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January 14, 2013

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Oregon Public Utility Commission
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**Re: UM 1182 Phase II
In the Matter of PORTLAND GENERAL ELECTRIC COMPANY'S
REPLY TESTIMONY**

Attention Filing Center:

Enclosed for filing in the above-captioned docket, please find the following:

Original and five copies of Reply Testimony of:

- **Darrington Outama, Ty Bettis, Jaisen Mody and Patrick G. Hager (PGE / 200)**
- **Jonathan Jacobs (PGE / 300)**

Three copies on CD of:

- **Work Papers (non-confidential)**

These documents are being filed electronically with the Filing Center. Hard copies will be sent via US Mail. An extra copy of this cover letter is enclosed. Please date stamp the extra copy and return it to me in the envelope provided.

Sincerely,

Patrick G. Hager
Manager, Regulatory Affairs

PGH:kr

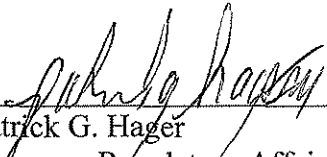
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CERTIFICATE OF SERVICE

I hereby certify that I have this day caused **PORTLAND GENERAL ELECTRIC COMPANY'S REPLY TESTIMONY (PGE / 200, 300)** to be served by electronic mail to those parties whose email addresses appear on the attached service list from OPUC Docket No. UM 1182.

DATED at Portland, Oregon, this 14th day of January, 2013.



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**BEFORE THE PUBLIC UTILITY COMMISSION
OF THE STATE OF OREGON**

UM 1182 Phase II

PORTLAND GENERAL ELECTRIC COMPANY

Reply Testimony and Exhibits of

Darrington Outama

Ty Bettis

Jaisen Mody

Patrick G. Hager



Portland General Electric

January 14, 2013

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I. Introduction

1 **Q. Please state your names and positions at Portland General Electric (PGE).**

2 A. My name is Darrington Outama. I am the Manager of Origination, Structuring, and
3 Strategic Analysis.

4 My name is Ty Bettis. I am the Manager of Merchant Transmission and Resource
5 Integration.

6 My name is Jaisen Mody. I am the General Manager of the Generation Projects
7 department.

8 My name is Patrick G. Hager. I am the Manager of Regulatory Affairs.
9 Our qualifications are provided in PGE Exhibit 100.

10 **Q. Please describe how your reply testimony is organized?**

11 A. Section II responds to NIPPC's direct testimony, with particular attention to problems
12 with NIPPC's adder approach (Section II.A), heat rate degradation (Section II.B), cost
13 under-runs and over-runs (Section II.C), capacity factors for wind resources (Section
14 II.D), and counterparty risk (Section II.E). Section III responds to Staff's direct
15 testimony.

II. Response to NIPPC's Direct Testimony

A. Problems with the "Adder" Approach

1 Q. From your reading of NIPPC's testimony, what problem is the adder approach
2 intended to address?

3 A. Our understanding is that adders are intended to correct for alleged flaws in the scoring of
4 benchmark bids. In this case, the alleged flaws are purported to favor the benchmark bids
5 by either understating the cost or overstating the performance for a benchmark bid.

6 Q. If the work of the Independent Evaluator (IE) results in accurately scored bids,
7 would bid adders improve the evaluation process?

8 A. No. The evaluation criteria used to score bids are developed to account for the benefits,
9 costs, and risks associated with each proposal. If a robust bid evaluation methodology is
10 utilized, there is no remaining need for adders. Improving the evaluation method and
11 process directly would be a superior approach to the application of rigid and probably
12 inaccurate bid adders.

13 Q. In practice, will the adder approach as suggested by NIPPC be effective in detecting
14 and correcting deficiencies in the scoring methodology?

15 A. No, the bid adders NIPPC proposes will likely introduce systematic flaws in the scoring
16 methodology that could lead to the selection of bids that reflect higher cost alternatives
17 for our customers. To obtain adders, NIPPC relies on historical data to establish both the
18 existence and magnitude of the alleged bias. The averages that NIPPC computes are
19 statistical estimates and are, therefore, subject to sampling errors. However, NIPPC
20 provides no measures of the accuracy (standard errors) of their estimates. Without
21 measures of accuracy, we cannot conclude with any confidence that bias even exists,

1 much less applies in the scoring of any particular benchmark resource or competitive bid.

2 This deficiency applies to each of the proposed adders NIPPC calculates for heat rate
3 degradation, construction cost over-runs, and wind capacity factor.

4 **Q. What are the problems with the data that NIPPC uses to calculate their adders?**

5 A. We discuss specific data problems in more detail in Section II.B, II.C and II.D as does
6 Dr. Jonathan Jacobs (PGE Exhibit 300), but in general NIPPC's estimated adders suffer
7 from one or more of the following data deficiencies:

- 8 • Small sample size (construction cost overruns);
- 9 • Non-representative samples (heat rate degradation, construction cost overruns, wind
10 capacity factors);
- 11 • Use of an irrelevant historical period (heat rate degradation, construction cost
12 overruns, wind capacity factors).

13 **Q. Are the plant observations that NIPPC uses in their empirical work representative
14 of PGE's experience or practice?**

15 A. No. Dr. Jacobs (PGE Exhibit 300) demonstrates that the data sets that NIPPC witness
16 Monsen relied upon are flawed and do not yield appropriate adjustments to the
17 methodology for scoring competitive bids. For wind capacity factors, we show in
18 Section II.D below that NIPPC's analysis uses data that are neither appropriate nor
19 representative of wind forecasting accuracy on a going forward basis.

20 **Q. Is there an inherent problem in NIPPC's use of historical data to adjust the scoring
21 of benchmark and competitive bids?**

22 A. Yes. To be useful in the evaluation of bids, data must reflect current (and expected)
23 technology, market conditions, and regulatory practice. A historical data series long

1 enough to produce a “reasonable” sample size is likely to extend back to a time period
2 that is not relevant if the forecasting methodology is changing.

3 **Q. Is it possible that the use of adders could provide an incentive for IPP's to increase**
4 **their PPA bids and thereby make the RFP process less competitive?**

5 A. Yes. In developing generation projects, IPPs and utilities essentially face the same set of
6 costs, challenges, and opportunities. If IPPs recognize that they are competing against a
7 benchmark resource that is subject to artificial cost adders, they would have an
8 opportunity and incentive to increase their bid price because the benchmark resource will
9 be handicapped by the adder and therefore less competitive. A less competitive bidding
10 process is not in the best interest of our customers.

11 **Q. What assumptions does NIPPC use to justify the use of an adder?**

12 A. NIPPC’s proposal is based on the presumption that a utility can understate its costs and
13 later “game” the rate-making process to recover its true costs. Such a presumption
14 incorrectly discounts the rigor and scrutiny of the ratemaking process in Oregon.

15 **Q. Are there other reasons for rejecting an adder approach?**

16 A. Yes, the adder approach would unduly complicate the scoring process. The role of the
17 Independent Evaluator (IE) is to ensure that bids are assessed accurately using a uniform
18 set of criteria. The adders have unknown accuracy and unknown variation, making the
19 assessment of the relative merits of alternative bids and resources more difficult, not less.

B. Heat Rate Degradation

20 **Q. Dr. Jacobs (PGE Exhibit 300) has concerns regarding NIPPC’s study on heat rate**
21 **degradation. Can you briefly summarize his concerns?**

1 A. Yes. Dr. Jacobs correctly points out that in order to determine whether or not a heat rate
2 degradation bias exists for benchmark bids “the analysis would have to use appropriate
3 data, correct as necessary for plant commitment and dispatch, and have a baseline similar
4 to the benchmark bids themselves” (PGE Exhibit 300, page 10 at 3-5). Dr. Jacobs goes
5 on to add “[t]he baseline for an analysis like NIPPC’s if the goal were to determine
6 whether the utility forecasts were biased, would be a hypothetical forecast constructed for
7 each plant using the utilities’ methods and contemporary manufacturer information”
8 (PGE Exhibit 300, page 11 at 21-23).

9 **Q. What is PGE’s position on heat rate degradation?**

10 A. We believe that the appropriate heat rate to utilize in scoring all bids, not just benchmark
11 bids, is a long-run degraded heat rate based on information provided by the turbine
12 manufacturers. As work papers stated in our direct testimony, PGE’s benchmark bids
13 already incorporate the long-run degraded heat rate. If a bid includes a heat rate that is
14 lower (better) than the long-run degraded heat rate, then the bid’s score should be
15 adjusted to reflect a long-run degraded heat rate.

C. Cost Under-Runs and Over-Runs

16 **Q. Do you agree with Dr. Jacob’s observation that the data that Mosen used in his**
17 **analysis does not satisfy the conditions needed to conduct an analysis for potential**
18 **bias?**

19 A. Yes. We also note that the PGE benchmark resources that were previously selected
20 through competitive bidding processes have been completed at less than expected cost.

D. Capacity Factors for Wind Resources

1 **Q. NIPPC advocates using an adder approach to correct wind capacity forecasting**
2 **errors. Is this approach valid?**

3 A. No. NIPPC derives its adder using historical wind data, which is based on forecasting
4 models that were developed when wind power was in its infancy in the United States. At
5 that time, extensive wind data were not available and models were thus developed using
6 small samples, which were the only data available. Over time as more data became
7 available, inefficiencies in wind data and modeling techniques were observed and
8 corrected for to produce more accurate and robust forecasts. Any reasonable forecaster
9 would consider their prior inaccuracies, incorporate additional data, and revise
10 forecasting models and methods to improve their forecasts over time.

11 In short, with time and greater experience, forecasting techniques and results
12 improved. Accordingly, using past forecasting errors to predict future forecasting
13 accuracy is a flawed approach because it assumes that forecasters will not learn and
14 improve over time which Monsen implicitly assumed. He provided no evidence to
15 support this faulty assumption.

16 **Q. Is there a need for a capacity factor adder for wind plants?**

17 A. No. Monsen's approach is to reduce the expected capacity factor for benchmark wind
18 plants. The logic for this reduction is based on research he performed on twelve wind
19 plants owned and operated by PacifiCorp. He claims that PacifiCorp's plants have
20 underperformed their expected capacity factors by a weighted average of 11.7%.

21 The Monsen study compared only a few years of operating data to the projected
22 capacity factor for the entire life of a project. Relying on a data set with a very limited

1 number of wind years and plants is not likely to lead to a robust and unbiased estimate of
2 capacity factors. Calculating and utilizing an adder based on a very limited amount of
3 data, especially for a resource where the fuel availability (wind) is recognized to vary
4 considerably from year to year, is likely to lead to an estimate with a high probability of
5 departing materially from the true value.

6 **Q. You mentioned that prior years' wind capacity factor forecasts are not a reliable**
7 **indicator of future forecasting accuracy. Would you please explain?**

8 A. Certainly. The wind industry in the U.S. is still relatively immature. Original plant
9 assessments that were being developed in the early 2000's for plants coming on-line in
10 2006 -- 2008 were using higher assumed plant availability numbers that were derived
11 from European wind plants. European wind plants were used, even with different turbine
12 technology and turbine concentration / dispersal than in the United States, because there
13 was no significant experience in the U.S. However, in the past 5 – 10 years, the U.S.
14 wind industry has made substantial increases in wind plant deployments, gained
15 operating experience, and made significant strides in improving wind capacity factor
16 assessments.

17 GL Garrad Hassan (GLGH), a leading company in the field of wind generation
18 assessments, has been very transparent regarding their experience over the past ten years.
19 The accuracy of their wind assessments has significantly improved due to major advances
20 in knowledge, experience, and technology. Monsen, however, implicitly assumes, by
21 using predictions derived from historical, obsolete forecasting techniques, that forecasters
22 will not take advantage of these and future improvements to generate more accurate
23 performance forecasts.

1 Given the increased deployments and operating experience of the U.S. wind fleet,
2 and the advances in forecasting methods and technologies over the past several years,
3 Monsen's suggested approach of reducing the expected capacity factor for benchmark
4 resources is significantly flawed, and if implemented would result in a less accurate bid
5 evaluation process.

6 **Q. Are there factors that Monsen failed to incorporate in his analysis of PacifiCorp's**
7 **capacity factor forecasts?**

8 A. Yes, Monsen's analysis and testimony excluded several factors that undermine his
9 conclusion.

10 (1) The first two years of over-estimates were calculated based on the
11 performance of a 41.1 MW wind plant in Southeastern Wyoming (Foote Creek) that was
12 put into service in 1999. This plant was PacifiCorp's first wind plant and was developed
13 when wind energy production was in its infancy in the U.S. Logically, this plant should
14 be removed from the data set as not representative.

15 (2) It appears that Monsen did not appropriately account for seasonality. For
16 example, he includes the addition of the 100.5 MW Leaning Juniper 1 Wind Farm in
17 2006 even though it did not reach full commercial operations until September 2006. It is
18 common knowledge that wind capacity factors vary by month and season, and that wind
19 farms in the lower Columbia River Gorge region exhibit higher capacity factors in the
20 late spring and summer while the fall and winter months generally exhibit lower wind
21 production. By utilizing 2006 output in the analysis for Leaning Juniper I, the annual
22 capacity factor results were skewed due to the inclusion of a partial year that was

1 represented by lower production months. He should have removed the 2006 data for
2 Leaning Juniper I.

3 This same data error applies to several of the data points Monsen relies upon. The
4 Marengo Wind Farm came online in August 2007, and in 2008, PacifiCorp added another
5 381 MW of wind, with 164 MW of this wind coming online during the summer of 2008
6 and another 217 MW in December 2008. All of these annual data points should be
7 removed from the analysis.

8 Given these infirmities and the limited data set used, the results of Monsen's
9 analysis are not reliable and should not be used.

10 **Q. Mr. Monsen uses data from different plants over different time periods in three**
11 **different wind regimes (Columbia River Gorge, Southeast Wyoming, and Eastern**
12 **Wyoming) to derive a single "lifetime" capacity factor adder. Is this a reasonable**
13 **methodology? (NIPPC Exhibit 100, pg. 30 at 5-9)?**

14 A. No. As we have already discussed, the data set has significant deficiencies, which will
15 lead to inaccurate and biased estimates.

16 **Q. What is a minimum reasonable sample size of annual capacity factor overestimates**
17 **to determine a "lifetime" capacity factor overestimate?**

18 A. This is difficult to say because the data must incorporate the weather oscillations that
19 drive the wind. The Pacific Decadal oscillation and inter-decadal oscillation (that are
20 often used for forecasting wind) show that climatic patterns can persist for 20-30 years
21 (Pacific Decadal oscillation) or 8-12 years (inter-decadal oscillation). To filter out time
22 series patterns such as these, one generally needs a dataset with a larger sample size than
23 the length of these known periodic signals. Stated differently, conducting an analysis to

1 determine whether capacity factor bias exists by using a few years of actual data for an
2 asset with a life of more than 30 years is unlikely to lead to accurate predictions of future
3 forecasting errors.

4 **Q. Has a solution been presented to mitigate the risk that a utility could overestimate**
5 **the projected wind capacity factor for benchmark resources?**

6 A. Yes. PacifiCorp states that after development of their shortlist, they retain “a qualified
7 and independent third-party technical expert (the Capacity Factor Expert) to assess the
8 expected wind resource capacity factor associated with each alternative on the initial
9 short list, including the cost-based utility ownership benchmark resource” (PAC
10 Exhibit 100, pg. 6 at 16-18).

11 **Q. Would PGE support this practice in their RFP scoring process?**

12 A. Yes. In fact, PGE is using an independent wind expert in our current renewable resource
13 RFP to address this risk and to ensure that the capacity factor estimates for all bids,
14 including the benchmark proposal, are independently reviewed.

E. Counterparty Risk

15 **Q. NIPPC has provided testimony from Collins regarding counterparty risk (NIPPC**
16 **Exhibit 200). Did this testimony address the relevant time period?**

17 A. No. The testimony focused on the period between 1992 and 1997, the years during
18 which Collins worked in the industry. During that period however, credit concerns and
19 risk evaluation methods were less developed in the wholesale electricity industry than is
20 true today.

21 **Q. How have credit concerns and risk evaluation methods changed since 1997?**

1 A. After the energy crisis of 2001, and the subsequent bankruptcies of several large energy
2 merchant and independent power producers (e.g., Enron, National Energy Group, NRG,
3 Mirant, Calpine and Dynegy), the financial world became more conservative regarding
4 the “credit worthiness” of a counterparty. This trend has continued as a result of the
5 more recent financial market crisis and subsequent “Great Recession”. Today, a
6 counterparty with less than investment-grade credit rating would not be accepted as a
7 trading partner for spot market power sales much less a long-term PPA. This has been
8 the wholesale energy industry standard for several years.

9 **Q. NIPPC criticizes the credit rating agencies. Are their arguments persuasive?**
10 **(NIPPC Exhibit 200, pg.8 at 21)**

11 A. No. PGE agrees that the rating agencies are implementing reforms, but these reforms are
12 not moving in the direction of relaxing credit requirements for project financing and
13 power plant owners. In fact, the reforms are heading the exact opposite direction: rating
14 agencies are looking past the plant owner to external factors in their assessment of risk.
15 For example, Standard & Poor’s *Principles of Credit Ratings* (see PGE Exhibit 201) state
16 that “[i]n addition to our assessment of an obligor’s stand-alone creditworthiness,
17 Standard & Poor’s analysis considers the likelihood and potential amount of external
18 support (or influence) that could enhance (or diminish) the obligor’s creditworthiness.”

19 In addition, debt ratings are more stringent now than when NIPPC’s witness
20 worked in the industry. Not only do rating agencies rate the debt that is directly on the
21 entity’s balance sheet, they impute additional debt depending on the length and contract
22 terms of long-term contracts that the entity has executed.

1 **Q. NIPPC claims that “no lender makes a commitment to a loan or can obtain a rating**
2 **for a loan structure when the PPA terms are still being negotiated” (NIPPC Exhibit**
3 **200, pg. 4 at 18). Is this relevant?**

4 A. No. It is the utility that is ultimately responsible for the delivery of reliable power to its
5 customers. Under a typical PPA, a utility would likely have limited recourse against a
6 seller in the event of contract default because the seller’s ownership structure is typically
7 a Special Purpose Entity (“SPE”). However, performance risk can be mitigated in part by
8 the seller / SPE obtaining financial assurances from interested parties (such as a bank or
9 parent company) to support the contract, and through negotiated contract rights such as
10 step-in provisions.

11 **Q. How does PGE assess credit and counterparty performance risk in its RFP**
12 **evaluation?**

13 A. In its RFP scoring process, PGE applies the following industry standard approach to
14 assessing credit and counterparty performance risk:

- 15 • Sellers must be investment grade in order to participate. As an alternative, they can
16 obtain parental guarantees or a Letter of Credit from a qualified financial institution
17 to alleviate this obligation.
- 18 • A seller with better credit rating and financial ratios will score better. This is to
19 recognize that, all else being equal, purchasing from a AAA-rated company is
20 inherently less risky than purchasing from a company with a more limited or poor
21 credit history.

22 **Q. Do you agree with Collins’ statement that “By raising credit quality as a concern,**
23 **but not proposing any RFP terms that level the playing field for the regulatory**

1 **compact’s relative financial security, utilities can inadvertently drive up the cost of**
2 **IPP’s signaling to IPP creditors how hostile the franchise is to them” (NIPPC**
3 **Exhibit 100, pg. 8 at 10-13)?**

4 A. No. In the case of limited liability entities such as SPEs, PGE’s RFP does include terms
5 that allow entities to demonstrate their credit worthiness. These mechanisms include
6 providing a credit support provider guarantee and posting collateral. In addition, the
7 ownership and capital structure of the project company associated with a PPA is a
8 business choice by the bidder. Limited liability ownership structures (SPE) are often
9 chosen by project developers in lieu of balance sheet financing in order to protect the rest
10 of the organization from specific project default risk and in some cases lower the amount
11 of equity funding required. However, such structures are inherently more risky as they
12 do not retain the financial strength and capability of a larger entity beyond the assets
13 within the project. Additionally, not all bids are based on an SPE ownership structure,
14 and in fact, some RFP sellers exhibit robust levels of credit worthiness that are
15 comparable to, or superior to, that of an investor-owned utility.

16 **Q. Do IPPs themselves have credit requirements in order to enter into bi-lateral**
17 **transactions?**

18 A. Yes. PGE participates in the Northwest wholesale electricity market and several
19 counterparties are IPPs. These parties commonly negotiate credit and collateral
20 requirements as part of the enabling agreements necessary for purchasing and selling
21 wholesale gas and electricity. These enabling agreements have margin call calculations,
22 credit thresholds, and triggers that reflect the credit ratings and financial wherewithal of
23 the counterparties. IPPs operating in the wholesale energy markets have been

1 comfortable executing and abiding by the credit requirements within these enabling
2 agreements.

3 **Q. Collins asks, “Why, given the accrual of experience with IPP’s, would one presume**
4 **a utility needed to have and should pay for investment grade credit from**
5 **suppliers?” (NIPPC Exhibit 200, pg. 9 at 6-7). How do you respond?**

6 A. Our response is that the IPPs themselves and all other participants in the wholesale
7 energy markets require a standard of credit protection that starts with investment grade
8 credit (or equivalent credit enhancements) from their trading partners. PGE is simply
9 implementing these industry best practices in the RFP scoring.

10 **Q. Collins found that PGE’s credit scoring approach was lacking on 5 premises**
11 **(NIPPC Exhibit 200, pg. 8-9). Do you agree?**

12 A. No. We discuss each of the five premises below.

13 **Q. Do you agree with Collins’ first objection to PGE’s use of credit rating agencies**
14 **credit assessments?**

15 A. No. PGE agrees that the rating system is undergoing reform. Nevertheless, these
16 reforms are tightening credit requirements and grading criteria, not relaxing the
17 requirements. PGE is applying the industry best practices standard for assessing credit
18 risk in using counterparty debt ratings as a proxy.

19 **Q. Do you agree with Collins’ second objection that PGE’s use of credit ratings is not**
20 **transparent?**

21 A. No. In fact, PGE’s 2007 RFP (see PGE Exhibit 202) specified in detail how PGE
22 conducts its credit evaluation:

1 **“Credit Evaluation**

2 This category scores the creditworthiness of the Bidder. We will take into
3 account the following credit considerations in our scoring:

- 4 • Debt and equity ratings
- 5 • Performance assurance
- 6 • Financial ratio analysis
- 7 • Default risk
- 8 • Credit concentration and liquidity
- 9 • Enforceability of contractual credit terms
- 10 • Bidder revisions to contract templates that may affect credit requirements”

11 **Q. Do you have any comments on Collins' third complaint regarding the alleged lack of**
12 **a sufficient nexus to the credit of the entity that will hold the asset?**

13 A. Yes. PGE believes that there is a direct nexus between a bidder’s current financial status
14 and their ability to obtain project financing for a prospective power plant from a financial
15 institution. PGE’s RFP scoring accurately reflects the financial wherewithal of the seller
16 at the time of bid submittal, so the scoring does capture the counterparty risk associated
17 with the ability of the seller to complete the project and perform the obligations under a
18 proposed PPA.

19 **Q. Fourth, Collins questions "what is it worth in price received to have triple A credit**
20 **instead of triple B?" (NIPPC Exhibit 200, pg. 9 at 5-6). Can you respond?**

21 A. Yes. PGE does, in fact, differentiate between seller credit grades in its bid evaluation
22 process by allocating more points to entities with a higher credit score. In addition, the
23 credit rating is ~~not the~~ only metric PGE uses to measure creditworthiness and allocate

1 points. Points associated with credit rating are combined with points allocated to the bid,
2 based on certain financial ratios and other credit risk attributes.

3 **Q. Do you agree with Collins' fifth complaint that the PPA Seller submitting a bid may**
4 **not be the entity owning the facility over time?**

5 A. Yes, but it does not diminish the validity of scoring counterparty risk as part of the bid
6 evaluation process. While PGE acknowledges that there is a potential for changes in a
7 seller's ownership and creditworthiness after short-list selection, this risk is largely
8 mitigated by the terms and conditions that would be required during contract negotiations
9 and prior to deal execution. Any changes in seller ownership and credit status after the
10 deal is executed are mitigated through typical wholesale energy contract terms and
11 conditions such as "material adverse change" clauses that address deterioration of credit
12 quality through increases in collateral requirements and / or events of default. Other
13 common contractual provisions that would mitigate the risk associated with seller change
14 in ownership include:

- 15 • Buyer approval requirements for any sale or transfer of generation assets
16 supporting the PPA;
- 17 • Requirements that generation assets can only be sold or transferred to a
18 company with equal or better credit rating than the original owner / bidder;
- 19 • Buyer right to step-in and operate the plant assets in the event of default; and
- 20 • Requirement that seller must maintain a guaranty from an investment grade
21 credit support provider.

1 Moreover, the fact that ownership of a PPA generation asset or project company could
2 change over time does not diminish the prudence of assessing the seller's credit and
3 performance risk at the time of bid evaluation.

4 **Q. Do you believe that changes in the evaluation of counterparty risk are appropriate?**

5 A. Yes. PGE proposes to more fully incorporate counterparty risk in our scoring matrix for
6 future RFPs.

7 **Q. Can you summarize the changes you would propose that would address
8 counterparty risk more accurately?**

9 A. Certainly. In addition to PGE's current RFP scoring method, we propose working with
10 the IE to include in our non-price scoring section any changes in risk allocation that may
11 result from bidder exceptions to our proposed standard contract template terms and
12 conditions. More specifically, we propose as the following:

- 13 • Non-Negotiable contract terms for certain key contract terms and conditions: in order
14 to account for the fact that bidder's edits of the template contract also affect
15 counterparty risks, some contractual terms should be deemed non-negotiable. The
16 acceptance in whole of these non-modifiable contract terms should be a threshold
17 requirement for participating in the RFP. These non-modifiable contract terms could
18 include, but may not be limited to:

19 Seller must offer buyer the right to:

- 20 ▪ Step into the role of builder, operator or owner in the event of seller
21 default;
- 22 ▪ Approve a merger or transfer of plant assets underlying a PPA;
- 23 ▪ Events of default and security requirements.

- 1 • Score all bids to reflect redline edits / exceptions of terms and conditions in the template
2 contracts. The methodology for evaluating and scoring any contract exceptions would be
3 done by the utility in consultation with the IE. Changes to contract provisions to be
4 considered for score adjustment could include, but may not be limited to:
- 5 ○ Change in law;
 - 6 ○ Change in regulation;
 - 7 ○ Addition of condition precedent or contract effectiveness;
 - 8 ○ Addition of no-fault termination clauses or conditions that alter or limit seller
9 performance obligations ;
 - 10 ○ Conditions or provisions that establish caps or limitations on damages or remedies
11 for performance failures;
 - 12 ○ Changes in events constituting *force majeure*;
 - 13 ○ Changes in performance assurance provisions;
 - 14 ○ Provision of environmental attributes.

15 The inclusion of the above recommendations would improve scoring of the bids
16 to more appropriately reflect counterparty risk without prejudice to the structure of the
17 proposal.

18 In addition, PGE suggests that the RFP process should be modified to include an
19 evaluation and scoring of all bid structures that result in a direct balance sheet impact to
20 the utility, such as capital lease treatment, consolidation of variable interest entities (VIE)
21 or any other GAAP rule that may directly affect the financial statements of the buyer, and
22 which are not currently captured in RFP scoring.

III. Response to OPUC Staff's Testimony

1 Q. Do you agree with Staff that the focus of the risk analysis in Phase II is risk to
2 customers? (Staff Exhibit 100, pg. 3 at 21-22)

3 A. Yes. PGE agrees that the focus of the risk analysis in Phase II is the risk to utility
4 customers. We, therefore, believe that an effective bid scoring process is one that is
5 designed to ensure that all proposed resources are assessed against a consistent set of
6 criteria that captures all relevant factors which affect customer costs and benefits.

7 Q. Do you agree with Staff that “the Commission has concluded that the bid evaluation
8 process is biased towards the benchmark resource bid”? (Staff Exhibit 100, pg. 4
9 at 13-19)

10 A. No. Staff alludes to the Averch-Johnson Effect that there is a tendency by the utility to
11 substitute capital for labor in its profit-maximizing calculations. However, this is a high-
12 level theoretical result that makes no specific forecast that would imply a tendency by a
13 utility to bias its bid evaluations. We believe that the Commission concluded that utilities
14 have an incentive to select benchmark resources because of the ability to place these
15 investments in rate base, but the Commission did not conclude that the competitive bid
16 evaluation methodology or process was itself biased. It is the goal of this phase of the
17 docket to determine whether any improvements can be made in the evaluation and
18 analysis of utility benchmark and competitive bids.

19 Q. Do you agree with Staff that the current bid evaluation process could treat the
20 benchmark resource bid and the IPP bid differently and that the different treatment
21 would not be evidence of bias? (Staff Exhibit 100, pg. 6 at 3-13)

1 A. Yes. As we have emphasized in our testimony, the bid evaluation process should be
2 designed to assess the value and risk of each unique resource and bid, independent of
3 ownership status. Further, we agree with Staff that “[i]f it turns out that the existing bid
4 evaluation criteria reasonably account for differences in risk between two bids, then that
5 is evidence that the bid evaluation criteria are free of bias.” (Staff Exhibit 100, pg. 7 at
6 10-12)

7 **Q. Do you agree with Staff’s criticisms of NIPPC’s recommendations for a cost over-**
8 **run adder? (Staff Exhibit 100, pg. 10 at 22-23)**

9 A. Yes. Specifically, we agree that any recommendation must include an examination of the
10 current bid evaluation guidelines. It makes no sense to propose a cure when a diagnosis
11 has not been made. As we have emphasized in our testimony, if the RFP scoring
12 recognizes the value of construction cost guarantees (and conversely the risk of the
13 absence of such guarantees), then construction cost risk has been accounted for, and an
14 adder is not necessary. We believe that RFP bid scoring should favor proposals
15 incorporating cost guarantees regardless of ownership status.

16 **Q. Do you share Staff’s concerns regarding NIPPC’s data and the methodology they**
17 **used to derive the cost over-run adder? (Staff Exhibit 100, pg. 8 at 9-14)**

18 A. Yes. Staff is concerned about the small sample size used to derive this cost over-run
19 adder (Staff Exhibit 100, pg. 11 at 17-23). We share these concerns. As we discussed
20 earlier, Dr. Jacobs (PGE Exhibit 300) provides a detailed discussion of the data NIPPC
21 used to calculate the cost over-run adder. He concludes that their data set is not adequate
22 to distinguish the relevant cost estimates used to justify the bid and the final construction
23 cost.

1 **Q. Staff has expressed concerns about the data set and methodology used to calculate**
2 **NIPPC’s heat rate degradation adder (Staff Exhibit 100, pg. 15 at 4-10). Do you**
3 **share these concerns?**

4 A. Yes. Staff notes that the vintages of the plants in the data set “are not representative of
5 gas plants that will be bid into future RFPs” (Staff Exhibit 100, pg. 15 at 12-13). Staff
6 also recognizes that the data set consists primarily of observations for Simple Cycle
7 Combustion Turbines and concludes that it is unlikely that valid inferences could be
8 drawn from this data set for Combined Cycle Combustion Turbines. Dr. Jacobs’
9 testimony includes a detailed critique of NIPPC’s heat rate degradation estimates and
10 provides alternative estimates based on a better data set. He concludes that, even with a
11 superior data set, it is not possible to derive useful estimates of long-run heat rate
12 degradation from available data.

13 **Q. Do you concur with Staff’s analysis of heat rate degradation?**

14 A. Yes, for the most part. We agree with Staff’s concerns regarding Monsen’s analysis and
15 proposed adders. As an example, we both believe that Monsen’s analysis has data and
16 methodological issues. While recognizing the limitations in the data set utilized by
17 Monsen and, subsequently, Staff, we again point out that the correct metric against which
18 to measure heat rate degradation is the “as bid” heat rate. As we stated in our direct
19 testimony, the appropriate heat rate for bid evaluation is the long-run average degraded
20 heat rate. We also note that Staff’s analysis did not adjust for changes in plant dispatch /
21 operation.

22 Staff supports the results from their “Alternative Five” which indicates a 0.11%
23 degradation in heat rate, which is effectively zero. However, if Staff had the data needed

1 to adjust for changes in plant dispatch and operation, and were able to compare actual
2 heat rates to the projected lifetime average heat rate, we believe that their analysis would
3 not support a heat rate adder.

IV. Summary and Conclusions

1 Q. Please summarize your recommendations to the Commission.

2 A. First, we recommend that the Commission reject the use of adders for heat rate
3 degradation, cost under / over runs, and wind capacity factor.

4 Second, we recommend that the Commission continue to maintain its focus on
5 potential improvements in the analytic framework and process for RFP bid evaluation.

6 Finally, if the Commission finds that improvements are needed for any of the four
7 issues identified for this phase of the docket, we recommend that the Commission direct
8 the utilities to work with the IE and Staff to develop scoring criteria to address the costs,
9 risks and benefits imposed on utility customers for each specific issue.

List of Exhibits

<u>PGE Exhibit</u>	<u>Description</u>
201	Standard & Poor's <i>Principles of Credit Ratings</i>
202	PGE 2007 Request for Proposals

**BEFORE THE PUBLIC UTILITY COMMISSION
OF THE STATE OF OREGON**

UM 1182 Phase II

PORTLAND GENERAL ELECTRIC COMPANY

Direct Testimony and Exhibits of

Jonathan Jacobs



Portland General Electric

January 14, 2013

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I. Introduction

1 **Q. State your name, title, and business address.**

2 A. My name is Jonathan M. Jacobs. I am a Managing Consultant for PA Consulting Group,
3 Inc. My business address is One California Plaza, Suite 3840, 300 South Grand Avenue,
4 Los Angeles CA 90071.

5 **Q. Please describe your educational background.**

6 A. I hold the degree of Doctor of Philosophy in Mathematics, awarded by the University of
7 Wisconsin-Madison in 1985. My undergraduate degree is an S.B. from the University of
8 Chicago with a concentration in mathematics.

9 **Q. Please briefly describe your professional background.**

10 A. I have been in the electricity industry for almost twenty-three years, and have been with
11 PA Consulting Group for the last twelve years. I have advised electric utilities, merchant
12 power producers, system operators, and other clients on various aspects of energy
13 procurement, resource planning, ISO markets, pricing and forecasting. Of particular
14 interest to this proceeding, for the last six years I have been an Independent Evaluator of
15 resource solicitations for San Diego Gas & Electric Co.

16 Prior to joining PA I was employed for three years by PG&E Energy Services,
17 where I was Director-Market and Financial Modeling. Before that I worked for seven
18 years for Pacific Gas & Electric Company, generally on mathematical modeling for
19 electricity supply planning.

20 My curriculum vita is provided in Section III.

21 **Q. On whose behalf are you testifying?**

22 A. I am providing reply testimony on behalf of Portland General Electric Company.

1 **Q. To what testimony are you replying?**

2 A. I am primarily replying to the testimony of Northwest and Intermountain Power
3 Producers Coalition (NIPPC). I will also refer to the testimony of the Oregon PUC Staff,
4 both for context and to address the meaning of “risk” and “bias”, which are two key
5 terms used in this Proceeding. I will also reference some earlier comments submitted by
6 NIPPC, to which the Staff testimony makes reference.

7 **Q. What are the conclusions of your testimony?**

8 A. I conclude that the heat rate degradation and cost over-run adders proposed by NIPPC’s
9 witness, William Monsen, are not appropriate.

10 His historical analysis of heat rate degradation used inappropriate data and had
11 computational flaws, including a failure to correct for plant dispatch. I did a similar
12 analysis correcting for these issues and obtained no evidence of statistically significant
13 heat rate degradation unexplained by dispatch. The baseline of his computation does not
14 appear to represent how Oregon utilities estimate heat rates of benchmark bids.

15 In addition, Mr. Monsen’s estimate of a cost over-run adder is based on historical
16 data from California. It is difficult to find historical comparisons for cost over-runs
17 relative to benchmark bids, because again the historical baselines are often not
18 comparable to benchmark bids. That is the case for the first of Mr. Monsen’s examples
19 that I examine. The second example seems to be a type of over-run that is not specific to
20 utility benchmark bids; if a similar situation occurred to an IPP project, I think it is likely
21 that the Power Purchase Agreement would permit the IPP to pass the cost on to
22 customers. Finally, I do not believe that he has provided sufficient justification for
23 including “deferred construction costs” in such an adder.

II. Bias, Risk and Bid Adders

1 **Q. According to Staff, what are the goals of this phase of UM-1182?**

2 A. Basically, this phase of the Proceeding is intended to analyze four different risks: Cost
3 Over-run/Under-run Risk; Wind Capacity Factor Risk; Counterparty Risk; and Heat Rate
4 Degradation Risk. Staff states that the goals are to determine how utilities currently
5 account for each risk in the bid evaluation process; whether the evaluation methods are
6 biased in favor of benchmark or IPP bids; and to make recommendations as to how to
7 remove any bias.

8 **Q. Generically, how do utilities evaluate bids?**

9 A. The most natural way to think of evaluating bids is to assign a numerical value to each
10 bid (a score) and to choose the bid with the highest score. This appears to be the way that
11 Portland General and PacifiCorp both evaluate bids, based on their direct testimony. The
12 numerical score is supposed to fairly represent the value of each bid. It can be easy to see
13 the intended relationship between score and value, if for example “value” is taken to
14 mean expected cost to customers. The numerical score can also include various
15 qualitative factors.

16 If a score is based on multiple attributes or factors, the utility can try to combine
17 them, usually by adding them up. For example, PacifiCorp states it adds together a price
18 score on a scale of 1-70, and a non-price score on a scale of 1-30. In doing this, the
19 utility is explicitly or implicitly scaling the different factors.

20 Finally, the utility could determine not to scale the factors, in which case the bids
21 would not be measured against a single standard of value. If the utility took this
22 approach it would look for “non-dominated” bids. A bid is “dominated” if there is

1 another bid which is at least as good on every factor, and better on one. This strategy
2 may be unobjectionable if the utility seeks to contract with several resources, but if the
3 utility has to choose only some of the non-dominated resources, it still needs to determine
4 and explain a selection rule.

5 **Q. What does it mean to say an evaluation method is biased?**

6 A. Generally speaking, an evaluation method is biased if it does not consistently rank bids
7 according to their value. For example, if the method were based on numerical scores and
8 the method assigned a higher score to bid A than to bid B even though B was more
9 valuable, the method would be biased.

10 There is also a statistical definition of bias which is particularly relevant here.
11 Suppose that the utility scored bids based on a formula that is based on certain parameters
12 of the bids, and those parameters are not known with certainty. The utility or the bidder
13 has to estimate those parameters. When dealing with an unknown parameter it is
14 customary to assume it has some kind of a probability distribution – for example, to
15 assume that the price of natural gas per MMBtu, while unknown, has a 30% chance of
16 being under \$3.00, a 40% chance of being between \$3.00 and \$4.00, and a 30% chance of
17 being above \$4.00. An unbiased parameter estimate is, by definition, one that equals the
18 expected value of its probability distribution. Of course, the “true” probability
19 distribution is also unknown, but it is common to use the observed frequency distribution
20 of similar data, such as historical gas prices, as a proxy.

21 Using this definition one would say that an evaluation method based on a
22 numerical formula is (systematically) biased if it is based on parameters estimated in such
23 a way as to differ from their expectations. It would be biased in favor of a particular type

1 of resource (such as utility benchmark resource) if the estimates of some parameter(s) of
2 those resources were systematically skewed in such a way that bids received higher
3 scores than if the expected parameter values had been used instead.

4 **Q. Does opening testimony reflect this view of “bias”?**

5 A. Yes. In particular, NIPPC’s direct testimony alleges that utility estimates of several bid
6 parameters – capital cost, wind capacity factor and heat rate degradation are biased.

7 **Q. What is a “bid adder”?**

8 A. A bid adder would be a rule for adjusting a bid score. A bid adder defined in this docket
9 would be intended to eliminate bias. A bid adder could also introduce a preference for
10 bids with certain characteristics either intentionally or unintentionally.

11 **Q. Does any respondent propose bid adders?**

12 A. NIPPC proposes an adder to the capital costs of utility benchmark bids. They also
13 propose a modification to be applied to the heat rates of utility benchmark bids, which is
14 more of a bid modification than an adder.

15 **Q. Is there another use for the statistical definition of bias?**

16 A. Yes. Bid evaluation often involves a variety of uncertainties, including for example
17 hydro conditions, fuel prices, and load. Each combination of possible outcomes of those
18 uncertainties is a scenario. A bid score formula can involve these values, and could
19 produce a single numerical answers for each scenario. In principle one could form the
20 expected value of those answers. The expected value is the weighted average of the
21 different possible numerical values, each one weighted by its probability. Of course,
22 even if the numerical formula correctly reflects the resource value under each scenario,
23 its expected value may not correctly reflect the resource value in an uncertain world.

1 **Q. How could the expected value not correctly reflect the resource value?**

2 A. It is often assumed that we prefer certainty to uncertainty and therefore would prefer a
3 resource whose cost or score did not vary much from scenario to scenario, even if the
4 expected value of its score were the same or less than that of another resource with much
5 more score variability. In finance that variability is called “risk”. If customers prefer a
6 less variable cost, an evaluation method that does not penalize variability would be
7 systematically biased.

8 **Q. Is it necessary to address this kind of “bias”?**

9 A. I don’t believe that any of the opening testimony addresses the presence or absence of
10 such bias in utility bid evaluation. The Staff Opening Testimony specifically faults
11 NIPPC’s submitted “Phase II Report” for not addressing “risk” as variability. Staff bases
12 this on the Commission’s language in Order 12-324, which characterizes the four issues
13 to be addressed as “risks”.

14 I believe from context that the Commission used the word “risk” in a common
15 language sense, to identify these as issues of concern. In the materials that I have
16 reviewed, I have not seen the need to address variability as a consistent theme. It is true
17 that NIPPC states that fixed-price power purchase agreements are valuable because they
18 eliminate upside cost risk, and that utilities believe traditional cost-based ratemaking is
19 valuable because it allows customers to capture cost reductions. But no party appears to
20 present any evidence as to whether utility customers value the reduction of uncertainty,
21 and how much they would pay for a certain amount of cost reduction.

A. Heat Rate Degradation

1 **Q. What is the issue of heat rate degradation?**

2 A. The heat rate of a power plant is the ratio of fuel use to electric energy production.
3 Greater efficiency is associated with using less fuel to produce the same amount of
4 electric energy, so lower heat rates are associated with higher efficiency. Heat rates of
5 fossil-fired power plants generally increase over time, although regular maintenance can
6 restore most of the lost efficiency. Again, heat rate degradation is an increase in the
7 numerical value of the heat rate, relative to a particular value associated with the
8 generator. I call that value the “baseline” for the degradation. Photovoltaic plants
9 similarly decline over time, in that case manifested as a reduction in capacity factor for
10 constant insolation, but I believe that only thermal plants are at issue here. Heat rate
11 degradation due solely to aging will reduce the net benefit (increase the cost) of a power
12 plant.

13 **Q. Are there any other reasons for the observed efficiency of a power plant to change**
14 **over time?**

15 A. Yes, especially if one is referring to the average efficiency. Average heat rates can
16 increase due to changes in dispatch (the plant can be dispatched away from its optimal
17 operating point, for example to provide ancillary services) or the addition of emissions
18 controls. If the heat rate degrades due to plant dispatch, it is likely that the system
19 operator has determined that the operational benefits of the revised dispatch exceed the
20 costs and a cost-only computation will not capture that. If a heat rate degrades due to an
21 additional emissions control not connected with new regulations, the control may well

1 have been installed or imposed to avoid unexpectedly high emissions costs from other
2 resources, and again the benefit (avoided emissions cost) should exceed the cost increase.

3 **Q. What does the utilities' opening testimony say about heat rate degradation?**

4 A. Portland General describes their benchmark bids as based on "long-term degraded heat
5 rates". They say that they base their heat rate estimate on the "new and clean" heat rate
6 and "non-recoverable degradation" information provided by turbine manufacturers, as
7 well as "long-run average degradation" over the maintenance cycle. PacifiCorp says they
8 begin with a "new and clean" heat rate and apply "the heat rate degradation curve, which
9 is supplied by the OEM." Idaho Power says they assume heat rates degrade "based on
10 the manufacturer's specifications."

11 **Q. Given the utilities' description of how heat rate degradation is accounted for in their
12 current bid evaluations, could there be sources of bias in the heat rate estimates?**

13 A. Yes, possibly. The manufacturers' estimates of heat rates could be incorrect or based on
14 unrealistic operating conditions. And, the utilities could have applied the manufacturers'
15 degradation forecasts incorrectly. Although the effects of these potential sources of bias
16 seem to be small, I can see the value of testing them.

17 **Q. What does other parties' opening testimony say about heat rate degradation?**

18 A. William Monsen, testifying on behalf of NIPPC, described an analysis he conducted,
19 from which he estimated the average degradation in heat rate for utility-owned generation
20 to be 8%. He recommended that this be used as the floor for the average degradation in
21 the heat rate of any utility benchmark resource over time.

22 Robert J. Procter testified on behalf of the Commission Staff. He first noted that
23 "none of [NIPPC's] adders consider how the utility's current bid evaluation methodology

1 addresses the four risks” (Staff Exhibit 100, pg. 10 at 19-20) This is in a section about
2 the Cost Over-run / Under-run analysis but it is clearly meant to apply to NIPPC’s
3 analysis as a whole, that is, to the Heat Rate Degradation as well. In other words, there
4 have been no alternatives presented to the utilities’ descriptions of how they estimated
5 heat rates. Mr. Procter then critiqued an earlier presentation of Mr. Monsen’s historical
6 analysis.

7 **Q. What does Mr. Monsen recommend in his testimony?**

8 A. He recommends that the IE ensure that utility benchmark bids enforce an average heat
9 rate degradation of 8%.

10 **Q. Does he recommend an 8% adder to heat rates in benchmark bids?**

11 A. No, he recommends that the IE ensure that heat rates incorporate 8% average
12 degradation.¹ I infer that he means the average heat rate should be 8% larger than the heat
13 rate in an un-degraded state. He also says that if a benchmark bid can be shown to
14 incorporate some degradation, the adder should only be large enough to bring the average
15 degradation up to 8%. He does not give any evidence as to whether benchmark bids
16 already include degradation.

17 **Q. On what basis does he make this recommendation?**

18 A. He says it is based on analysis of historical figures for power plant heat rate degradation.
19 He does not compare the accuracy of his method to the accuracy of the forecasts in utility
20 benchmarks, in other words, does not determine which method is really less biased.

¹ Northwest & Intermountain Power Producers Coalition (NIPPC), *Technical Approach to Developing Bid Adders for Utility-Owned Generation Proposals*, Jan. 31, 2012. It appears as attachment 1 to Northwest & Intermountain Power Producers Coalition (NIPPC), “Comments of the Northwest and Intermountain Power Producers Coalition” in Docket UM-1182, March 19, 2012.

1 **Q. Could an analysis like NIPPC's be used to reveal biases in the estimation of heat**
2 **rate degradation in utility benchmark bids?**

3 A. Conceivably, but the analysis would have to use appropriate data, correct as necessary for
4 plant commitment and dispatch, and have a baseline similar to the benchmark bids
5 themselves.

6 **Q. What do you mean by "appropriate data"?**

7 A. Historical degradation in power plant heat rates might imply that the heat rates of new
8 power plants will degrade by a like amount, but only if it is reasonable to think that the
9 plants are similar. There is no evidence that the majority of plants whose heat rates
10 NIPPC studied were similar to modern gas-fired combined cycle plants. Furthermore,
11 there is no evidence that the maintenance practices applied to the power plants in the
12 NIPPC database are similar to those in use today. "Appropriate data" would mean a
13 database consisting of modern gas-fired combined cycle plants with Long Term Service
14 Agreements, which have been maintained according to manufacturers' recommendations.

15 **Q. What do you mean by "correct as necessary for plant commitment and dispatch"?**

16 A. NIPPC has only considered average heat rates, which depend heavily on operation. A
17 turbine-generator's heat rate varies with its operating level. The average heat rate also
18 depends on the number of times a plant is started, because each start requires an
19 additional amount of fuel that is not effectively converted to electricity. Finally, a
20 combined-cycle plant often contains one or more combustion turbines and one or more
21 heat recovery steam generators, and the efficiency of the plant can vary with the number
22 of components operating. These factors are controlled by the plant's commitment and
23 dispatch – roughly, by the amount of electricity it produces, or is scheduled to produce.

1 If the plant is scheduled to produce less electricity it will probably operate less
2 efficiently, at a higher heat rate. That less efficient operation should not be interpreted as
3 degradation of the plant. I illustrate, later in my testimony, an analysis that tries to
4 correct for changes in dispatch, or more specifically, for changes in electricity
5 production.

6 **Q. Is it possible also to correct for maintenance?**

7 A. Perhaps one could control for the amount of time since the last major maintenance, if it
8 were possible to get data about historical maintenance schedules. But if you already had
9 historical maintenance data, you should be able to limit your database to plants that have
10 been maintained at proper intervals.

11 **Q. What do you mean by “a baseline similar to the benchmark bids themselves”?**

12 A. NIPPC reviews a set of power plants and compares historical annual average heat rates
13 for each plant to a baseline which is not necessarily consistent with the way utilities
14 estimate heat rates; it is not derived from manufacturer’s data for heat rates and their
15 degradation. (There are actually different baselines in the original NIPPC analysis and
16 the testimony, as I will explain.) To evaluate the potential bias in the utilities’ values one
17 would have to compute the average difference between the “actual” heat rates and those
18 utility forecasts, or between the “actual” heat rates and proxy forecasts based on
19 manufacturers’ data. Different turbine models or configurations could have different
20 amounts of degradation, which could still be accounted for in the manufacturers’ data.
21 The baseline for an analysis like NIPPC’s, if the goal were to determine whether the
22 utility forecasts were biased, would be a hypothetical forecast constructed for each plant
23 using the utilities’ methods and contemporary manufacturer information.

1 **Q. Do you agree with Staff's critique of the original NIPPC analysis?²**

2 A. Staff raised three issues. The first, as I understand it, was that NIPPC did not address risk
3 in the sense of variability. I am not in a position to know if it was the Commission's
4 intention that they do so. Second, Staff raised data and methodological concerns with
5 NIPPC's analysis, which I share. As will be seen below, I have conducted my own
6 analysis and do not agree with NIPPC's results. Third, Staff said that NIPPC did not
7 "consider how contract terms of specific bids influence ratepayer exposure." This could
8 mean the consideration of whether manufacturers' guarantees or improved maintenance
9 scheduling, reflected in long-term service agreements mitigate any degradation that had
10 been observed in the historical record; I did not go beyond the historical record in this
11 testimony to estimate such mitigations.

12 **Q. What were Staff's data and methodological concerns?**

13 A. That the NIPPC dataset included data as early as 1981 and power plants that could have
14 come online even earlier; that the dataset is dominated by simple cycle combustion
15 turbines; that NIPPC did not perform adequate sensitivity analysis; and that NIPPC did
16 not weight the results by generation.

17 **Q. What did the original NIPPC analysis conclude?**

18 A. In that analysis, NIPPC stated that they observed an average heat rate increase of 5.6%.

19 **Q. Does the original NIPPC analysis support Mr. Monsen's figure of 8% for heat rate**
20 **degradation?**

² The term "original NIPPC analysis" will be used to refer to the analysis described in the *Technical Approach* document cited in note. This is done to distinguish it from the analysis underlying NIPPC's opening testimony, which appears to have been slightly different but is not completely specified.

1 A. Not directly, since he recommends a figure that is almost half again as large as the
2 average degradation stated by earlier analysis. Either the analysis has changed, or he has
3 another justification for the larger adder.

4 **Q. Are the NIPPC opening testimony and the original NIPPC analysis based on the**
5 **same dataset?**

6 A. It appears so.

7 **Q. What is that dataset?**

8 A. NIPPC's dataset was developed from Federal data and used for an econometric analysis
9 published in an academic paper. According to that paper the dataset was based on
10 temporally aggregated data collected by Federal agencies, in other words, generation and
11 fuel usage over periods of months or years, irrespective of the operating tempo during
12 those periods.³

13 **Q. Are there any differences in definitions or assumptions between the original NIPPC**
14 **analysis and the opening testimony?**

15 A. There appear to be at least two.

16 1. The "degradation" is estimated differently. In the original NIPPC
17 analysis, the "degradation" of an individual power plant's heat rate was estimated using
18 the average of the ratios of annual observed heat rates to the earliest heat rate for that
19 plant in the database. From that averaging, any improvements of 7.1% or greater were
20 excluded. Mr. Monsen's testimony says that instead of using the first observed heat rate
21 as a base, he used the absolute minimum observed for the plant, which could be an
22 observation prior to some of the "non-recoverable degradation" or at the most efficient

³ Fabrizio, K.R., N.L. Rose and C.D. Wolfram, "Do Markets Reduce Costs? Assessing the Impact of Regulatory Restructuring on U.S. Electric Generation Efficiency," *American Economic Review* 97 (September, 2007), pp. 1250-1277.

1 point in a maintenance cycle. In this case Mr. Monsen does not say that he eliminated
2 any improvements of 7.1% or greater.

3 2. The original NIPPC analysis used a simple average of all the heat rate
4 changes. In his testimony, Mr. Monsen says he used a weighted average, weighed by
5 capacity factor.

6 **Q. Did the NIPPC opening testimony obviate any of Staff's data and methodological**
7 **concerns?**

8 A. Perhaps. NIPPC changed from using a simple average to an average weighted by
9 capacity factor (not the same as generation). According to Staff's testimony, weighting
10 by generation would have reduced the estimated average degradation from 5.6%
11 to 0.11%; yet NIPPC has instead increased its recommendation to 8%.

12 **Q. Is the NIPPC dataset appropriate?**

13 A. No. It was dominated by plants using old technology, and does not reflect current
14 maintenance practices. Furthermore, NIPPC does not appear to have made any attempt
15 to take account of historical dispatch or operating tempo.

16 **Q. Did you attempt to construct a more appropriate database?**

17 A. Yes. PA Consulting staff under my direction extracted a database similar to the one used
18 by NIPPC, suited for this particular purpose, from data collected by the Energy
19 Information Administration (EIA) in the Annual Electric Generator Report (Form 860)
20 and Power Plant Operations Report (Form 923) for 2003-2011. This database contains
21 monthly volumes of fuel used and net electric generation (net means, after subtracting
22 station service, which means the power used in the plant itself). From those monthly
23 volumes one can also accumulate annual volumes. The average heat rate in any period is

1 the ratio of fuel used to net electric generation. We then computed period-to-period
2 changes in the heat rate and generation volume (deltas) for each plant.

3 **Q. What power plants did you include in your analysis?**

4 A. We included all combined cycle plants in the EIA database, other than single-shaft plants.
5 The EIA dataset separately identifies the combustion turbine part of the plant from the
6 steam turbines, labeling them by prime mover codes “CT” and “CA”. We aggregate the
7 fuel use and generation from the CT and CA parts of each plant. We excluded any units
8 labeled with other prime mover codes, including “CS” (single-shaft combined cycle).
9 Finally, we eliminated plants that were identified as combined heat and power (CHP).

10 **Q. Why did you exclude single-shaft plants?**

11 A. There are considerably fewer of them, and the mechanical differences could change their
12 heat rate degradation.

13 **Q. Did you exclude plants of capacity less than 150 MW, as NIPPC did in its original
14 analysis?**

15 A. No. NIPPC did not explain this limitation; I thought it may have been put in to exclude
16 aeroderivative gas turbines. I did not think it was necessary to include such a limit in
17 PA’s dataset as we were restricting to non-CHP CCCT units, and because it was not a
18 precise specification – NIPPC did not state whether they meant “summer capacity” or
19 “winter capacity”. I felt it was more important to exclude plants with low net generation,
20 since then a disproportionate amount of their fuel is used for startups or for station
21 service. Therefore, in my analysis of annual data I excluded years in which the plant
22 generated less than 13,000 MWh – that’s about a 1% capacity factor for a 150 MW plant.
23 In my analysis of monthly data, which I will describe later, I excluded months in which a

1 plant generated less than 1,200 MWh, which would have been the production of a
2 150 MW plant running at full load for eight hours.

3 **Q. If you excluded a single month or year for a power plant, in the middle of the period**
4 **of record, how did this affect your computation of the deltas?**

5 A. I computed the change from one period to the next period in the database, skipping any
6 periods that were excluded.

7 **Q. Did you also exclude non-utility power plants, as in the original NIPPC analysis?**

8 A. No. I don't think NIPPC has claimed that utility plant heat rates degrade more than non-
9 utility plants, only that utility benchmark bids may not reflect "typical" degradation for a
10 class of plants.

11 **Q. Did you exclude any particularly large heat rate improvements, as in NIPPC's**
12 **original analysis, or particularly large degradations?**

13 A. No. Large improvements could be attributable to operation at more efficient dispatch
14 points with fewer starts. Similarly, large degradations could be attributable to operation
15 at less efficient dispatch points for shorter periods, with more startups. Neither can be
16 automatically attributed to data errors.

17 **Q. You computed period-to-period changes, while Mr. Monsen's testimony addressed**
18 **differences from a single baseline for each generator. How does that affect the**
19 **quality of the analyses?**

20 A. Aging, and wear and tear on a power plant, are not the only reasons average heat rates
21 change. Plant efficiency also responds to other factors, including dispatch. The best
22 observed heat rate, which was the baseline in Mr. Monsen's testimony, does not
23 necessarily represent the least degraded state; it could represent the period in which other

1 influences were most favorable. If so, the choice of baseline would magnify the impact
2 of extraneous factors. The period-to-period change reflects both degradation and those
3 extraneous factors, but we can correct for at least one (quantity of production) and try to
4 average out the rest. I believe that correction for variation in production eliminates one
5 potential bias in the analysis, and averaging out the impacts of random changes in
6 extraneous factors eliminates another. Then if the average period-to-period degradation
7 is significantly different from zero, one can compound it over multiple periods to
8 estimate the average degradation over the plant lifetime.

9 **Q. Please provide some general descriptive statistics of your dataset.**

10 A. There were a total of 281 power plants in the database. The database covered 9 years
11 (108 months) so there were at most 9 annual observations (deltas) per plant. There were
12 a total of 1814 annual deltas in the database (6.5 per plant).

13 **Q. What did you observe about heat rate changes?**

14 A. The average annual heat rate change was 0.53%, but the average logarithmic difference
15 was -0.35%, which is an improvement rather than a degradation.

16 **Q. What do you mean by the logarithmic difference?**

17 A. The logarithmic change in heat rate is the natural logarithm of the ratio of two successive
18 heat rates. If the two heat rates are r_1 and r_2 , their difference is $r_2 - r_1$ but the
19 logarithmic difference is $\ln(r_2/r_1)$. It happens also to equal the difference of their
20 logarithms, $\ln r_2 - \ln r_1$.

21 **Q. Is there an advantage to using the logarithmic difference?**

22 A. Yes. We are really interested in the effects of multiple years of heat rate changes, to get a
23 long-run average. It's a compounding process, like compound interest. It is relatively

1 easy to compound logarithmic changes, you just add them. Averaging the percentage
2 changes in heat rates gives you a biased estimate because it overemphasizes positive
3 changes (increases), as the example I am about to give will show. There is a cycle over
4 which heat rate degradation is at least partly repaired by maintenance. Suppose that in
5 one year, the heat rate goes up by 50%; and in the next year, thanks to maintenance, it
6 goes back where it was. That's only a 33% decrease (from 150% back to 100%) so the
7 average of those changes would be 8.7% (the average of +50% and -33%); yet the total
8 change is zero. The logarithmic changes will have an appropriate zero average, because
9 the logarithms of 1.5 and 0.67 are equal and opposite.

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19 average of those changes would be 8.7% (the average of +50% and -33%); yet the total
20 change is zero. The logarithmic changes will have an appropriate zero average, because
21 the logarithms of 1.5 and 0.67 are equal and opposite.

22 **Q. But is this really comparable to NIPPC's analysis? They only looked at power**
23 **plants in non-restructured states.**

1 A. NIPPC took their data from a paper that tried to show that electric industry restructuring
2 was associated with increased power plant efficiency. That paper emphasized that it was
3 addressing labor and non-fuel operation and maintenance costs, not heat rates – in fact
4 the authors say “the data do not suggest gains in fuel efficiency from restructuring within
5 our sample.”⁴ But, their dataset only extends to 1999, and restructuring had only just
6 started at that point. A more recent paper studies data up to 2006 and concludes that
7 there is a difference in fuel efficiency between power plants in restructured and non-
8 restructured states – but only if “restructuring” includes significant retail load migration,
9 not just ISO wholesale markets.⁵ That paper includes a table indicating the first year in
10 which a market is considered fully restructured, by state. I excluded data from power
11 plants in restructured states, beginning in each case with the first year in which the state’s
12 electricity market was considered restructured. This left me with 1,040 annual
13 observations. The mean of the annual heat rate changes in this restricted dataset was
14 0.26%, and the mean of the annual logarithmic changes was -0.09%.

15 **Q. You mentioned earlier that heat rate changes can be caused by the way units are**
16 **dispatched, and that you tried to correct for that. How does your dataset allow you**
17 **to take account of dispatch?**

18 A. Primarily by using monthly observations. The fundamental problem is that when plants
19 are dispatched below their optimal operating points their efficiency is reduced.
20 Furthermore, when plants are committed for short periods of time, their fixed operating
21 costs (startup fuel) become more significant and increase the average heat rate. It is hard

⁴ Fabrizio et al., *ibid.*, p. 1269. They go on to say that their annual data are too aggregated to provide precise estimates of heat rate changes.

⁵ Craig, J.D. and S.J. Savage, “Market Restructuring: Competition and the Efficiency of Electricity Generation: Plant-level Evidence from the United States 1996 to 2006,” *Energy Journal* 34.1 (2013) pp. 1-31.

1 to detect these effects using annual data, because dispatch can vary significantly from
2 season to season. Historical monthly generation and capacity factors show much more
3 variability than annual figures. PA's dataset includes monthly generation and fuel
4 consumption, from which we were able to compute monthly heat rates. Dispatch
5 happens over shorter intervals, but the dispatch over a month should still be more
6 uniform than over a year.

7 **Q. Are there any other advantages to using monthly data?**

8 A. Yes. I believe we are better able to test for heat rate degradation during the first year of
9 operation (the "non-recoverable degradation" that Portland General Electric mentions,
10 and to which NIPPC also alludes in their January report). That is because the first year of
11 operation is often not coincident with a calendar year.

12 **Q. Do you conduct any special data modeling to capture dispatch effects?**

13 A. Yes. If heat rates are related to the level of generation, then the change in heat rate from
14 month to month should be related to the change in generation. Heat rate degradation
15 associated with aging would be unexplained by the change in generation. Therefore, I fit
16 a model of the logarithmic change in heat rate as a function of the logarithmic change in
17 generation.

18 **Q. You mentioned "logarithmic change" earlier; why did you choose to use only the
19 logarithmic change in this analysis?**

20 A. Suppose that the amounts of electricity production in two successive periods are g_1
21 and g_2 . The simple difference is $\Delta g = g_2 - g_1$, but if we just use Δg we have a scaling
22 problem: large plants will have large generation amounts and large Δg values just
23 because of their size, not because of the relative magnitude of the change. In that case

1 one might want to measure the change in generation by the ratio of one period's amount
 2 to the next: $pg = g2/g1$. Unfortunately, ratios are not symmetric: if you go from 1,000
 3 MWh generation to 2,000 MWh the ratio is 2, but if you go from 2,000 MWh to 1,000
 4 MWh the ratio is 0.5. Now suppose the heat rate at 2,000 MWh of generation is 8000
 5 Btu/kWh but at 1,000 MWh it degrades to 8,100 Btu/kWh. Using ratios it is hard to
 6 define a consistent relationship between the change in heat rate and the change in
 7 generation:

Table 1. Heat rate vs. generation, using ratios

		Generation (MWh)	Heat rate (Btu/kWh)	Impact*
(a)		1000	8100	
		2000	8000	
	Ratio	2	0.988	0.494
(b)		2000	8000	
		1000	8100	
	Ratio	0.5	1.0125	2.025

*Impact of generation on heat rate, measured as ratio of heat rate change to generation change.

8 The asymmetry in the ratios is eliminated by taking their logarithms. In the example
 9 illustrated below:

Table 2. Heat rate vs. generation, using logarithms

		Ln Generation	Ln Heat rate	Impact*
(a)		6.907	9.000	
		7.601	8.987	
	Difference	-0.694	0.013	-0.019
(b)		7.601	8.987	
		6.907	9.000	
	Difference	0.694	-0.013	-0.019

*Impact of generation on heat rate, measured as ratio of logarithmic heat rate change to logarithmic generation change.

1 Using logarithms has the additional value that “no change” is represented by a
 2 value of 0 rather than 1. It is quite common to use logarithmic changes in models like
 3 this.

4 **Q. What is the specific form of the mathematical model used to correct for dispatch?**

5 A. Mathematically we fit a model of the following form:

6 (1) $\Delta \text{Ln HR} = \alpha \Delta \text{Ln Gen} + \Gamma$

7 Where $\Delta \text{Ln HR}$ is the change in the natural logarithm of a power plant’s heat rate from
 8 month to month (logarithm of the ratios of successive monthly generation numbers),
 9 $\Delta \text{Ln Gen}$ is the change in the logarithm of its generation.

10 **Q. What do the coefficients of model (1) mean?**

11 A. The coefficient α controls the size of the correction for the generation level. The term
 12 $\alpha \Delta \text{Ln Gen}$ represents how much of the monthly heat rate change – degradation or
 13 improvement – we are able to attribute to the change in average generation level. The
 14 intercept Γ , is the average of what’s left. NIPPC’s analysis produces an average of the
 15 heat rate degradation, but in this model Γ is the average after eliminating the effect of

1 changes in the generation level. I would call it the underlying heat rate degradation. I
2 have used monthly data and logarithms, so to get the annual underlying heat rate
3 degradation you have to multiply by 12 and undo the logarithm. What you get is $e^{12\Gamma}$.

4 **Q. Does this approach correct for all the effects of dispatch and commitment?**

5 A. No. This is a partial solution. To get a better correction, though, one would need data at
6 a very fine time scale, or at least the number of startups in each period.

7 **Q. Please provide some general descriptive statistics of your monthly dataset.**

8 A. There were a total of 269 power plants in the dataset, fewer than in the annual dataset. I
9 mentioned above that I eliminated any months in which a plant generated less than
10 1,200 MWh; in addition, I ignored any plant for which there were less than 23 months of
11 valid values for $\Delta\text{Ln HR}$ and $\Delta\text{Ln Gen}$ left after all the other exclusions. The dataset
12 covered 9 years (108 months) so there were at most 107 observations (deltas) per plant.
13 The average number of observations was 80.5; 27 plants had the maximum possible
14 number of observations (107) and four had the minimum allowed (23). 107 of the plants
15 were installed during the period of record of the database and therefore had their first full
16 year of operation during that period. The variables of interest are $\Delta\text{Ln HR}$, $\Delta\text{Ln Gen}$ and
17 FirstYear, which equals 1 for the first 12 months after the plant begins operation and 0
18 thereafter:

Table 3. Statistics of the monthly observations

Variable	Observations	Mean	Sample standard deviation
<u>Overall</u>			
ΔLn HR	21,663	-0.000679	0.1238
ΔLn Gen	21,663	0.00598	0.8569
First Year	21,663	0.0665	0.2492
<u>Non-restructured</u>			
ΔLn HR	12,345	-0.00029	0.1260
ΔLn Gen	12,345	0.00303	0.8647
First Year	12,345	0.0715	0.2577

1 This seems to imply that heat rate doesn't degrade but improves (decreases) over time;
 2 however, this value is not really significant. The 2-sided t-test indicates that based on the
 3 entire dataset we can only be about 60% confident that the mean of ΔLn HR is nonzero,
 4 and based on the data from "non-restructured" states we can only be 20% confident in
 5 such a statement.

6 **Q. What were the results of model (1)?**

7 A. Recall the structure of model (1):

8
$$(1) \quad \Delta \text{Ln HR} = \alpha \Delta \text{Ln Gen} + \Gamma$$

9 The regression statistics are as follows:

Table 4. Model (1) applied to the full monthly dataset

R ²	16.9%			
Degrees of freedom	21,662			
	Coefficients	Standard Error	t Stat	Confidence
Γ	-0.00032	0.00077	-0.4223	33%
α	-0.05937	0.00089	-66.359	100%

1 In the coefficient table, “Confidence” means how confident we can be that the
 2 given coefficient is really nonzero based on the p-value of a 2-sided t-distribution. The
 3 R² value means there is a great deal of unexplained variation in ΔLn HR, but the high
 4 confidence that α is significant indicates that the mean degradation (the apparent monthly
 5 heat rate improvement) is explained by the change in generation.

6 Using the data from “non-restructured” states yielded similar results, with even
 7 less confidence that the corrected estimate Γ is different from zero:

Table 5. Model (1) applied to non-restructured states

R ²	17.0%			
Degrees of freedom	12,344			
	Coefficients	Standard Error	t Stat	Confidence
Γ	-0.00011	0.00103	-0.1046	8%
α	-0.06003	0.00120	-50.227	100%

8 **Q. Did you also attempt to test for “non-recoverable degradation” as opposed to other**
 9 **heat rate degradation?**

10 A. Yes. I tried several approaches. First, I computed the mean logarithmic change in heat
 11 rate for data from the first year of plant operations (where the variable FirstYear equals 1)

1 and compared it both with the average over all data, and with a similar average restricted
 2 to plants installed during the first year of record (“new plants”). I computed these
 3 averages using the entire dataset and also using only the data from non-restructured
 4 states:

Table 6. Comparing heat rate changes in first year of operation to other years

	All states	Non-restructured states
All plants, all months	-0.00068	-0.00029
New plants, all months	-0.00164	-0.00251
First year only	-0.00842	-0.01346

5 Secondly, in order to correct for the fact that generation should be increasing as a
 6 plant enters operation I estimated a similar model that includes the variable FirstYear to
 7 identify degradation during the first year of operation:

8 (2) $\Delta \ln \text{HR} = \alpha \Delta \ln \text{Gen} + \beta \text{FirstYear} + \Gamma$

9 With this model, the corrected heat rate change would be $\beta + \Gamma$ per month during
 10 the first year of operation, and Γ per month thereafter.

11 I estimated model (2) on the full monthly dataset and got the following regression
 12 statistics:

Table 7. Model (2) applied to the full monthly dataset

R ²	16.9%			
Degrees of freedom	21,662			
	Coefficients	Standard Error	t Stat	Confidence
Γ	8.33x10 ⁻⁵	0.00079	0.1050	8%
α	-0.05935	0.00089	-66.339	100%
β	-0.00612	0.00308	-1.9896	95%

1 I also estimated model (2) restricted only to data from non-restructured states:

Table 8. Model (2) applied to the non-restructured states

R ²	17.0%			
Degrees of freedom	12,344			
	Coefficients	Standard Error	t Stat	Confidence
Γ	0.00016	0.00107	0.1484	12%
α	-0.06002	0.00120	-50.2174	100%
β	-0.00374	0.00401	-0.9320	65%

2 Finally, in order to check the possibility that the non-recoverable degradation
 3 impacts the structure of the heat rate curve (changes α) I used an alternative specification
 4 based on the product of FirstYear and the logarithmic change in generation:

5 (3) $\Delta \text{Ln HR} = \alpha \Delta \text{Ln Gen} + \chi (\text{FirstYear} \times \Delta \text{Ln Gen}) + \Gamma$

6 I also estimate model (3) both on the full dataset and on the non-restructured
 7 states:

Table 9. Model (3) applied to the full monthly dataset

R^2	17.0%			
Degrees of freedom	21,662			
	Coefficients	Standard Error	t Stat	Confidence
Γ	-0.00029	0.00077	-0.3739	29%
α	-0.05811	0.00093	-62.474	100%
χ	-0.01678	0.00338	-4.95985	100%

Table 10. Model (3) applied to the non-restructured states

R^2	17.0%			
Degrees of freedom	12,344			
	Coefficients	Standard Error	t Stat	Confidence
Γ	-0.00011	0.001033	-0.10775	8%
α	-0.06017	0.001244	-48.3728	100%
χ	0.00185	0.004489	0.41251	32%

1 **Q. What do these results show about non-recoverable degradation?**

2 A. These results are counter-intuitive. From Table 6 it appears that heat rates in the first
 3 year of operation improve rather than degrade. Model (2) implies the same thing, using
 4 both the full dataset and the data from non-restructured states. Model (3), when
 5 estimated using the full dataset, also appears to imply heat rates degrade less in the first
 6 year of operation than in subsequent years, but when estimated using the data from non-
 7 restructured states does show a (not statistically significant) degradation in the first year.

8 One way to interpret these counter-intuitive results is that time-averaged data,
 9 even using monthly period, are not precise enough to estimate the “non-recoverable

1 degradation". This would also call into question the ability to make other quantitative
2 statements about heat rate degradation based on average data, especially annual averages
3 as Mr. Monsen used. Two other hypotheses would be that there is actually "breaking-in"
4 period over the first year during which heat rates improve, or that the difference between
5 "new and clean" heat rates and heat rates observed during actual operation has to do with
6 the difference between a test environment and an operational environment rather than
7 physical changes during the first year. I do not believe I have evidence to support either
8 of those two positive statements.

9 **Q. What are your conclusions?**

10 A. It is my opinion that there are flaws both in the appropriateness of the data used for
11 NIPPC's original analysis, and in aspects of the analysis itself. Mr. Monsen's testimony
12 is apparently based on a similar analysis, but not exactly the same since his recommended
13 "heat rate adder" is more aggressive than the original analysis would support. Still, even
14 if we were to assume that NIPPC's analysis were correct and produced a number
15 compatible with Mr. Monsen's recommendation, it does not by itself support the
16 recommendation. The baseline of the NIPPC analysis is not consistent with the way that
17 utilities estimate the heat rate in a benchmark bid.

18 Dispatch is clearly a significant variable to explain the historical changes in heat
19 rates from one period to the next. Heat rate depends both on the level at which a plant is
20 dispatched and the frequency of startups. The startup frequency is not captured in the
21 publicly available data and the actual dispatch level cannot be precisely derived from data
22 representing periods of a month or longer.

1 I attempted to correct publicly available heat rate data for changes in dispatch
2 (generation). I found that the average corrected monthly degradation could not be
3 distinguished from zero. Furthermore, Portland General Electric's opening testimony
4 notes that turbine manufacturers give figures for "non-recoverable" heat rate degradation,
5 which usually occurs during the first year of operation. Average heat rate changes during
6 the first year of operation for plants installed during the years 2003-2010 are actually
7 more negative than average changes for the same plants in later years, both before and
8 after correction for changes in generation. Negative changes would indicate
9 improvement, not degradation. Yet, turbine manufacturers apparently caution purchasers
10 to expect degradation in the first year.

11 I conclude that publicly available heat rate and generation data for combined
12 cycle units are unlikely to provide useful estimates of heat rate degradation. The figures
13 that NIPPC derived in their original analysis are probably attributable to several of the
14 data and methodological problems that Staff identified – vintage of the power plants,
15 large number of simple cycle gas turbines, questionable eliminations – as well as the fact
16 that NIPPC did not try to correct for generation changes. A better estimate might be
17 derivable from a historical database of the results from actual efficiency tests; however,
18 constructing such a database would be expensive and time-consuming and I am unaware
19 of an existing and available data source.

B. Cost Over-Run / Under-Run

20 **Q. What is the issue of cost over-runs and under-runs?**

21 A. Cost over-runs and under-runs are simple examples of estimation error. A cost over-run
22 in a power plant occurs when the actual installed cost is greater than the estimate that

1 formed the basis of the decision to construct the plant. An under-run similarly occurs
2 when the actual cost is less than the estimate. If power purchase agreements with new
3 power plants included ironclad assurance of delivery at a fixed price, but utility projects
4 built based on benchmark bids passed actual costs to customers, then customers would be
5 much more affected by cost over-runs and under-runs for utility benchmark bids than by
6 those for projects with PPAs. Finally, if utility projects more frequently have over-runs
7 than under-runs, then the estimation error could be considered to represent estimation
8 bias.

9 **Q. Has any party's opening testimony alleged the existence of an estimation bias in**
10 **utility benchmark bids?**

11 A. William Monsen, testifying on behalf of NIPPC, has suggested that utilities tend to
12 underestimate costs and therefore experience cost over-runs. He has provided an analysis
13 in his testimony of eleven UOG projects in California, and computed an average cost
14 over-run of 7%.

15 **Q. Could an approach such as Monsen's be used to reveal estimation bias in utility**
16 **benchmark bids?**

17 A. It is theoretically possible to estimate an "expected" cost over-run, e.g., the average
18 percentage by which total costs placed into ratebase for individual projects tend to exceed
19 the cost estimates. It would be very difficult to develop an appropriate database of
20 project estimates and costs. The data would have to satisfy several conditions:

- 21 • The cost estimates would have to be verifiably those that were used to support a
22 decision to pursue a utility project, or (preferably) a decision by a regulator to
23 approve a utility's plan to construct such a project.

- 1 • The cost estimate should have been used as a basis for comparing the utility
2 project with other alternatives, in a situation where the other alternatives could
3 have been chosen had they been more cost-effective.
- 4 • The cost estimate should have been subject to the kind of scrutiny and mitigation
5 that would be applied by an Independent Evaluator in Oregon.
- 6 • The actual construction cost should not have been increased by extraordinary
7 costs which could have allowed for repricing of a competitive contract.
- 8 • Any past cost over-runs should be adjusted for the effects of price inflation, if the
9 estimate was to be used in a solicitation whose model contract adjusted prices for
10 inflation between project selection and commercial operation.

11 **Q. Why do you say it would be difficult to develop such a database?**

12 A. With effort, one can find estimates that supported the decision to pursue some utility
13 projects. Mr. Monsen says he has done so for the eleven California projects. But it may
14 not be possible to find enough to produce a good estimate. In particular, Mr. Monsen
15 cites his own difficulty in obtaining the cost estimates for a number of Oregon projects.
16 Even in cases where he found publicly available cost estimates, he appropriately
17 questioned whether those cost projections were the ones that had been used to compare
18 the utility project with other alternatives.

19 PA Consulting staff under my direction did a quick scan of publicly available
20 data, looking for datasets of actual and estimated costs of gas-fired power plants. PA was
21 unable to find good consistent data.

22 **Q. Do the project cost estimates in the Monsen testimony satisfy the conditions listed**
23 **above?**

1 A. No. In general there is no evidence provided that the estimates were subject to the same
2 kind of scrutiny as would have been applied in Oregon, nor that they were used as the
3 basis for a competitive comparison with other alternatives. I can provide more detail
4 about several of these projects.

5 **Q. Please discuss the project(s) with the largest cost over-run.**

6 A. The largest cost over-run in the NIPPC testimony is associated with four SCE peaking
7 power plants (Barre, Center, Grapeland and Mira Loma). In Decision (D.) 10-05-008,⁶
8 the California Public Utilities Commission allowed SCE to recover \$260 million in costs
9 for these four plants. Mr. Monsen quotes an estimate of \$250 million for five peaking
10 power plants, and prorates that to obtain an estimate of \$200 million for four. The
11 approved cost represents a 30% cost increase.

12 **Q. Why is that cost estimate not a good basis for comparison with Oregon benchmark
13 resources?**

14 A. SCE was ordered to install UOG peaking resources by Commissioner Michael Peevey.⁷
15 This order did not reference any cost estimate. The plants are not placed in competition
16 with any other resources, and in fact are exempted from an ongoing solicitation process.
17 SCE was permitted to establish a memorandum account in which to record the costs for
18 later reasonableness review with no explicit standard, and D.10-05-008 does not compare
19 actual costs to any prior estimate. I have not found the original source of the
20 \$250 million estimate, but it is referenced in the Resolution approving the establishment
21 of the memorandum account and in some SCE testimony in the proceeding in which the

⁶ Exhibit 103 to NIPPC's opening testimony.

⁷ Assigned Commissioner's Ruling addressing electric reliability needs in Southern California for Summer 2007, in Rulemakings 05-12-013 and 06-12-013, August 15, 2006.

1 costs were approved for ratebasing.⁸ That testimony notes that the original estimate was
2 made under conditions of “limited time”.

3 I do not believe the \$250 million estimate was intended to be rigorous or binding.
4 There was no reason for SCE to produce a rigorous cost estimate as the Commission had
5 ordered SCE to build the plants. The order was not based on the cost estimate. None of
6 the documentation I have reviewed indicates the costs were reviewed by an Independent
7 Evaluator or other party. The estimate was not used as a basis for comparison with other
8 alternatives. This is not the kind of estimate that Oregon utilities would be expected to
9 provide for a benchmark bid and the fact that it was over-run is not relevant.

10 **Q. Please provide another example.**

11 A. The NIPPC testimony includes a 26% over-run on PG&E’s Gateway plant. An earlier
12 NIPPC document in this proceeding identified this over-run as attributable to a new State
13 Water Board requirement for dry cooling.⁹ An IPP plant in a similar situation would
14 have been subject to the same unanticipated requirement, and would have incurred the
15 same additional cost. Furthermore, it is likely that the Power Purchase Agreement would
16 have assigned such Change in Law risk to the utility offtaker. In other words, the
17 increase in actual construction cost would probably have allowed for repricing of a
18 competitive contract. Even if the original cost estimate had been comparable to a utility
19 benchmark bid, the over-run is one that would have impacted ratepayer costs whether it
20 had been associated with a benchmark bid or an IPP contract. Therefore it should not be
21 included in determination of a benchmark-specific bid adder.

⁸ Note 12 in MRW & Associates, LLC, *Leveling the Bidding Field: Some Initial Steps Toward Fairly Comparing Proposals for Utility-Owned Generation and Independent Power Projects*, Oakland CA, Nov. 16, 2011, p. 11, submitted in this proceeding.

⁹ MRW & Associates, op. cit., p. 11.

1 **Q. Did you also review Mr. Monsen's statements about deferred construction costs?**

2 A. Yes, I did.

3 **Q. What is the gist of Mr. Monsen's position?**

4 A. Mr. Monsen correctly notes that utilities – indeed all plant owners – make capital
5 expenditures during the entire lifetime of a power plant. He provides an example of a
6 single power plant which faced a \$14 million capital expenditure after commercial
7 operation, which he attributes to a construction defect. He states that this should be
8 considered a construction cost and hence an over-run relative to the reported cost. He
9 then states that any capital expenditure shortly after commercial operation, which exceeds
10 the annual depreciation on the plant, is really a deferred construction cost and should
11 count as an over-run.

12 **Q. Do you agree with Mr. Monsen's position?**

13 A. No. A single example of a sizable capital expenditure made after commercial operation
14 but which arguably should have been made during construction hardly proves that all
15 post-operation capital expenditure are deferred construction costs. Furthermore, by
16 trying to use this to justify an adder to be included in benchmark bids, Mr. Monsen
17 implicitly assumes that the financial models from which the benchmark bids are derived
18 do not contain any capital expenditures during the first five years that exceed
19 depreciation.

20 **Q. How does Mr. Monsen attempt to estimate the size of such an adder?**

21 A. He says that he computed capital expenditures by plant for the set of projects he used in
22 his analysis of cost over-runs, by comparing the values in line 17 of p. 402 of the utility
23 owner's FERC Form 1. He added depreciation, based on depreciation rates found on

1 p.337 of the FERC Form 1, and expressed the result as a percentage of the cost of the
2 plant in its first year of service, which I assume he got from p. 402, line 17 of that year's
3 FERC Form 1. He then computed the average percentage for each power plant, and
4 created a capacity-weighted average over all the plants.

5 **Q. Was he able to apply this method to all of the power plants whose construction costs**
6 **he had studied?**

7 A. No, he says he was unable to get good data for the Mountainview and Humboldt plants.

8 **Q. What does he say was his numerical result?**

9 A. He proposes an adder of 5.7% per year for the first five years of operation, which would
10 be a total of 28.5% of the initial cost. By comparison, the \$14 million expenditure he
11 used as his justification, which I would assume is an extreme case chosen to make a
12 point, is only 23% of that plant's \$60 million construction cost.

13 **Q. Is Mr. Monsen's five-year period justified?**

14 A. It appears to be chosen arbitrarily. I am not sure why one should assume that "deferred
15 construction expenditures" persist for five years. Furthermore, he has data for only one
16 plant, Palomar Energy Center, which came online before 2009. As the latest FERC
17 Form 1s currently available are for 2011, he can have no more than two observations of
18 capital expenditure for a plant that came online in 2009 or later.

19 **Q. Were you able to construct a dataset similar to the one used by Mr. Monsen?**

20 A. Yes. I was able to obtain copies of the utilities' FERC Form 1s and do the computations
21 Mr. Monsen describes, with one exception.

22 **Q. What was the exception?**

1 A. San Diego Gas & Electric's FERC Form 1 reports both Miramar plants together. I can
 2 therefore compute the figures for Miramar I through 2008, and for the combination of
 3 Miramar I and II thereafter.

4 **Q. Did you verify Mr. Monsen's 5.7% figure?**

5 A. No. Table 11 shows the annual values in line 17 of FERC Form 1 page 402.

Table 11. Total plant capital costs by year, from FERC Form 1

Plant	Total Plant Cost (K\$)						
	2005	2006	2007	2008	2009	2010	2011
Barre			62,887	65,990	70,731	72,922	77,028
Center			64,387	68,384	72,294	75,181	79,333
Grapeland			56,701	62,992	65,340	67,831	71,715
Mira Loma			55,716	59,081	61,700	62,175	68,758
Gateway					438,069	445,555	447,237
Miramar I	33,436	33,885	33,885	34,184			
Palomar		467,681	469,618	490,518	502,795	509,111	539,398
Colusa						645,020	651,885
Miramar I + II					83,668	84,621	88,845

6 Next, Table 12 displays the changes in those values relative to the first-year
 7 values, averaged by plant and then converted to a capacity-weighted average. I note that
 8 the percentage changes after 2009 in the total cost of Miramar I and II are assumed to
 9 apply to the first years of Miramar II as well as the fourth and fifth years of Miramar II:

Table 12. Annual capital cost increases relative to first-year cost

Plant	Capacity	Relative Cost Increase (%) -- NOT net of depreciation					AVERAGE
		Y1	Y2	Y3	Y4	Y5	
Barre	49	4.93%	7.54%	3.48%	6.53%		5.62%
Center	49	6.21%	6.07%	4.48%	6.45%		5.80%
Grapeland	49	11.10%	4.14%	4.39%	6.85%		6.62%
Mira Loma	49	6.04%	4.70%	0.85%	11.82%		5.85%
Gateway	580	1.71%	0.38%				1.05%
Miramar I	48	1.34%	0.00%	0.89%	1.14%*	5.05%*	1.69%
Palomar	566	0.41%	4.47%	2.63%	1.35%	6.48%	3.07%
Colusa	659	1.06%					1.06%
Miramar II	48	1.14%*	5.05%*				3.09%
Capacity-weighted average							2.12%
Miramar I + II		1.14%	5.05%				

*-Values taken from Miramar I and II

1 **Q. Did you adjust these figures for depreciation?**

2 A. No, I believe Mr. Monsen erred in doing so. It had been my understanding that “Cost of
 3 Plant”, as reported on lines 13-17 of page 402 of the FERC Form 1, referred to
 4 undepreciated cost – it should not be necessary to add back depreciation. I asked
 5 Portland General Electric to ask staff involved in the production of FERC Form 1 what
 6 those lines represented and they reported back to me that they represented undepreciated
 7 costs.

8 **Q. Did you attempt any independent verification?**

9 A. Although the FERC Form 1 tables do not appear to break out depreciation by plant, such
 10 a breakout can sometimes be found in other materials included in the Form 1. For

1 example, I looked at the Notes to Financial Statements in Southern California Edison's
 2 2010 Form 1 (page 123.38). That included a breakout of SCE's original investment in
 3 four jointly-owned plants and the accumulated depreciation against those investments. I
 4 computed the depreciated net investment and compared it to the total costs from line 17
 5 of page 402. The results are given in Table 13.

**Table 13. Undepreciated and depreciated investments in SCE
 jointly owned plants, in \$Millions**

Plant	Investment in Facility	Depreciated investment	net	Total Cost (p 402, line 17)
Four Corners Units 4 and 5	596	97		583
Mohave	347	35		313
Palo Verde	1,899	356		1,829
San Onofre	5,369	1,289		4,745*

*-Units 2 and 3

6 It is clear that the costs on page 402 are of the same magnitude as the
 7 undepreciated investments, and much greater than the depreciated values. The difference
 8 for San Onofre is probably owing to the absence of Unit 1. Remaining differences
 9 between the "Investment in Facility" and "Total Cost" columns probably represent
 10 differences between GAAP and FERC accounting.

11 **Q. You have displayed a figure of 2.12% for annual capital expenditures for this set of**
 12 **power plants. Do you believe it is an accurate and unbiased estimate of "deferred**
 13 **construction costs"?**

14 **A.** No. Most of the above-average annual increases are attributable to the SCE peakers on
 15 the fifth year of operation. I have already explained that the SCE peakers were developed
 16 quickly under CPUC order so it is not surprising that some construction costs were

1 incurred after operation; they are not representative. Second, the choice of a five-year
2 period was arbitrary and in both cases where five years of data are present (Miramar and
3 Palomar) the year 5 costs are higher than any other year. Year 5 also does not appear to
4 be representative, which highlights the arbitrariness of the choice to use a five-year
5 average.

6 **Q. What is your conclusion?**

7 A. I conclude that there are few good examples of utility projects, with cost estimates that
8 are of the same nature as utility benchmark bids in Oregon, that can be used to create an
9 expected figure for utility cost estimation bias. Such bias is probably best eliminated by
10 the Independent Evaluator's active scrutiny of the benchmark bids.

11 Historical examples of cost over-runs do not by themselves imply that utility
12 benchmark bids are consistently underestimated. Oregon utilities may apply a level of
13 rigor to their benchmark bids not present in estimates presented to regulators for other
14 utility power plants. I examined the examples presented in Mr. Monsen's testimony.
15 The largest cost over-runs are associated with combustion turbines that were built on an
16 expedited basis. I do not think that the estimate that Mr. Monsen has characterized as the
17 initial cost for those plants was intended as the basis for project approval, but instead
18 there is evidence that the California Public Utilities Commission set up a mechanism
19 explicitly to account for estimation error.

20 Furthermore, cost over-runs are not unique to utility projects, as is shown by Mr.
21 Monsen's second example. Mr. Monsen attributes the over-run in the PG&E Gateway
22 project to a regulatory change. Under many PPAs that risk would be borne by customers.

1 Finally, Mr. Monsen suggests an adder for “deferred construction costs” that is
2 about five times the size of the proposed over-run adder. The justification for the adder is
3 weak, and I believe the computations that Mr. Monsen says he made to support it rely on
4 an erroneous interpretation of FERC Form 1 data.

III. Credentials

Jonathan Jacobs

Curriculum vita



Jonathan Jacobs has twenty-seven years' experience as a consultant, manager and tool builder, including twenty-two years' experience in the electricity industry on both the regulated and unregulated sides. Dr. Jacobs is an expert in the use of mathematical models and quantitative tools in electricity operations and markets. He manages PA's US practice in the review and certification of ISOs' market models, processes and systems, which provides and evaluation of the dispatch and pricing of \$20 billion worth of energy as well as ancillary services. He understands the issues around integrating renewables into the grid, having served as an Independent Evaluator of renewable-specific contract solicitations by a California utility and contributed key intellectual content to a new renewable-markets venture. He is a skilled communicator, having provided testimony to state and Federal regulators on the reasonableness of utility practices, appropriate risk metrics, and other topics. Prior to joining PA he served as Director, Market and Financial Modeling for a deregulated energy retailer. In that position he was responsible for the methods and tools the company used to price retail power contracts, for competitive analysis of utility rates, for monitoring the progress of electric industry restructuring in various jurisdictions to identify profitable target markets, and for forecasting the company's own load obligations. He has worked on diverse modeling applications including hydro forecasting and scheduling, contract valuation, portfolio analysis and system reliability modeling. He has developed large and small decision support systems, developed algorithms and managed projects of sizes ranging from \$10,000 to \$2,000,000. His doctorate is in mathematics.

Primary expertise	Related experience	Qualifications	Client list
<ul style="list-style-type: none"> • Energy market structures • Renewable capacity requirements and markets • Resource planning and procurement • Mathematical and statistical modeling 	<ul style="list-style-type: none"> • Software certification • Full-requirements pricing • California energy markets 	<ul style="list-style-type: none"> • PhD, Mathematics • Member, Institute for Operations Research and Management Sciences (INFORMS) and International Association for Energy Economics (IAEE) 	<ul style="list-style-type: none"> • Ontario IESO • ISO New England • San Diego Gas & Electric • City of Los Angeles •

Primary expertise

Resource planning and procurement – Independent evaluation of solicitations for renewable energy; procurement process design and implementation; resource portfolio modeling

Organized power markets – Review and certification of software used to clear and settle energy and capacity markets; consultation on market design and structure

Capacity requirements – Renewable portfolio standards and market-based approaches to fulfilling those standards

Mathematical modeling of energy assets – Use of optimization and simulation models to project asset operation and cash flows.

Key related achievements

San Diego Gas & Electric Company (March, 2006- present)– Since 2006, Jonathan has been the Independent Evaluator of resource procurements for a US utility with approximately 4500 MW peak load, including an all-source procurement (2006), a peaking capacity procurement (2006), procurements particularly directed at small-scale renewables, both by power purchase and turnkey construction, and renewable-specific procurements (2006, 2008, and 2009) through which the utility sought to increase its renewable procurement from approximately 5% of load (800,000 MWh) to 20%. He has also been advisor on procurement design and execution, and evaluator both of individual offers and of the utility's conduct of the solicitation.

Multiple clients (April - November, 2012) – Jonathan developed and led a benchmarking study of the efficiency and cost of energy procurement -- both purchased power and fuel for generation -- across a diverse set of utilities. The respondents were diverse in terms of geography, size and ISO participation.

Pacific Gas & Electric Co. (2012) –Jonathan advised the utility on practices elsewhere in several areas of utility energy procurement including integrated resource planning, price-based maintenance schedule, cost forecasting and procurement incentive mechanisms.

ISO New England (July, 2004 - present) – As Assignment Manager since 2005, Jonathan has certified the correctness of multiple components of the dispatch and pricing software, as well as auctions for forward obligations for operating reserves and capacity credits. In particular, the Forward Capacity Auction is a descending-clock auction with locational computation of market-clearing prices. During 2004, as a technical expert, Jonathan personally validated a particularly complex mathematical model used to impose security constraints upon the commitment and dispatch used to compute Locational Marginal Prices.

City of Los Angeles (April, 2011 - present) – Jonathan has managed an ongoing review, on behalf of the City Council, the cost structure of the Department of Water and Power including all aspects of water and electricity supply and the Department's Integrated Resource Plan. Jonathan also reviewed proposed rate structure changes. This work built on a prior assignment in which PA examined the Department's energy costs (under a pass-through clause) and an associated rate proposal.

EDI Holdings (Pty) Ltd. and the National Energy Regulator of South Africa (April - October, 2008) – Jonathan developed a report on design considerations for a contract based market structure to support the development of independent power plants, possibly involving a "Single Buyer" of new capacity.

Additional experience

New York Independent System Operator (April, 2006 - Oct. 2010) – As Assignment Manager, Jonathan certified the correctness of several releases of NYISO's settlements engine and market software (including real-time security-constrained economic dispatch, day-ahead security-constrained unit commitment and locational pricing). These systems are used to manage day-ahead and real-time scheduling of over 165,000 GWh of energy annually, and to settle transactions with an annual value of almost \$10 billion. Certification ensures that these complex systems and the mathematical models they embody are faithful to their formulations and to the market structure defined in the tariff. Certification includes the design and conduct of appropriate system tests

Eskom Holdings (South Africa) (Oct. - Dec. 2011) – Jonathan managed PA's engagement, as a subcontractor, to provide expertise in ISO structure and markets around the world. The larger assignment in which PA was involved was a due diligence study of the impacts on the incumbent utility of the potential implementation of an Independent System and Market Operator.

Independent Electricity System Operator (Ontario) (Dec. 2010 - Nov. 2011) – Jonathan led the testing and certification of substantial changes to the software that implements the day-ahead commitment market – referred to as Enhanced Day-Ahead Commitment (EDAC).

EDI Holdings (Pty) Ltd. (Oct. 2010 - Jan. 2011) –Jonathan led a project to assess European models for the restructuring of the South African electricity industry, with an emphasis on metropolitan and municipal distributors.

Independent Electricity System Operator (Ontario) (Dec. 2010 - Nov. 2011) – Jonathan led the testing and certification of substantial changes to the software that implements the day-ahead commitment market – referred to as Enhanced Day-Ahead Commitment (EDAC). Jonathan had previously led a review of the detailed design of the market changes, in order to ensure it was consistent with the plan reflected in the modifications made to

the IESO tariff. The design included rules for market clearing and 120 pages of settlement equations and procedures, suggesting many improvements and corrections.

Independent Electricity System Operator (Ontario) (April 2009 - Jan. 2010) –Jonathan led a review of the detailed design of the market changes, in order to ensure it was consistent with the plan reflected in the modifications made to the IESO tariff. The design included rules for market clearing and 120 pages of settlement equations and procedures, suggesting many improvements and corrections.

Confidential client – From 2007 to 2009, Jonathan generated and elaborated key intellectual capital for a new business venture to streamline renewables procurement and facilitate the development of renewable generation.

New York Power Authority – In 2010, Jonathan led a project to develop a business case for a system to model generation and develop “ideal” schedules as a basis for performance metrics. He had previously led a project to identify and characterize appropriate metrics for generation bidding. In 2008, as the leader of a team of PA consultants Jonathan assessed the need for and potential benefit of an optimal bidding system for the Niagara Power Project, and presented the assessment to NYPA management.

Confidential client – Jonathan assisted the utility to prepare for a regulatory audit of its load forecasting and supply planning functions by conducting a series of interviews and identifying gaps relative to best practices.

National Grid – In 2009, Jonathan developed and delivered technical training in the concept and operation of a mathematical model for valuing a portfolio of fuel and transportation contracts

Pacific Gas & Electric Co. –Jonathan advised the utility on the design of tools and processes to monitor ISO dispatch and prices and on the potential enterprise risk exposure from its commodity operations. In 2008, he provided confidential regulatory analysis. In an earlier assignment, he evaluated the design of an auction for an “open season” on transmission capacity to identify any unanticipated problem areas or gaming opportunities

Calpine Corp. –Jonathan led the development of a suite of individual cash flow models for power contracts (mostly cogenerators) to support the valuation and restructuring of those contracts

Portland General Electric –Jonathan specified and prototyped a cost simulation model, compatible with an in-house spreadsheet-based simulation model, that could be used to estimate the statistical properties of the distribution of PGE’s net variable power costs

Alberta Independent System Operator –Jonathan reviewed the data and processes used by this Canadian transmission operator to forecast load. He also surveyed similarly situated entities, in the US and internationally to ascertain best practices

Puget Sound Energy –Jonathan developed and implemented a method to weather-normalize demand and usage forecast by customer category, for use in rate setting. He supported utility staff in dealing with regulators and in subsequent workshops on how to regularize the use of weather normalization in the future

Cinergy / CG&E – Jonathan wrote a report explaining the bases for pricing default or provider of last resort (POLR) service in multiple US jurisdictions where the electricity for that service is bought from competitive markets, through auctions or Requests for Proposal, and compared the cost to the consumer in those jurisdictions. The work was used in a regulatory proceeding to set the utility’s own default or POLR pricing

Xcel Energy – Jonathan testified twice on behalf of this utility: FERC testimony on the utility’s use of a unit commitment model to estimate the avoided cost of power transactions, for use in computing charges and credits under a fuel clause; and Colorado PUC testimony on the utility’s use of a rule-based model for allocated power costs between native load and external transactions, in comparison with the use of unit commitment and production simulation models

MBIA Corp. – Jonathan advised on developments in the California energy markets and California regulation and analyzed the impacts of those developments (and of related bankruptcy filings) on the finances of Pacific Gas & Electric, Southern California Edison, the California Independent System Operator and the California Department of Water Resources

La Paloma Generating Co. – Jonathan developed a suite of optimization models for plant scheduling against price expectations, taking into account a complex tolling agreement. The models were fielded in Excel spreadsheets

Southern California Edison – Over an extended period in 2002 and 2003, Jonathan assisted the utility in

preparing to resume power procurement, with a particular emphasis on valuing wholesale power contracts and tolling agreements. He designed and implemented a contract valuation tool; helped evaluate tools for short-term operations planning and scheduling; and developed a contingency plan to be executed if vendor-supplied tools for risk management and settlement failed

Statoil ASA – Jonathan estimated the value of a "redirection option" for liquefied natural gas shipments, specifically, whether as part of an LNG shipping plan it was worthwhile to procure capacity at a second terminal, in a distinct geographic market

PSEG Power – Jonathan projected the likely clearing price for New Jersey's auctions of the responsibility for providing basic generation service, based on forward prices of energy, capacity and ancillary services, and a consistent set of risk premia. The initial forecast, prior to any such auctions having been conducted, tracked quite closely to actual results.

Additional experience

PG&E Energy Services Co. – Director, Market and Financial Modeling – 1997–2000 – Led a staff of 9 professionals, including engineers, computer modeling specialists, economists and forecasters, based in two separate locations, with the responsibility for analysis and mathematical modeling to support the company's commodity power sales, and participate in the development of strategy and tactics for the retail power business. Developed PG&E Energy Services' computer tool for pricing retail power contracts, which allowed the company to originate one of the largest contract portfolios in California. Directed the development of computer modeling for electricity pricing in emerging markets outside California.

Pacific Gas & Electric Co. – Project Manager / Team Leader – 1990–7 – Led a group of regular and contract employees responsible for the specification, selection, design, implementation and integration of mathematical models and decision support systems for medium- and long-term planning for PG&E's electricity supply business. Led the development of an Integrated Generation, Transmission and Distribution (IGTD) planning model aimed at distributed generation, and of an "energy reliability" model. Technical lead for hydroelectric scheduling and stream flow modeling.

Education

Ph.D., Mathematics, The University of Wisconsin, Madison.

M.S., Mathematics, The University of Wisconsin, Madison.

S.B., Mathematics, The University of Chicago.
