



JOHNS HOPKINS

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ENERGY INSTITUTE

# Risk Management and Clean Energy Transition

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# EPICS



Imperial College  
London



# Energy at Hopkins

# What Matters for Efficient Risk Management?

## Make Better Decisions with Less

### Better:

Cost, reliability, emissions  
Physics-informed models  
Market suitable  
Fast computations  
Insight discovery

### Less:

Data requirements  
Computational costs  
Malfunction  
Distrust

# A Typical Decision-Making Pipeline

Consider a decision-making problem (SCUC/SCED):  $\min_{x \in \mathcal{X}(\omega)} C(x, \omega)$

$C(\cdot)$  objective (cost) function

$\mathcal{X}$  feasible solution space

$x$  vector of decision variables

$\omega \in \Omega$  vector of uncertain parameters

## Bottlenecks for risk management:

- Scenario generation
- Accelerated computations
- Risk representation

## Concerns:

- Do we have enough data?
- What about guarantees and solution accuracy?
- Do we target the right risk?
- Who will use it?

## Current markets:

- No differentiation between extreme and normal reserve
- No direct interface for risk management (VB, CfD, etc)

If extremes are not considered, why should producers care?

# How to Fit Risk Into This Pipeline?

Consider a decision-making problem (SCUC/SCED):  $\min_{x \in \mathcal{X}(\omega)} C(x, \omega)$

$C(\cdot)$  objective (cost) function                       $\mathcal{X}$  feasible solution space

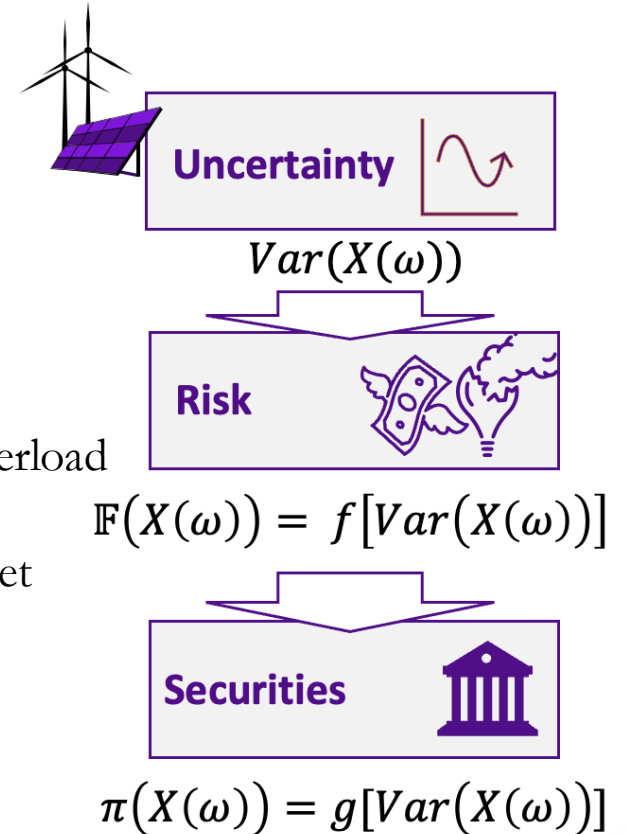
$x$  vector of decision variables                       $\omega \in \Omega$  vector of uncertain parameters

## Uncertainty arises from

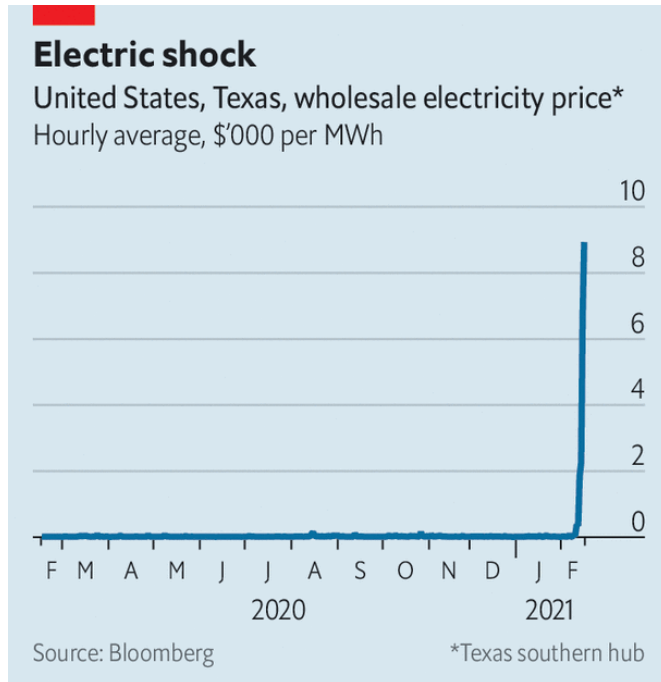
- Forecast errors (renewables and demand)
- Sudden unavailability of resources, full or partial
- *Complex* events

## Risk arises from uncertainty, regardless of its source

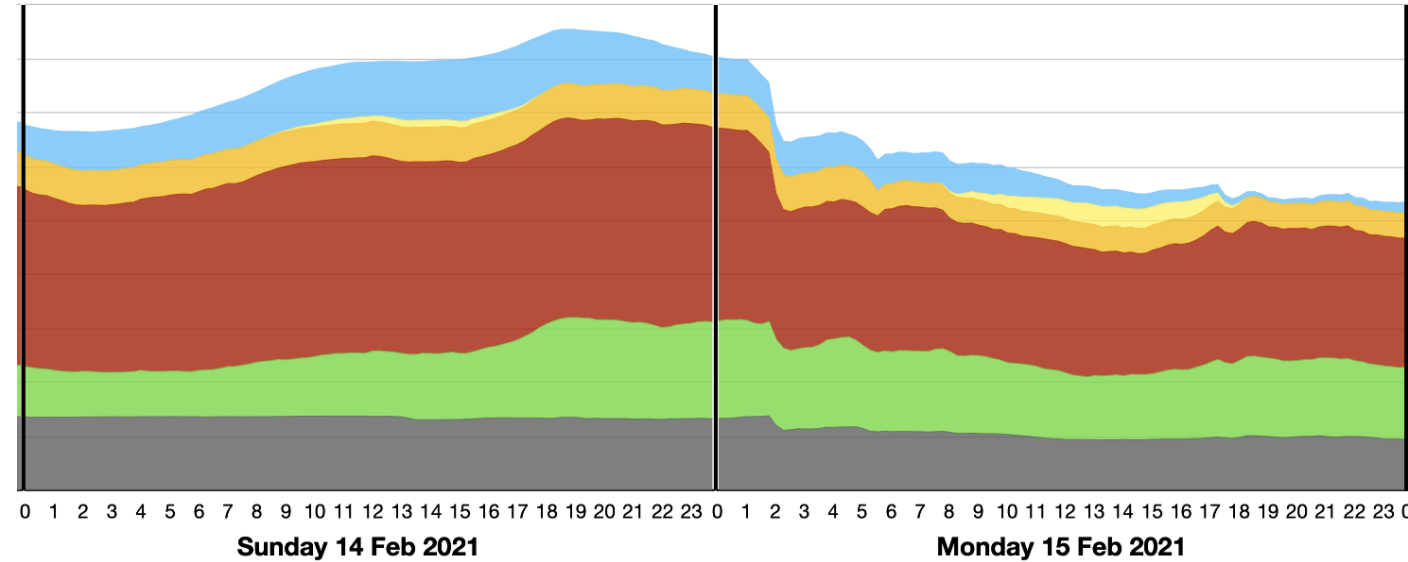
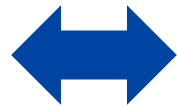
- Physical risk (e.g. constraint violation or operational infeasibility – overload or power mismatch)
- Financial risk (e.g. shortfall from physical risk – profit, liquidity, market losses)
- Risk = Probability  $\times$  Consequence
- Various risk metrics exist (CVaR, VaR, CoVaR, etc)



# Bottlenecks Cause “Missing Money” Effects: A Weather Example



Low availability  
led to high  
prices



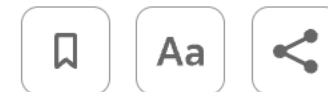
The Economist

Commodities

## Texas cuts \$9,000 power price cap after February freeze

Reuters

December 3, 2021 11:53 AM EST · Updated 2 years ago



# Bottlenecks Cause “Missing Money” Effects: A Weather Example

Texas did the wrong thing (once again), it is obvious:

- Reduced prices cap led to less incentives to improve availability and participate in the market, especially for high-stake hours
  - High demand
  - Low availability of supply
  - A combination of both

The important and overlooked argument is that market prices didn't help

- Let's discuss ideal-world prices

“ideal” price  $\longrightarrow \lambda_{i,j} = \lambda_{i,j}^e \cdot \pi^0 + (\lambda_{i,j}^{max} - \lambda_{i,j}^e) \cdot (1 - \pi^0)$

price cap  $\longleftarrow$

marginal price of energy  $\longleftarrow$  probability of normal operations  $\longleftarrow$  probability of an “oops”

# Out of Market Interventions Can be Much Worse

**What is the spot market for wholesale electricity, and how will AEMO's decision to suspend it affect consumers?**

By Tobias Jurss-Lewis

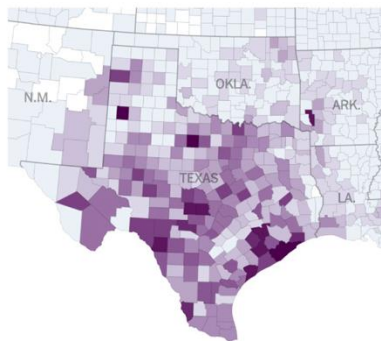
Energy Industry

Wed 15 Jun 2022

STATE

**Texas Supreme Court sides with state regulators on \$16 billion winter storm overcharges**

Tuesday, 12:15 p.m.  
4.4 million customers (35.1%)



**\$1B+ in losses/hr**



Several important “market” limitations:

- Limited demand elasticity + shielded consumption
- No incentives to stay in the market, especially during critical hours
  - High demand
  - Low availability of supply
  - A combination of both
- Self-commitment is always an option

**1GW producer lost opportunity of ~\$100K/hr**

# “Missing Money” as a Problem Statement

Private risk evaluation of “firm” agent  $i$

$$\begin{aligned} \max_{\alpha_i, p_{G,i}} \mathbb{R}_i [\lambda_i p_{G,i}(\omega) + \chi_i \alpha_i] \\ p_{G,i}(\omega), \alpha_i \in \mathcal{O}_i \end{aligned}$$

Private risk evaluation of “variable” agent  $j$

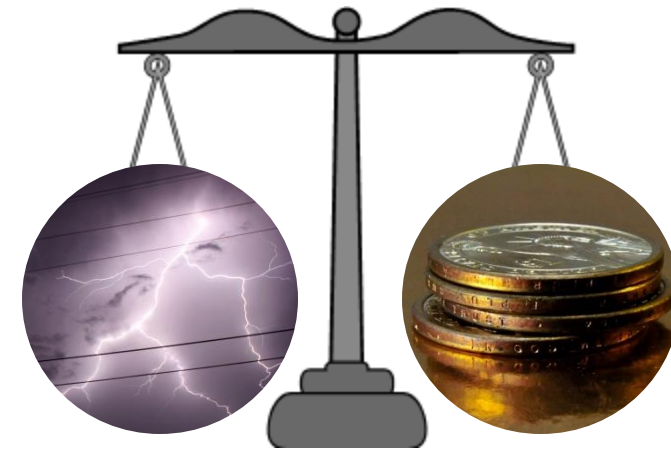
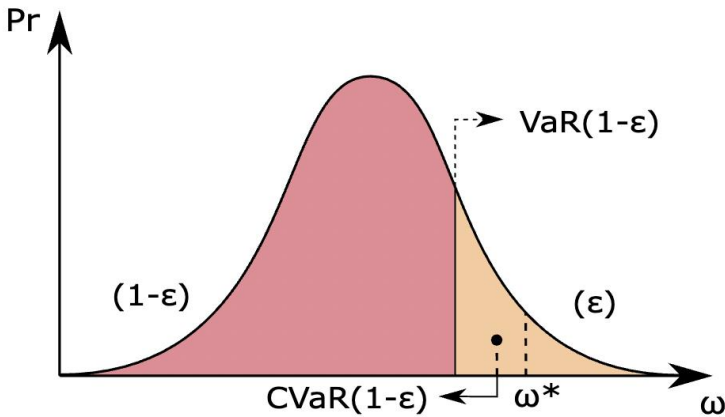
$$\begin{aligned} \max_{\alpha_j, p_{U,j}} \mathbb{R}_j [\lambda_{i:j} p_{U,j}(\omega) - \chi_{i:j} \alpha_j] \\ p_{U,j}(\omega), \alpha_j \in \mathcal{O}_j \end{aligned}$$



Social risk evaluation



$$\begin{aligned} p_{U,j:i} + p_{G,i} &= p_{D,i} && :(\lambda_i) \\ \sum_{i:j} \alpha_{i:j} &= 1 && :(\chi_{i:j}) \end{aligned}$$



What is “worst-case”  $\omega^*$  and “optimal”  $\epsilon$ ?

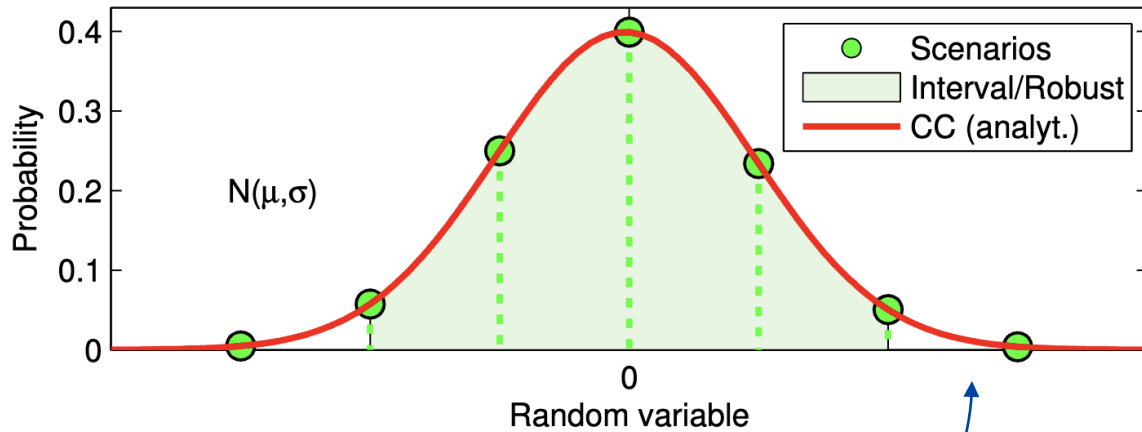
Two missing parts: (i) **extremes** and (ii) **volatility**



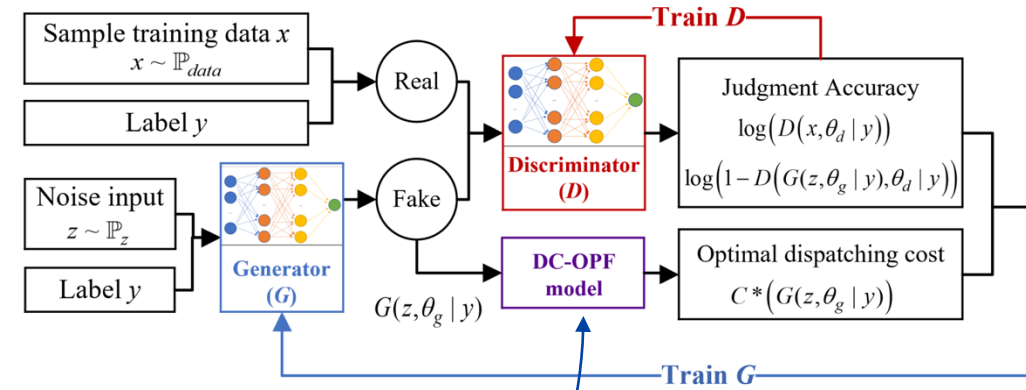
## What is Needed to Overcome “Missing Money” Effects?

**Risk management** requires thinking across three dimensions of representation of risks – (i) **modeling extreme outcomes and variability**, (ii) **endogenizing them into decision support tools** and (iii) **alignment of incentives and risk management goals**.

# Aligning Private and Social Risks: Represent Extreme Outcomes



Hard to sample extreme events



Hard to implement physics of extreme events

Risk modeling is largely informed by financial engineering, benefiting from:

- Lack of physical complexity
- Randomness
- Rather “smooth” jumps

$$C = N(d_1)S_t - N(d_2)Ke^{-rt}$$

where  $d_1 = \frac{\ln \frac{S_t}{K} + (r + \frac{\sigma^2}{2})t}{\sigma\sqrt{t}}$

and  $d_2 = d_1 - \sigma\sqrt{t}$

# Aligning Private and Social Risks: Endogenize Extreme Outcomes

More generic and nuanced approaches to sample *less-likely* events is needed

$$\begin{array}{l} \underset{u \in \mathcal{U}}{\text{minimize}} \quad J(u) \\ \text{subject to} \quad \mathbb{P}(F(u, \xi) \geq z) \leq \alpha, \quad \text{where } \alpha \ll 1. \end{array}$$

Traditional risk  
constraints



$$\begin{array}{l} \mathbb{P}(F(u, \xi) \geq z) \\ \xi^* \in \underset{\xi \in \Xi}{\text{argmin}} \{I(\xi) : F(u, \xi) \geq z\}. \end{array}$$

Risk constraints with  
large-deviation theory

Challenges:

- Depend on the indicator function  $I(\cdot)$
- Includes inner minimization for  $\xi^*$
- Not clear how to compute the optimal solution
- Likely to be extremely conservative

# Aligning Private and Social Risks: Endogenize Extreme Outcomes

$$\begin{aligned} & \underset{u \in \mathcal{U}}{\text{minimize}} && J(u) \\ & \text{subject to} && P_k(u, z, \xi^*) \leq \alpha \\ & && \xi^* \in \underset{\xi \in \Xi}{\text{argmin}} \{I(\xi) : F(u, \xi) \geq z\} \end{aligned}$$

Employ a three-step strategy:

- Reformulate the inner level using first-order conditions
- Postulate the indicator function
- Obtain single-level approximation

# Aligning Private and Social Risks: Endogenize Extreme Outcomes

$$\begin{aligned} & \underset{u \in \mathcal{U}}{\text{minimize}} && J(u) \\ & \text{subject to} && P_k(u, z, \xi^*) \leq \alpha \\ & && \xi^* \in \underset{\xi \in \Xi}{\text{argmin}} \{I(\xi) : F(u, \xi) \geq z\} \end{aligned}$$

Cannot be solved at scale and efficiently

$$\begin{aligned} & \underset{u, \xi^*, \lambda}{\text{minimize}} && J(u) \\ & \text{subject to} && u \in \mathcal{U}, \xi^* \in \Xi, \lambda \in \mathbb{R}_+ \\ & && P_k(u, z, \xi^*) \leq \alpha, \\ & && F(u, \xi^*) = z, \\ & && \nabla I(\xi^*) = \lambda \nabla_{\xi} F(u, \xi^*). \end{aligned}$$

Note that  $\geq$  is replaced with  $=$

- No convexity of  $F(\cdot)$  is required
- Indicator function is low

$$\begin{aligned} & \underset{u, \xi^*, \eta^*, \lambda}{\text{minimize}} && J(u) \\ & \text{subject to} && u \in \mathcal{U}, \xi^* \in \Xi, \eta^* \in \mathbb{R}^n, \lambda \in \mathbb{R}_+ \\ & && P_k(u, z, \xi^*) \leq \alpha, \\ & && F(u, \xi^*) = z, \\ & && \eta^* = \lambda \nabla_{\xi} F(u, \xi^*), \\ & && \xi^* = \nabla S(\eta^*). \end{aligned}$$

Single-level approximation

- Ensures convexity
- Provides guarantees
- Captures extreme cases
- Can be used for pricing

# Aligning Private and Social Risks: Endogenize Extreme Outcomes

$$\begin{aligned} & \underset{u \in \mathcal{U}}{\text{minimize}} && J(u) \\ & \text{subject to} && P_k(u, z, \xi^*) \leq \alpha \\ & && \xi^* \in \underset{\xi \in \Xi}{\text{argmin}} \{I(\xi) : F(u, \xi) \geq z\} \end{aligned}$$

Cannot be solved at scale and efficiently

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Treatment of risk via Taylor's expansion

$$P_1(u, z, \xi^*) = \Phi(-\sqrt{2I(\xi^*)}) = \Phi(-\|\xi^* - \mu\|_{\Sigma^{-1}}).$$

or

$$P_2(u, z, \xi^*) \approx \Phi(-\|\xi^* - \mu\|_{\Sigma^{-1}}) \det_{\perp \hat{n}}(H)^{-\frac{1}{2}}$$

# Aligning Private and Social Risks: Example

Chance-constrained ED

$$\begin{aligned}
 & \min_{\Xi} \mathbb{E}_{\omega} \left[ \sum_{n \in \mathcal{G}} C_n(p_n(\Omega)) \right] \\
 & \text{s.t.} \quad \sum_{n \in \mathcal{G}} p_n + \hat{W} - d = 0 \\
 & \quad p_n(\Omega) = p_n + \delta_n(\Omega) \quad \forall n \\
 & \quad \mathbb{P}_{\omega} [p_n^{\min} \leq p_n(\Omega) \leq p_n^{\max}] \geq 1 - \epsilon \quad \forall n \\
 & \quad \sum_{n \in \mathcal{G}} \delta_n(\Omega) - \Omega = 0
 \end{aligned}$$



LDT-constrained ED

$$\begin{aligned}
 & \min_{p, \alpha, \beta} \mathbb{E}_{\omega} \left[ \sum_n C_n(p_n, \alpha_n, \beta_n) \right] \\
 & \text{s.t.} \quad \alpha_n, \beta_n, p_n \geq 0 \quad \forall n \\
 & \quad \mathbb{P}_{\omega} [p_n^{\min} \leq p_n - \alpha_n \omega \\
 & \quad \quad \leq p_n^{\max}, \forall n] \geq 1 - \epsilon \\
 & \quad \mathbb{P}_{\omega} [p_n^{\min} \leq p_n - \alpha_n \omega - \beta_n \omega \\
 & \quad \quad \leq p_n^{\max}, \forall n] \geq 1 - \epsilon^{\text{ext}} \\
 & \quad \sum_n p_n + \hat{W} - d = 0 \\
 & \quad \sum_n \alpha_n = 1 \\
 & \quad \sum_n \beta_n = 1
 \end{aligned}$$

$$\begin{aligned}
 & P_k(p_n, \alpha_n, \beta_n) \leq 1 - \epsilon^{\text{ext}} \quad \forall n \\
 & \omega^* \in \arg \min_{\omega} \{ I(\omega) : p_n^{\min} \leq \\
 & \quad p_n - (\alpha_n + \beta_n) \omega \leq p_n^{\max} \}
 \end{aligned}$$

# Aligning Private and Social Risks: Example

Single-level LDT-constrained ED

$$\begin{aligned} \min_{p, \alpha, \beta, \omega^*, \lambda^*} \quad & \mathbb{E}_\omega \left[ \sum_n C_n(p_n, \alpha_n, \beta_n) \right] \\ \text{s.t.} \quad & \alpha_n, \beta_n, p_n \geq 0, \lambda^* > 0 \quad \forall n \\ (\delta_n^+) : \quad & p_n - p_n^{max} + \alpha_n \hat{\sigma}_n \leq 0 \quad \forall n \\ (\mu_n^+) : \quad & -p_n^{max} + p_n - (\alpha_n + \beta_n) \omega^* = 0 \quad \forall n \\ (\nu) : \quad & \Sigma^{-1/2} \omega^* + \Phi^{-1}(\epsilon^{ext}) \leq 0 \\ (\xi_n) : \quad & \Sigma^{-1} \omega^* - (\alpha_n + \beta_n) \lambda^* = 0 \quad \forall n \\ (\pi) : \quad & \sum_n p_n + \hat{w} - d = 0 \\ (\rho) : \quad & \sum_n \alpha_n - 1 = 0 \\ (\chi) : \quad & \sum_n \beta_n - 1 = 0 \end{aligned}$$

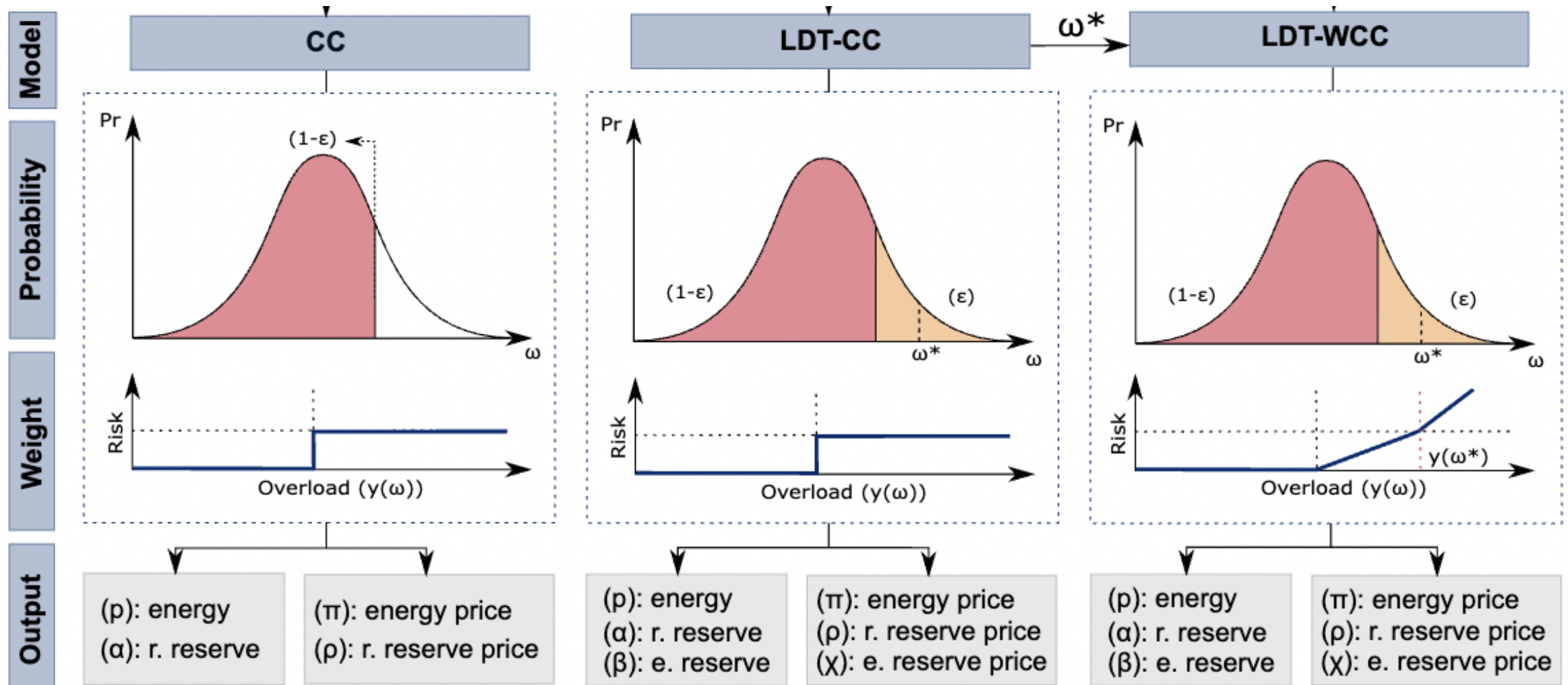
Extremely conservative!

Here is an idea: use weighted chance constraints to alleviate conservatism!

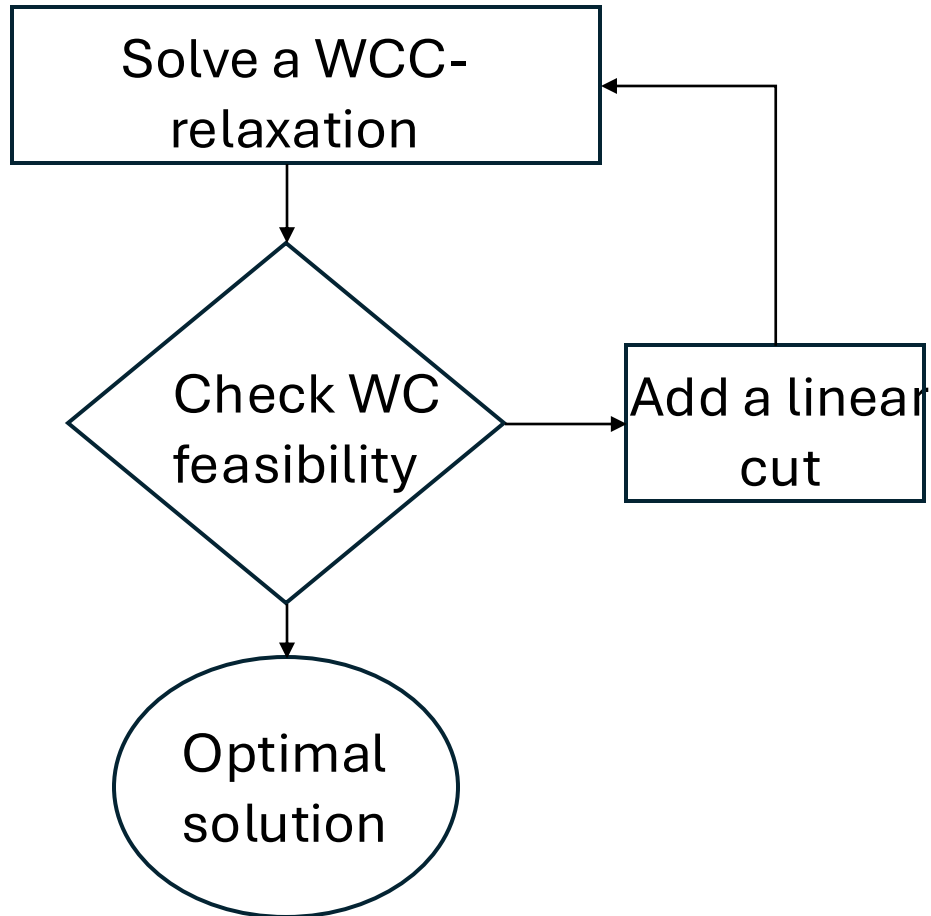


# Aligning Private and Social Risks: Example

Weighted chance constraints to avoid conservatism by regulating your rate of response (risk)



# Aligning Private and Social Risks: Computable Equilibrium!



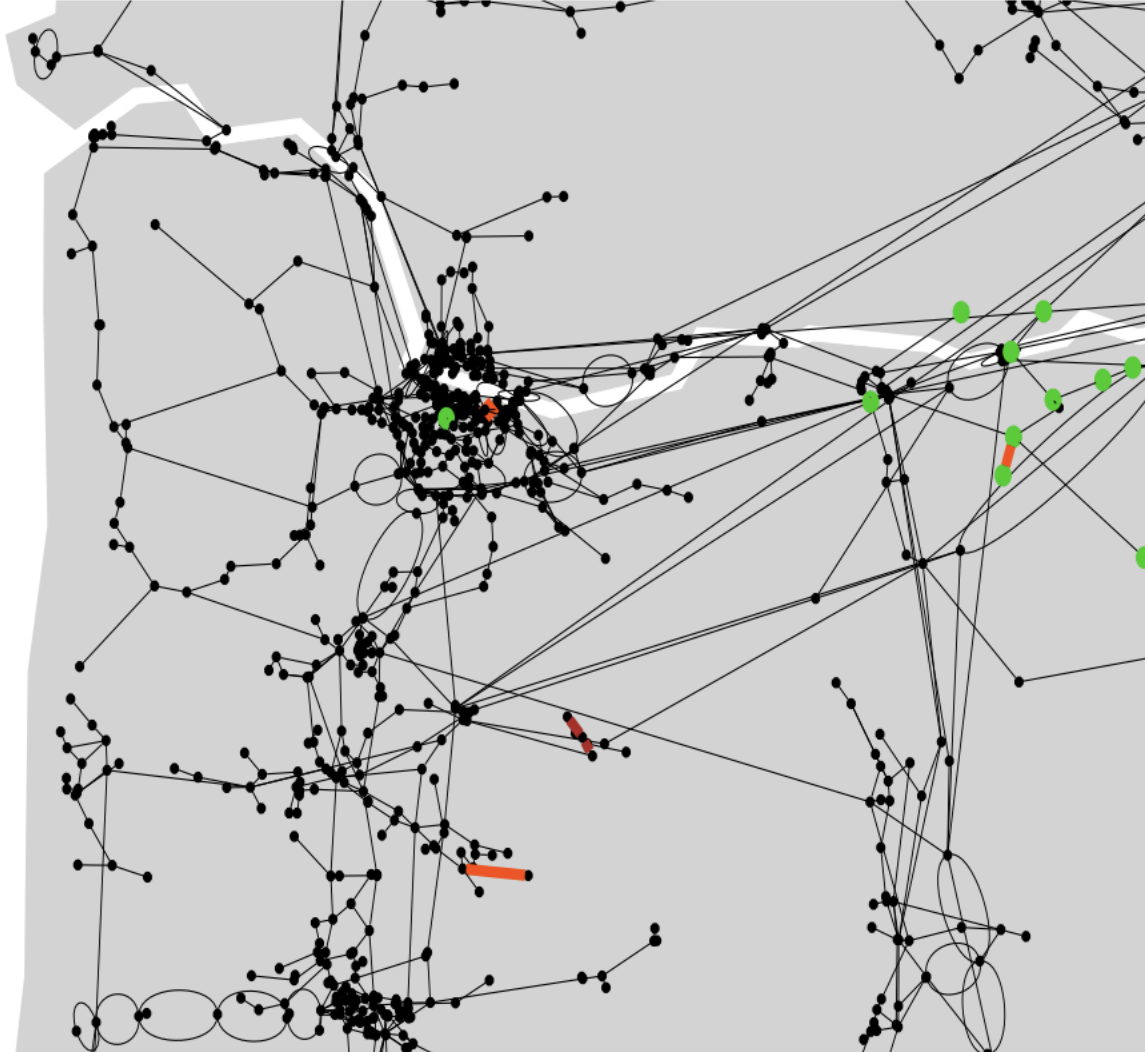
**1) Theorem 1:** *Equilibrium payments:* Let  $\{p_n^*, \alpha_n^*, \beta_n^*, \omega^*, \lambda^*\}$  be the optimal solution of the problem [ref] and let  $\{\pi, \rho, \chi\}$  be the dual variables. Then,  $\{p_n^*, \alpha_n^*, \beta_n^* \forall n\}, \omega^*, \lambda^*, \pi, \rho, \chi\}$  constitutes a market equilibrium.

- The market clears at  $\sum p_n - \hat{W} = d$ ,  $\sum \alpha_n = 1$ , and  $\sum \beta_n = 1$
- Each producer maximizes its profit under the payment  $\Gamma_n = \pi p_n + \rho \alpha_n + \chi \beta_n$

First given a  $(\omega^*, \lambda)$  if  $\{p_n^*, \alpha_n^*, \beta_n^* \forall n\}$  is feasible and solved to optimality, optimal values  $\{p^*, \alpha^*, \beta^* \forall n\}$  must satisfy equality constraints. And as the result  $\sum p_n^* - \hat{w} = d$ ,  $\sum \alpha_n^* = 1$ , and  $\sum \beta_n^* = 1$

**Important:** completes market with risk, while ensuring cost recovery and revenue adequacy.

# Aligning Private and Social Risks: Results



Solvable for a realistically large instance:

- 2209 nodes
- 2866 nodes
- June 2022 data set

Solver: Gurobi

One instance w/out cutting planes: 19.4 s

One instance w/cutting planes: 3.7 s

Cost savings (relative to non-WCC case) – 3.9%

# Aligning Private and Social Risks: Results

Considering extreme events does change dispatch and reserve allocation

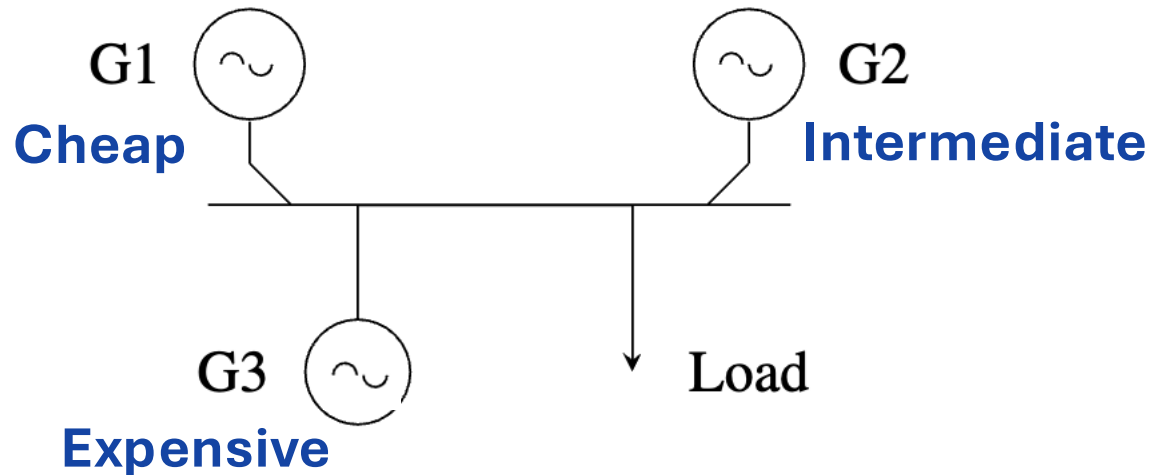


TABLE I: Optimal primal results

Model	Energy ( $p$ ) & Reserves ( $\alpha, \beta$ ) dispatch			
	Variables	G1	G2	G3
CC	$p^*$ [MW]	40	45	30
	$\alpha^*$ [%]	0	0	1
LDT-WCC	$p^*$ [MW]	40	36.05	38.95
	$\alpha^*$ [%]	0	0.04	0.96
	$\beta^*$ [%]	0	1	0
LDT-CC	$p^*$ [MW]	40	36.45	38.55
	$\alpha^*$ [%]	0	0	1
	$\beta^*$ [%]	0	0.68	0.32

The most diversified reserve portfolio ←

Considering extreme events does change reserve, not energy prices

TABLE II: Optimal dual results and total cost

Model	Energy ( $\pi$ ) & Reserves ( $\rho, \chi$ ) prices			
	$\pi^*$ [\$/MW]	$\rho^*$ [\$/MW]	$\chi^*$ [\$/MW]	T. Cost [\$]
CC	35.30	89.99	-	3502.18
LDT-WCC	35.46	80.42	100.75	3591.03
LDT-CC	35.41	90.55	120.00	3621.31

# Aligning Private and Social Risks: Larger Instances

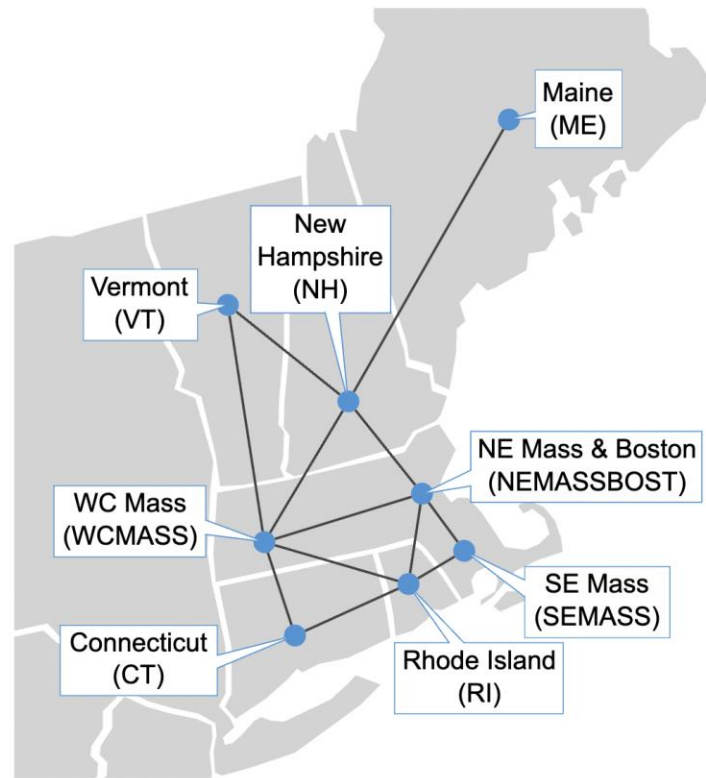


TABLE IV: Optimal dual results [\$/MW]

	Price	CC	LDT-WCC	LDT-CC
Energy	$\pi_{CT}^*$	55.84	56.44	56.18
	$\pi_{ME}^*$	125.61	126.29	126.01
	$\pi_{NEMASSB}^*$	141.93	142.98	142.32
	$\pi_{NH}^*$	125.61	126.29	126.01
	$\pi_{RI}^*$	51.50	52.19	51.90
	$\pi_{SEMASS}^*$	207.02	207.63	207.39
	$\pi_{VT}^*$	118.08	118.77	118.48
	$\pi_{WCMASS}^*$	106.78	107.48	107.19
Reg. Res	$\rho^*$	280	850	1049.22
Ext. Res	$\chi^*$	-	1500	1699.224

Pay more upfront to avoid being sorry

Consistent zonal **energy prices** and **monotonic (to risk) reserve prices**.

## So What?

- Let's discuss ideal-world prices:

$$\lambda_{i:j} = \lambda_{i:j}^e \cdot \pi^0 + (\lambda_{i:j}^{max} - \lambda_{i:j}^e) \cdot (1 - \pi^0)$$

- We found the best proxy by introducing an additional (extreme) reserve product and completing market design with risk:

$$\lambda_{i:j} = \lambda_{i:j}^e \cdot \pi^0$$
$$\chi \sim \mathbf{E}_{\pi^0} [(\lambda_{i:j}^{max} - \lambda_{i:j}^e(\pi^0)) \cdot (1 - \pi^0)]$$

- Still, it doesn't solve the problem of price volatility

# Price Volatility

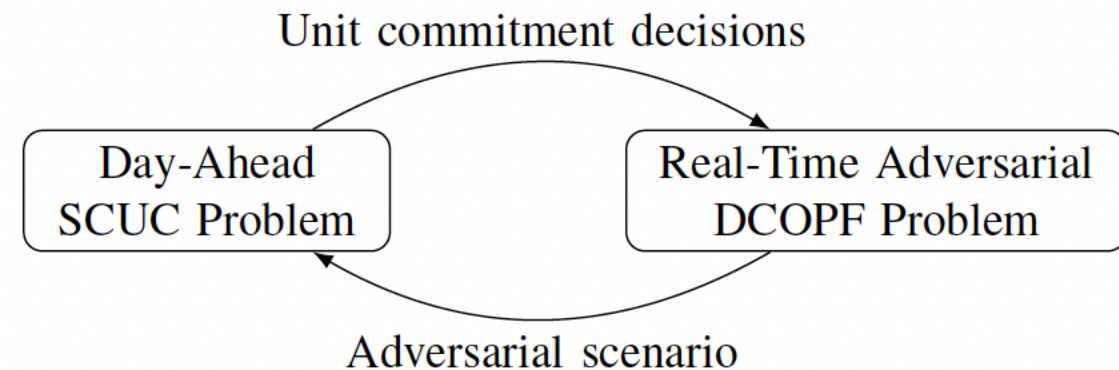
Price volatility drives consumer's risk exposure (recall largely inelastic demand):

- Volatile prices can still be efficient though
- Hedges against volatility exists (e.g., VB)
- Important point: we do not seek to eliminate volatility
- Goal: Complete markets with information about volatility

RT Load – DA Load

		<b>HIGH</b>	<b>VERY HIGH</b>
High			
Medium			<b>HIGH</b>
Low or negative			
	Low or negative	Medium	High

RT LMP



# Price Volatility

Adversarial problem can come in a variety of forms, but there are two conditions:

- Must be internalized with current market designs
- Must be “computable”
- Must be “priceable”

$$\min \sum_{t \in \mathcal{T}^{\text{DA}}} \left( \sum_{g \in \mathcal{G}} (h_{gt} + C_g^{\text{Start}} v_{gt} + C_g^{\text{Down}} w_{gt}) + \sum_{i \in \mathcal{N}} C^{\text{VOLL}} p_{it}^{\text{Unmet}} \right) + \rho \hat{V}(\mathbf{y})$$

Typical  
SCUC/SCED  
constraints

$$\hat{V}(\mathbf{y}) = \max_{\omega \in \Omega} \frac{1}{N^{\text{tp}}} \sum_{t \in \mathcal{T}^{\text{RT}}} \sum_{i \in \mathcal{N}} \lambda_{it}(\mathbf{y}, \omega) (d_{it}^{\text{RT}} - \bar{D}_{it})^+$$

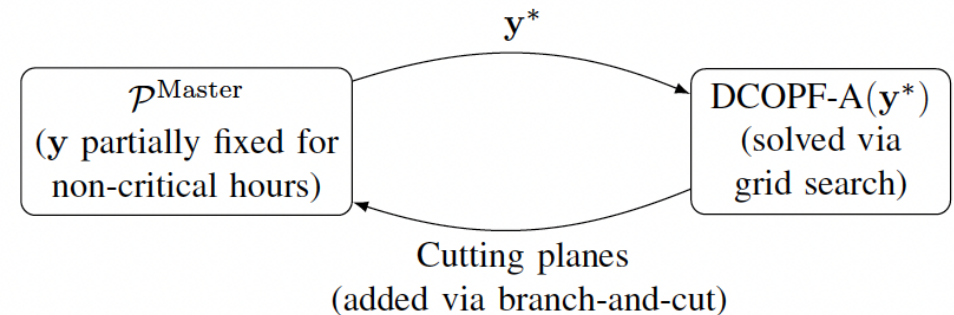
Proxy for  
consumer risk  
exposure

RT price

Power  
mismatch

Decomposable problem:

- No-good, L-shaped, LBBD cuts
- Solves as quick as SCUC





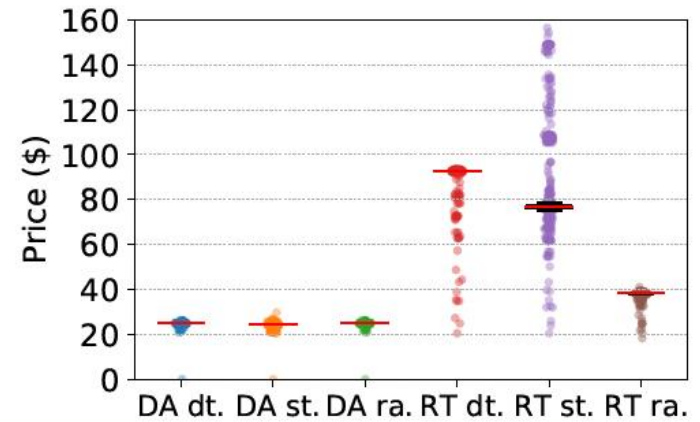
# Price Volatility

TABLE I: COMPARISON OF RISK-AWARE AND DETERMINISTIC SCUC PROBLEMS FOR  $\rho = 1$

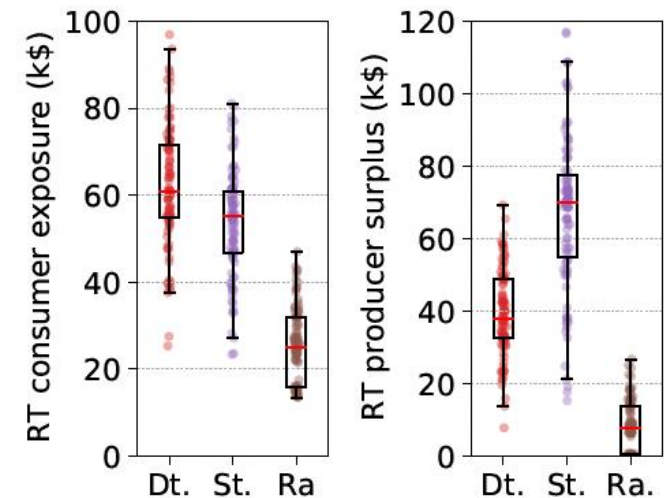
$R^d$	$R^w$	Save (k\$)	Deter. cost (M\$)	Cost red. (%)	DA cost diff (\$)	Consr. exp. (k\$)
0.1	0.2	0.00	5.37	0.00	0.00	8.53
0.1	0.4	0.12	5.37	0.00	116.16	8.53
0.1	0.6	0.00	5.37	0.00	0.00	8.89
0.1	0.8	42.29	5.41	0.78	116.16	8.89
0.1	1.0	42.28	5.41	0.78	123.50	8.89
0.2	0.2	114.71	5.50	2.09	116.16	17.78
*0.2	0.4	114.14	5.50	2.08	688.03	17.78
0.2	0.6	115.74	5.50	2.11	1446.22	16.76
0.2	0.8	108.58	5.50	1.98	8130.20	17.78
*0.2	1.0	115.64	5.50	2.10	1072.06	17.78

\* Instance solved with 3 root cuts.

- Consumer risk exposure is reduced at no expense to the system efficiency.
- Risk management is not orthogonal to efficiency.
- Dramatic reduction in consumer exposure



(a)



## Concluding Thoughts

- Risk management scales to realistically large networks
- Extreme outcomes and volatility are considered endogenously and without computationally intensive sampling
- We provide a robust pricing framework that captures option-value of resiliency
- This framework can be adapted to other applications

# Thank you! Questions? Suggestions? Feedback?

- We are constantly looking for Ph.D. applicants
- Reach out to us at [ydvorki1@jhu.edu](mailto:ydvorki1@jhu.edu)

- Collaborators:



Tomas  
Tapia



Zhirui  
Liang



Daniel  
Bienstock



Cheng  
Guo