

Distribution Forecasting Working Group

Uncertainty and Proposals to Improve DER Methods

Meeting 2: May 2nd 2018



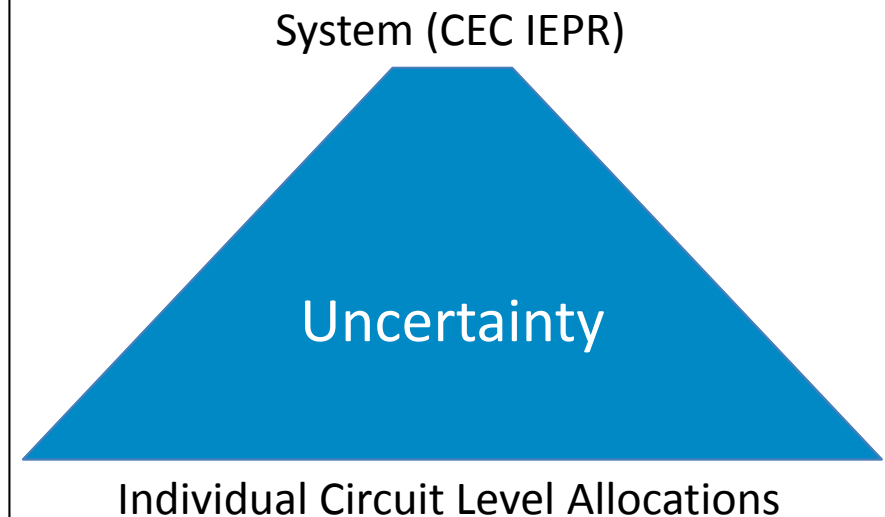
Key Sources of Uncertainty in DER Forecast Allocation

<p>Universal Sources of Uncertainty:</p> <ul style="list-style-type: none"> • Uncertainty in the system level forecast propagates down and compounds • Data availability and quality • Amount of recorded historical adoption • Methodology • Impact profiles/customer dispatch behavior • Weather • Adoption not necessarily independent between DERs • Technical potential uncertain • ... 	<p>DER Specific Drivers of Uncertainty in the Allocations:</p> <table border="1"> <tr> <td data-bbox="1024 483 1108 695">PV</td> <td data-bbox="1119 483 1967 695"> <ul style="list-style-type: none"> • “Lumpy” non-res adoption • Rapidly changing policy/economic landscape • Limited information on roof type/shading </td> </tr> <tr> <td data-bbox="1024 703 1108 849">EV</td> <td data-bbox="1119 703 1967 849"> <ul style="list-style-type: none"> • Location of most EV customers unknown • Driving & charging patterns • Bias toward large battery size adopters </td> </tr> <tr> <td data-bbox="1024 857 1108 1109">EE</td> <td data-bbox="1119 857 1967 1109"> <ul style="list-style-type: none"> • Impact shape varies by measure adoption • Unknown distribution of “upstream” measures • Customer class definition between CEC and PG&E </td> </tr> <tr> <td data-bbox="1024 1117 1108 1287">DR</td> <td data-bbox="1119 1117 1967 1287"> <ul style="list-style-type: none"> • Load consumption variation (significant deviation of actual load may underestimate or overestimate DR projections) • Fluctuating enrollments and engagement </td> </tr> <tr> <td data-bbox="1024 1295 1108 1401">Storage</td> <td data-bbox="1119 1295 1967 1401"> <ul style="list-style-type: none"> • Extremely limited historic data • Dispatch behavior not guaranteed </td> </tr> </table>	PV	<ul style="list-style-type: none"> • “Lumpy” non-res adoption • Rapidly changing policy/economic landscape • Limited information on roof type/shading 	EV	<ul style="list-style-type: none"> • Location of most EV customers unknown • Driving & charging patterns • Bias toward large battery size adopters 	EE	<ul style="list-style-type: none"> • Impact shape varies by measure adoption • Unknown distribution of “upstream” measures • Customer class definition between CEC and PG&E 	DR	<ul style="list-style-type: none"> • Load consumption variation (significant deviation of actual load may underestimate or overestimate DR projections) • Fluctuating enrollments and engagement 	Storage	<ul style="list-style-type: none"> • Extremely limited historic data • Dispatch behavior not guaranteed
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Fundamentally, we are trying to predict human behavior (customer level and policy)

Range of Uncertainty

- Uncertainty is smallest at the system level
- Any forecast uncertainty at the system level is magnified as the forecast is allocated
 - CEC to discuss uncertainty/calibration process at the system level in Mtg 3
- Factors which create more uncertainty at the local level
 - Number of observations
 - Potential for offsetting error is reduced
 - More impact from individual customer decisions
- Local Planners and an annual distribution plan update serve as a feedback loop to mitigate uncertainty



Degree to which Evaluation can Reduce Uncertainty

- Uncertainty is unavoidable
- Improved/additional data can mitigate to some extent
- IOUs have identified the following datasets which could support reduction of uncertainty:
 - DMV data for Electric Vehicles
 - Actual DER output data (Currently most DER are not metered separately)
- We can also reduce uncertainty by continuously improving our allocation methods as we get more data, more adopters, and more experience using the allocations in distribution planning



Proposals to Improve Methods in 2018/19 Planning Cycle through actual data

- Each year IOUs refresh latest actual adoption data
 - Models are trained on the latest data
 - Independent variables tested for significance (where relevant to model)
- Continued collaboration with the CEC is required
 - There can be a disconnect between latest actual adoption trends and the prior CEC forecast due to the lag between forecast development and use for Distribution planning

