







# Easing the Squeeze in DSM Portfolio Planning: A Quantitative Analysis of Potential Impacts of Feedback/ Behavior-Based Programs

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#### **Agenda**

- Initial Big Squeeze Analysis (2011)
- Big Squeeze II: Objectives & Methodology
- Feedback Program Types Five Scenarios
- Uncertainty Analysis Monte Carlo Simulation
- Probability Distributions of Required Inputs Parameters
  - Annual Household Electricity Savings
  - Participation Rate
  - Useful Life
- Potential Impacts on a Real DSM Portfolio Electricity Savings

# **Initial Big Squeeze Analysis on Feedback Programs (2011)**

### **Initial Big Squeeze Analysis (2011)**



- Objective: To assess the "savings gap" between state-mandated saving targets, i.e. EERS, and current DSM portfolios, and examine options to fill it
- Methodology: Modeled a typical DSM portfolio against a typical EERS target
  - ICF's EEPM model used as the "engine"
  - Built a generic DSM portfolio based on ICF client experience
  - Used ACEEE data to calculate a typical EERS target
  - Established a baseline scenario and "savings gap"
  - Re-estimated the baseline and gap with federal lighting and appliance standards
  - Developed several "gap-filler" scenarios

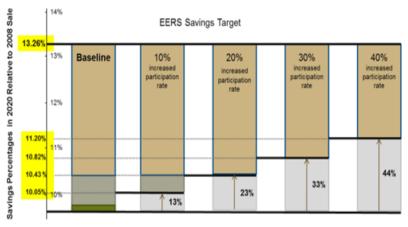


#### "Big Squeeze II":





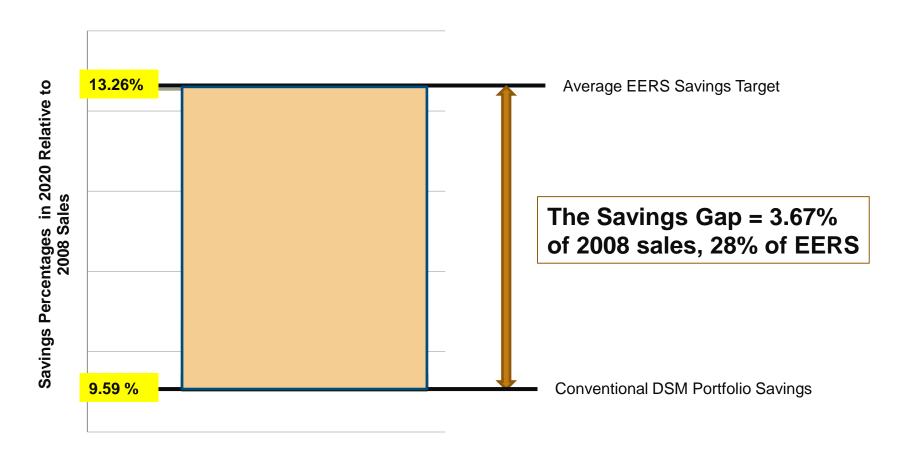
- Thorough review of existing literature and recent evaluation data
- Used a more robust statistical technique Monte Carlo simulation through @Risk software
- Quantified how various feedback types can fill the savings gap estimated previously



DSM Planning Portfolio Savings

### The Savings Gap





☐ By 2020, conventional DSM would likely fall 28% short of an average EERS target

### "Big Squeeze II": Overview & Objectives



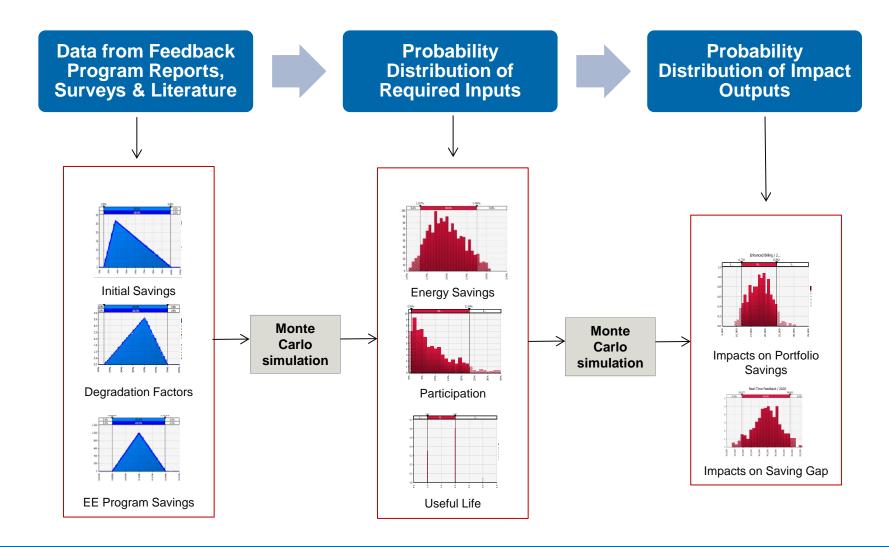
- Feedback/behavior-based programs have been recognized as a mechanism to provide deeper energy savings and higher customer satisfaction.
- Significant uncertainty associated with their performance due to the limitations of robust ex post program evaluation data

#### Objectives:

- Quantify the impacts of various feedback programs within a larger DSM portfolio while explicitly accounting for uncertainties associated with their performance.
- Assess how these emerging programs can fill the gap between the projected savings from typical DSM portfolios and state-mandated saving targets

### **Uncertainty Analysis - Monte Carlo Simulation**





# Feedback Program Types – Five Scenarios



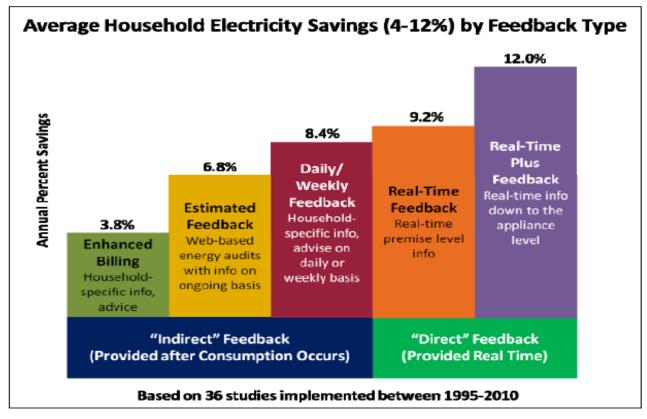




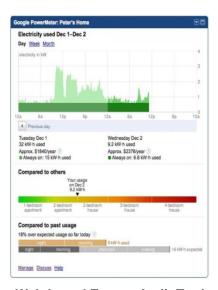


### 2010 ACEEE Study – A Meta-Review









**Web-based Energy Audit Tool** 



In-Home Energy Display Device

### **Five Feedback Program Scenarios**



Scenario No.	Scenario Name	Feedback Type	Participation Plan	Description
Scenario 1	Enhanced Billing	Indirect	Opt-out	Household Specific Information and advice
Scenario 2	Estimated Feedback	Indirect	Opt-in	Web-based energy audits without info on ongoing basis
Scenario 3	Daily/Weekly Feedback	Indirect	Opt-in	Household specific info & advice on daily/weekly basis
Scenario 4	Real-Time Feedback	Direct	Opt-in	Real time consumption & cost info at the aggregated level
Scenario 5	Real-Time Plus Feedback	Direct	Opt-in	Real time consumption & cost info disaggregated at appliance level

### Opt-out Participation Plan Broad program reach, shallow savings High continuing costs to maintain savings e.g. mailers

### Opt-in Participation Plan Narrow program reach, deep savings Upfront cost to acquire, low continuing cost e.g. give email address



**Web-based Energy Audit Tool** 

# **Electricity Savings Associated with Various Feedback Programs**



### **Total Electricity Saving Impact**



 It is widely accepted that the behavior-based programs reduce energy use by

a) Direct Energy-Use Reduction (Change in Energy-Use Behaviors)



b) Increased Participation in Existing Residential Energy Efficiency Programs/Technologies (Energy Efficiency Investment Behaviors)

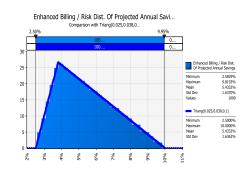


### a) Direct Energy-Use Reduction – Initial Assumptions



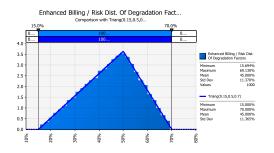
1. Used saving ranges from the 2010 ACEEE study as the basis

Household Electricity Saving Ranges from 2010 ACEEE Meta-Review of 36 Study				
No.	Name	Min	Average	Max
Scenario 1	Enhanced Billing	2.50%	3.80%	10.00%
Scenario 2	Estimated Feedback	5.00%	6.80%	8.50%
Scenario 3	Daily/Weekly Feedback	4.00%	8.40%	19.00%
Scenario 4	Real-Time Feedback	0.50%	9.20%	18.00%
Scenario 5	Real-Time Pluse Feedback	9.00%	12.00%	18.00%



Degraded the saving estimates by the following factors to match the assumptions with estimates from recent evaluation studies on existing feedback programs

Degradation Factors				
No.	Name	Min	Most-Likely	Max
Scenario 1	Enhanced Billing	45%	65%	70%
Scenario 2 & 3	Estimated & Daily/Weekly Feedback	30%	50%	60%
Scenario 4 & 5	Real Time Feedback	20%	40%	60%

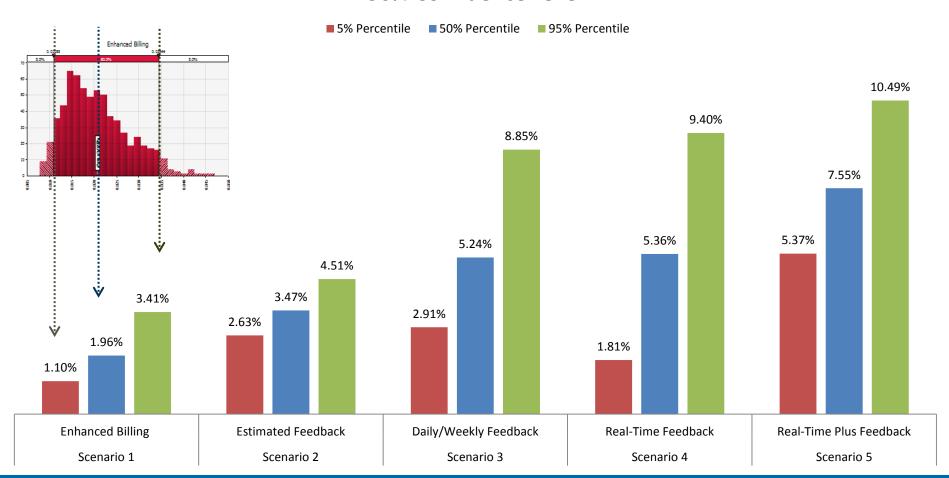


 Subtracted range of 0 – 2% to remove the potential savings resulting from increased participation in existing EE program

### a) Direct Energy-Use Reduction Distributions



### Household Electricity Saving Ranges for Various Feedback Program Types at 90% Confidence Level



# b) Increased Participation in Existing EE Programs



 Assumptions for increased participation for Enhance Billing scenario in 2013 & the annual growth based on the OCD study

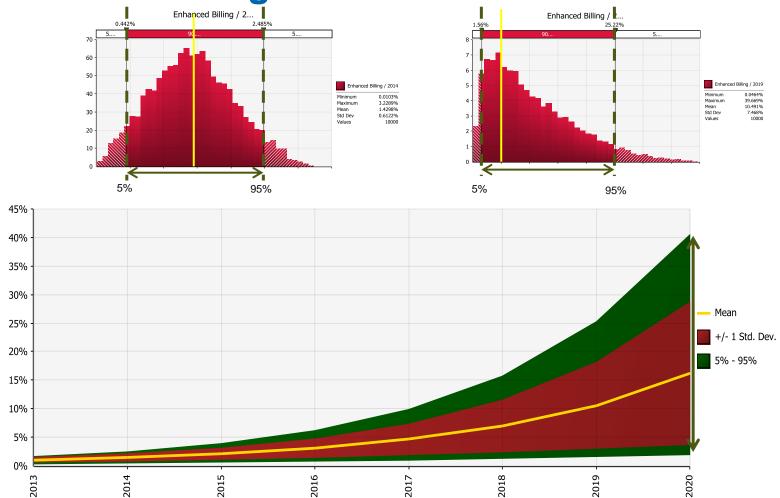
Enhanced Billing	Min	Most Likely	Max
Increased Participation in Existing Residential EE Programs in 2013	0%	1.0%	2.0%
Annual Participation Growth Rate	0%	67%	67%

- 2) Data from a study by Karen Ehrhardt-Martinez, presented in 2012 ACEEE Summer Study, as well as average savings for each EE program type were used to estimate a weighted average for the impact of each feedback programs. By comparing the estimated weighted averages, we conclude that
- On average customers with online feedback are about 6% more likely to participate in EE programs
- On average customers with real-time feedback are about 45% more likely to participate in EE programs

Likeliness of Increased Participation in Existing EE Programs				
No.	Min	Average	Max	
Scenario 1	100%	100%	100%	
Scenario 2 &3	100%	106%	120%	
Scenario 4 & 5	120%	145%	200%	

Increased Participation Input Distribution – Enhanced Billing



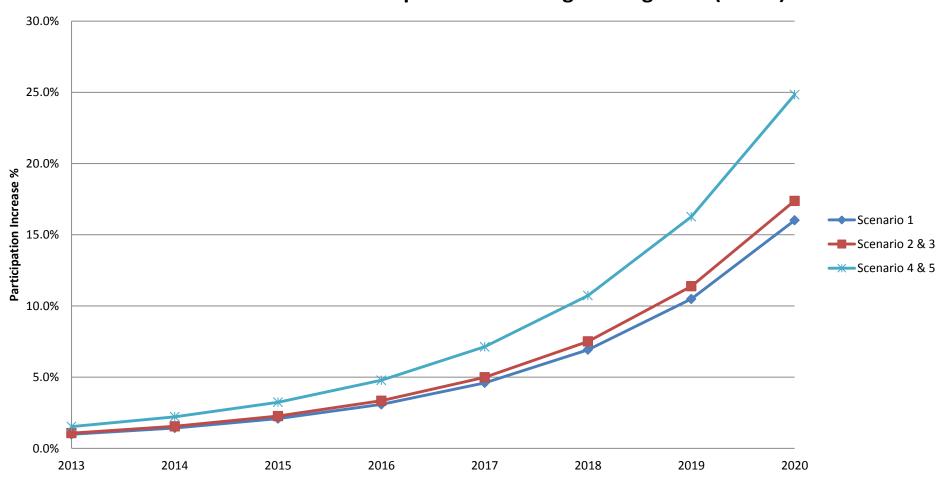


Enhanced Billing can be expected to increase EE program participation by a range of 2%-40% with the mean of 16% at the 90% confidence level

# b) Increased Participation in Existing EE Programs



#### **Increase in Annual Participation in Existing EE Programs (Mean)**



# Feedback Program Participation & Useful Life Input Distributions

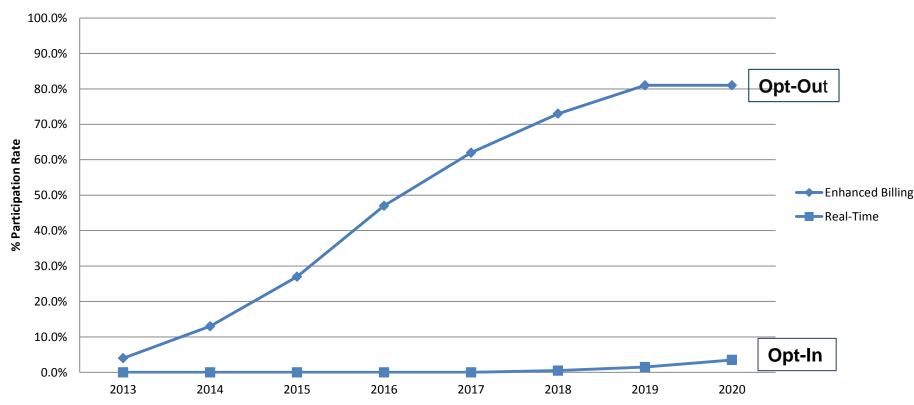


### **Program Participation Inputs – Scenario 1**



Participation in Enhanced Billing starts at 3%-5% of household in 2013 and ramp up to 74%-87% by 2019 & Real-Time Feedback starts at 0%-1% in 2018 and ramp up to 2%-5% by 2020

#### **Cumulative Participation Rate (Mean) - Scenario 1**

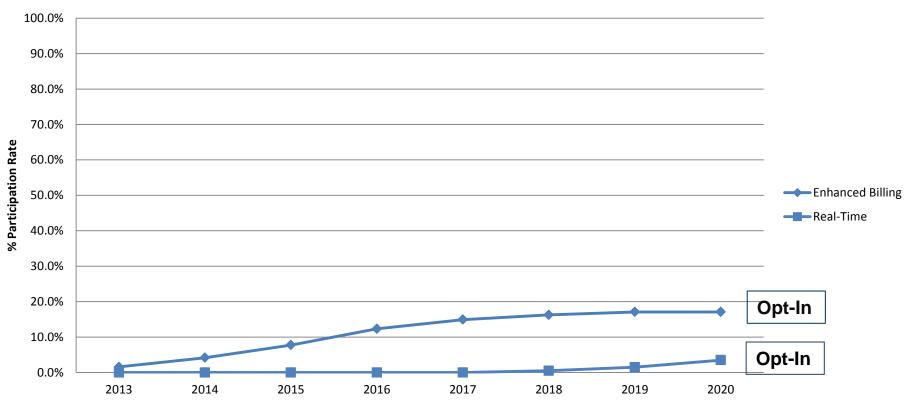


#### **Program Participation Inputs – Scenario 2&3**



□ Participation in Estimated Feedback or Weekly/Daily Feedback starts at 1%-2% of household in 2013 and ramp up to 14%-20% by 2019 & Real-Time Feedback starts at 0%-1% in 2018 and ramp up to 2%-5% by 2020

#### **Cumulative Participation Rate (Mean) - Scenario 2 &3**

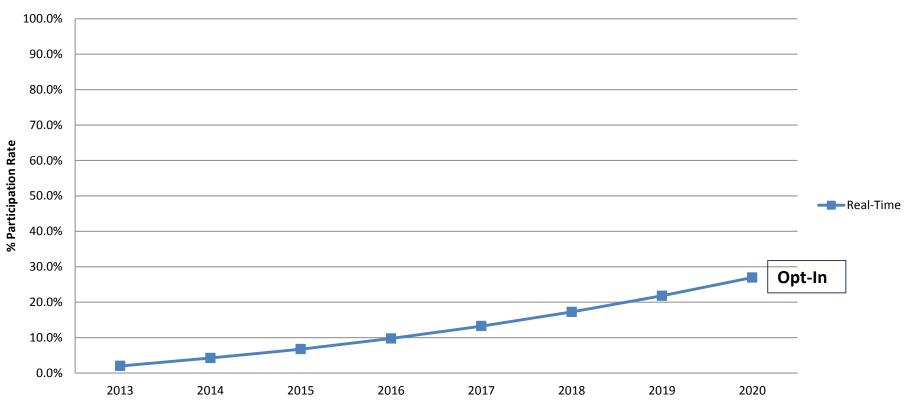


### **Program Participation Inputs – Scenario 4&5**



□ Participation in Real-Time Feedback starts at 1%-3% in 2013 and gradually ramps up to 19%-34% by 2020

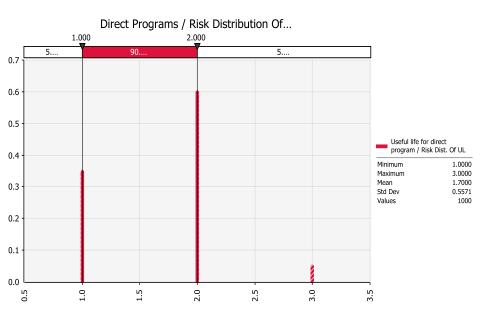
#### **Cumulative Participation Rate (Mean) - Scenario 4 & 5**

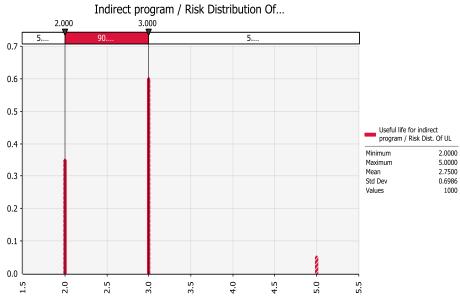


### **Useful Life Input Distributions**



Feedback Program Useful Life	Min	Most Likely	Max
Indirect Programs	1	2	3
Direct Programs	2	3	5
Probability of Occurrence	35%	60%	5%





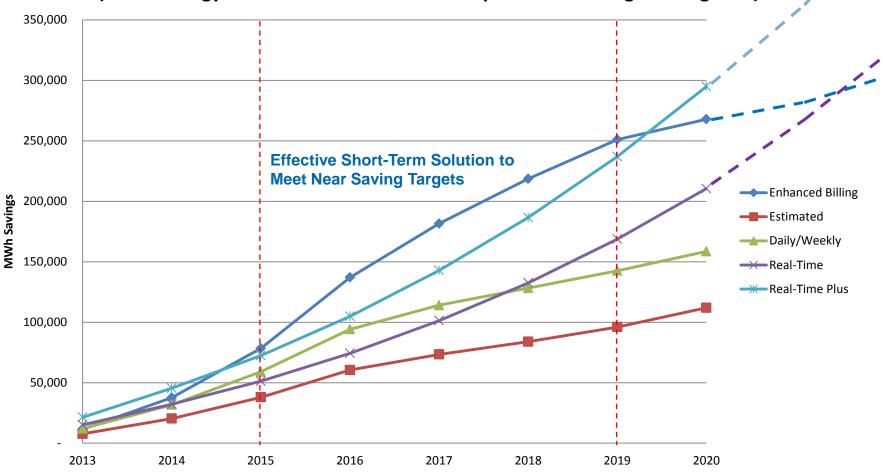
# Potential Impacts on a DSM Portfolio Electricity Savings



### **Total Cumulative Electricity Savings Impacts**



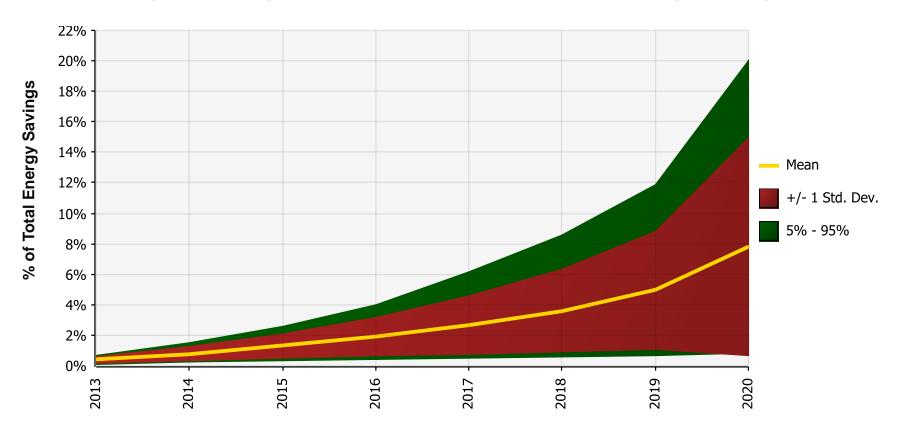




### Impact of Increased Participation in EE



#### **% Savings Resulting from Increased Participation in Existing EE Programs**

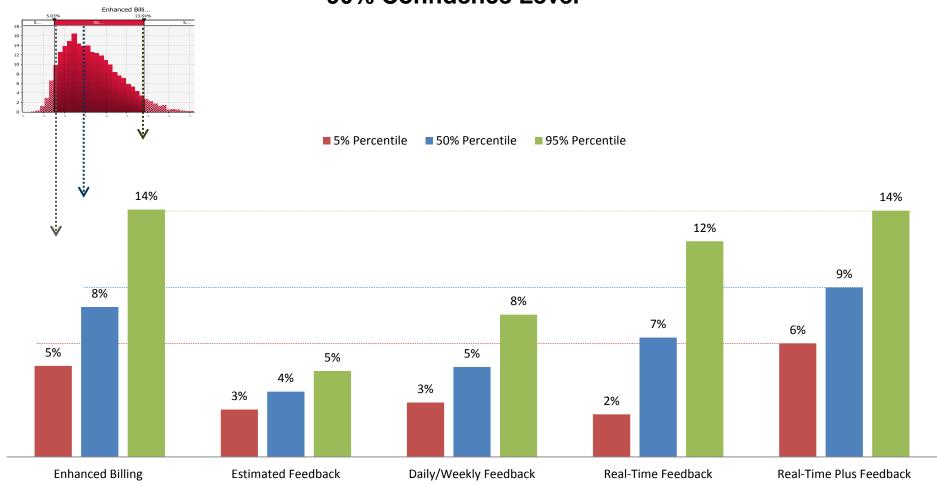


The impact of increased participation on existing EE program savings starts from **0.1%-0.7%** with the mean of **0.4%** of total program savings in **2013** and increases to **1%-20%** with the mean of **5.7%** of total program savings in **2020**.

# Impact of Feedback Programs on Total Portfolio Savings



Increased Total Portfolio Savings% in 2020 by adding Feedback Programs - 90% Confidence Level



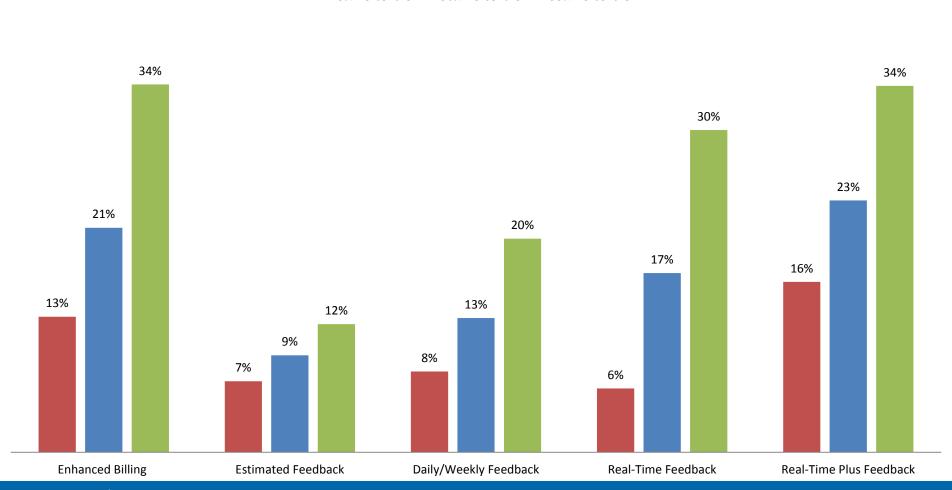
# Impact of Feedback Programs on Residential Portfolio Savings

■ 5% Percentile ■ 50% Percentile



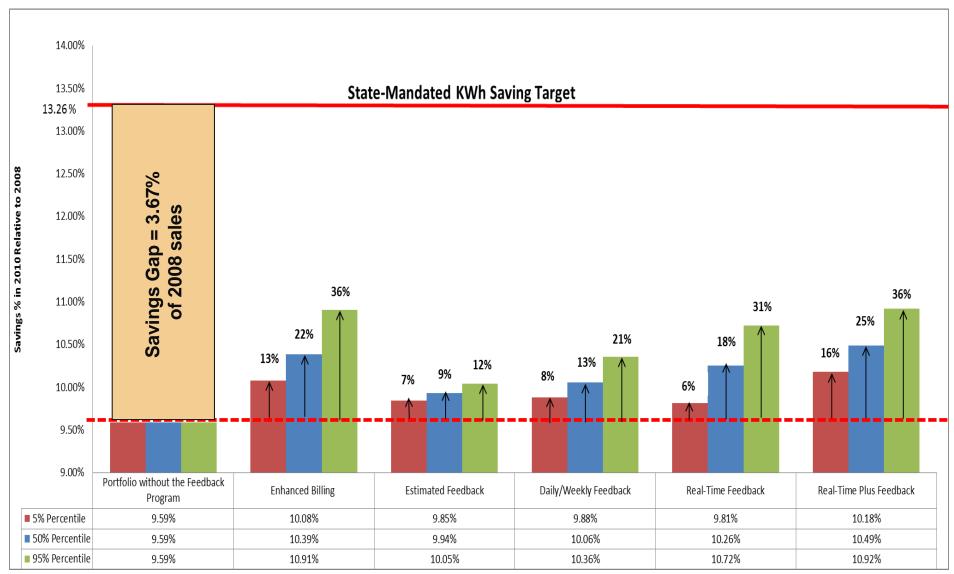
Increased Residential Portfolio Savings% in 2020 by adding Feedback Programs - 90% Confidence Level

■ 95% Percentile



# Percentage of Savings Gap Potentially Filled by Feedback Programs





#### **Conclusions**



- This "Big Squeeze II" analysis confirms previous studies on the potential contributions of residential feedback programs to portfolio savings.
- By 2020, there is a 90% chance that feedback programs increase total DSM portfolio electricity savings by 2% 14% with the average of 7%.
- By 2020, there is a 90% chance that feedback programs increase residential portfolio electricity savings by 6% - 35% with the average of 17%.
- Program planners/administrators can fill 7%-36% (average of 17%) of the 2020 EERS savings gap by integrating residential feedback programs into DSM portfolio planning.
- The impact of Enhanced Billing could be significant in **short-term** compared to other feedback types. Given its lower cost of the implementation, this approach could be a very effective short-term solution to meet the near state-mandated saving targets.





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### **Next Steps**

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- Cost effectiveness analysis (e.g. TRC test)
- Further sensitivity analysis of saving distributions to input variables and identify the level of impact
- Thorough analysis of change in patterns of final savings across the range of input variables to better understand the impact of each input on savings
- Identify & prioritize the areas of focus for program planners and administrators to more effectively unlock the potentials of feedback programs

