://www.nerc.com/comm/PC/Integration%20of%20Variable%20Generation%20Task%20Force%20IVGT/Sub%20Teams/Probabilistic%20Techniques/IVGTF1-2.pdf>.

³⁸ Exhibit 3-MM, 20190916 data disks/Confidential Attachments 6.10-1/Data Disk 2/Data Disk 2_CONF/Assumptions + Inputs/ Load,CONF/Base Loads (Net of PG).csv, RMP's Responses to Vote Solar 6th Set Data Requests – Attach 6.10-1 (Aug. 26, 2019).

512



513 **Figure 10. Periods of ~~high-demand~~high demand that could put system at LOLP risk**

514 As shown in the figure, periods of high demand occur primarily during the peak months of
515 July and August, but lesser peak periods can also occur in May, June, and September.
516 PacifiCorp relied on a single-year assessment of LOLP, but as demonstrated in Figure 10,
517 LOLP-risk periods can vary year-to-year. When inter-annual variations are considered
518 more fully, the periods of LOLP risk will change from year to year.³⁹ Therefore,
519 PacifiCorp’s method utilizing a precise calculation, but based upon limited data, is likely
520 to miss periods of LOLP risk in a long-term planning study.

521 **Q. What method do you recommend be adopted for determination of the resource**
522 **capacity value?**

523 A. For my assessment of the capacity value of solar for RMP, I used the capacity factor
524 method including the top 10% of load hours as an input. I then used the solar energy

³⁹ Other analyses have found a much larger impact when multiple years of actual data are considered. Consider this quote: “Analysis that was undertaken for the California Energy Commission found that during an unusually late, hot summer period when many units were taken out of service for scheduled maintenance, the hourly LOLP in late September was nearly as high as during the peak summer period. Situations like this can result in a lack of recognition of the exposure of the power supply to potentially high levels of risk....” *Supra* n.36 at 8–9.

525 production in those hours to calculate the capacity factor—the ratio of the mean to the
526 maximum. This method is preferred in instances, such as this, where data is limited and
527 has been shown to reliably approximate ELCC.⁴⁰

528 Using this method, I found the capacity value of CG exports averages ~~29.51~~27.65% of the
529 rated installed capacity. This was calculated for each of the years 2021-2030, and varied
530 slightly from year to year, from 26.91% to 28.53% ~~to 30.39%~~. For 2038-2040, when
531 demand data was unavailable, I used the average of 2021-2037, which is ~~29.51~~27.65% of
532 rated capacity.

533 **B. COST OF CAPACITY**

534 **Q. What is the cost of capacity in the context of avoided generation costs?**

535 **A.** The cost of capacity is typically evaluated based upon the cost of a peaking
536 resource, which could be a combustion turbine, aeroderivative generator, or reciprocating
537 engines. Regardless of the specific technology used, the capacity resource selection
538 attempts to isolate the contribution of capacity from the contribution of energy or other
539 ancillary services. Peaking plants typically have a low cost in terms of \$/kW, but often
540 have high energy costs. Due to these characteristics, peaking plants are commonly used in
541 the power system industry as proxy resources in the evaluation of capacity.

542 **Q. What information did you review to determine the appropriate cost of**
543 **capacity to employ in your analysis?**

⁴⁰ Michael Milligan and Brian Parsons, *A Comparison and Case Study of Capacity Credit Algorithms for Intermittent Generators*, National Renewable Energy Laboratory, Mar. 1997, <https://www.nrel.gov/docs/legosti/fy97/22591.pdf>.

544 A. I consulted the 2019 IRP, Table 6.1, which showed costs of options for RMP's
545 preferred portfolio of resources.⁴¹ Because my method calls for a capacity resource, I
546 selected a low-cost capacity resource based on its combined base capital cost and fixed
547 O&M cost.

548 **Q. Which resource do you recommend be considered in the analysis of avoided**
549 **generation capacity costs of CG?**

550 A. I selected the CCCT Dry "J/HA.01", DF, 2x1, ISO resource because it is a low-cost
551 capacity resource, consistent with a least-cost planning process and consistent with the
552 objective of calculating an avoided capacity cost for CG. The net capacity of this resource
553 is listed at 126 MW, with base capital cost of \$316/kW, fixed O&M of \$4.05/kW-yr, with
554 a 40-year design life.

555 **C. CALCULATION OF AVOIDED GENERATION CAPACITY COST**

556 **Q. How did you calculate the avoided generation capacity cost associated with**
557 **CG exports in RMP's Utah service territory?**

558 A. I employed several assumptions to derive a levelized c/kWh rate for avoided
559 generation capacity costs associated with CG exports based on the above-described
560 findings for resource capacity value and generation capacity cost. To conduct the
561 calculation, I adopted a 9.39% carrying charge to convert the capital cost to an annual
562 \$/kW-year.⁴² I applied my findings that solar exports have an average ~~29.51~~27.65%

⁴¹ *Supra* n.12 at Table 6.1.

⁴² I selected the 9.39% carrying charge based on PacifiCorp's 20-year Generation Annual Economic Carrying Charge from a 2018 Marginal Cost Study filed in California. PacifiCorp, *Exhibit Accompanying Direct Testimony of*

563 resource capacity value to the resulting annualized costs to derive the avoidable capacity
564 value. In order to incorporate the effect of avoided line losses, I adopted a line loss factor
565 of 1.09080 as recommended by Mr. Volkmann. Finally, I adopted PacifiCorp's inflation
566 rate of 2.28% and discount rate of 6.92% and find that the levelized value of the CG solar
567 is ~~\$16.00~~14.80/MWh of the solar resource, or ~~1.60~~1.48 cents/kWh.

568 VI. AVOIDED CARBON EMISSIONS

569 **Q. How does CG allow the utility to avoid carbon emissions?**

570 A. Because solar energy has a marginal cost of \$0/MWh, it is always preferred in the
571 dispatch stack, as discussed previously in this testimony in Section 0. The process of
572 determining the dispatch stack—the combination of resources' deployed capacity to meet
573 demand—is a complex economic optimization that chooses the resource mix based upon a
574 least-cost dispatch, subject to a variety of physical and reliability constraints. Because the
575 marginal cost of solar is lower than any thermal resource, it will always be fully deployed
576 unless some constraint on the power system requires an uneconomic dispatch, which raises
577 operational costs of the power system. Distributed solar is not subject to the usual economic
578 dispatch performed by power system operators, but it has the same impact on the dispatch
579 stack. For every MW of solar power at a given time, one less MW of thermal generation is
580 needed, and therefore less fuel burn is required to meet demand; thereby, reducing

581 emissions. The specific emission reduction is a complex calculation, and the level of
582 emission reduction per MWh of solar generation can vary based upon system conditions.

583 **Q. What source of information did you use to evaluate the avoided carbon**
584 **emissions associated with CG exports in RMP’s Utah service territory?**

585 A. To evaluate the avoided carbon emissions associated with CG exports, I developed
586 a blended emissions rate based on actual emissions data from the U.S. Environmental
587 Protection Agency (“EPA”) and the U.S. Energy Information Administration.⁴³ This data
588 includes the annual energy generation from each thermal power plant in PACE, measured
589 both in electrical energy (kWh) and fuel (BTU – British Thermal Units, a measure of the
590 energy content in fuels). The data set also contains emission information for several
591 pollutants, including carbon dioxide (CO₂). Data on power plant emissions is specific to
592 each power plant, and I utilized data from 2017 and 2018 to construct average emission
593 rates for RMP.⁴⁴ Before calculating the emissions associated with RMP, I pro-rated RMP’s
594 ownership rate for all jointly-owned plants so that my calculations would apply only to
595 RMP. Because the 2019 IRP contains many coal plant retirements, this average emission
596 rate will change as plants retire and can either result in the increase or the decrease of
597 average emissions per unit of energy, depending on the relationship of the retired plants’
598 emission rates compared to the average. To account for this, I relied upon the Preferred

⁴³ *Power Plant Data Highlights*, U.S. Environmental Protection Agency, <https://www.epa.gov/airmarkets/power-plant-data-highlights>; *2017 National Emissions Inventory (NEI Data)*, U.S. Environmental Protection Agency (Feb. 2020), <https://www.epa.gov/air-emissions-inventories/2017-national-emissions-inventory-nei-data>.

⁴⁴ Emissions and production data from Naughton were not used because it appeared incorrect or incomplete in the EPA/EIA data.

599 Portfolio P-45CNW as described in the 2019 IRP.⁴⁵ Each year in which there was a thermal
600 plant retirement (or multiple retirements), I recalculated the emissions rate accordingly.

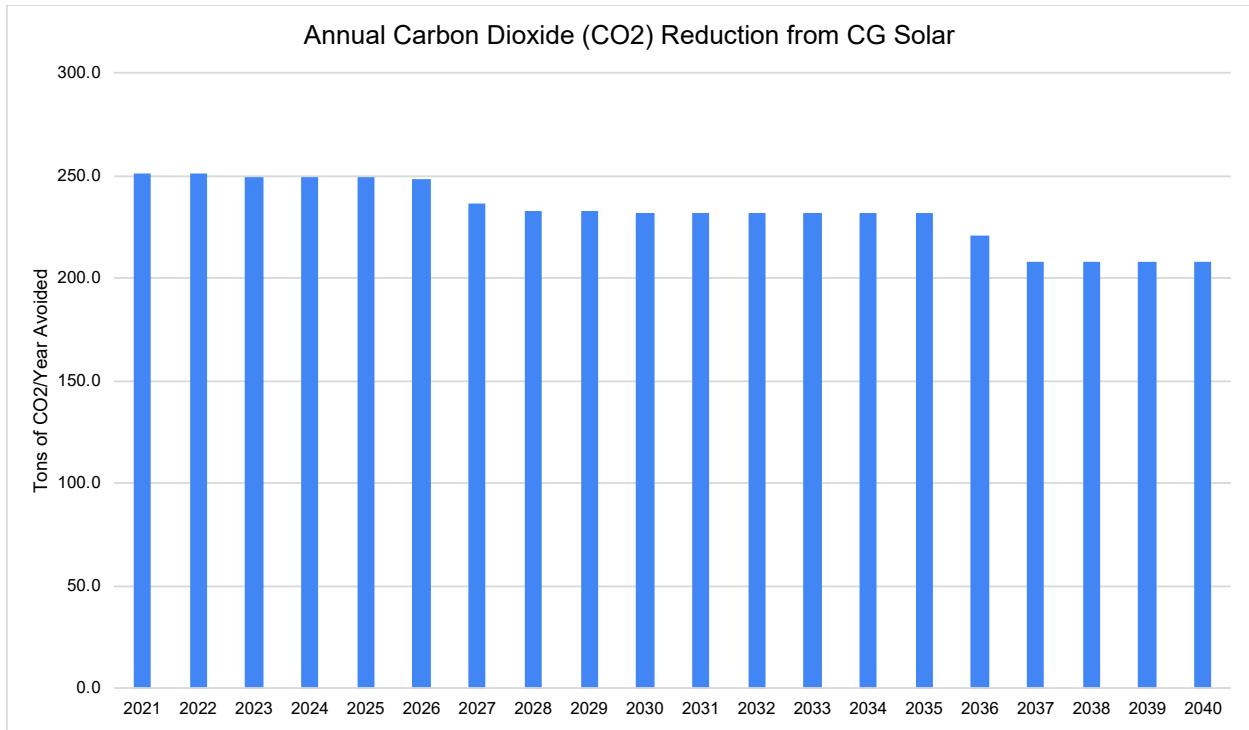
601 This analysis takes into account that CG exports will displace thermal generation and
602 incorporates coal retirements based on expectations set forth in the IRP. The only cases in
603 which CG exports would replace another renewable is if the system is not planned for
604 flexibility. Renewable curtailment generally occurs because thermal units are not flexible
605 enough to respond to changing net demand. PacifiCorp has demonstrated a keen interest,
606 backed by significant analysis, to ensure their system will have the needed flexibility.⁴⁶

607 **Q. How did you calculate avoided carbon emissions reductions?**

608 A. Using the average carbon emission rate for each year, I calculated carbon emission
609 reductions based the CG export profile provided by Dr. Lee in Exhibit 1-~~AJL~~AJL-
610 REVISED. Because of losses, one MWh of energy that is generated on the distribution
611 system requires more than one MWh of energy generated at a central power plant, I utilized
612 the same energy loss factor of 8.86% to calculate avoided carbon as used to calculate the
613 avoided energy in my earlier testimony based on the recommendation of Mr. Volkmann.
614 The annual results for the CG-related reductions in CO₂, are shown in Figure 11 where one
615 can discern a downward trend in the annual emissions reduction, caused by the relatively
616 early retirements of high-emission units. This causes the remaining operating thermal fleet
617 to have lower emissions than before, reducing the emission benefits from solar energy.

⁴⁵ *Supra* n.12 at 279.

⁴⁶ For example, Appendix F of the 2019 IRP is a study of flexibility. *See supra* n.12 at 77.



618

619

Figure 11. Emission reductions

620

Q. How was this information used in Vote Solar’s Value of CG analysis?

621

A. I provided Dr. Berry with data on the annual carbon emissions avoided by CG over the study period 2021-2040. It is my understanding that Dr. Berry has used this information to derive values associated with avoided carbon emissions.

622

623

624 **VII. SUMMARY OF RECOMMENDATIONS**

625

Q. Please summarize your recommendations.

626

A. Based upon my findings, I recommend that the following levelized avoided costs be adopted in this proceeding:

627

628

1. An avoided energy value of ~~3.65~~3.55 c/kWh;

- 629 2. An avoided line loss value of 0.31 c/kWh;
- 630 3. An avoided generation capacity value of ~~1.60~~1.48 c/kWh; and
- 631 4. Avoided carbon emissions based on my annual projections that average to
- 632 ~~229,097~~232 tons/year.

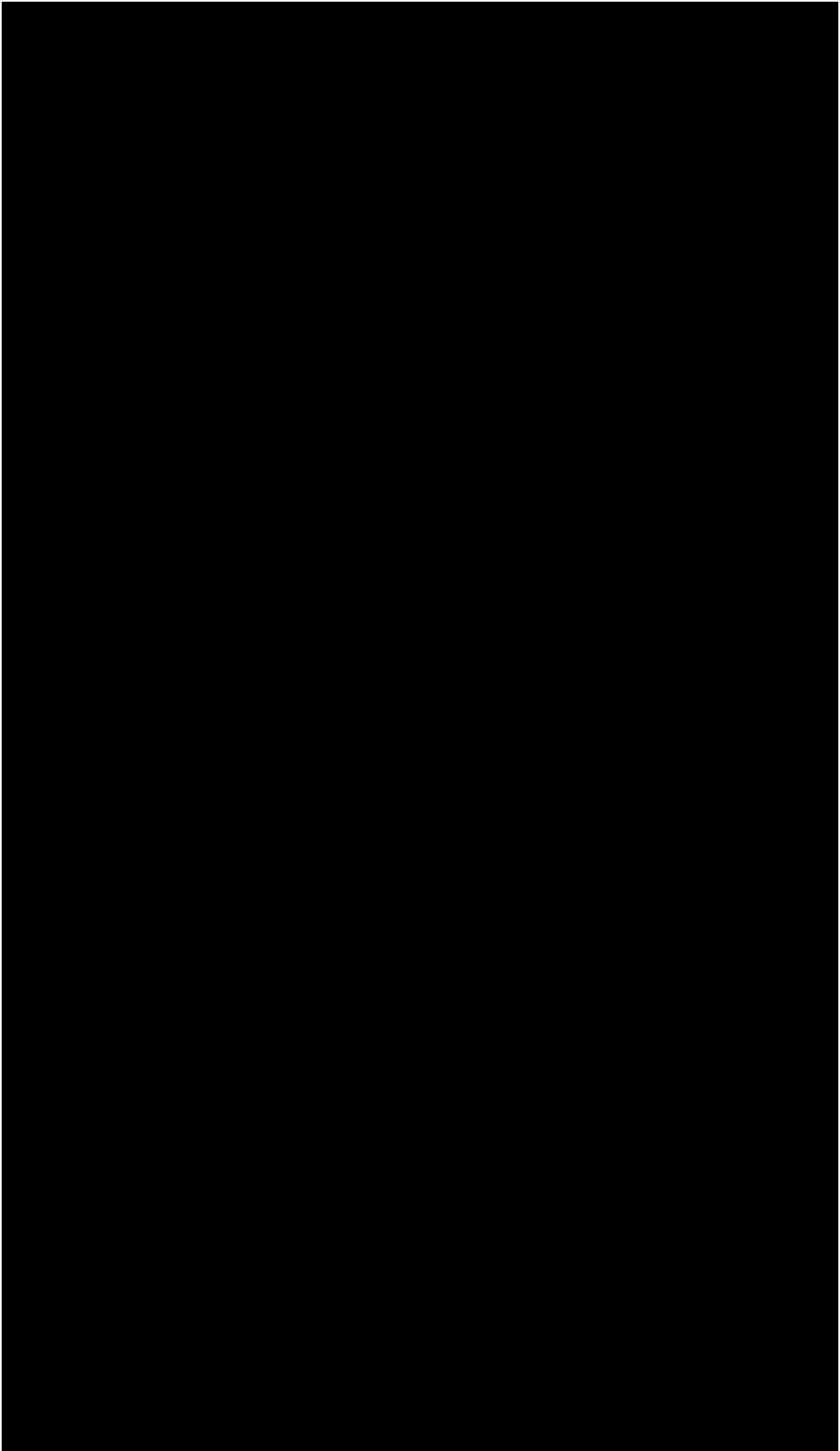
633 **Q. Does this conclude your revised testimony?**

634 A. Yes.

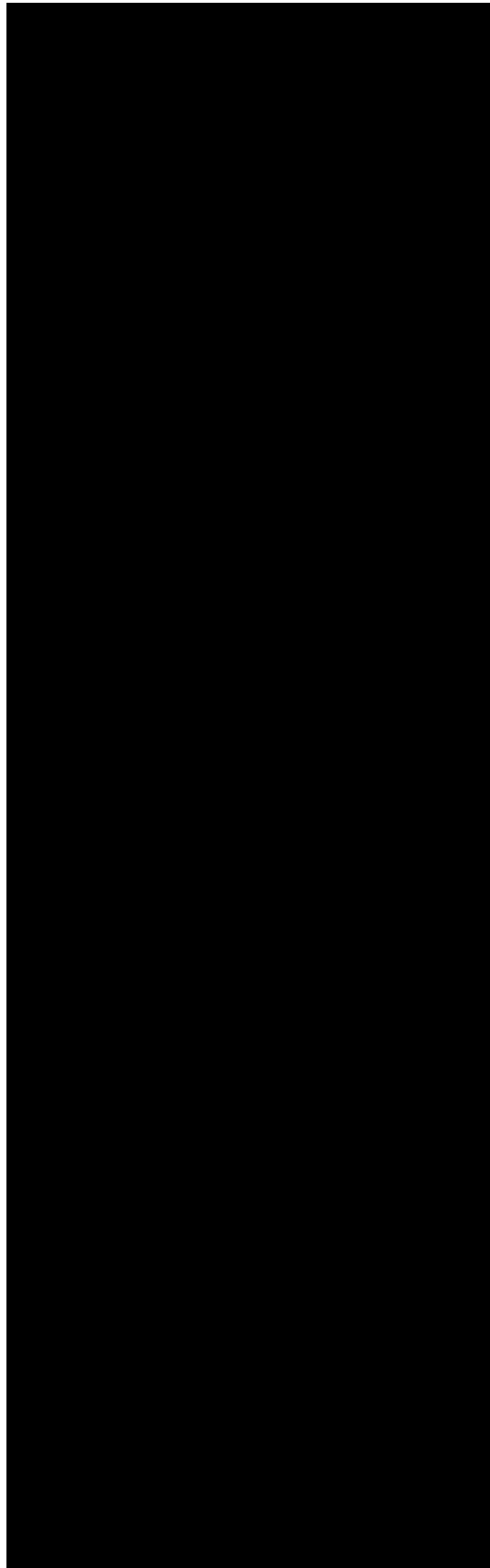
635 **VIII. APPENDICES**

636 Appendices to follow on next page.

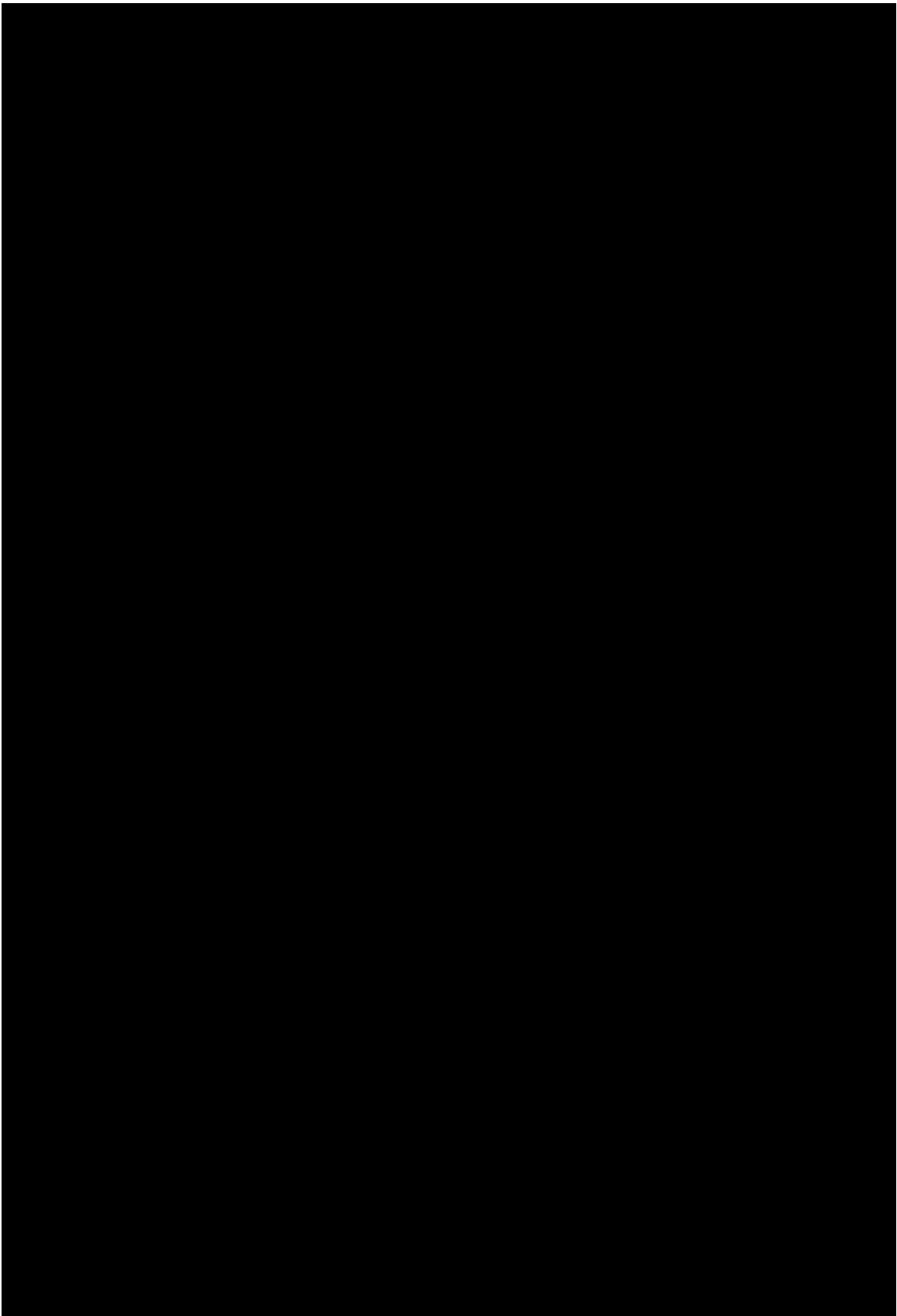
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Appendix Fig. 1. Comparison of 12x24 price blocks



Appendix Fig. 2. 12x24 Average Prices Across Three Hubs



642

643

Appendix Fig. 3. 12x24 Value of Solar Energy and Average 3-hub Prices

CERTIFICATE OF SERVICE

I hereby certify that on this ~~3rd~~^{8th} day of ~~March~~^{May}, 2020 a true and correct copy of the foregoing was served by email upon the following:

DIVISION OF PUBLIC UTILITIES:

Chris Parker	chrisparker@utah.gov
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