

A Bilevel QP-PLP Approach to Demand Response Modulation between Consumers and a Single Electricity Seller

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**Demand response:
What, why, when, how?**

What and why?

Demand Response (DR) refers to situations where *consumers readjust their behaviors in reaction to some signals of the electricity providers.*

DR helps in many aspects including.....

- flexibility issues
- reliability issues
- improving profits
- sustainability issues and
- reducing environmental impact

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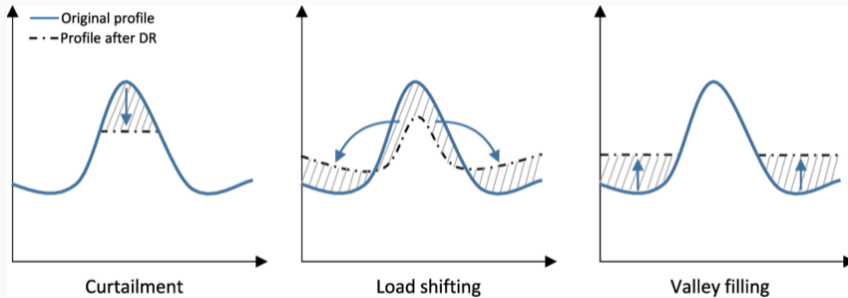
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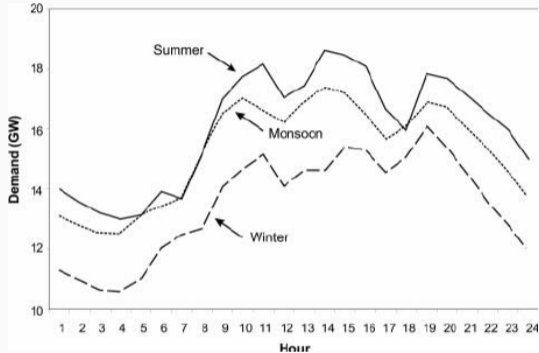
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What and why?



When?



In Thailand, the intensive use of electricity occurs during 9AM – 9PM. Hence the ideal solution could be to shift these loads to the off-peak hours.

How?

Q: How to force the consumers into the DR program ?

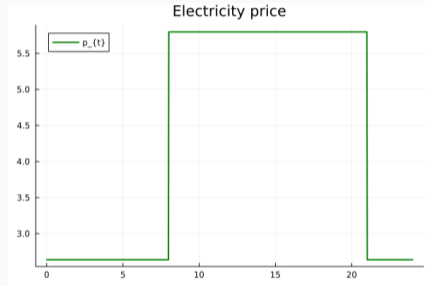
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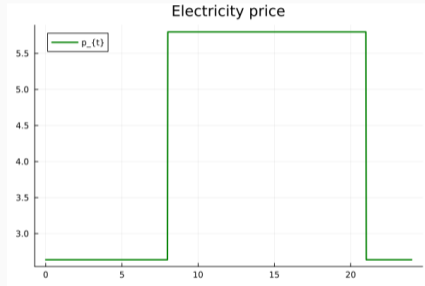
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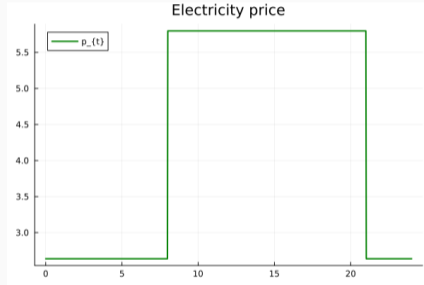
Higher on-peak prices forces the consumers to shift their loads... **HOPEFULLY!**

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Where do the missing demands go?

They need a new resources...

- the grid, when it is cheap,
- renewables,
 - △ solar panels
 - △ wind turbines
 - △ etc.
- exchange networks,
 - △ virtual power plant
 - △ storages
 - △ vehicles to grid
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Unilateral vs Bilateral Optimization

Attitude towards demand.

Q: Who are in charge of the demand?

A: The consumers. (Classical approach.)

The model: Find a strategy that maximizes their own utility.

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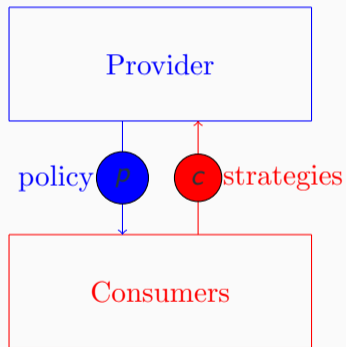
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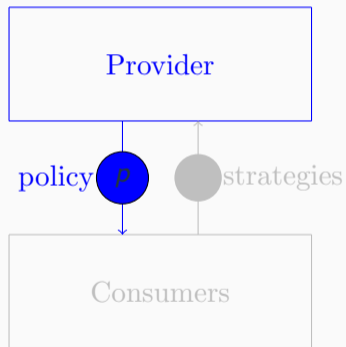
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The provider offers a policy.

The consumers response with their strategies.

The utilities of both actors were affected by both policy and strategies.

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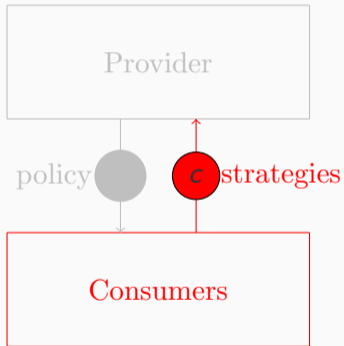
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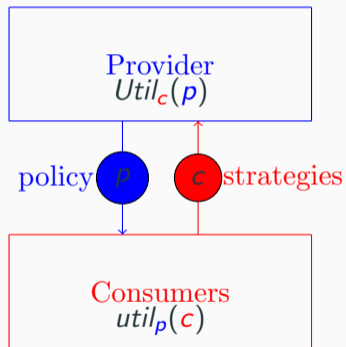
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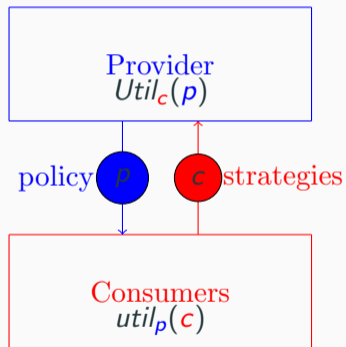
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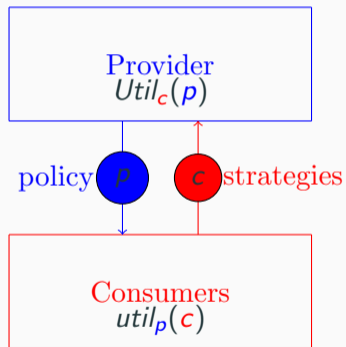


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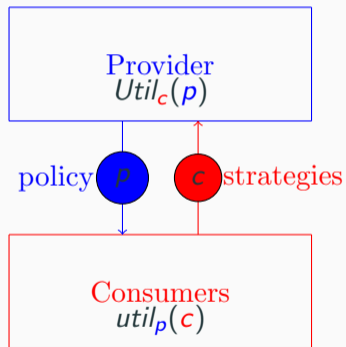
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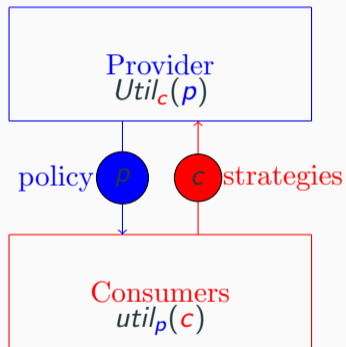
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Solution Method and Simulation

Solution method: Please bring your multipliers

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The parts in CC of the form

$$0 \leq F(u) \perp G(u) \geq 0$$

usually causes the constraint set to be nonconvexity and even nonconnected.

Here, we **branch** on the **perpendicularity** instead of having them as equality constraints.

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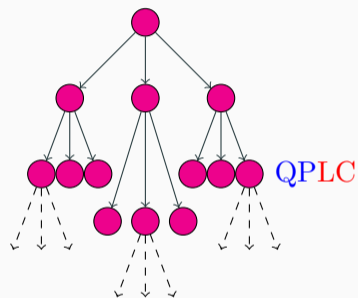
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Maximization on a tree

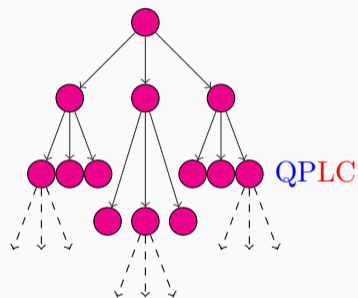


QPCC then reduces to a **maximization on a tree**, where each **node**...

- represents a case in the **perpendicularity**, and
- is a **QPLC**

This is **solvable** with the **fast pruning algorithm** of Gurobi.

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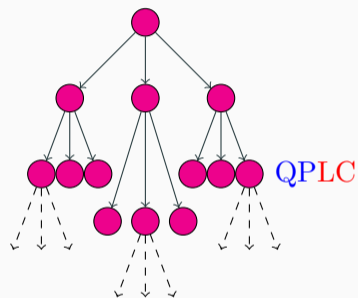


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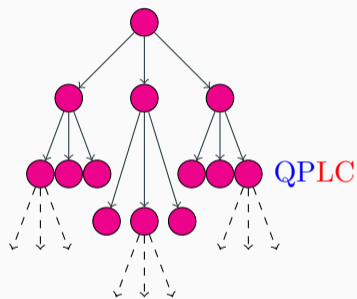


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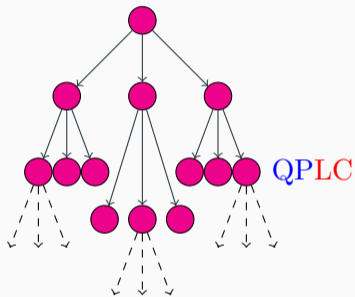


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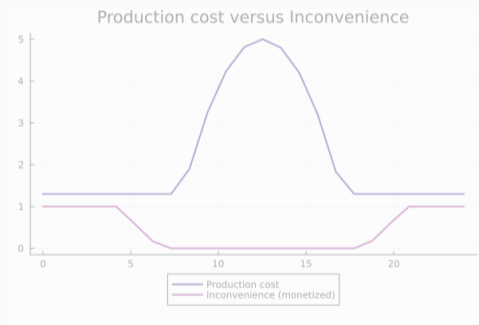
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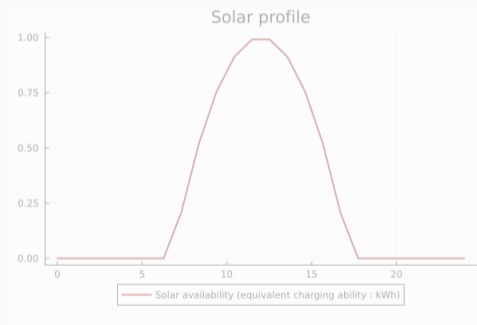
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We implement our model on a `tree` using Gurobi through the JuMP package of Julia.



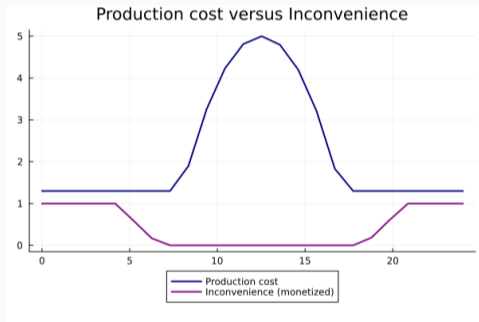
L: Cost and monetized inconvenience.



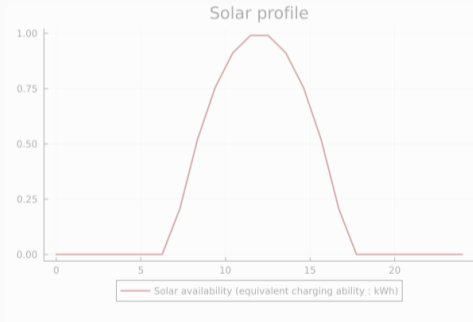
R: Solar availability.

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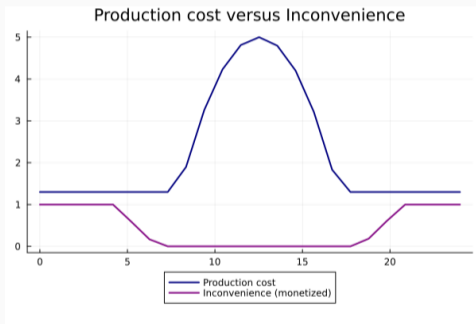
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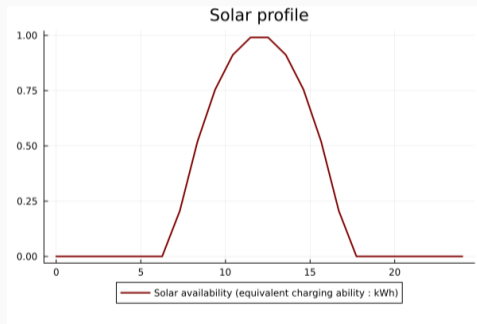
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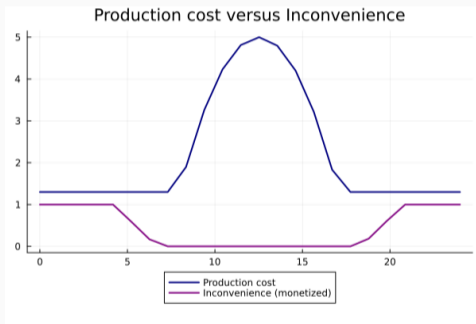
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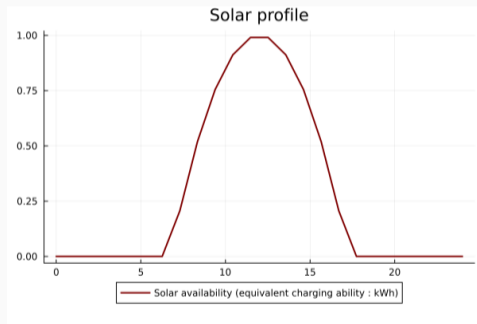
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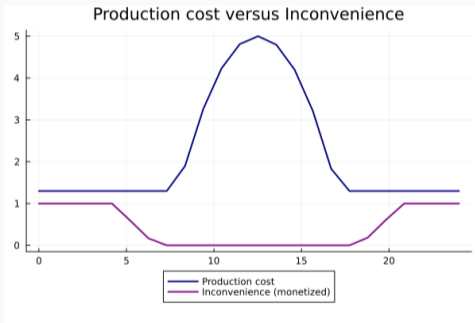
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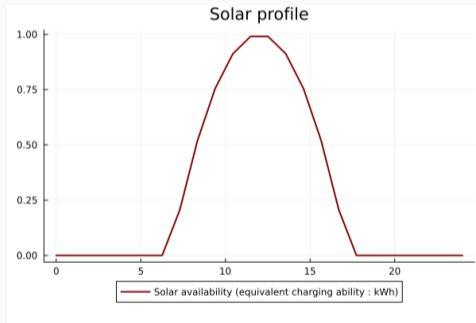
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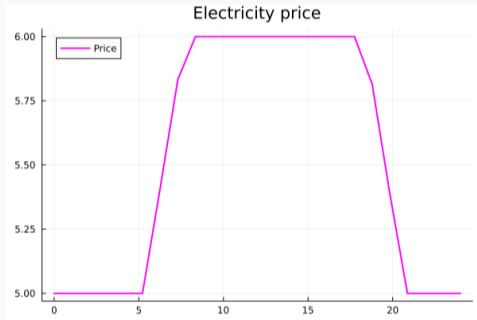


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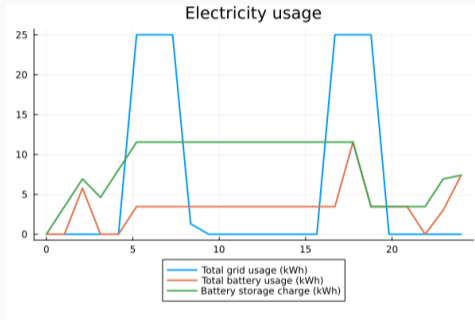


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Simulation results

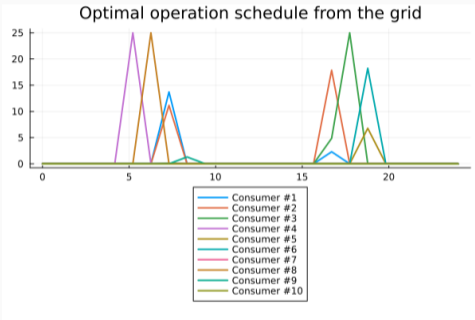


L: Optimal policy.

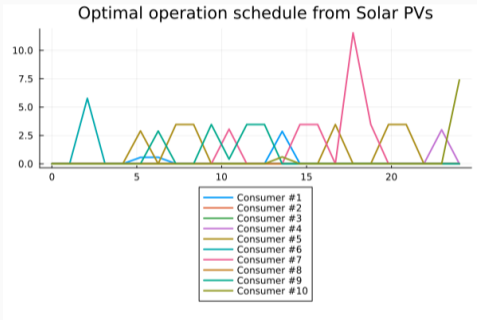


R: Optimal overall consumption.

Simulation results



L: Optimal grid consumption.



R: Optimal PV consumption.

Some remarks

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- Stochasticity is not considered in the simulations, **but this can be easily implemented under chance constraints framework and $\mathcal{N}(\mu, \sigma^2)$.**
- The price upper bound \bar{p} cannot be left off due to our **demand constraints**. **This is economical because the provider can raise the price as high as she wishes.**
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Thank you for being here. ;)