

# Household Water and Energy Use - the Nexus that Connects Us

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# Outline

- Perspectives on Water/Energy Analysis
- Purposes of Analysis
- Approaches to Analysis
- Alvar Escriva-Bou's modeling results
- Hard parts/Opportunities/Conclusions

# Water/Energy Analysis Perspectives

Water connects us all with each other and the world

- Household Perspective
- Water and Energy Utility perspectives
- Society perspective – local, regional, national
- Global perspective

Each perspective has different purposes

# Purposes of Analysis - Households

- Happiness
- Minimize cost
- Reduce energy and water use
- Drought management

Households don't pay for analysis.

# Purposes of Analysis - Water & Energy Utilities

- Minimize cost
- Demand management – peak energy/water use
- Reduce energy and water use
- Drought management
- Other extreme events – outages, etc.
- Use data from smart meters, network sensors
- Utilities pay for analysis (but not enough).

# Purposes of Analysis - Society

- Minimize cost
- Reduce energy and water use
- GHG emission reduction
- Drought management
- Utility and environmental regulation
- Societies pay little for analysis rarely.

# Approaches to Analysis - Empirical

- Big and medium data – Econometrics, regression, machine learning – “top down”
- Advantages – Real data & experiences, immediate applicability, common methods
- Disadvantages – Conditions of calibration data, future changes, causality unclear
- Examples – Price elasticity of demand studies since Gottlieb 1963; gobs of local & meta-studies

# Approaches to Analysis - Mechanistic

- Build demands from end-uses with behavioral assumptions, often optimization – “bottom up”
- Advantages – Mechanistic, detailed causal understanding of causes of demand and changes
- Disadvantages – Some end uses lack data; household motivations not completely clear; never completely mechanistic
- Examples – Rosenberg and Abdallah (2014); Escriva-Bou (2015); Lund (1995)



# Alvar Escriva-Bou's results +

Escriva-Bou, A., J. R. Lund, and M. Pulido-Velazquez (2015), Modeling residential water and related energy, carbon footprint and costs in California, *Environ Sci Policy*, 50(0), 270-281.

Escriva-Bou, A., J.R. Lund, M. Pulido-Velazquez, “Optimal residential water conservation strategies considering embedded energy in California,” *Water Resources Research*, Volume 51, Issue 6, pages 4482–4498, June 2015.

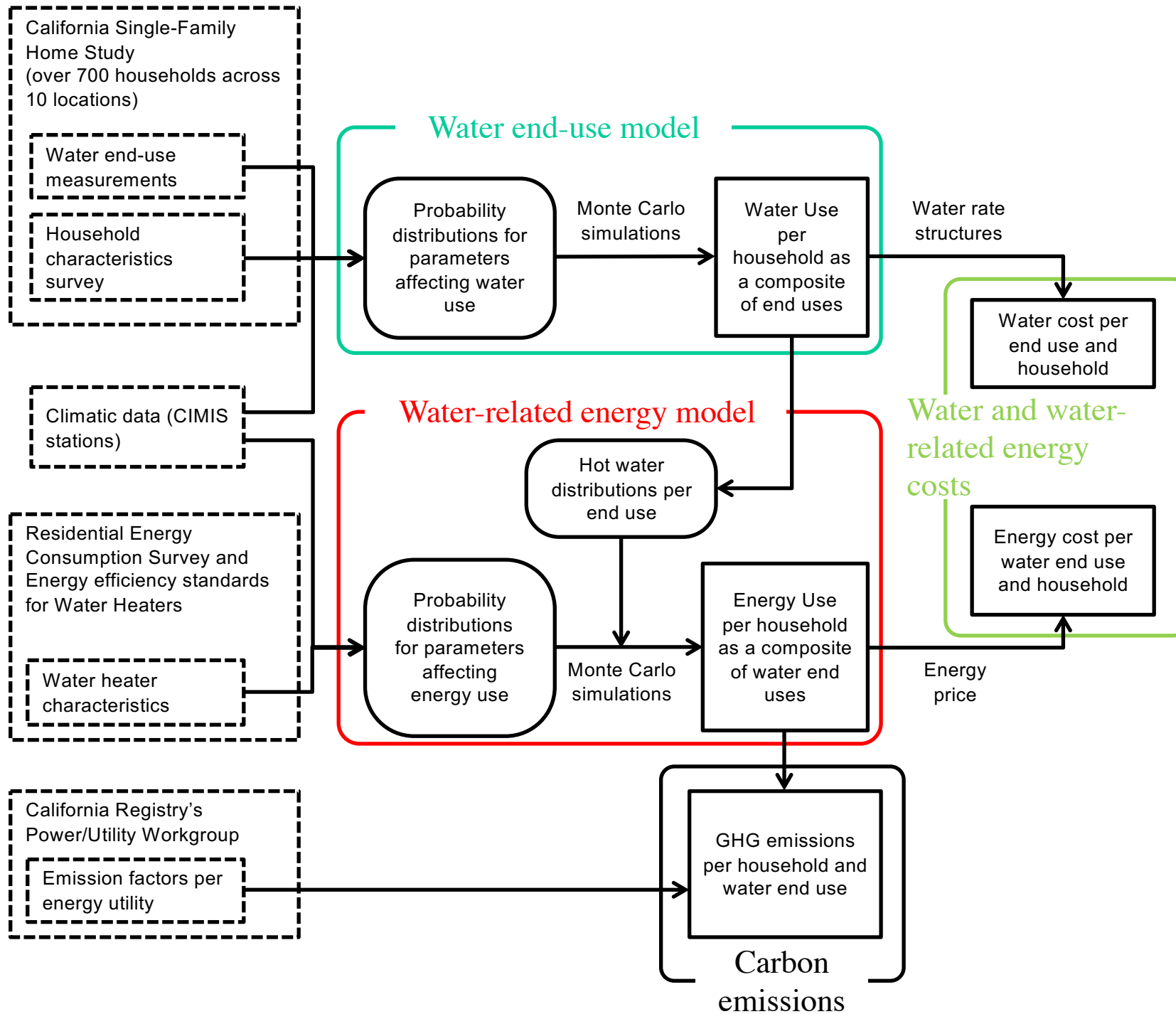
Following...

Rosenberg, D.E., T. Tarawneh, R. Abdel-Khaleq, and J.R. Lund, “Modeling Integrated Water-User Decisions in Intermittent Supply Systems,” *Water Resources Research*, Vol. 43, No. 7, July 2007.

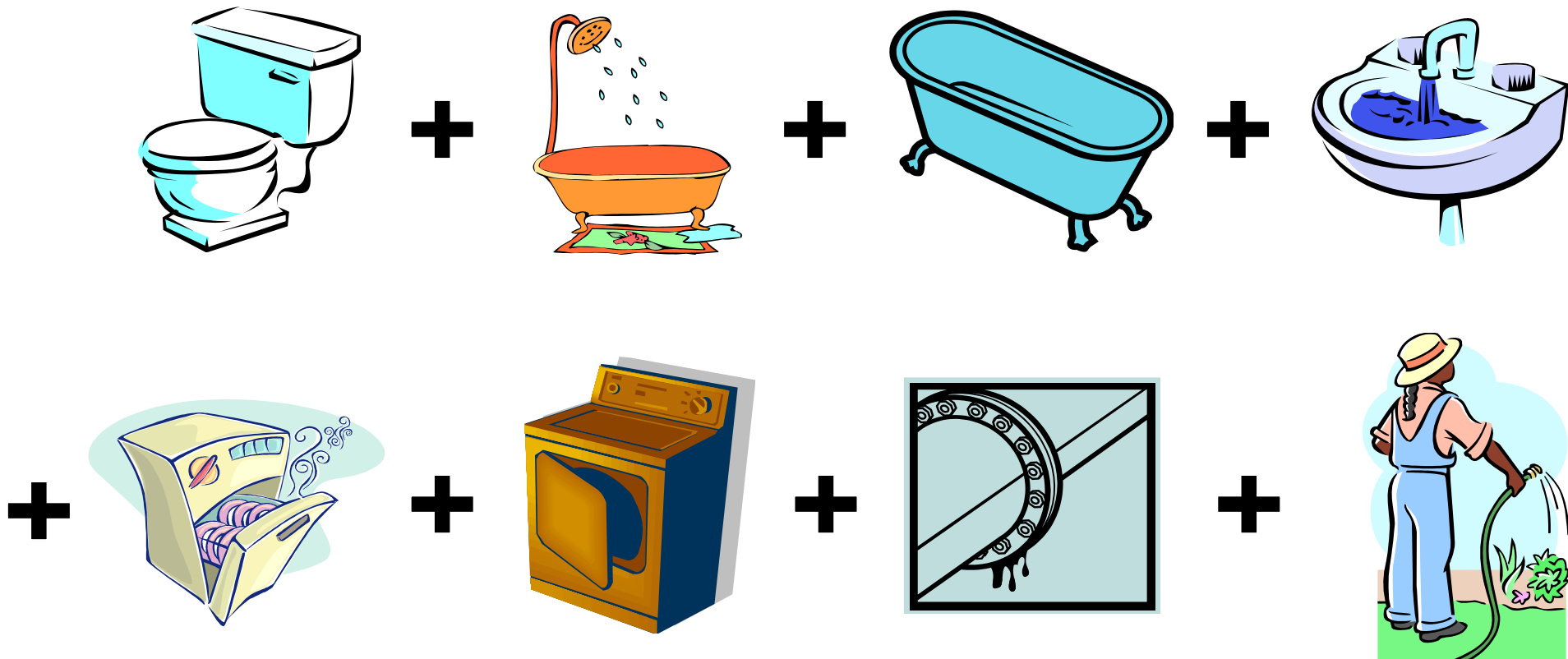
# 1) Residential water and related energy, carbon footprint and costs in California\*

- What energy and GHG emissions come from residential water end-uses?
- Does spatial variability and heterogeneity affect water and energy use?
- How do water and energy rate structures affect costs to households?

\*Escriva-Bou, A., J. R. Lund, and M. Pulido-Velazquez (2015), Modeling residential water and related energy, carbon footprint and costs in California, *Environ Sci Policy*, 50(0), 270-281.



# Water End-Use Model



# Water End-Use Models

$$Q_{shower} = \frac{(\#Std\ Shw) \cdot (Q_{std}) + (\#LowFlowShw) \cdot (Q_{LFShw})}{\#Showers} \cdot (Shower\ Length) \cdot (Shower\ Frequency) \cdot (\#Residents)$$

$$Q_{outdoor} =$$

$$ET \cdot (AreaLawn \cdot k_{cLawn} + AreaGarden \cdot k_{cGarden} + AreaPool \cdot k_{cPool}) \cdot ApplicationRatio$$

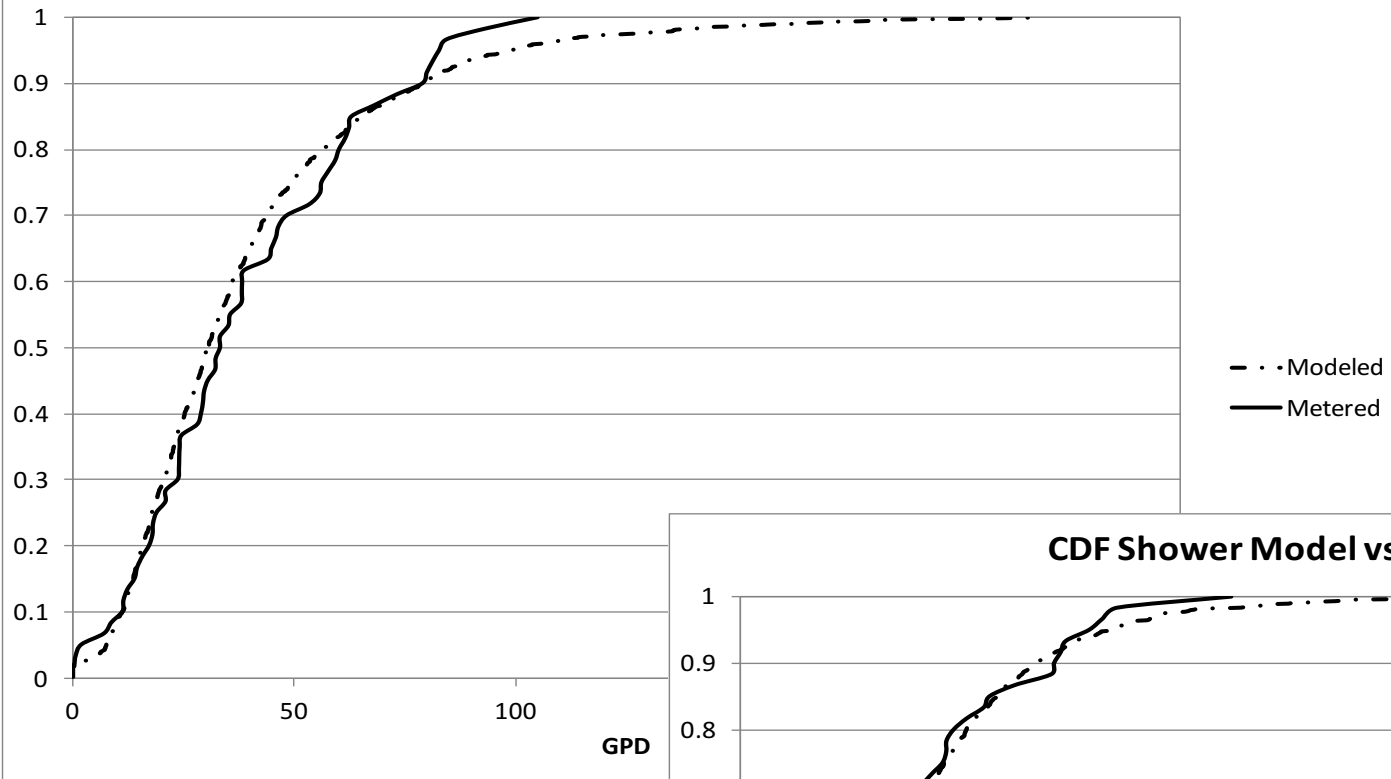
Monte Carlo analysis representing variability in →→→→→→→→→→

Household characteristics

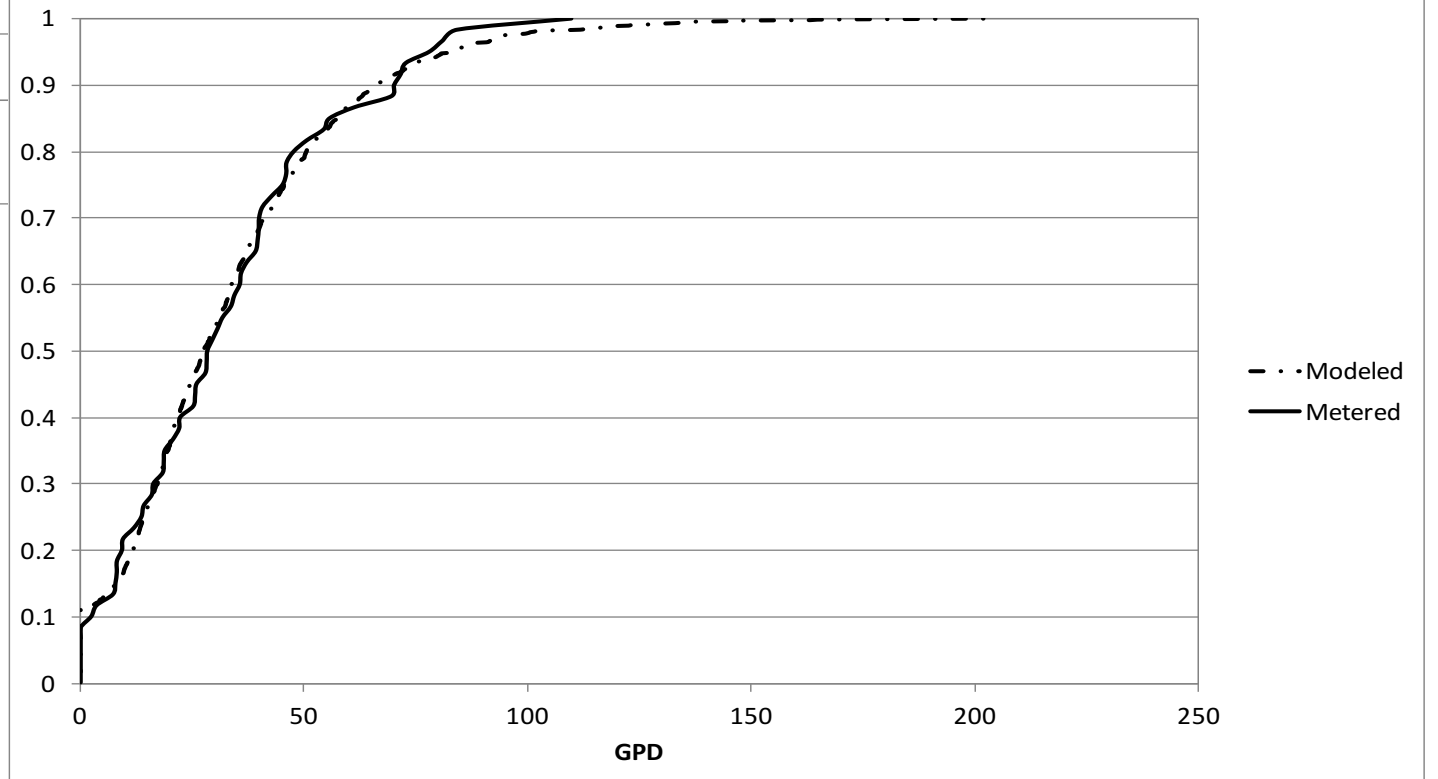
Users' behaviors

External conditions

**CDF Toilet Model vs. Real Data in Davis**



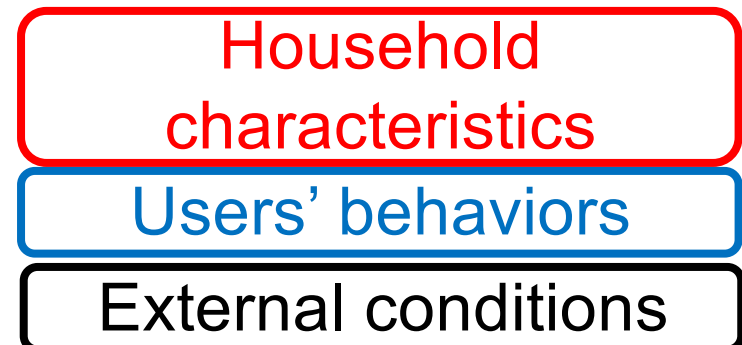
**CDF Shower Model vs. Real Data in Davis**



# Water-Related Energy End-Use Model

- From End-Water Uses → Hot water, using hot water prob. distributions per end-use (EBMUD, 2002).
- Energy Calculation - WHAM (Lutz et al., 1999):

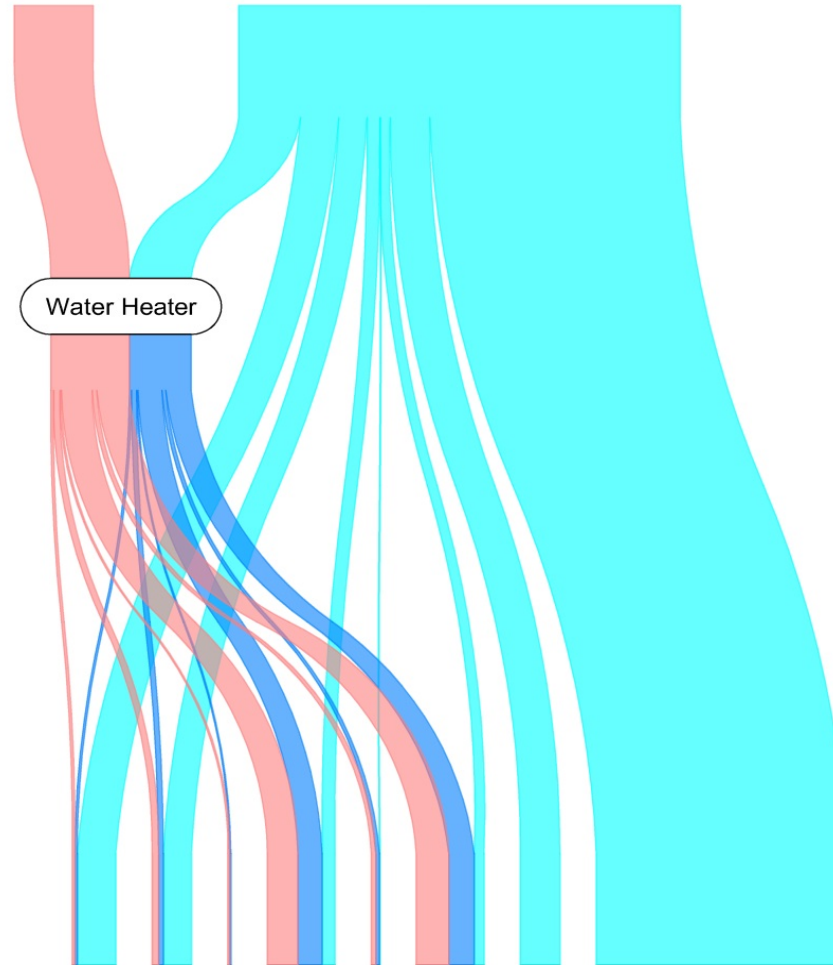
$$Q_{in} = \frac{vol \cdot den \cdot Cp \cdot (T_{tank} - T_{in})}{\eta_{re}} \cdot \left( 1 - \frac{UA \cdot (T_{tank} - T_{amb})}{P_{on}} \right) + 24 \cdot UA \cdot (T_{tank} - T_{amb})$$



# California overall results per household

Total Energy:  
10.4 kWh/day  
(728.1 kg CO<sub>2</sub>/year)

Total Water:  
364.0 GPD



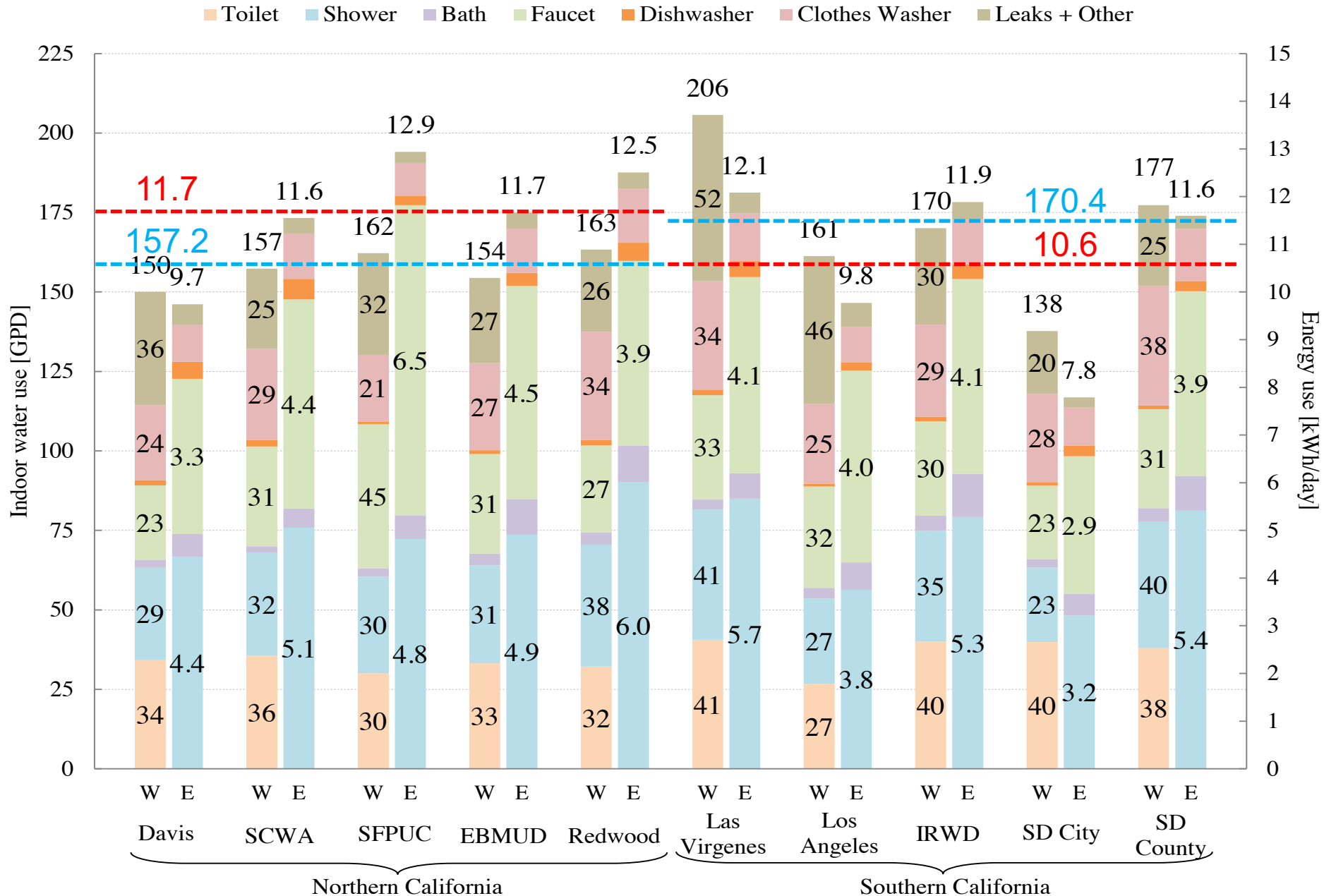
	L/O	CW	DW	Faucet	Bath	Shower	Toilet	Outdoor	Total
Cold Water (GPD)	31.7	23.1	0.0	10.6	0.4	8.1	32.5	206.6	313.0
Hot Water (GPD)	1.8	4.2	1.1	19.9	3.0	20.9	0.0	0.0	51.0
Energy (kWh/day)	0.4	0.8	0.2	4.0	0.6	4.3	0.0	0.0	10.4
CO <sub>2</sub> (kg/year)	26.2	58.8	16.2	283.4	42.6	300.8	0.0	0.0	728.1
Water cost (\$/month)	7.3	6.2	0.3	6.8	0.7	6.5	7.5	44.3	79.8
Energy cost (\$/month)	0.5	1.1	0.3	5.4	0.8	5.8	0.0	0.0	13.9
Total cost (\$/month)	7.8	7.3	0.6	12.2	1.6	12.6	7.5	44.3	93.7

80% of total  
water-related  
energy

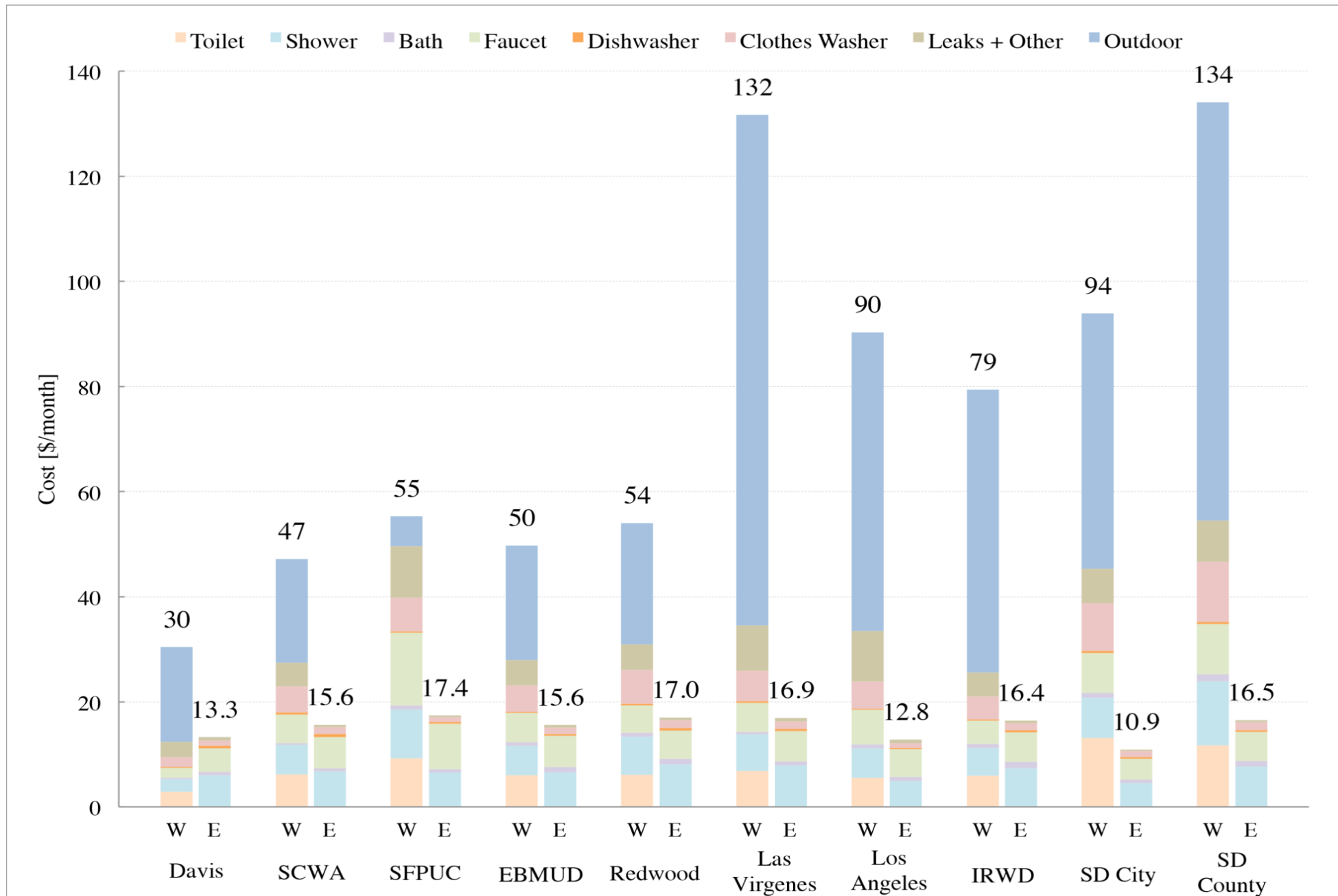
2% total per  
capita GHG  
emissions



# Household water and energy per city

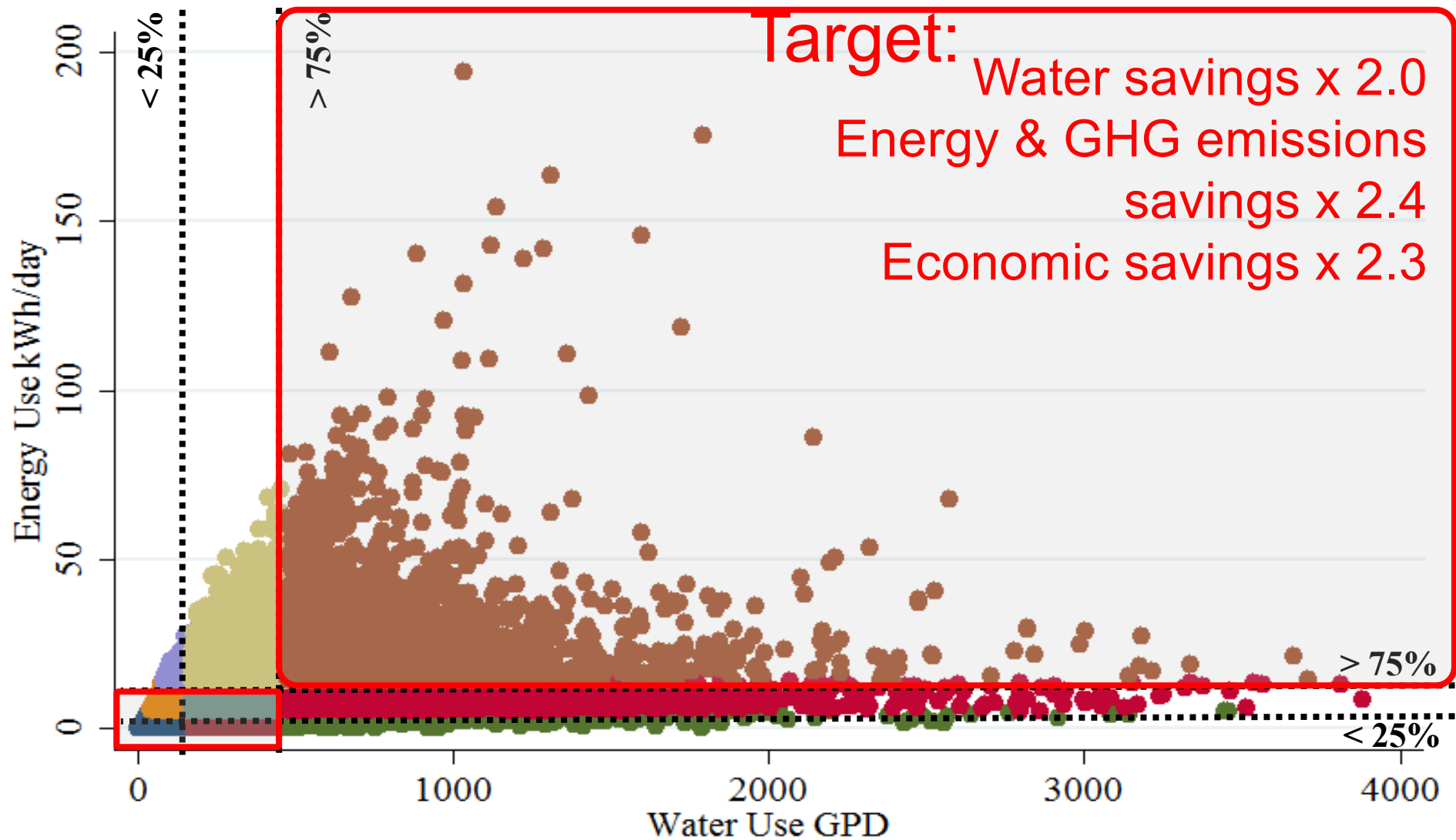


# Household water and energy costs per city

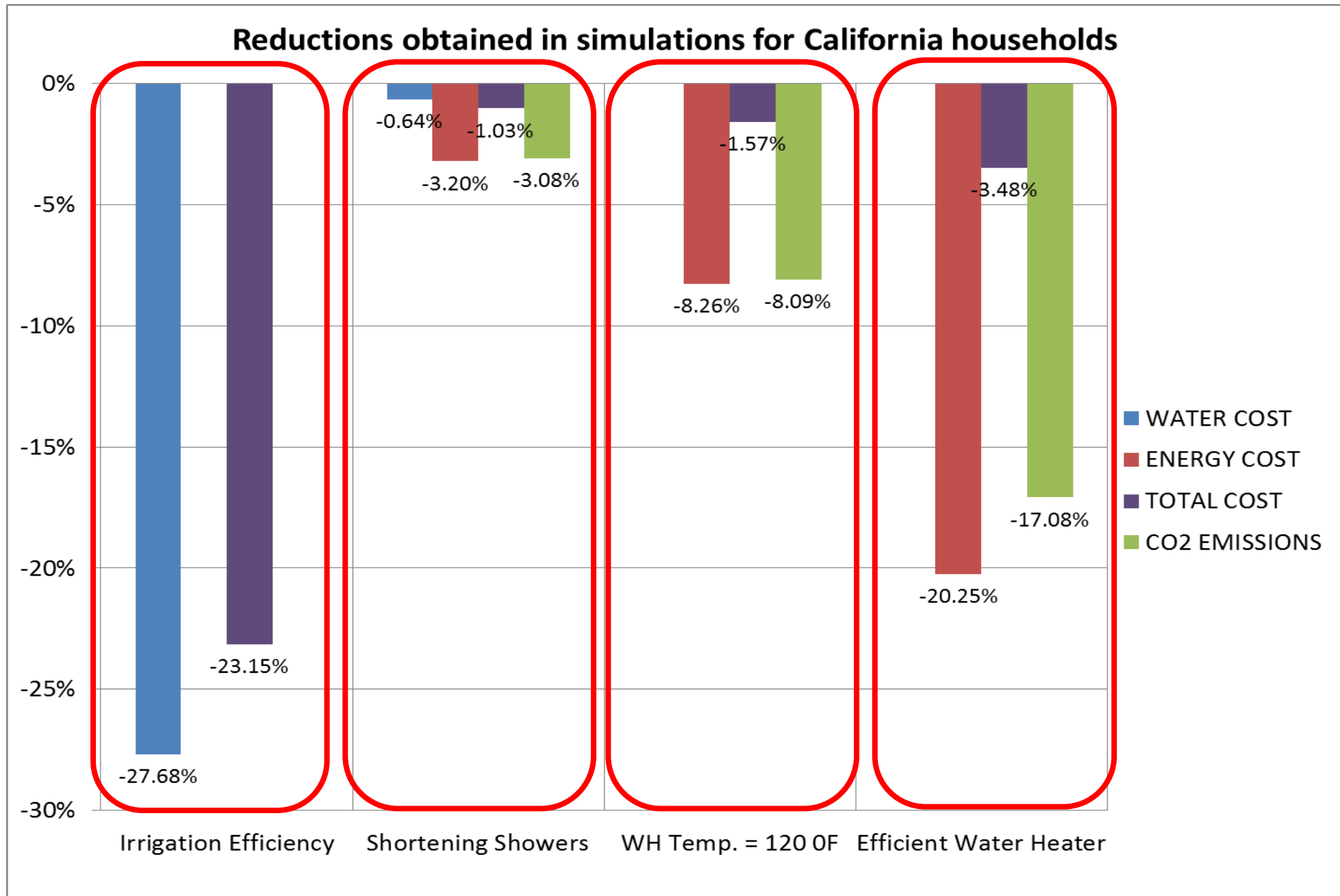


# Heterogeneity in consumption

Water and Energy Use per Household



# Results show potential for joint management



# Policy implications from mechanistic modeling

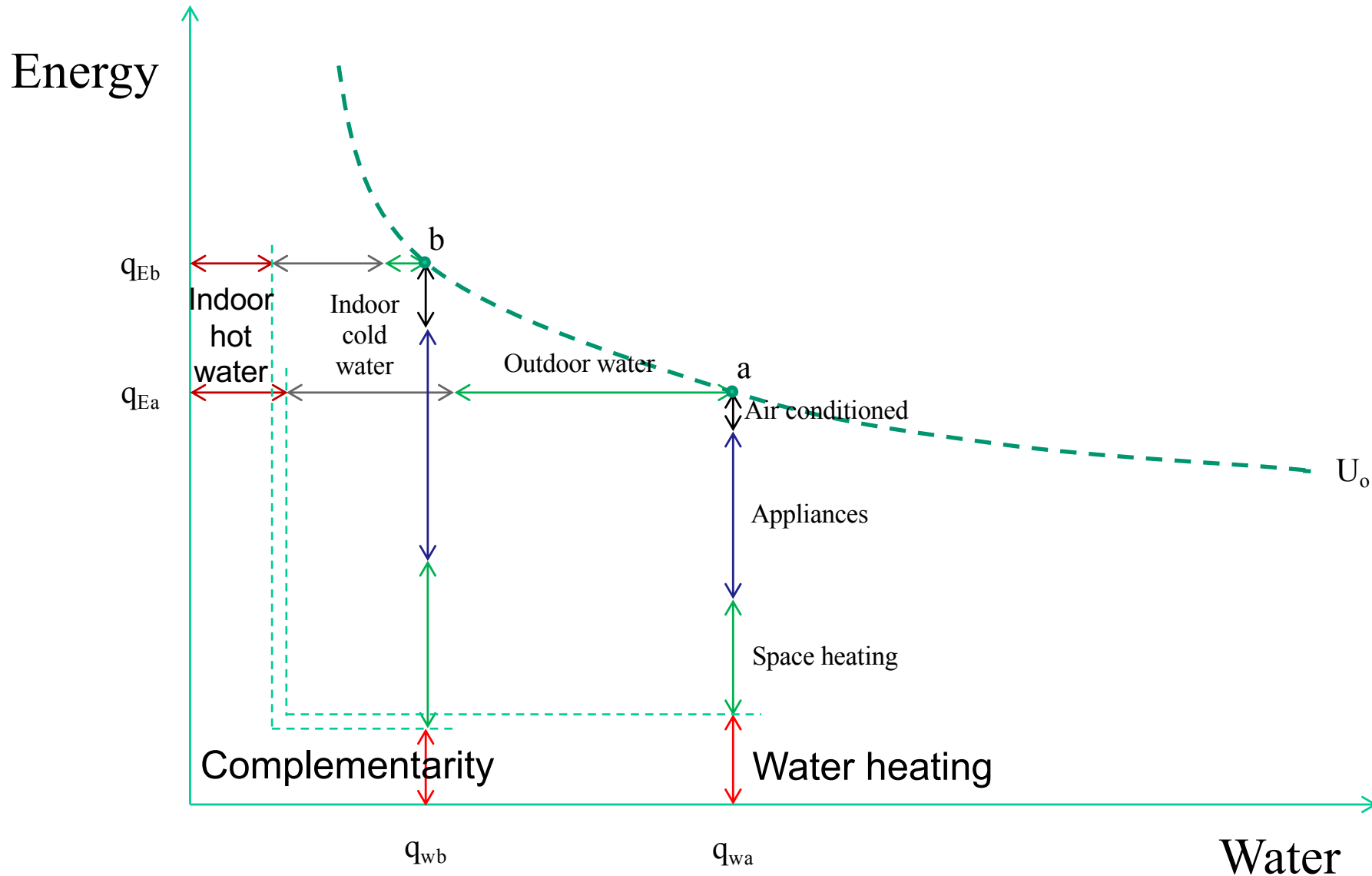
- Faucet + shower  $\approx$  80% water-related energy
- Air and inlet temperatures affect energy use
- “Willingness to adopt” conservation depends on:
  - Current consumption
  - Household stock
  - Water and energy prices
- Targeting
  - More than doubles cost-effectiveness of rebates

## 2) Least-cost water conservation mix for California households considering energy

- What is the least-cost water conservation mix for households, given water and energy prices?
- Does including energy affect willingness to adopt conservation actions?
- How significant are own- and cross-price elasticities?

\*Escriva-Bou, A., J. R. Lund, and M. Pulido-Velazquez (2015), Optimal residential water conservation strategies considering related energy in California, *Water Resources Research*.

# The economics behind the model: Complementarity of demand

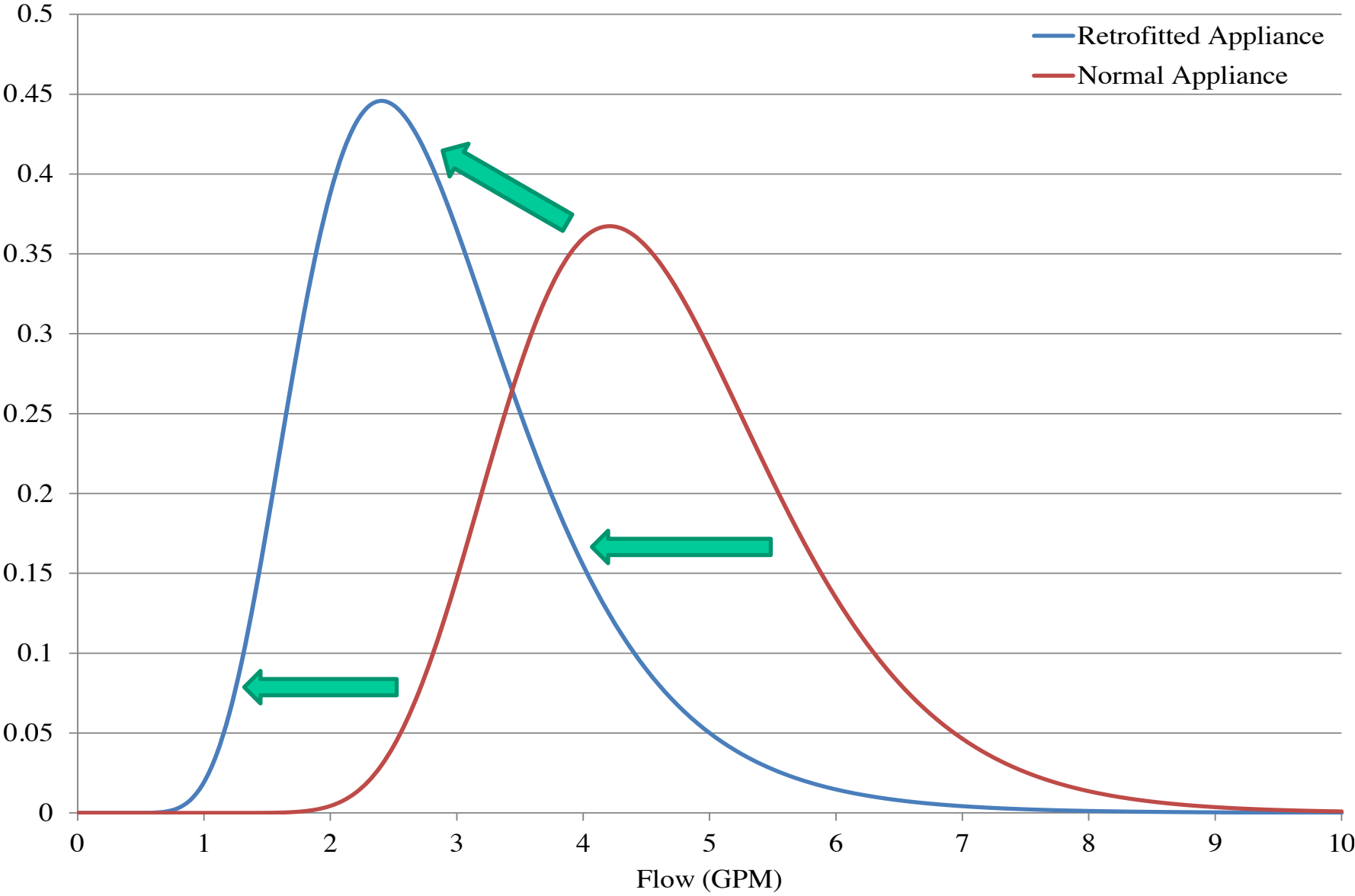


# Household optimization process

- Each household has conservation options
  - Long-term: Retrofits
  - Short-term: Behavioral
- Each action has
  - Cost
    - » Annualized costs for retrofits
    - » “Hassle costs” for behavioral changes
  - Effectiveness
    - » Water
    - » Energy
    - » Greenhouse gas emissions



# Conservation Actions: Savings and Technological Shifts



# Optimization Model

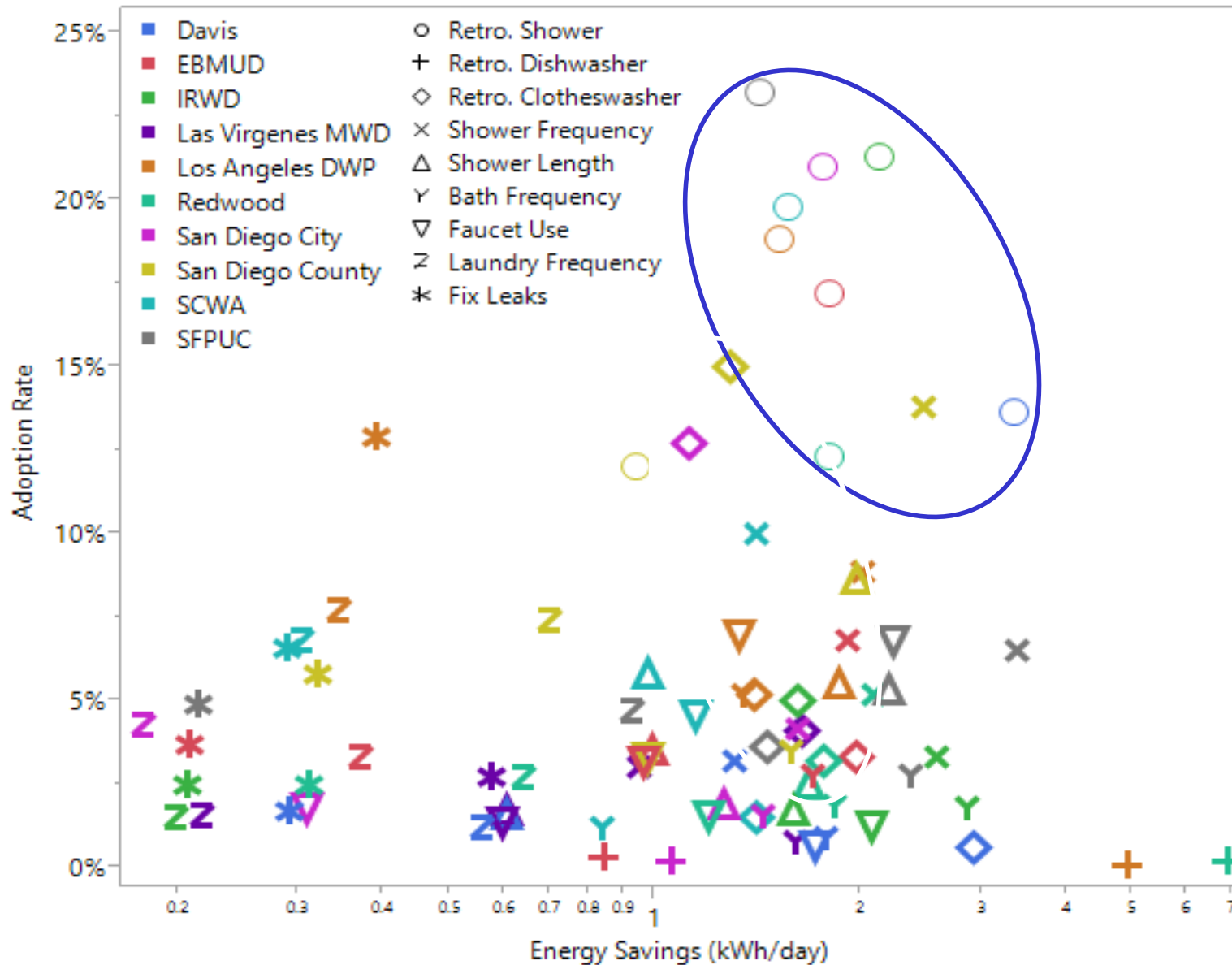
$$\begin{aligned}
 \text{Minimize } TOTAL\ COST &= \sum_{wlt} C_{wlt} \cdot X_{wlt} + \sum_{elt} C_{elt} \cdot X_{elt} + \\
 B \cdot \left[ \sum_{we} p_{we} \cdot \left( \sum_{ee} p_{ee} \cdot \left\{ D \cdot \left( \sum_{wst} C_{wst} \cdot X_{wst_{we,ee}} + \sum_{est} C_{est} \cdot X_{est_{we,ee}} \right) + B_{W_{we}} + B_{E_{ee}} \right\} \right) \right]
 \end{aligned}$$

Subject to:

- Decision variables are binary
- Savings are less than initial use (upper bound) and resource availability
- Mutually exclusive actions
- Interdependence among actions



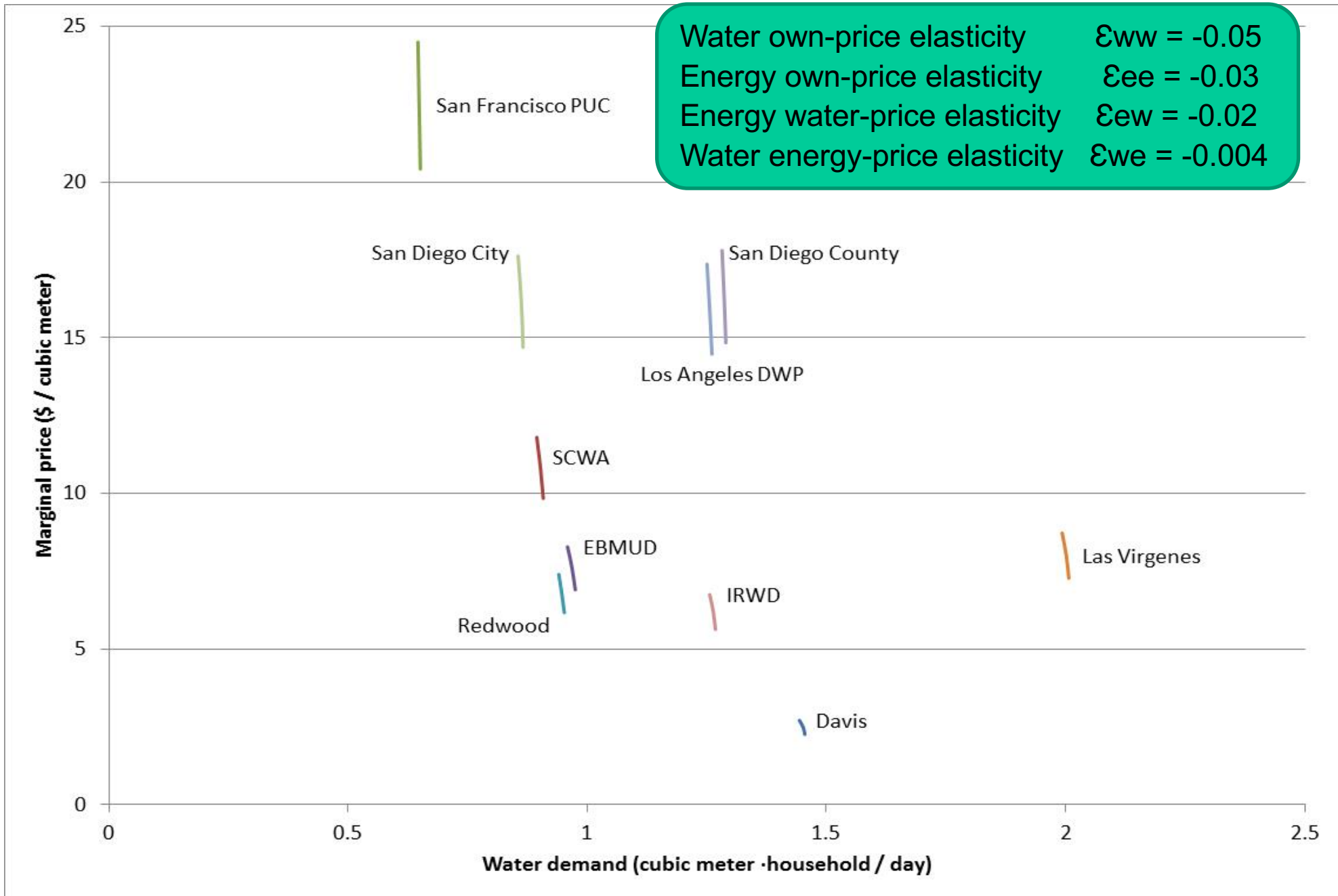
# Energy savings for water-related actions



# Increased conservation when energy is included

- Adoption rate:
  - Retrofit shower: +7.9%
  - Retrofit clothes washer: +1.7%
  - Reduce shower length: +3.2%
  - ...
- Increased savings:
  - Indoor water savings: +24%
  - Energy savings: +30%
  - GHG savings: +53%

# Demand functions and elasticities



# Policy implications from Mechanistic Modeling

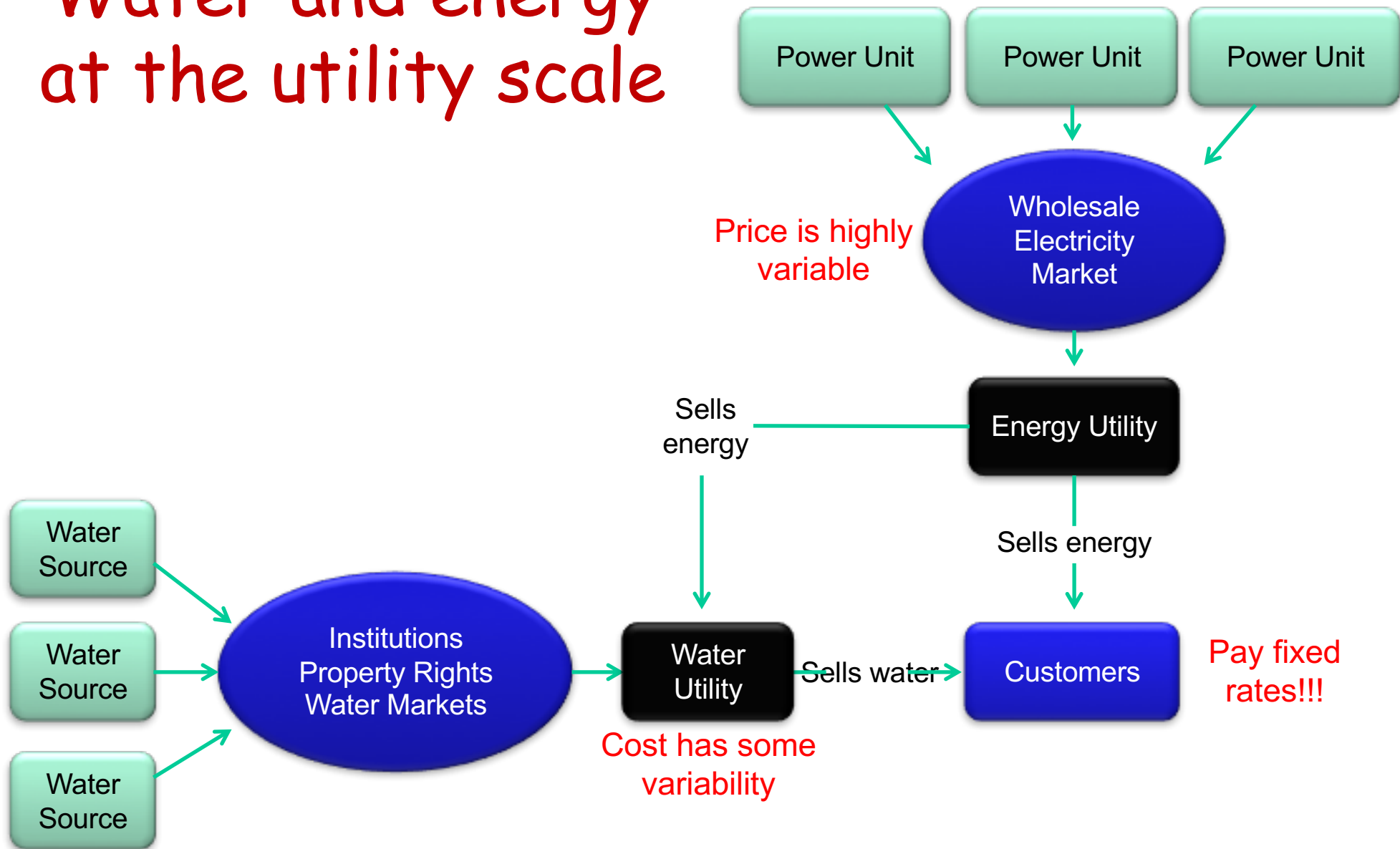
- Including water-related energy should increase water conservation (and energy and GHG savings).
- Outdoor and toilet save most water; shower, faucet and clothes washer better save energy.
- Behavioral actions: Much to do!

### 3) Coupling hourly end-use and utility-scale water-energy models

- How much energy and GHG emissions are embedded in urban water cycle?
- What are effects of water conservation on water and energy utilities?
- Are there synergies for water and energy utilities working together?



# Water and energy at the utility scale



# EBMUD Example



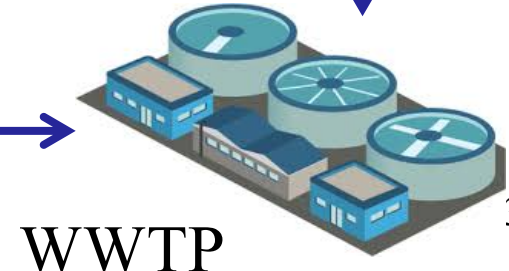
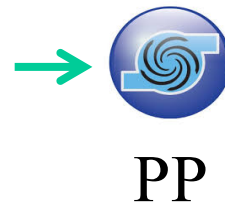
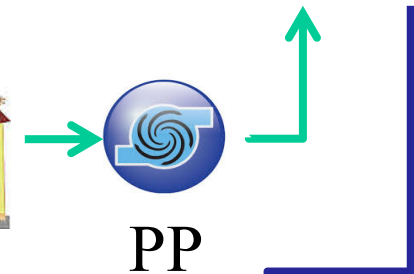
Pardee and  
Camanche  
Reservoirs

Total Supply:  
17604 MG/year  
(out of 64868 MG/year)

Leland  
Pop.  $\approx$  130,000  
6,391 MG/year  
Elevation:  
150 feet – 45 m

Danville  
Pop.  $\approx$  75,000  
3661 MG/year  
Elevation:  
350 feet – 107 m

San Ramon  
Pop.  $\approx$  150,000  
7553 MG/year  
Elevation:  
550 feet – 168 m



WTP

PP

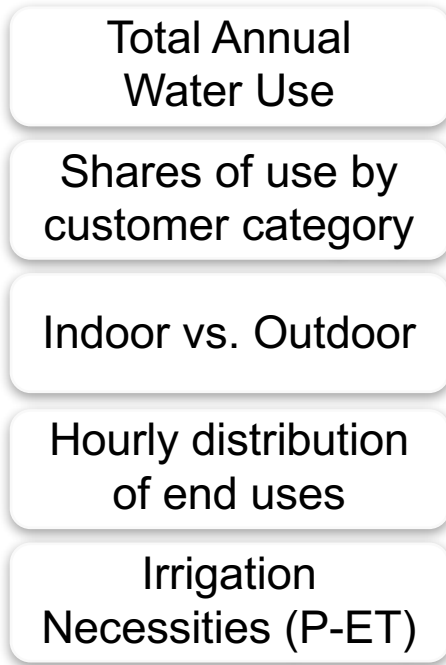
PP

PP

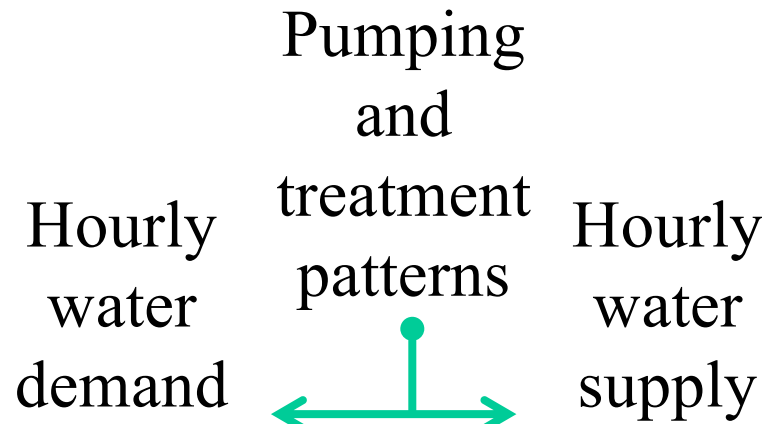
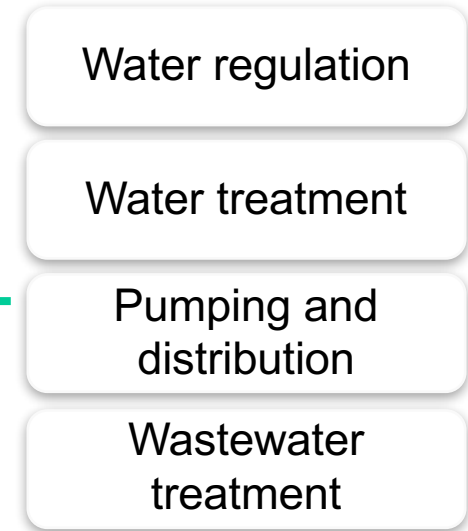
WWTP

# Assembling the model

Water users



Water utility



End-uses energy intensity

Water-related energy

Water-related energy

Regressions and pumping patterns

Natural Gas Utility

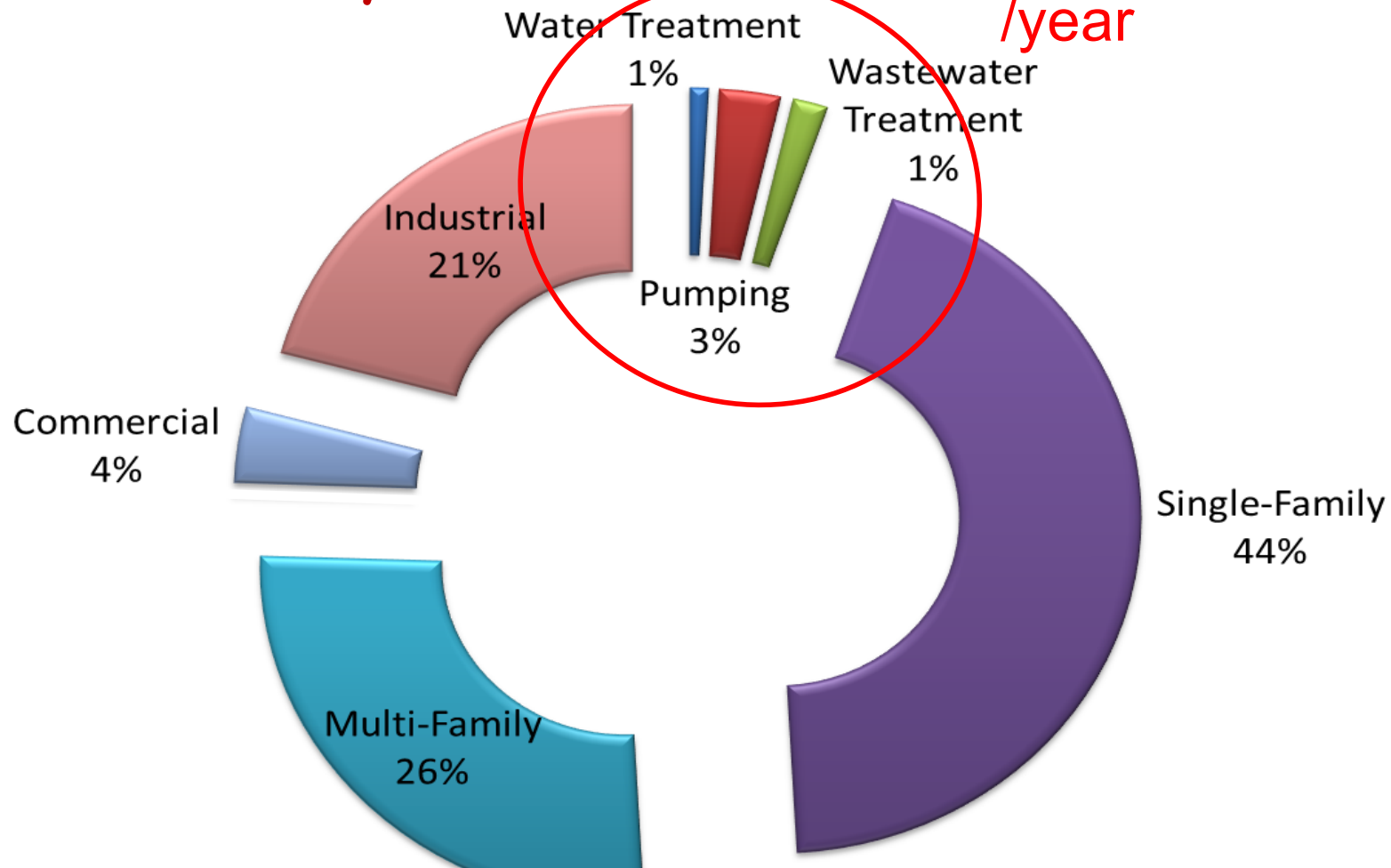
Electric Utility

Electricity price (TOU Tariff)

GHG emissions

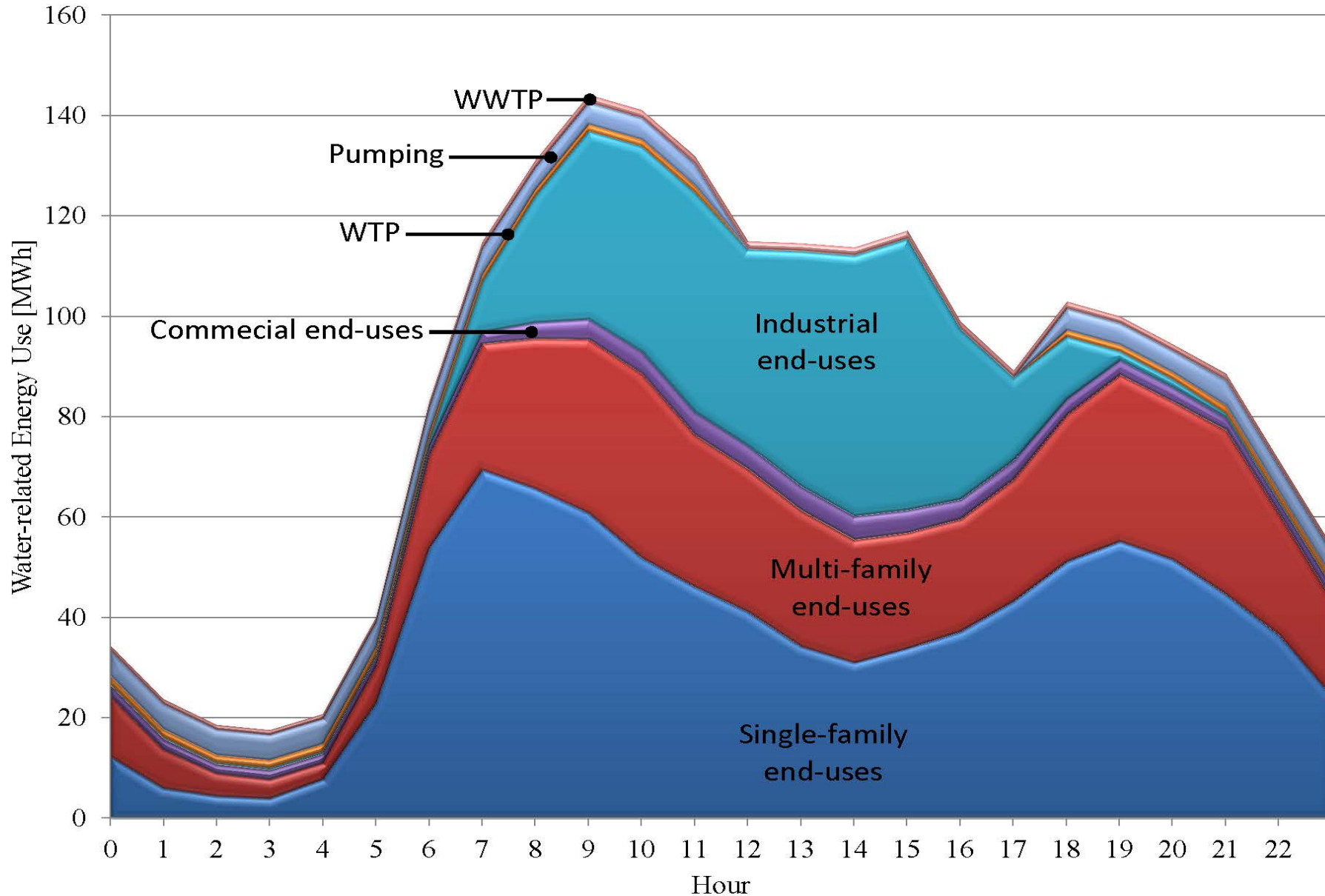
Electricity Market (Hourly Prices)

# Water-related energy consumption in the urban water cycle - EBMUD > \$12 million /year



Urban water cycle  
Total emissions per capita: 406 kg CO<sub>2</sub>/year  
4.5% total emissions per capita in CA

# Shifting Water Use Peaks to Off-Peak Energy Hours Has Economic Benefits



# Results

- Optimal water conservation
  - Water use: 6% reduction
  - Energy use: 5% reduction
  - GHG emissions: 5% reduction
  - Energy cost for water utility: 4.5% reduction



# Results

- Demand-response (peak shaving): Outdoor, clothes washer and dishwasher use are moved to off-peak hours.
  - Water use: Equal
  - Energy use: Equal
  - GHG emissions: Needs more discussion
  - Energy cost for water utility: 3% reduction
  - Energy cost for energy utility: 4% reduction

# Policy implications

- Saving water reduces some GHG emissions.
- Synergies exist for water and energy utilities working together.
- Temporal water demand management can be very effective to reduce energy peaks.

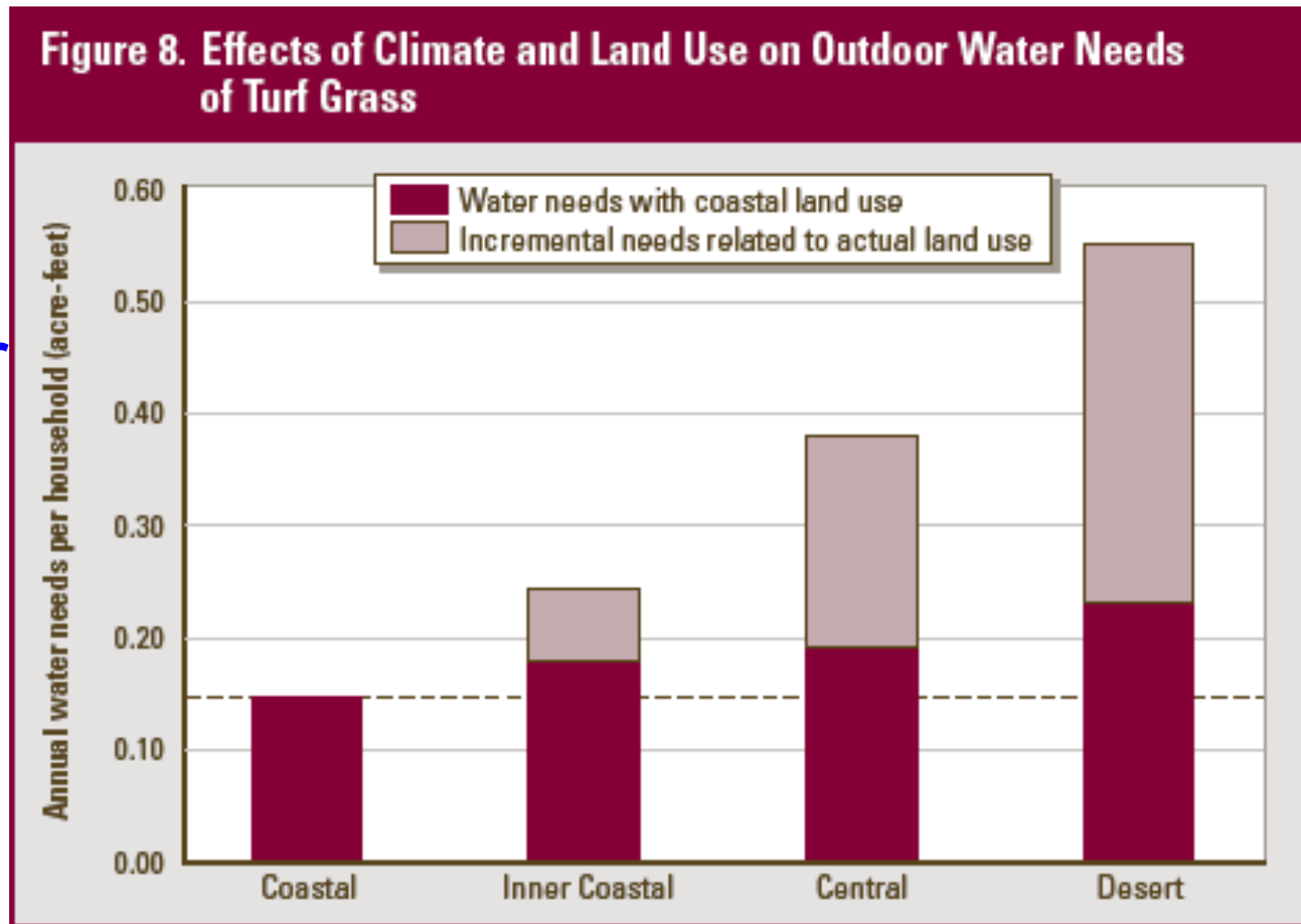


# Hard Parts Left to Do

- Outdoor water use and WTP estimation
- Monte Carlo modeling for outdoor use
- Energy - Hot water heater efficiency
- Getting data organized – starting to happen
- Testing models systematically and reconciling with empirical modeling

# Effects of Climate and Land Use

- Larger lawns & warmer, drier climate increase landscape water use
- Landscape type will also affect!



Hanak and Davis, 2009

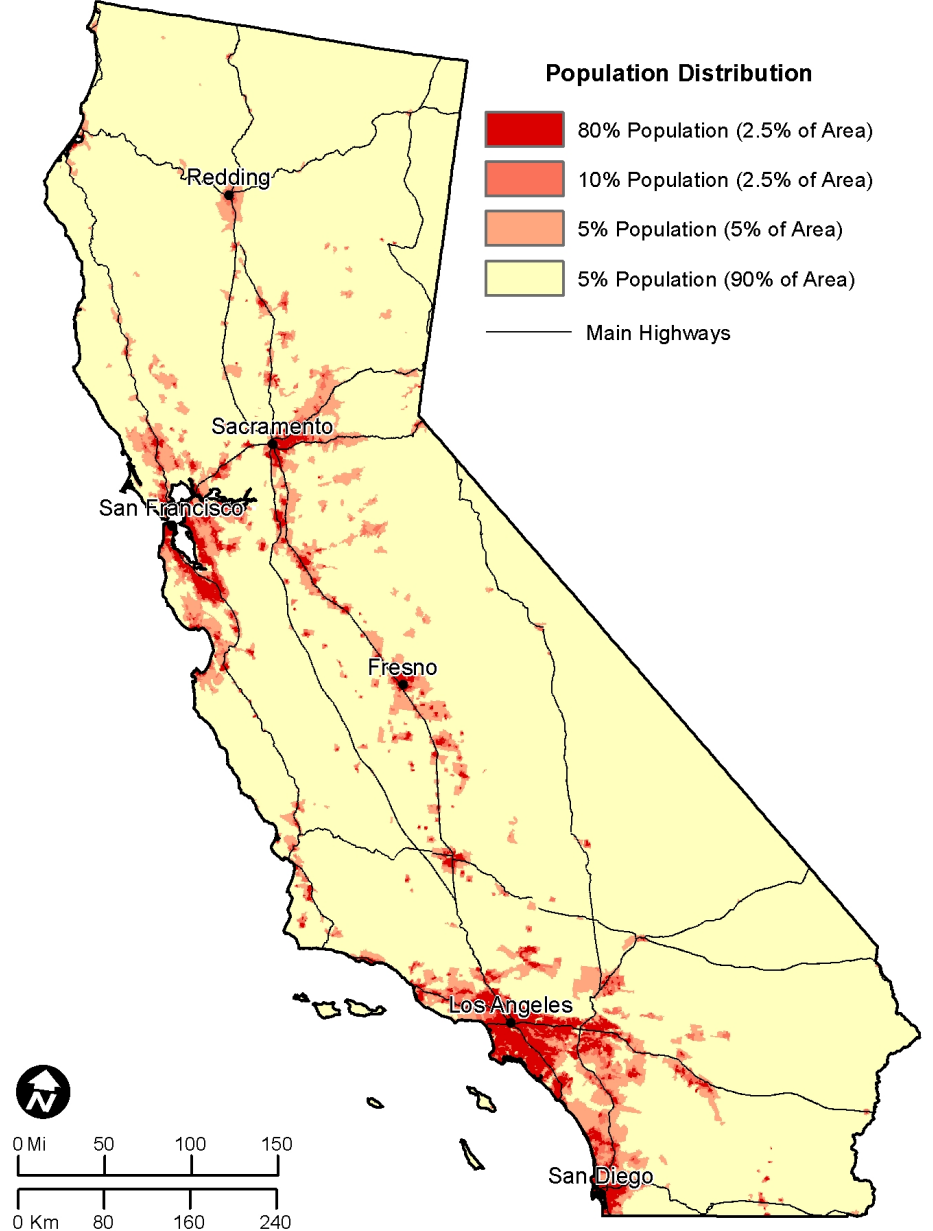
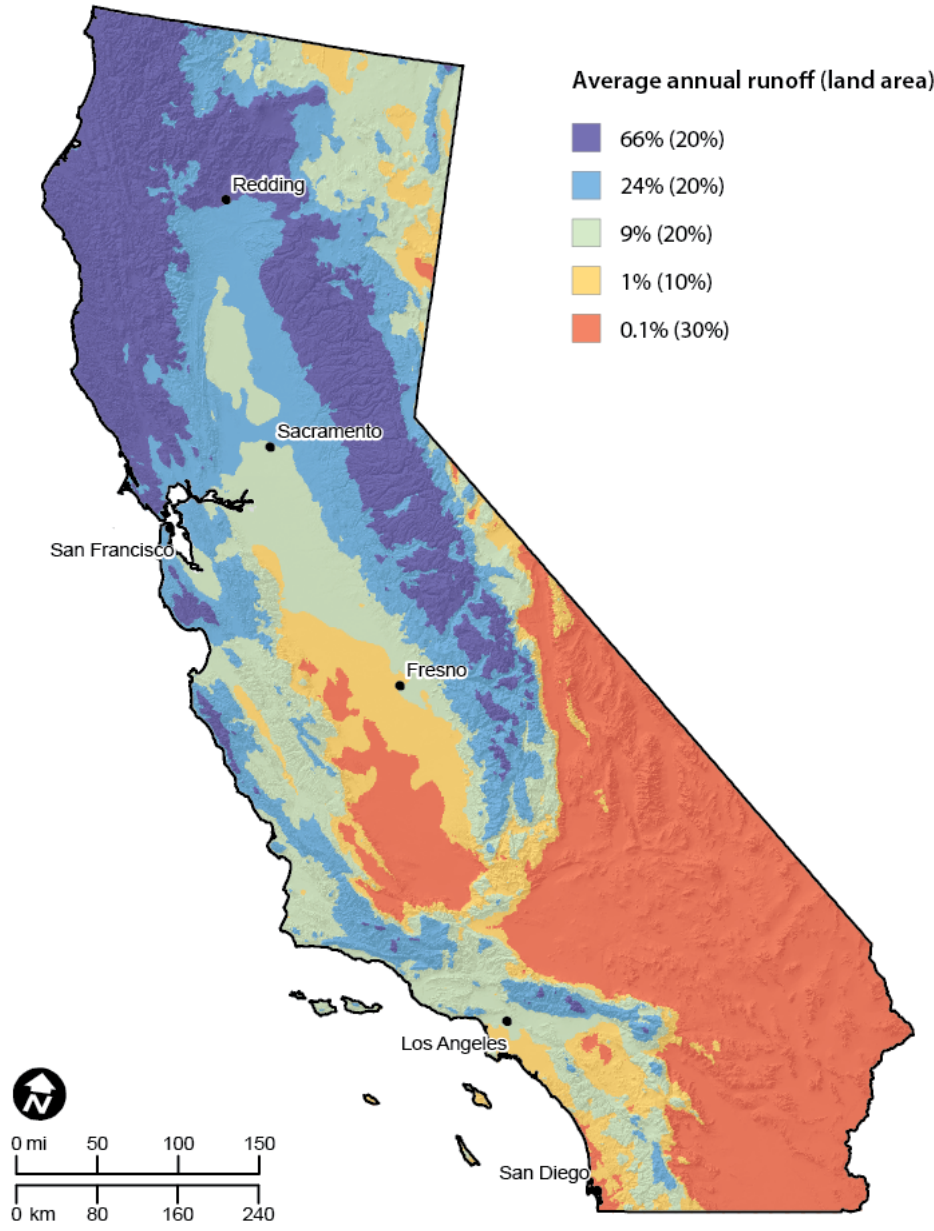
# Opportunities

- Smart meter data will drown us in data
- Commercialize mechanistic modeling
- Integrate with other supply and demand management activities at social, utility, and household scales
- Include more risk and financial analysis

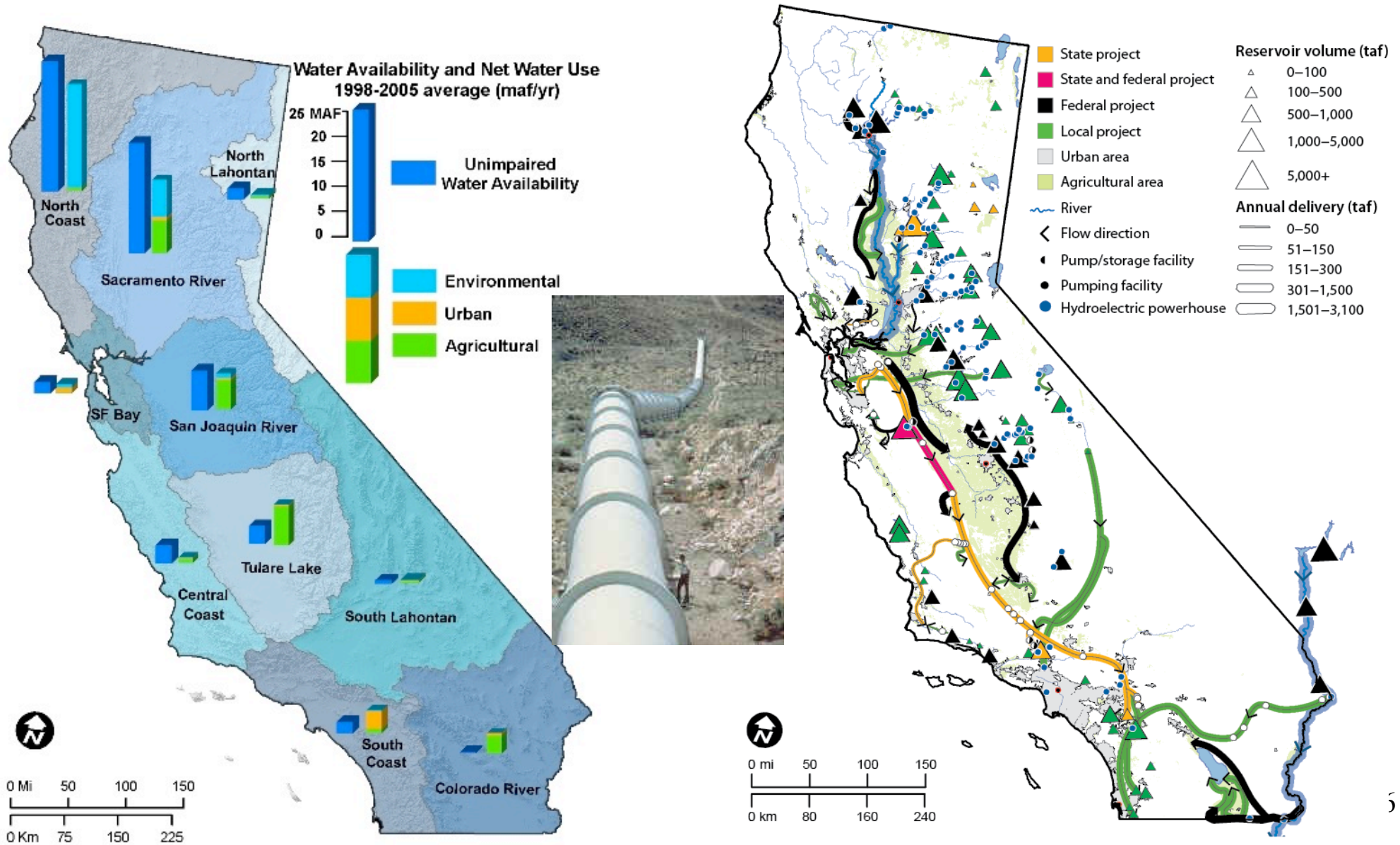
# Conclusions

- Nexus modeling is harder, but more interesting and useful than just yacking about X-Y-Z nexus
- Mechanistic modeling, better organizes problem with more flexibility and insights than empirical modeling alone
- Empirical and Mechanistic modeling should work better together
- Future looks bright for research and applications

# Water and People in California



# California depends on an engineered statewide network





# Effects of Climate and Land Use

- Average Water Requirements of Turf Grass for Small Single-Family Lots

Region	Yard Size (sf)	Weighted Average ET <sub>0</sub> (inches/year)	Annual Water Requirements (af)	Increase over Region with Lowest Need
San Francisco Bay Area	6,308	45.9	0.19	—
South Coast	7,623	49.8	0.25	31%
San Joaquin Basin	7,060	54.4	0.26	33%
Tulare Basin	7,711	56.2	0.29	50%
Sac. Metro region	8,129	56.8	0.31	59%
Inland Empire	8,858	56.2	0.33	72%

# AU and CA urban water use

- Urban CA could reduce use by 30-50+% with AU use rates.

Location	Residential Use, gpcd
Portland, OR	58
Albuquerque, NM	74
Tucson, AZ	97
Denver, CO	104
<b>California</b>	<b>104</b>
San Francisco	46
Oakland/East Bay	73-83
San Diego	73-92
San Jose	81-85
Los Angeles	91-99
<b>Sacramento</b>	<b>113-120</b>
<b>Australia</b>	<b>54</b>
Melbourne	40
Brisbane	45
Canberra	50
Sydney	55
Perth	75

Cahill and Lund, 2009



# Biggest difference in AU and CA use is usually outdoors

End Use	California				Australia					
	East Bay Area		California		Perth		Melbourne		Gold Coast	
	Use, gpcd	% of total	Use, gpcd	% of total	Use, gpcd	% of total	Use, gpcd	% of total	Use, gpcd	% of total
Toilet	20	21%	13	10%	9	9%	8	13%	5	13%
Shower/Bath	15	16%	13	10%	14	14%	14	24%	15	37%
Washing Machine	14	15%	10	8%	11	11%	11	19%	8	19%
Faucets	10	11%	11	9%	7	7%	7	12%	7	17%
Leaks	5	5%	10	8%	2	2%	4	6%	1/2	1%
Other	1	1%	2	1%	1	1%	1	1%	1/2	1%
<b>Outdoor</b>	<b>30</b>	<b>32%</b>	<b>67</b>	<b>53%</b>	<b>55</b>	<b>56%</b>	<b>15</b>	<b>25%</b>	<b>5</b>	<b>12%</b>
<b>Total</b>	<b>95</b>	<b>100%</b>	<b>126</b>	<b>100%</b>	<b>99</b>	<b>100%</b>	<b>60</b>	<b>100%</b>	<b>42</b>	<b>100%</b>

Cahill and Lund, 2006

Melbourne 25"

Queensland 48"

Perth 29"