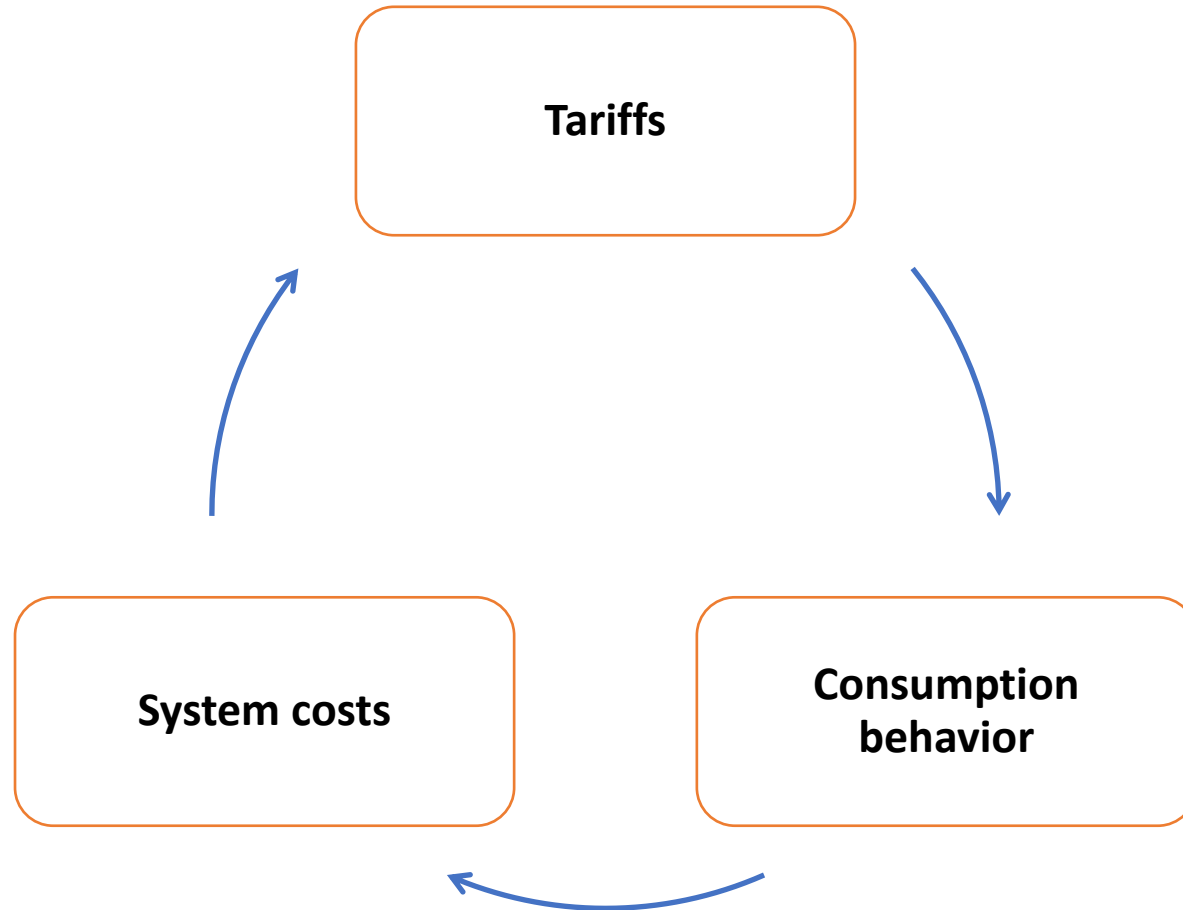


Economic and Social Effects of Residential Electricity Tariff Design

Scott Burger, Christopher Knittel, Ignacio Pérez-Arriaga, Ian Schneider, Frederik vom Scheidt

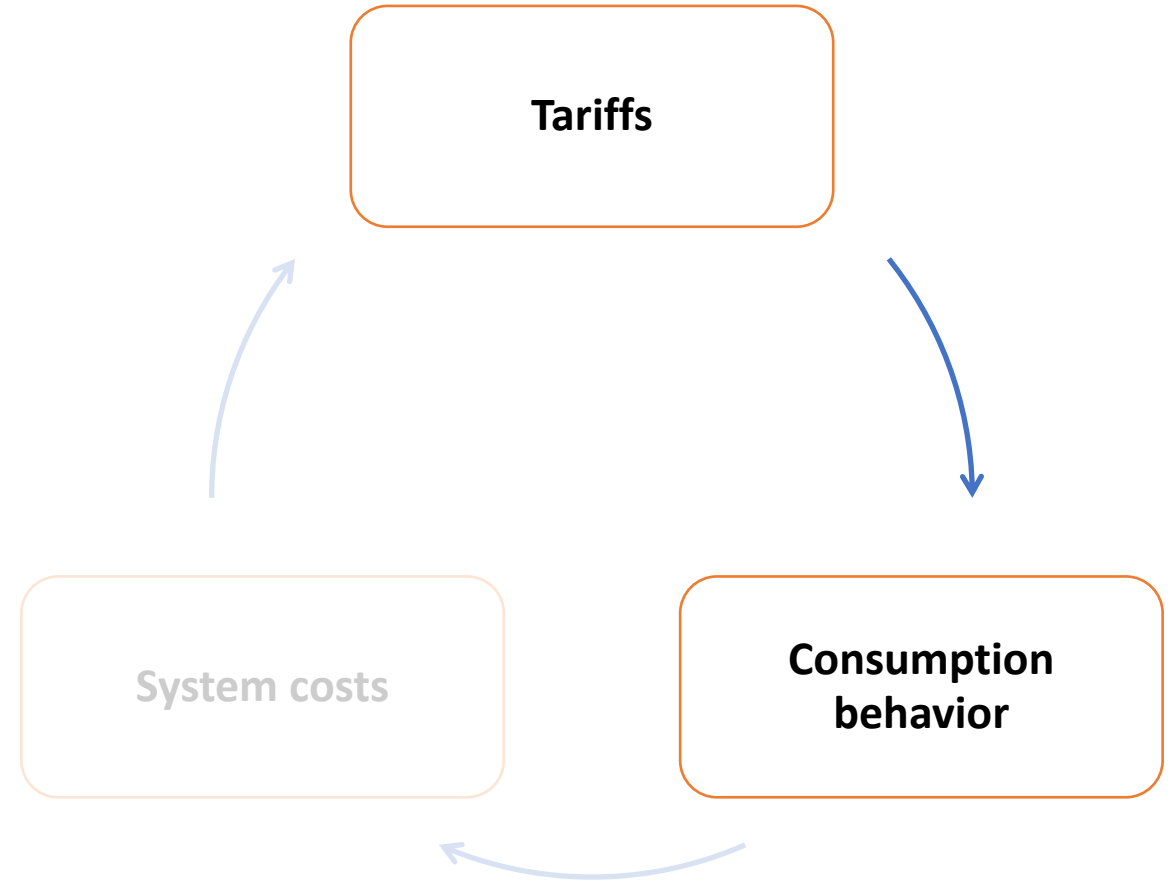


Electricity tariffs, customer behavior and system-wide costs are strongly connected



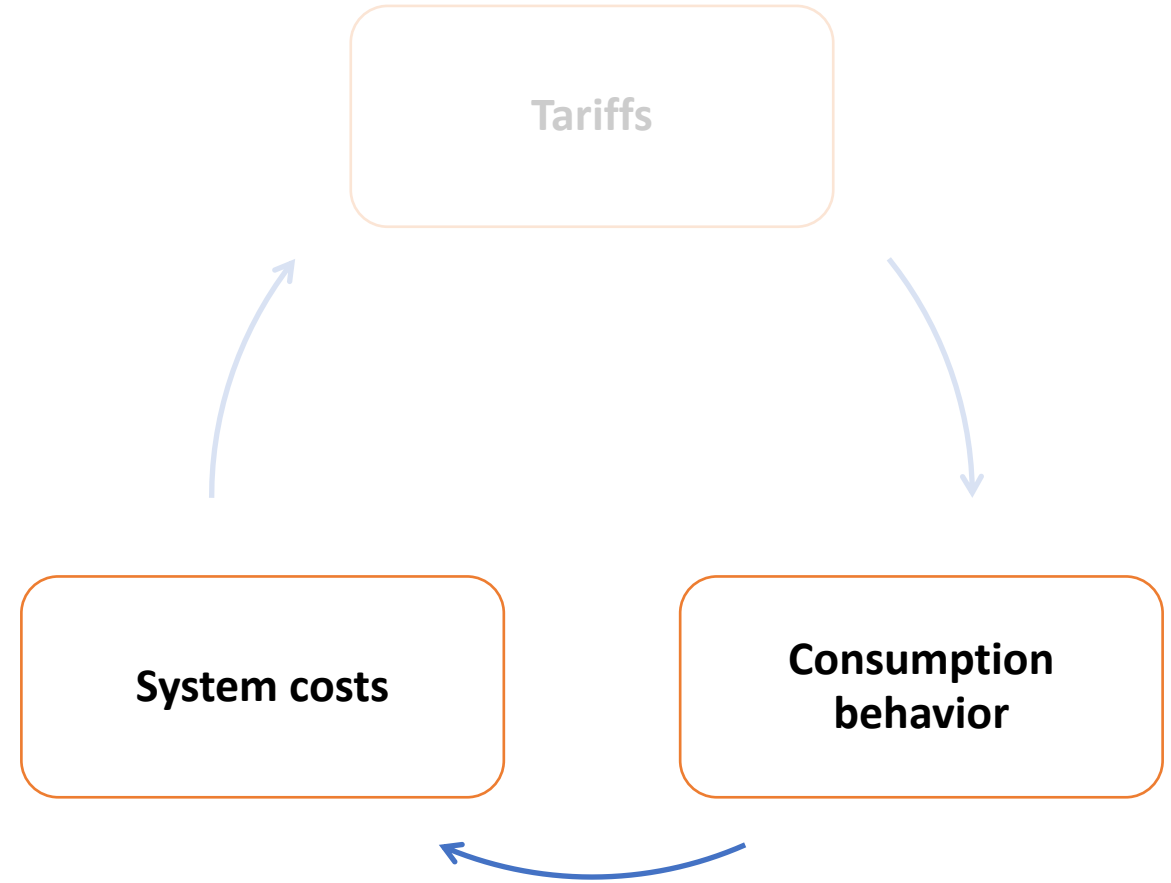
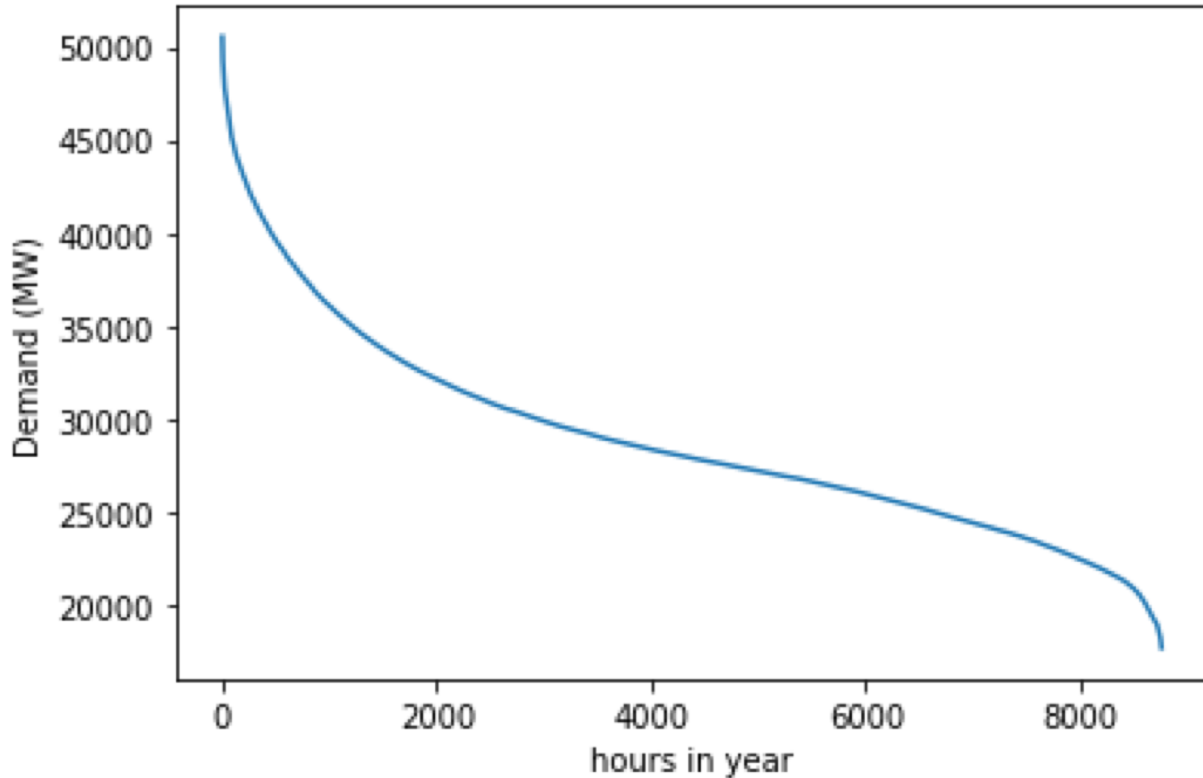
Prices influence how we consume electricity

- Meta analysis of time-varying tariffs [Faruqui et al. 2017]
 - 337 treatments
 - 63 tariff pilots
 - nine countries
- Over 94% of treatments finding non-zero customer response
- *“Price-based demand response is real and predictable”*



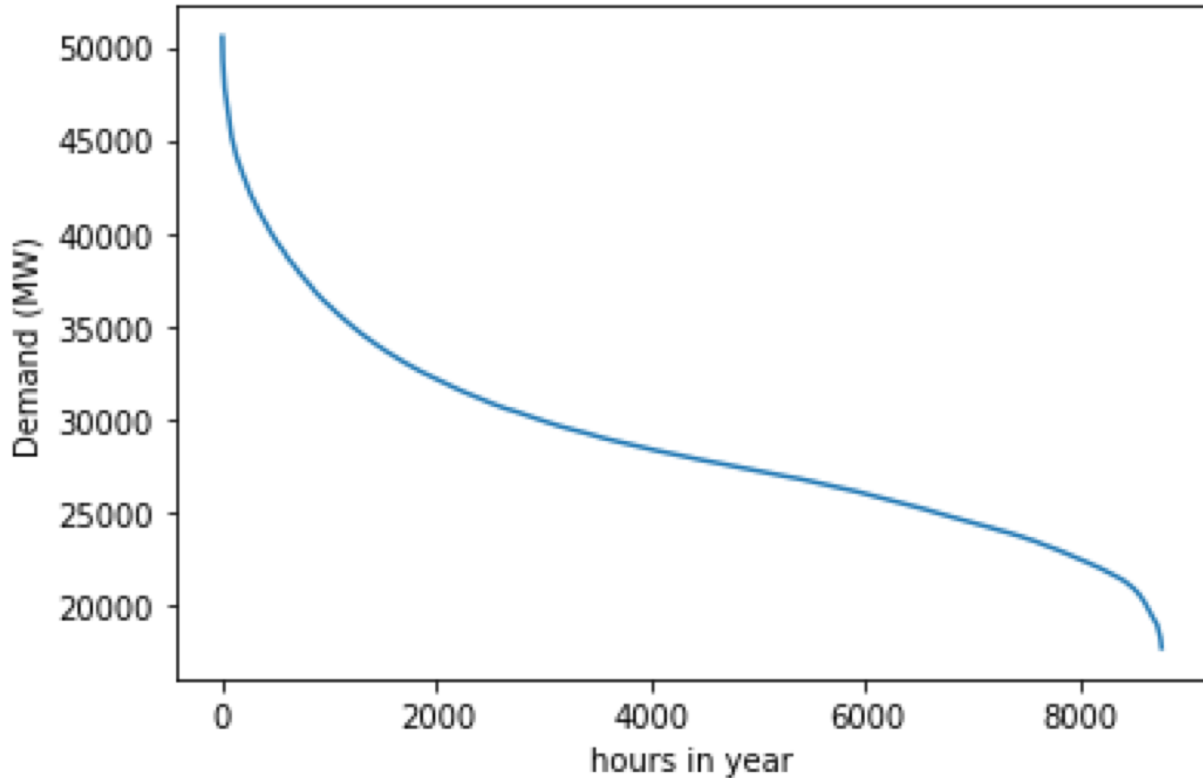
Consumption behavior determines system costs

Load Duration Curve in SPP

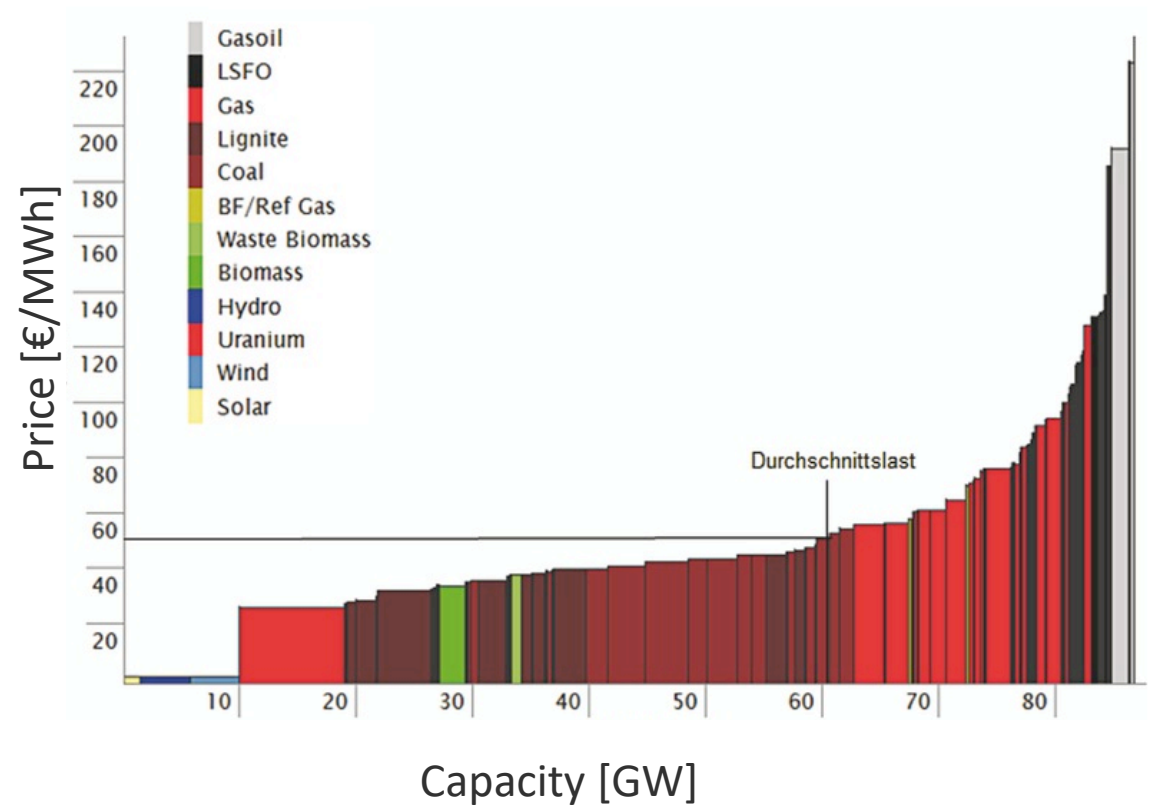


Consumption behavior determines system costs

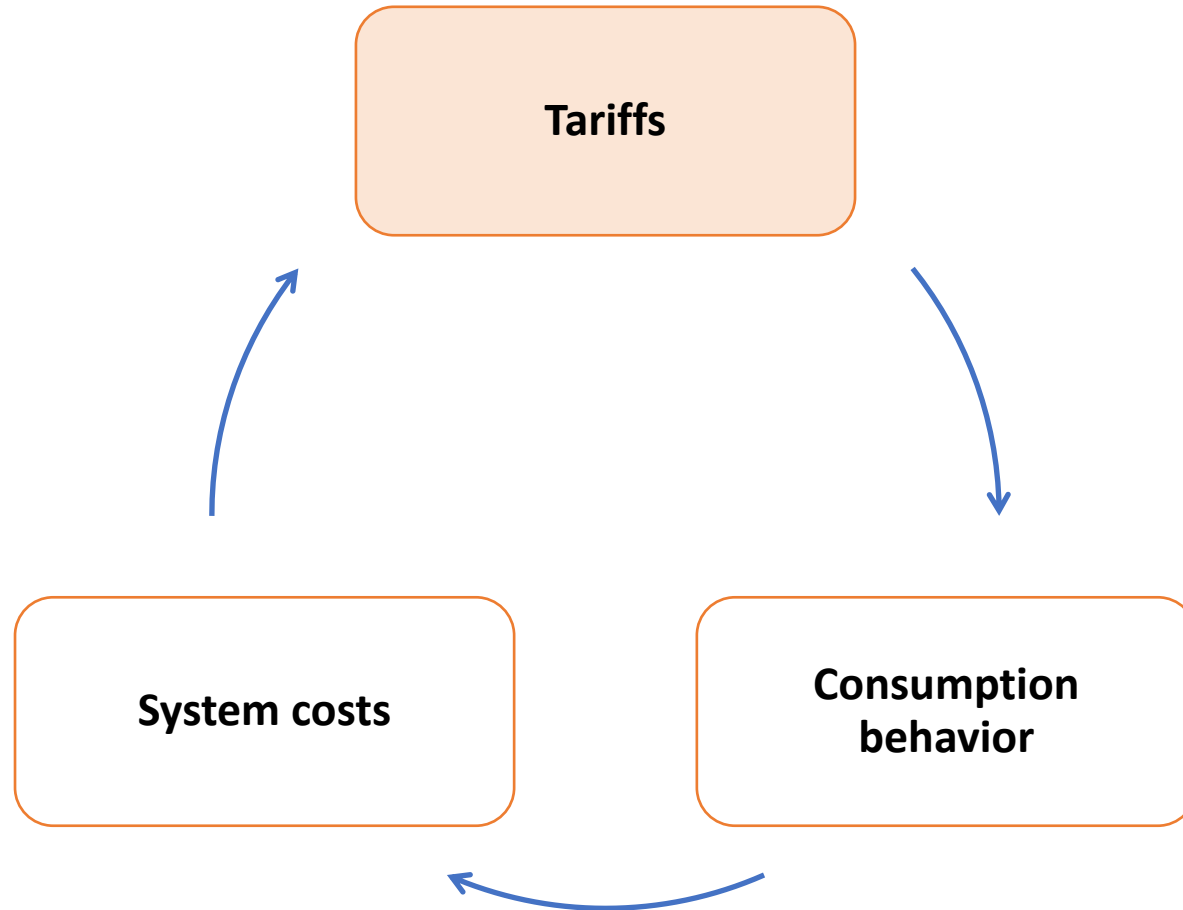
Load Duration Curve in SPP



Stylized marginal costs of generation

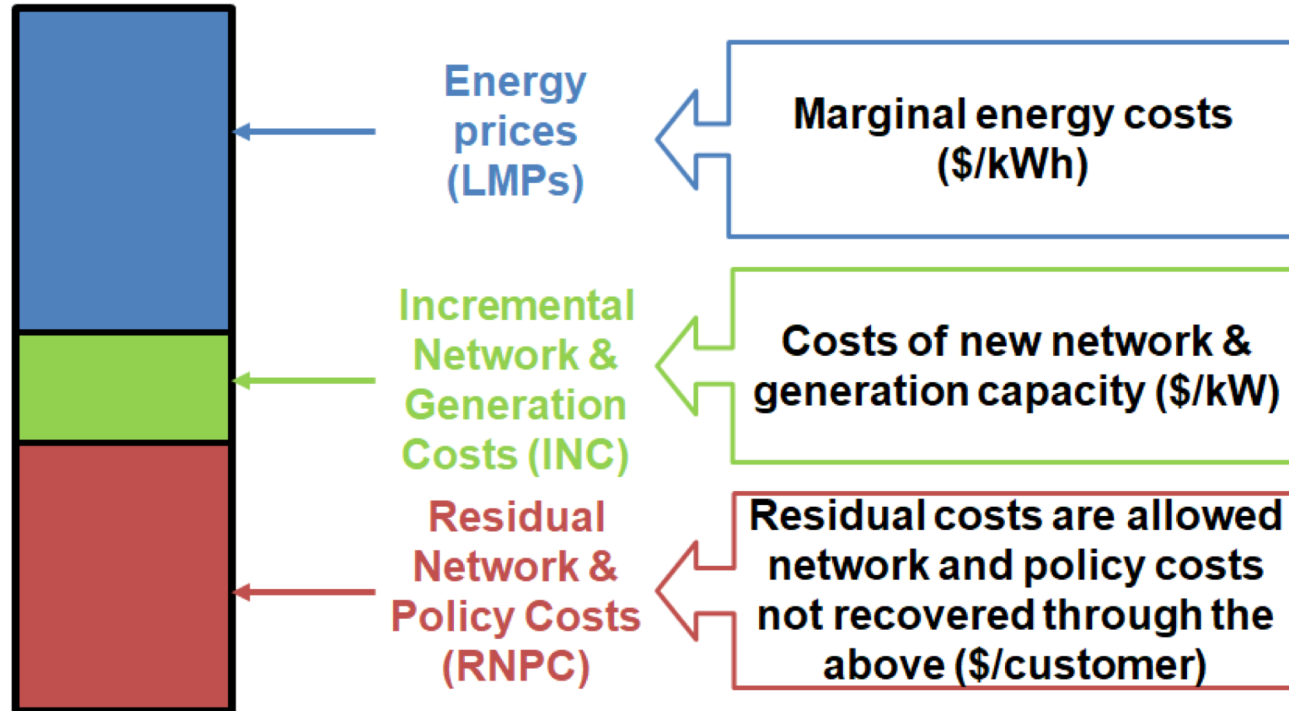


One key objective of tariffs design is to minimize overall system costs



But current tariff designs have inefficiencies that increase system costs

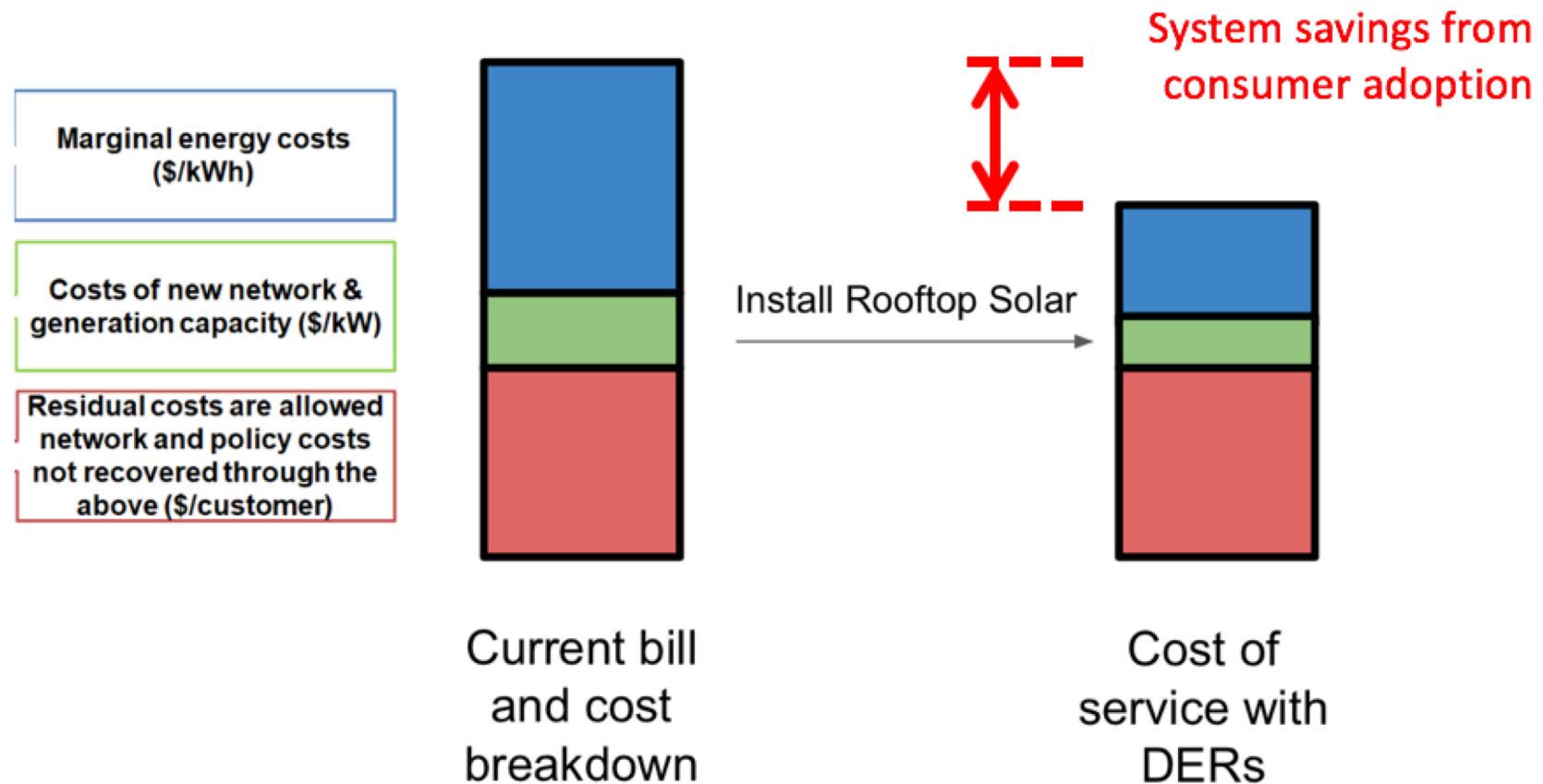
Efficient customer bill



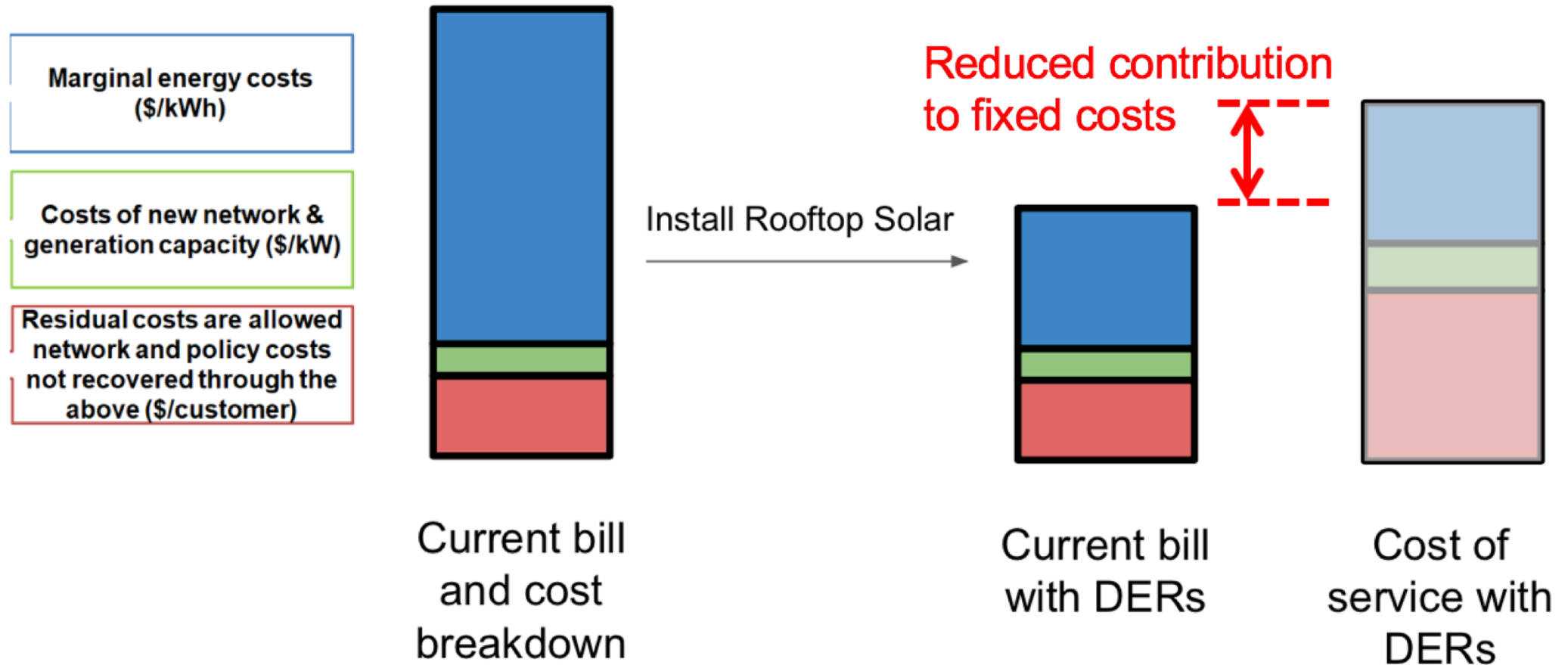
Three obvious inefficiencies with current rate design:

- Fixed costs recovered volumetrically
- Not time-based
- Not location-based

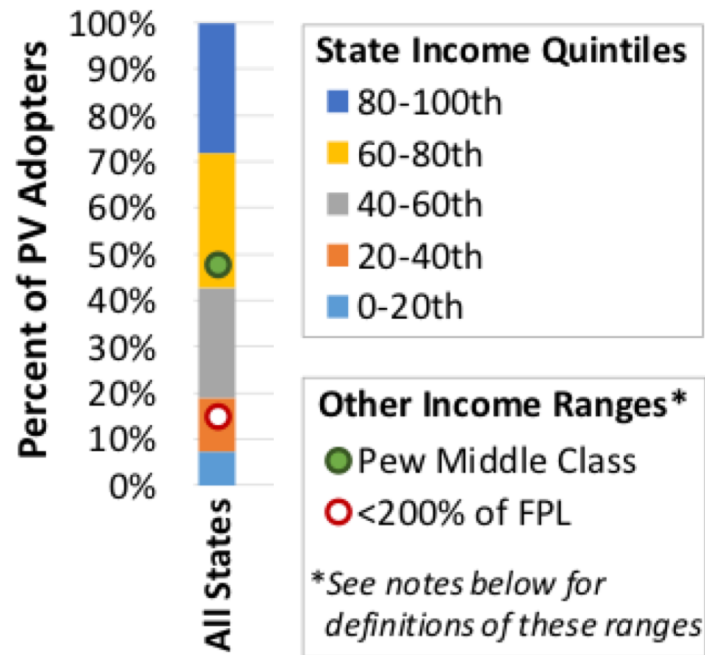
Dynamic inefficiencies are exacerbated by the growth of DERs



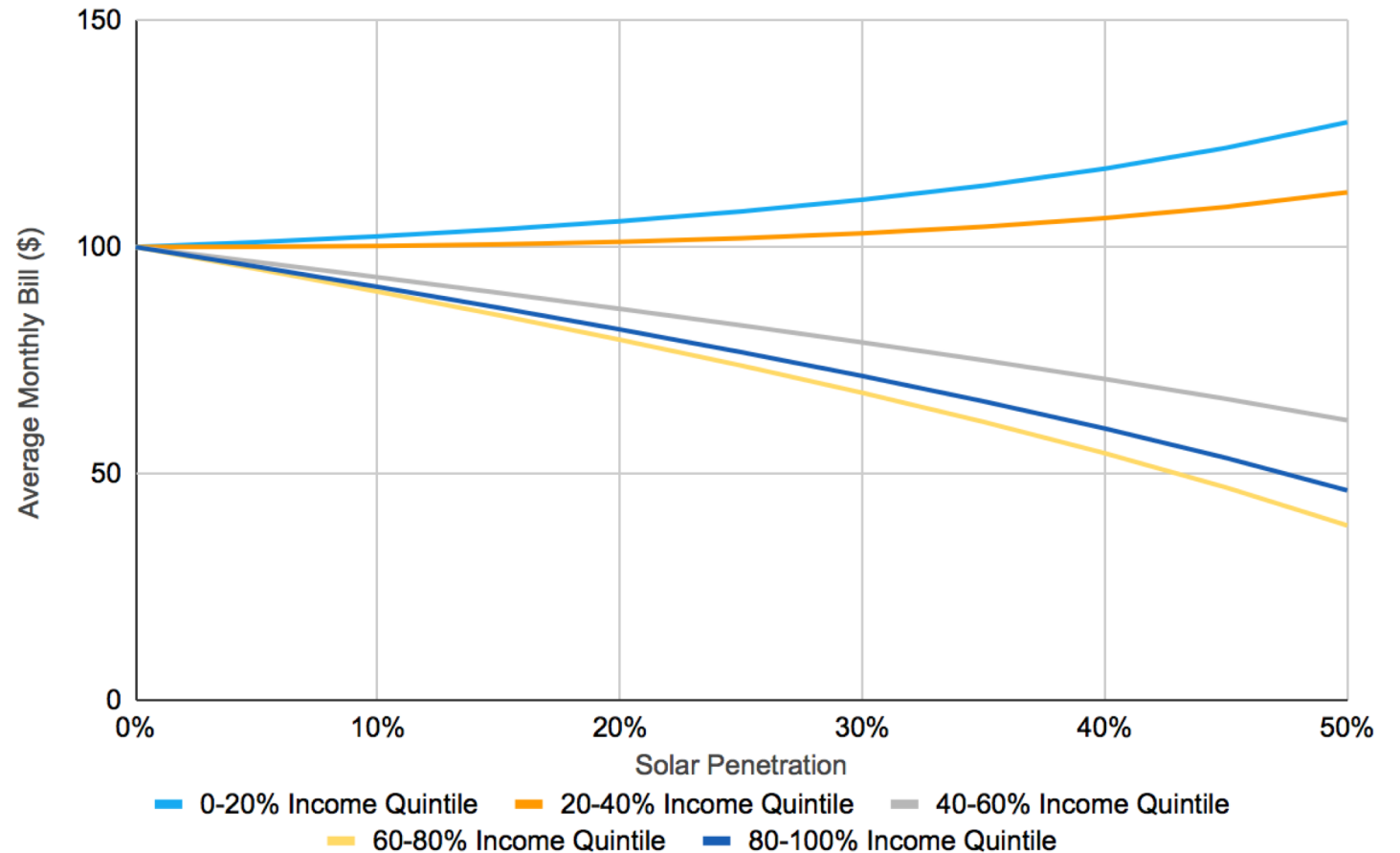
With inefficient tariffs, DER growth can raise or shift system costs



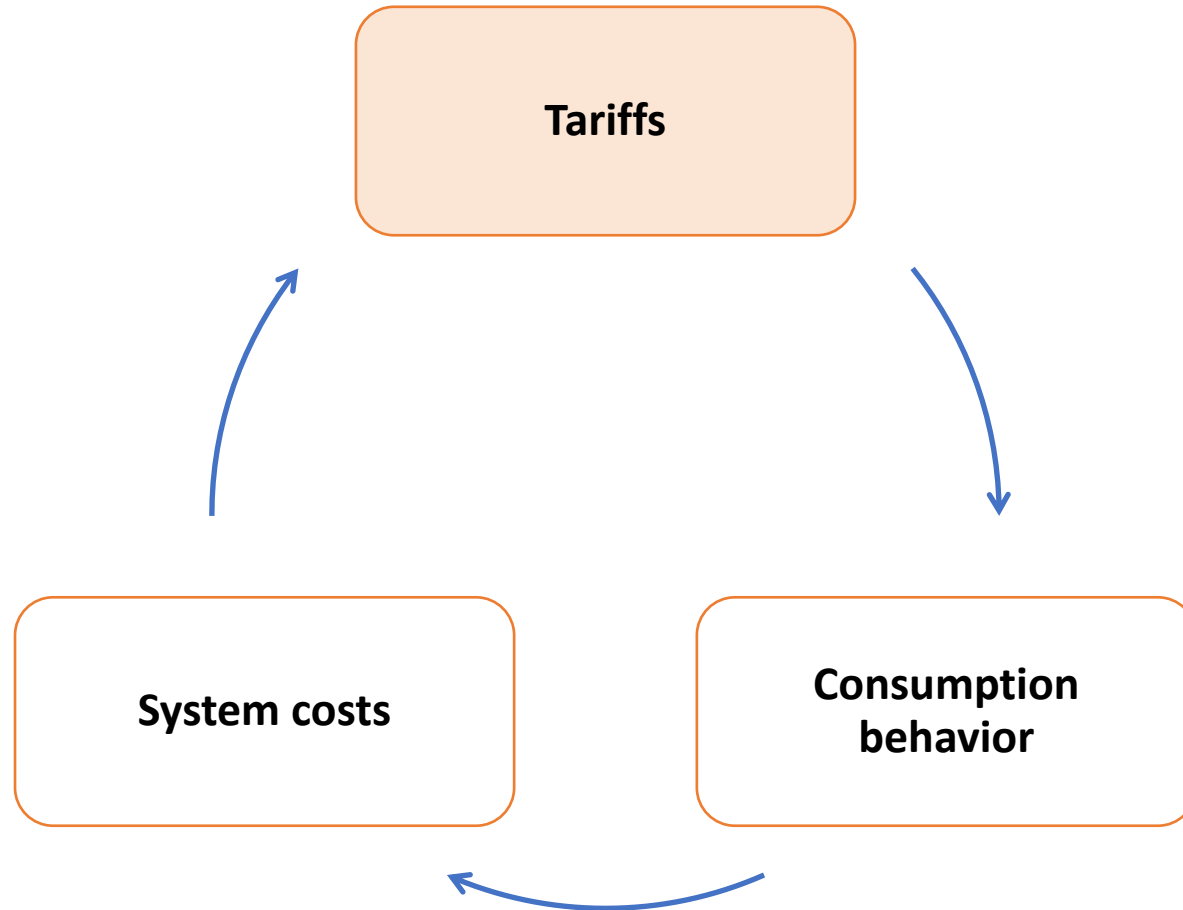
Inefficient tariffs have distributional impacts



Distributional Effects of Solar Adoption with Volumetric Tariffs



Can some tariff designs help improve welfare?



Can some tariff designs help improve welfare?

- Economic theory says yes. Many proposed improvements in existing literature.
- We test a few of these using hourly customer data.
- Then, we examine impacts on low-income customers and propose simple measures to mitigate impacts on low-income customers.

To evaluate alternative tariffs we use metering data from Chicago, USA



100.170 anonymized households



Consumption January-December 2016



30-minute smart meter readings



Housing type

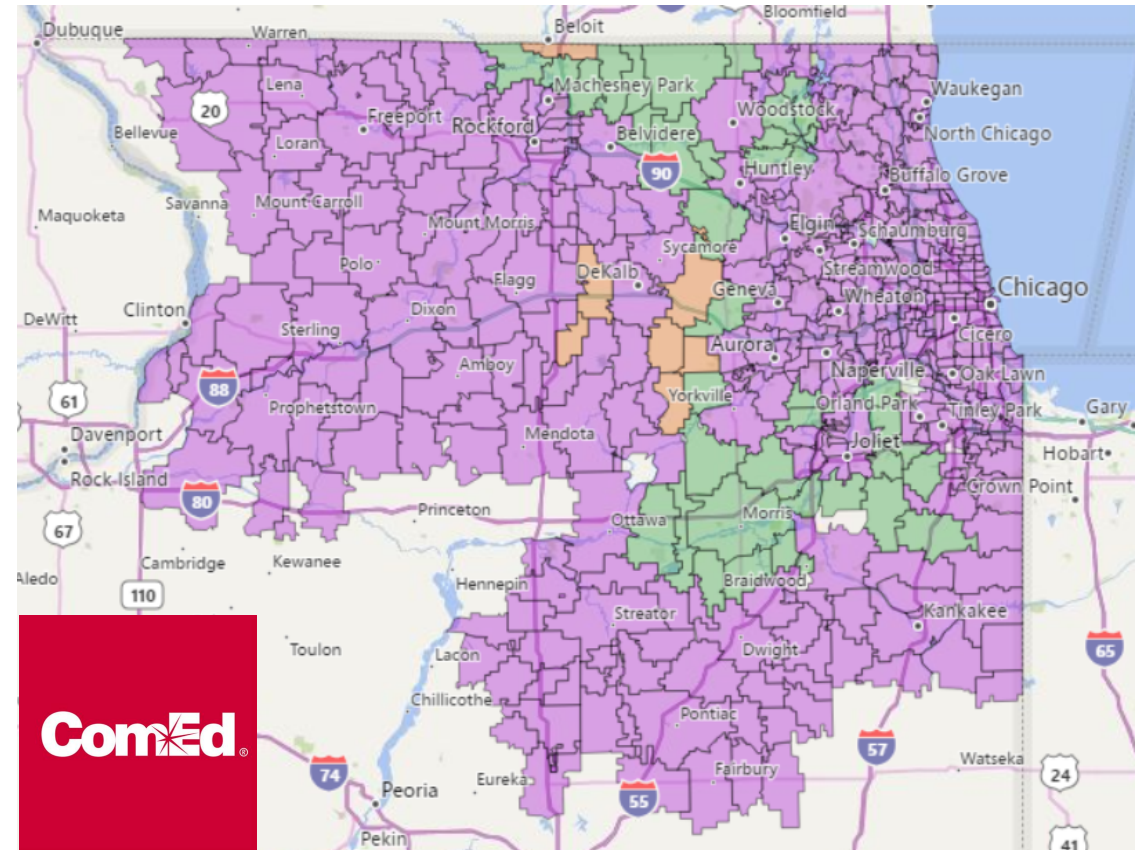


Heating type

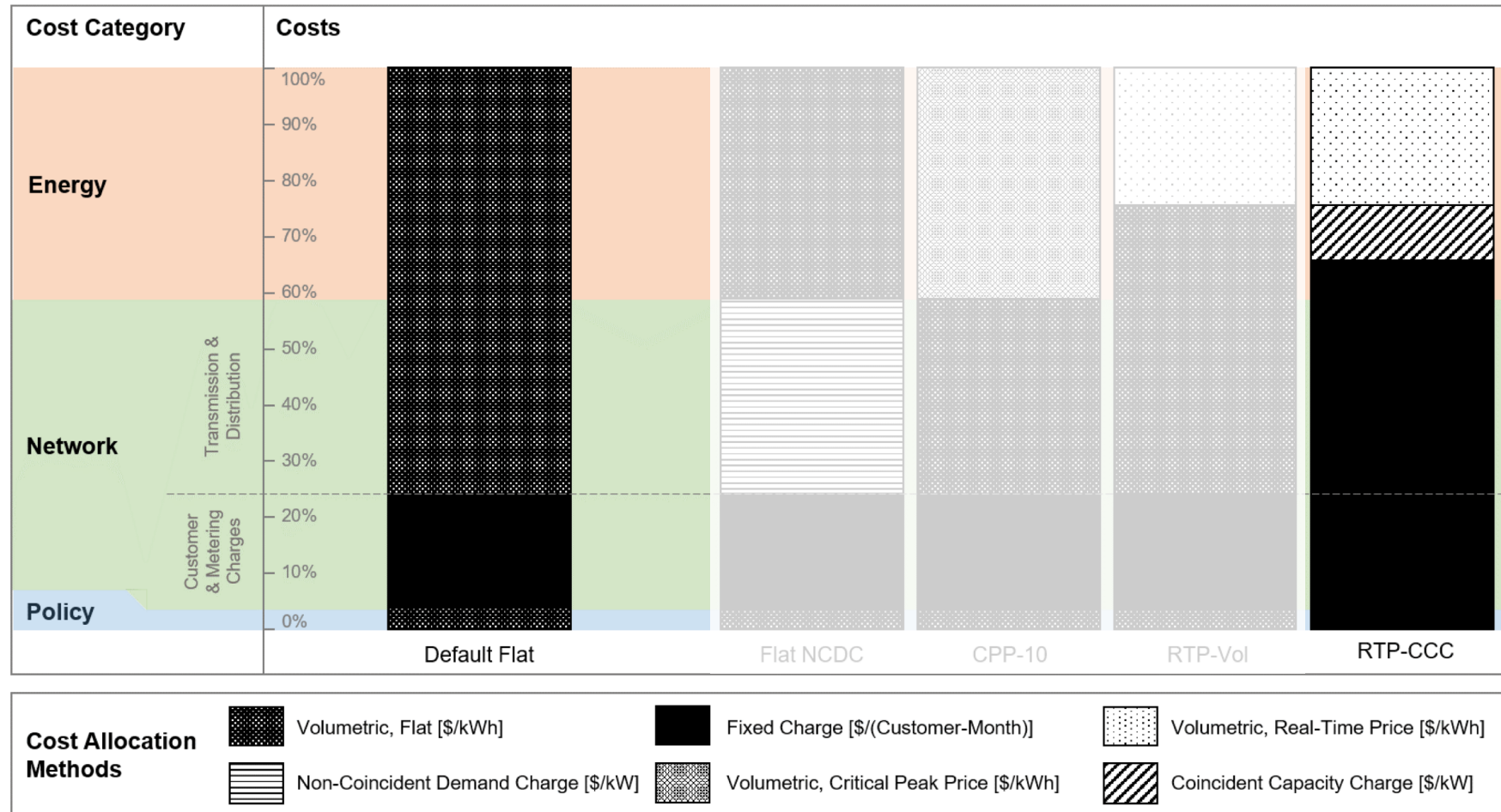


Geographic data: *9-digit zip*

Datenquelle: Commonwealth Edison, Citizens Utility Board Illinois



We create and evaluate five innovative tariffs designs



We compute tariff effects on customer expenditures and welfare for three scenarios

- **Elasticities**

1. $\varepsilon = 0$
2. $\varepsilon = -0,1$
3. $\varepsilon = -0,3$

- **Formula**

$$d_{i,h}^{new} = d_{i,h}^{old} * \left(\frac{p_h^{new}}{p_h^{old}} \right)^\varepsilon$$

d: demand, i: customer,

h: hour, p: price

- **Rebalancing**

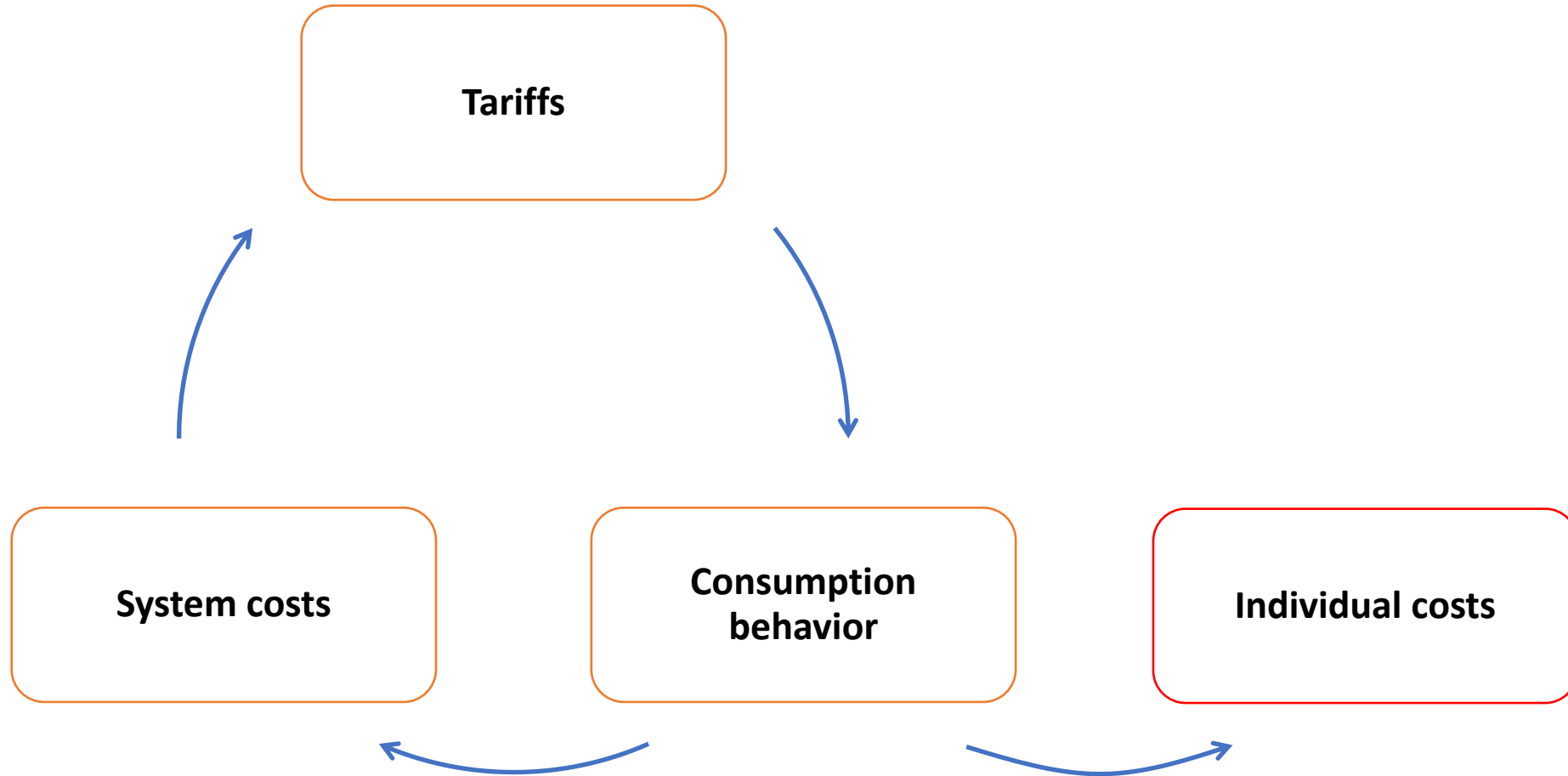
- Adjustment of fixed charges to ensure full cost recovery for non-energy costs

Table 4: Aggregate change in consumer surplus by tariff

Elasticity Case	Flat-NCDC	CPP-10	RTP-Volumetric	RTP-CCC
$\epsilon = -0.1$	\$983,429	\$445,683	\$125,181	\$10,036,693
$\epsilon = -0.3$	\$3,130,361	\$1,478,859	\$390,054	\$29,237,459

\$100-300 / household / year

Yet: minimizing overall system costs is not the only objective



Minimizing overall system costs is not the only objective



EU regulators: strong concerns regarding unknown distributional effects of new tariffs
[ACER 2016]

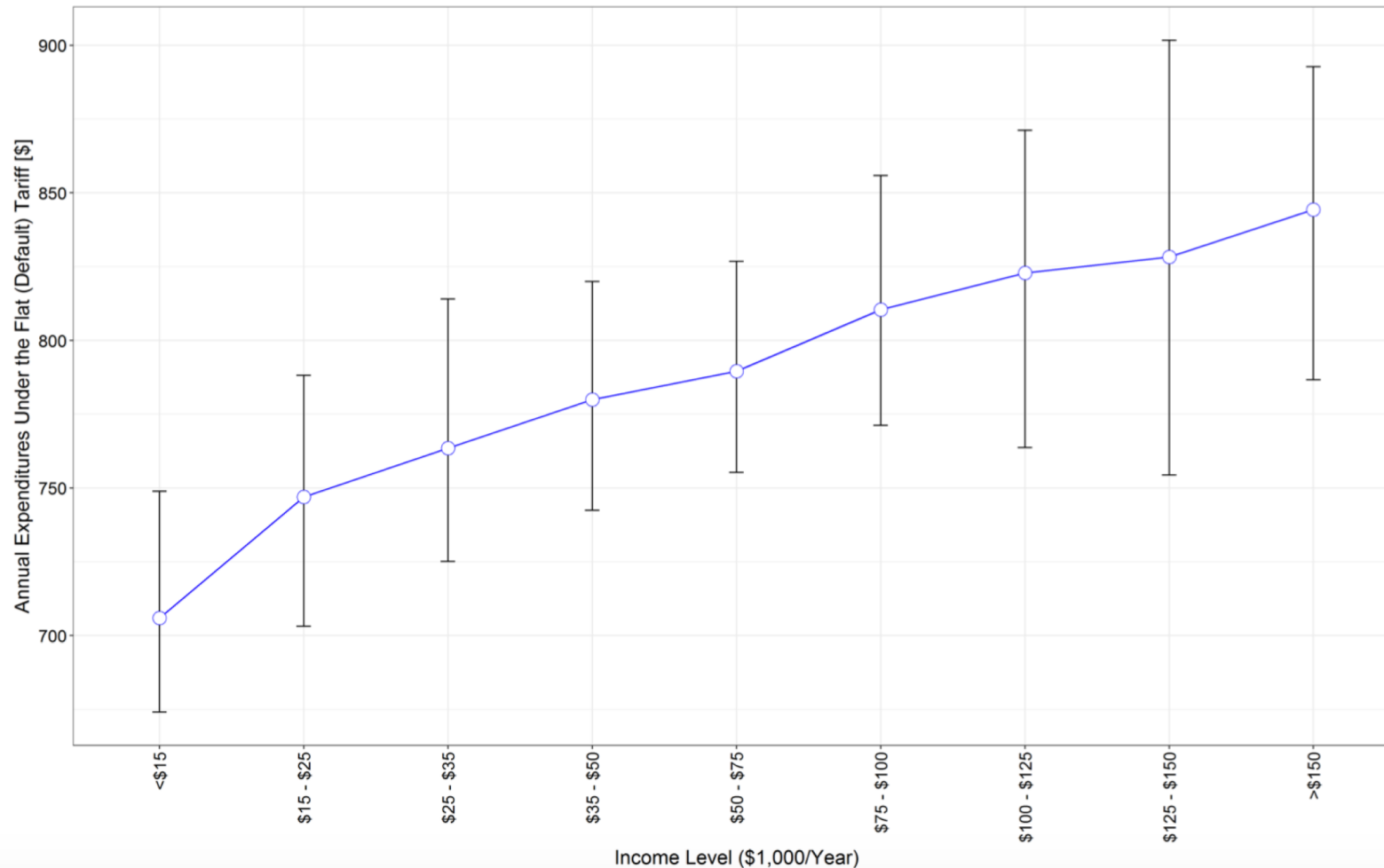


USA regulators: rejection of >80% of requests to increase fixed charges, frequently stating potential effects on low-income customers
[Trabish 2018], [Proudlove et al. 2018]

→ Importance of assessing socioeconomic effects of new tariffs

Current tariffs in many U.S. locations help keep rates low for low-income customers

Figure 1: Annual electricity expenditures under the Flat (default) ComEd tariff



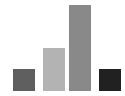
Matching consumption data with census data enables broad socioeconomic analyses



Socioeconomic data



Geographic data: Census Block Group (CBG)






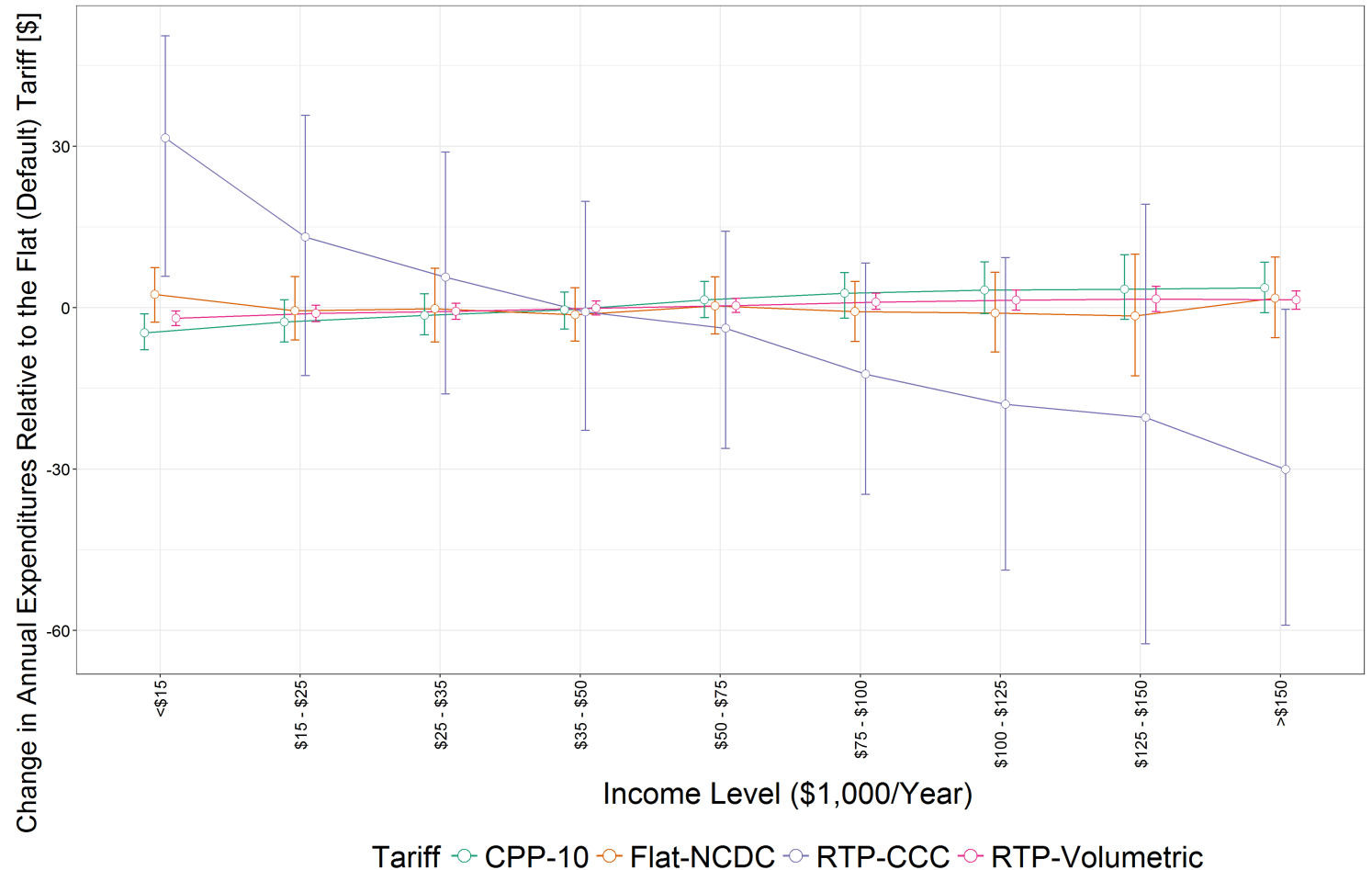
Distribution of household income in each Census Block Group



- Nine discrete income classes
- Assumption: same income probability distribution for all households
- Bootstrapping to determine confidence intervals of results

Effects of tariffs on electricity bills of low-income households (scenario: $\varepsilon = 0$)

Tariff changes	Effects on bills
Increased time-variability	
Increased fixed charges	
Capacity charges	



Proposals for mitigating bill impacts: Progressive Fixed Charges

- Objective: Maintain overall system savings while avoiding undesired social effects
- Idea: Differentiating fixed charges according to certain customer criteria
- Two proposals for discriminating variables:
 1. Customer demand characteristics
 2. Customer income

Progressive fixed charges based on customer demand characteristics

Table 5: Average Profile Variables by Income

Income (\$1,000 USD)	Average Monthly Consumption	Annual Peak Demand	Peak-To-Off-Peak Ratio	May Peak Demand	June Peak Demand	July Peak Demand	August Peak Demand	Consumption: 5:30PM-6:00PM	Consumption: 6:00PM-6:30PM	Consumption: 6:30PM-7:00PM
<\$15	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
\$15 – \$25	1.07	1.03	0.95	1.05	1.06	1.05	1.05	1.08	1.08	1.08
\$25 – \$35	1.10	1.06	0.95	1.09	1.09	1.09	1.09	1.12	1.12	1.11
\$35 – \$50	1.12	1.09	0.95	1.12	1.13	1.13	1.12	1.15	1.15	1.15
\$50 – \$75	1.14	1.13	0.97	1.17	1.17	1.17	1.16	1.18	1.18	1.18
\$75 – \$100	1.18	1.17	0.97	1.22	1.22	1.22	1.21	1.23	1.23	1.23
\$100 – \$125	1.20	1.19	0.97	1.25	1.26	1.25	1.25	1.26	1.26	1.26
\$125 – \$150	1.21	1.21	0.98	1.27	1.28	1.27	1.27	1.28	1.28	1.27
>\$150	1.25	1.29	1.02	1.36	1.35	1.34	1.33	1.32	1.33	1.32

Table 9: Average Profile Variables by Income

Income (\$1,000 USD)	Average Monthly Consumption	Annual Peak Demand	Peak-To-Off-Peak Ratio	May Peak Demand	June Peak Demand	July Peak Demand	August Peak Demand	Consumption: 5:30PM-6:00PM	Consumption: 6:00PM-6:30PM	Consumption: 6:30PM-7:00PM
<\$15	464.53	3.98	15.01	2.81	3.13	3.25	3.24	141.83	144.77	146.26
\$15 – \$25	496.02	4.11	14.31	2.94	3.30	3.42	3.40	153.56	156.47	157.87
\$25 – \$35	509.26	4.23	14.22	3.04	3.42	3.53	3.52	158.59	161.60	163.04
\$35 – \$50	521.05	4.33	14.22	3.13	3.54	3.65	3.63	163.53	166.58	167.96
\$50 – \$75	530.48	4.49	14.49	3.27	3.67	3.79	3.76	167.72	170.97	172.34
\$75 – \$100	546.66	4.63	14.51	3.41	3.83	3.94	3.92	174.55	177.91	179.21
\$100 – \$125	556.69	4.74	14.56	3.52	3.94	4.06	4.03	179.03	182.63	183.94
\$125 – \$150	561.76	4.82	14.73	3.58	4.01	4.12	4.10	181.42	185.09	186.39
>\$150	578.45	5.14	15.34	3.82	4.23	4.35	4.32	187.63	192.09	193.67

Progressive fixed charges based on customer demand characteristics



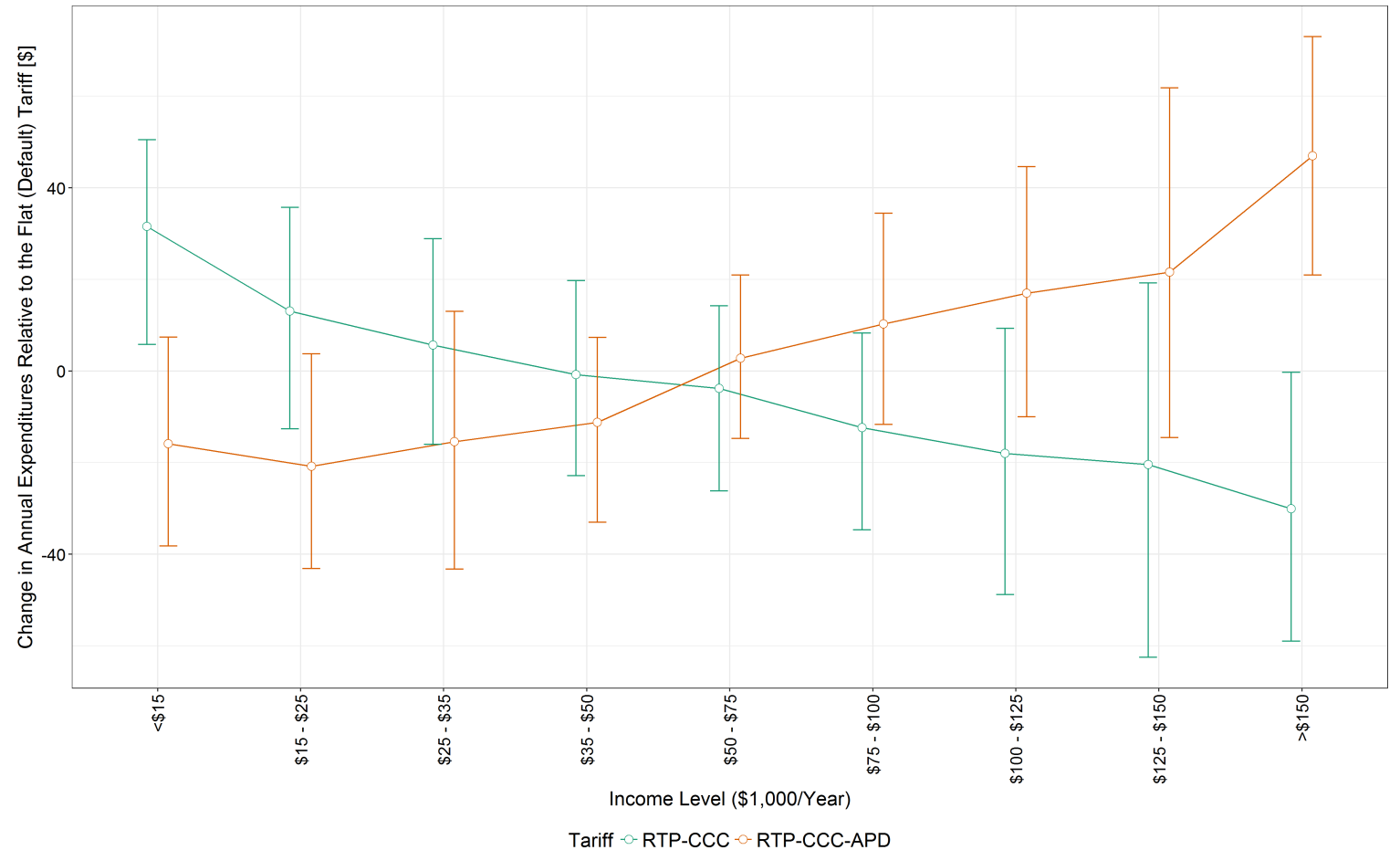
Feasible with existing and available data



Risk of Type 1 and Type 2 errors



Inefficient incentives when changed frequently



Progressive fixed charges based on customer income



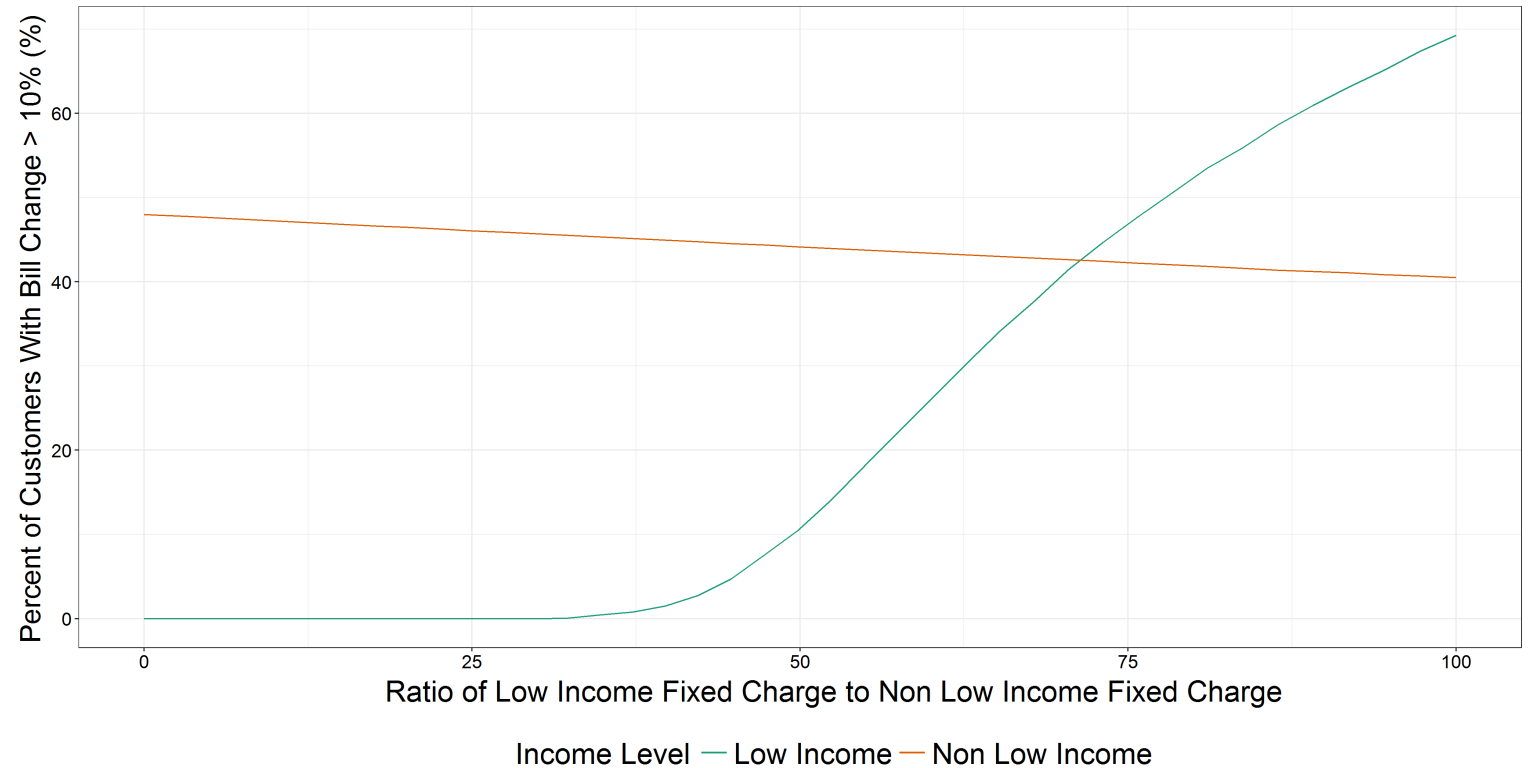
No Type 1 and Type 2 errors



Granular control over distributional effects

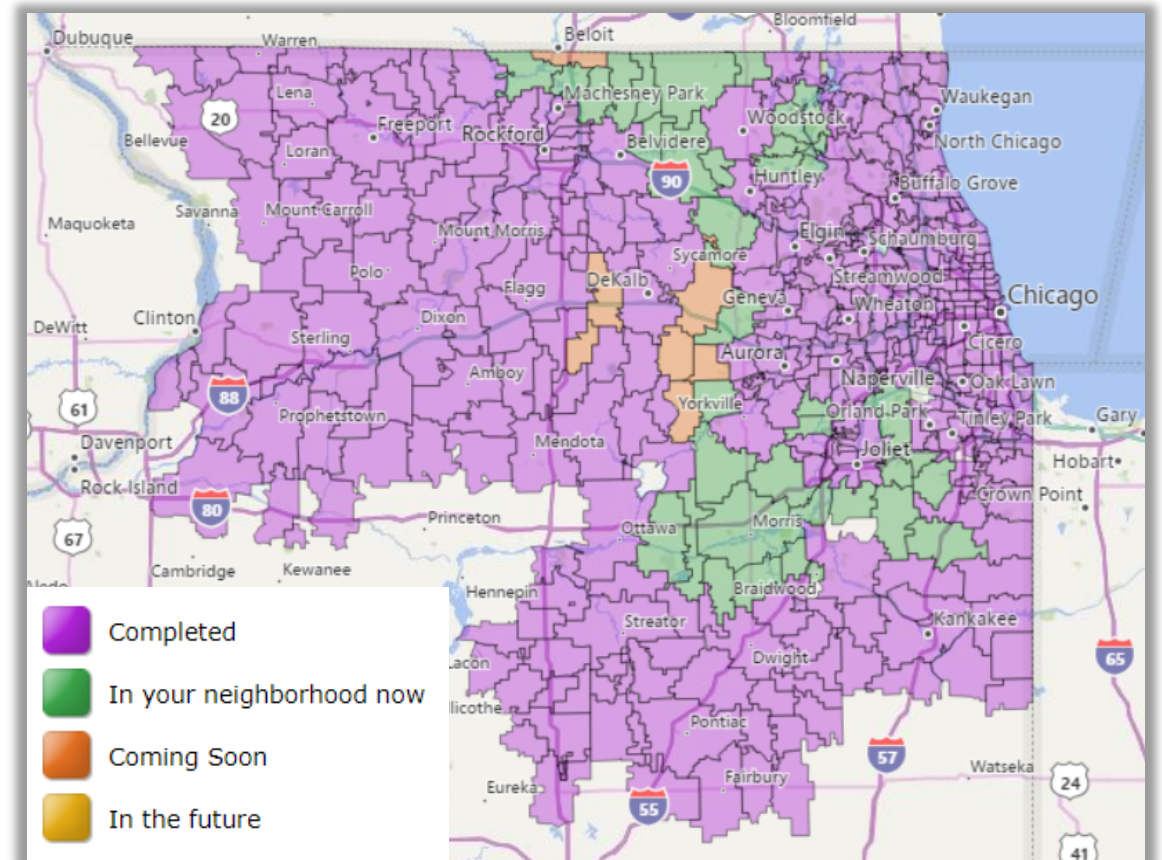


Additional sensitive customer data required



Limitations

- Consumption data
 - Cleaned according to “15/15 rule” before publishing
 - Not per se representative for US (or European) population
- Variable “household income” ignores number of residents in a household
- Assumptions for demand sensitivity:
 - All customer groups have the same elasticity
 - Customers react only to \$/kWh-prices
 - Cross-price elasticity is zero



Conclusion

1. Any transition to new tariffs creates winners and losers.
2. Moving volumetric components towards more time-varying prices benefits low-income customers (on average).
3. Transitioning to higher fixed charges causes higher average expenditures for low-income customers on average.
4. Differentiating fixed charges according to customer criteria can mitigate some or all of the undesirable distributional impacts while maintaining the desired economic efficiency benefits.

Thank you for your attention

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