

Modeling and Solving Battery Operation Problems

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Talk outline

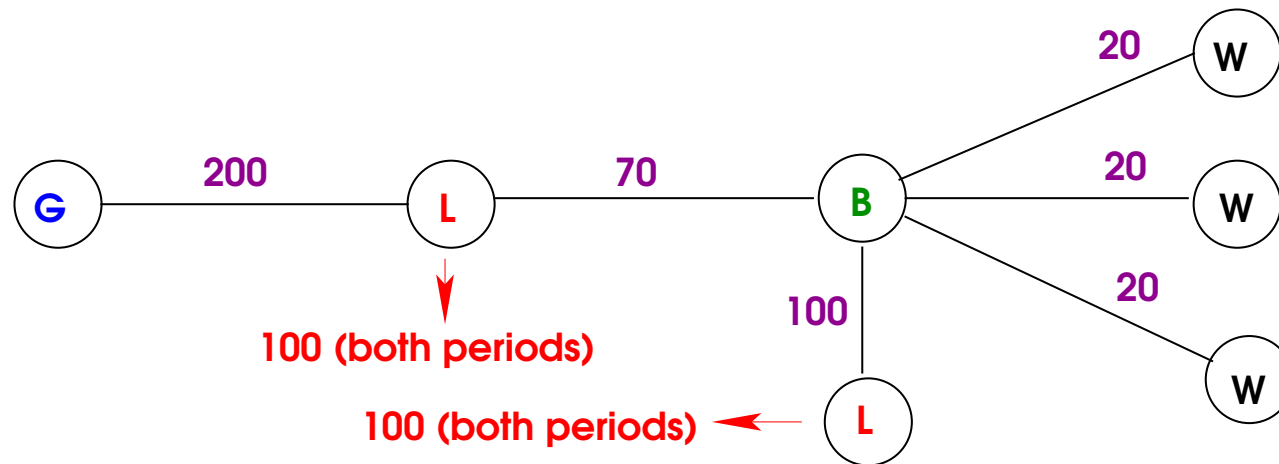
- Optimization framework for using storage to offset uncertainty in renewable forecast output
- “OPF-like” setup: minimize cost of generation
- Multiperiod model, with per-period renewable forecasts
- affine control for renewable output
- robust optimization used to handle forecast errors

Questions/issues

- Batteries not quite here yet?
- Transmission level or distribution level only?
- Can batteries be moved around? In what time frame?
- What is the cost of moving/installing batteries and how should that be factored into operations?
- What is the correct time frame for analyzing the benefit of using batteries?

A simple example on 2 periods

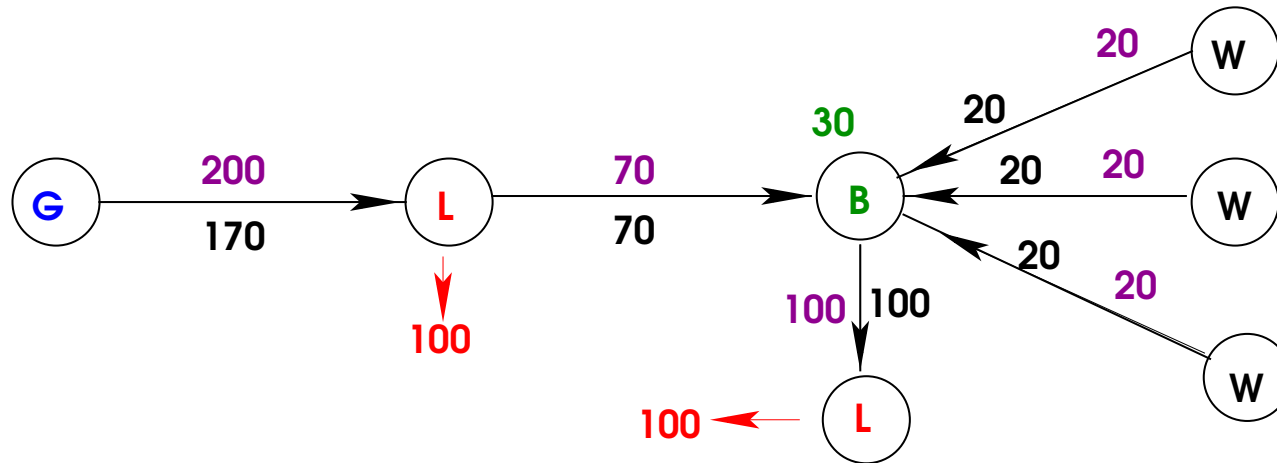
G = generator, L = load, B = battery, W = renewable



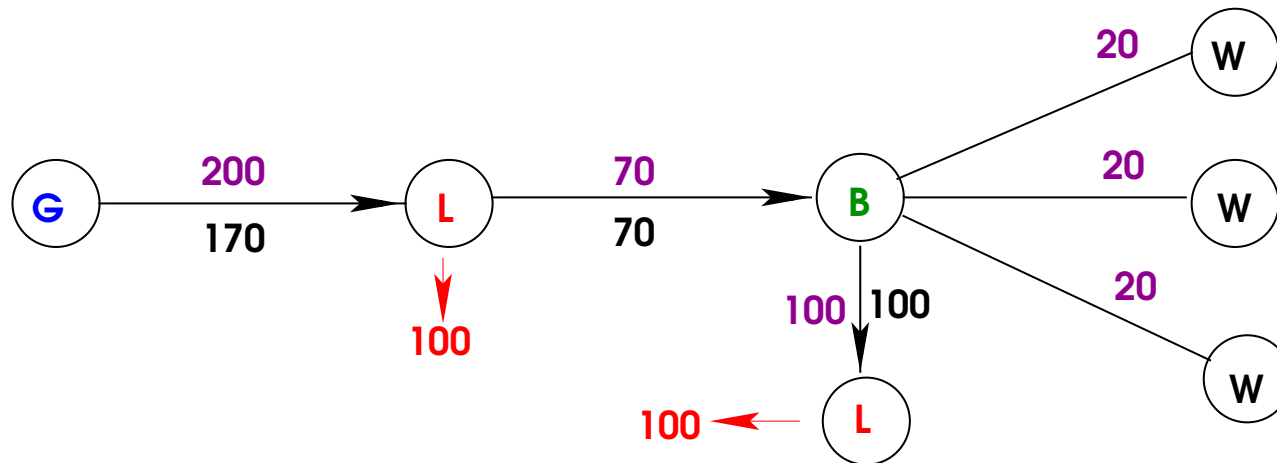
- **Period 1:** each renewable outputs **20**, no uncertainty.
- **Period 2:** each renewable outputs in the range **[0, 20]**.
- Battery and generator are large, but battery starts **drained**.

- **Period 1:** each renewable outputs **20**, no uncertainty.
- **Period 2:** each renewable outputs in the range **[0, 20]**.

Period 1:



Period 2:



Notation

1. T time periods of equal length Δ . Assume $\Delta = 1$.
2. $P_k^{g,t}$ = output of generator at bus k at time t (decision variable)
3. $\bar{w}_i^t + w_i^t$ = output of renewable at bus i at time t .
 \bar{w}_i^t = forecast, w_i^t = error (uncertain).
4. δ_j^t = output of battery at bus j at time t .
Assumption: all batteries at a given bus j are of similar type.
5. $P_k^{d,t}$ = load at bus k at time t (data).
6. **DC** power flow \rightarrow for all t , and all w ,

$$\sum_k P_k^{g,t} + \sum_i (\bar{w}_i^t + w_i^t) + \sum_j \delta_j^t = \sum_k P_k^{d,t}$$

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4. δ_j^t = output of battery at bus j at time t .
5. $P_k^{d,t}$ = load at bus k at time t (data).
6. For all t , and all w , $\sum_k P_k^{g,t} + \sum_i (\bar{w}_i^t + w_i^t) + \sum_j \delta_j^t = \sum_k P_k^{d,t}$

Generic affine control:

$$\delta_j^t = \bar{\lambda}_j^t - \lambda_j^t \sum_{i \in R(j)} w_i$$

- $\bar{\lambda}^t, \lambda^t$: decision variables
- $R(j)$: set of buses that battery at j responds to

Example 1: (bad?) a battery at each renewable, $R(j) = j$ for all j

Example 2: for all j , $R(j) =$ all renewables

Nominal case:

$$\sum_k P_k^{g,t} + \sum_i \bar{w}_i^t + \sum_j \bar{\lambda}_j^t = \sum_k P_k^{d,t}$$

Balance: $\sum_k P_k^{g,t} + \sum_i (\bar{w}_i^t + \mathbf{w}_i^t) + \sum_j \delta_j^t = \sum_k P_k^{d,t} \quad \forall \mathbf{w}$

Nominal case: $\sum_k P_k^{g,t} + \sum_i \bar{w}_i^t + \sum_j \bar{\lambda}_j^t = \sum_k P_k^{d,t}$

Generic affine control: $\delta_j^t = \bar{\lambda}_j^t - \lambda_j^t \left(\sum_{i \in R(j)} \mathbf{w}_i \right)$

This talk: $R(j) =$ all renewables, for each j .

Affine control: $\delta_j^t = \bar{\lambda}_j^t - \lambda_j^t (\sum_i \mathbf{w}_i)$

Balance: $\sum_k P_k^{g,t} + \sum_i \bar{w}_i^t + \sum_j \bar{\lambda}_j^t + (\sum_i \mathbf{w}_i) \left(1 - \sum_j \lambda_j^t \right) = \sum_k P_k^{d,t}$

Together with nominal case, implies: $\sum_j \lambda_j^t = 1$

Other considerations

- What is the **sign** of $\bar{\lambda}_j^t$? Of λ_j^t ?
- Require that they have the same sign? That they **all** have the same sign, for any given t ?
- Force $\lambda_j^t \geq 0$?
- Restrict the **number** of nonzero λ_j^t , for any t ?

Battery model

- Discharge rate bounds. We will want to lower- and upper-bound

$$\delta_j^t = \bar{\lambda}_j^t - \lambda_j^t \left(\sum_{i \in R(j)} w_i \right)$$

for all batteries i , time t and all w

- Energy state bounds. Let E_j^0 = initial energy state of battery at site i .
Then

$$E_j^t = E_j^0 + \Delta \sum_{h=1}^t \delta_j^h$$

is the energy state at the end of period t . (Δ = length of time periods)

Must have lower- and upper-bounds on E_j^t .

- Special bounds for E_j^T ?

Renewable forecast error model

Output of renewable i at time t : $\bar{w}_i^t + w_i^t$

- \bar{w}_i^t = forecast output
- w_i^t = error; $w \in \mathcal{W}$, where \mathcal{W} = uncertainty model.

Linear error model: $\mathcal{W} = \{w : Cw \leq d\}$

Example: concentration model

$$\begin{aligned} |w_i^t| &\leq \gamma_i^t, \quad \text{all } t \text{ and } i \\ \sum_i \alpha_i^t |w_i^t| &\leq \Gamma^t \quad \text{all } t \end{aligned}$$

Here, the γ_i^t , α_i^t and Γ^t are parameters. **Special case:** $\alpha_i^t = 1/\gamma_i^t$

Many variations, e.g. time-correlated models

$$\begin{aligned} \sum_t \alpha_i^t |w_i^t| &\leq \Gamma_i \\ \sum_i \sum_t \alpha_i^t |w_i^t| &\leq \Gamma \end{aligned}$$

Extension: histogram models.

Formulation

Optimization problem: minimize generation cost subject to being feasible under all modeled renewable outputs. **Variables:** $P^g, \bar{\lambda}, \lambda$

$$\min_{P^g, \bar{\lambda}, \lambda} \sum_t \sum_k c_k^t(P_k^{g,t})$$

s.t. the following constraints being feasible at all times t , for all $\mathbf{w} \in \mathcal{W}$:

$$B \boldsymbol{\theta}^t = P^{g,t} + \underbrace{\bar{\mathbf{w}}^t + \mathbf{w}^t}_{\text{renewables}} + \overbrace{\bar{\lambda}^t - \left(\sum_i \mathbf{w}_i^t \right) \lambda^t}_{\text{batteries}} - P^{d,t}$$

$$\frac{|\boldsymbol{\theta}_k^t - \boldsymbol{\theta}_m^t|}{x_{km}} \leq u_{km} \quad \text{for all } km \text{ (line limits at time } t)$$

$$r_j^{\min} \leq \bar{\lambda}_j^t - \lambda_j^t \left(\sum_i \mathbf{w}_i^t \right) \leq r_j^{\max} \quad \text{for all } j \text{ (discharge rate limits)}$$

$$0 \leq E_j^0 + \Delta \sum_{h=1}^t \left[\bar{\lambda}_j^h - \lambda_j^h \left(\sum_i \mathbf{w}_i^h \right) \right] \leq E_j^{\max} \quad \text{all } j \text{ (energy state)}$$

$$\min_{P^g, \bar{\lambda}, \lambda} \sum_t \sum_k c_k^t(P_k^{g,t})$$

s.t. the following constraints being feasible at all times t , for all $\mathbf{w} \in \mathcal{W}$:

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$$\frac{|\boldsymbol{\theta}_k^t - \boldsymbol{\theta}_m^t|}{x_{km}} \leq u_{km} \quad \text{for all } km \text{ (line limits at time } t)$$

$$r_j^{\min} \leq \bar{\lambda}_j^t - \lambda_j^t \left(\sum_i \mathbf{w}_i^t \right) \leq r_j^{\max} \quad \text{for all } j \text{ (discharge rate limits)}$$

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Is there a “compact” formulation?

- **What** is a compact formulation?
- For linear error model: $\mathcal{W} = \{w : Cw \leq d\}$, the answer is **yes**
- But it is going to large, and expensive.
- Factoid: modifications of DC model for large grids are dangerous hard for LP solvers.

How to build a cutting-plane procedure

Consider the balance equations at a given time t (removed from notation)

$$B \boldsymbol{\theta} = P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d$$

we can set $\boldsymbol{\theta}_\rho = \mathbf{0}$ where ρ = reference bus; flow balance can be rewritten

$$\boldsymbol{\theta} = V \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right]$$

here V is a **pseudo-inverse** of B

So for any line km ,

$$\boldsymbol{\theta}_k - \boldsymbol{\theta}_m = (v_k - v_m) \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right]$$

(v_h = row h of V)

How to build a cutting-plane procedure, 2

$$\theta_k - \theta_m = (v_k - v_m) \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right]$$

Suppose that $(P^{g*}, \bar{\lambda}^*, \lambda^*)$ is a **proposed** solution that is **infeasible**

i.e. there is a line km and an **error vector** $\hat{\mathbf{w}}$ such that

$$\frac{(v_k - v_m)}{x_{km}} \left[P^{g*} + \bar{w} + \hat{\mathbf{w}} + \bar{\lambda}^* - \lambda^* \left(\sum_i \hat{\mathbf{w}}_i \right) - P^d \right] > u_{km}$$

Then the inequality

$$\frac{(v_k - v_m)}{x_{km}} \left[P^g + \bar{w} + \hat{\mathbf{w}} + \bar{\lambda} - \lambda \left(\sum_i \hat{\mathbf{w}}_i \right) - P^d \right] \leq u_{km}$$

is **valid** and **cuts-off** $(P^{g*}, \bar{\lambda}^*, \lambda^*)$

A cutting-plane procedure

Start with a **relaxation** for the robust problem, e.g. the nominal problem (no errors), and then

1. Solve relaxation, with solution $(P^{g^*}, \bar{\lambda}^*, \lambda^*)$
2. **Play adversary:** find a worst-case distribution \hat{w} for $(P^{g^*}, \bar{\lambda}^*, \lambda^*)$

Comment: Requires solving small LPs

3. Say, e.g. for a given line km ,

$$\frac{(v_k - v_m)}{x_{km}} \left[P^{g^*} + \bar{w} + \hat{w} + \bar{\lambda}^* - \lambda^* \left(\sum_i \hat{w}_i \right) - P^d \right] > u_{km}$$

Comment: could be the reverse flow

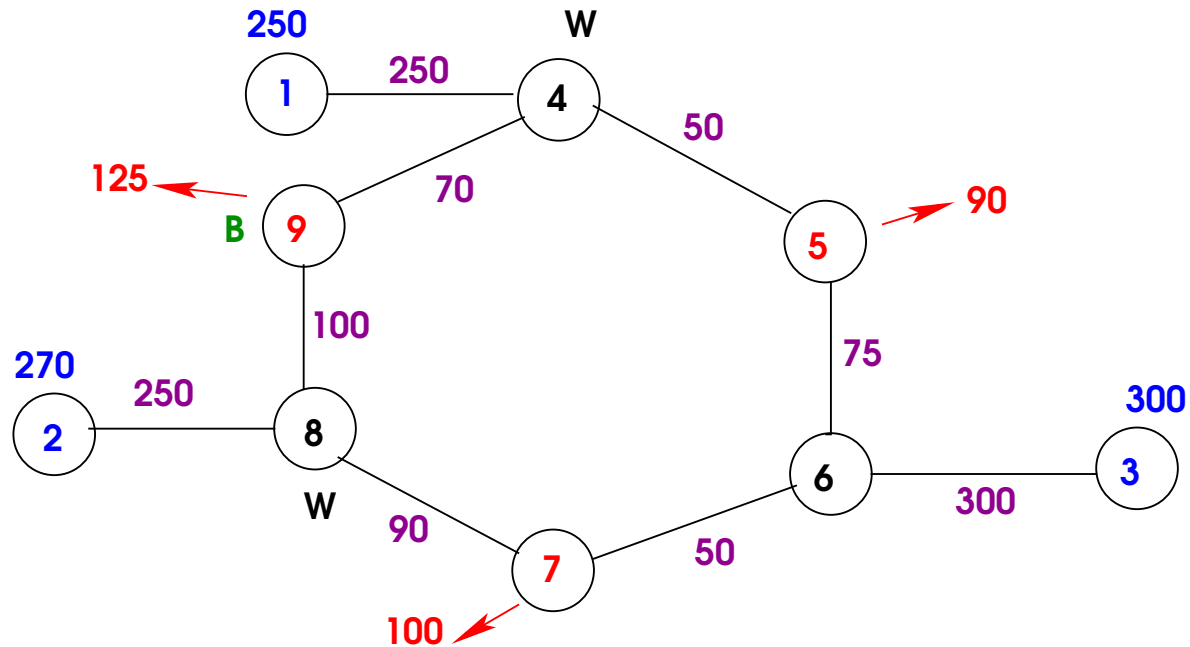
4. Then we add, to the relaxation, the cut

$$\frac{(v_k - v_m)}{x_{km}} \left[P^g + \bar{w} + \hat{w} + \bar{\lambda} - \lambda \left(\sum_i \hat{w}_i \right) - P^d \right] \leq u_{km}$$

and continue (goto **1**).

5. Else if the adversary fails, $(P^{g^*}, \bar{\lambda}^*, \lambda^*)$ is **optimal**

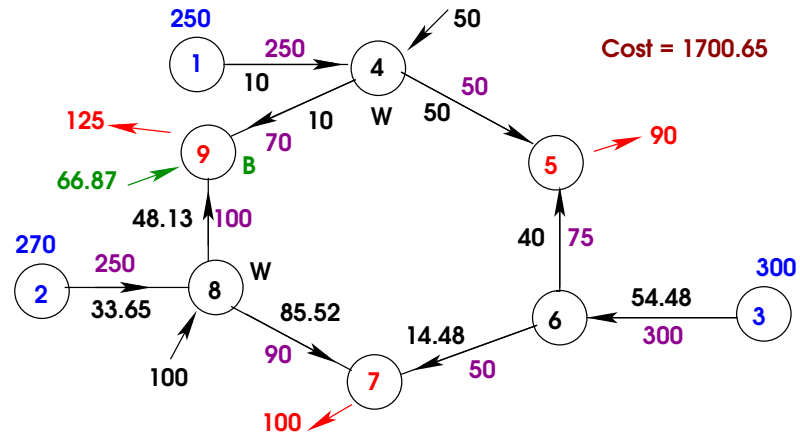
Modified Case9 example



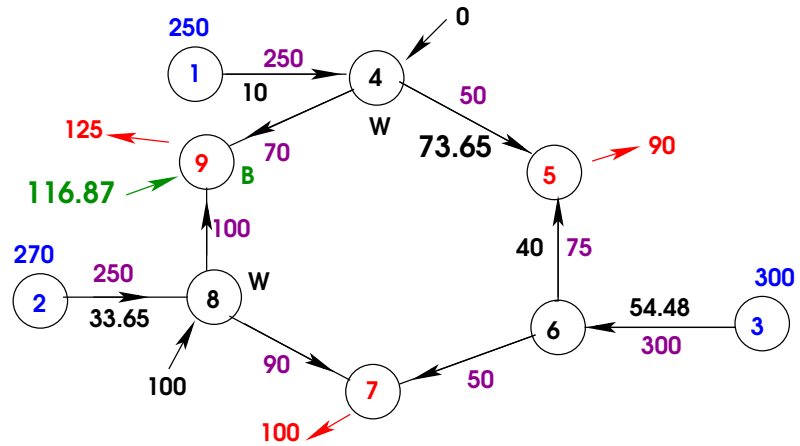
Renewables model:

- Output at 4: $50 + w_4$
- Output at 8: $100 + w_8$
- $-50 \leq w_4 \leq 0$, $-100 \leq w_8 \leq 0$,
- $2|w_4| + |w_8| \leq 100$

Nominal case

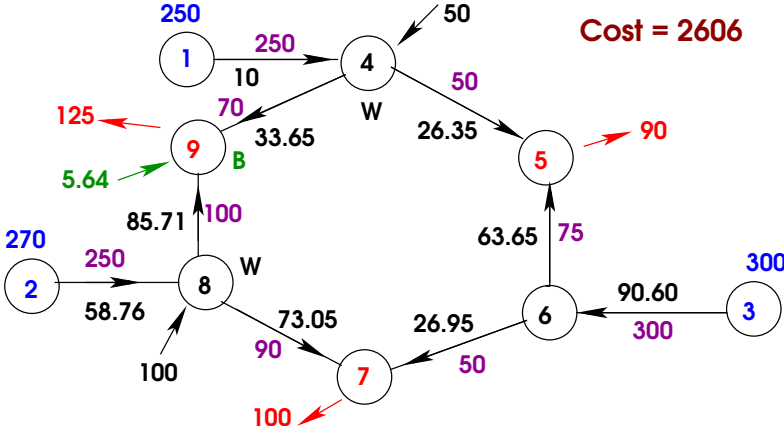


Worst case

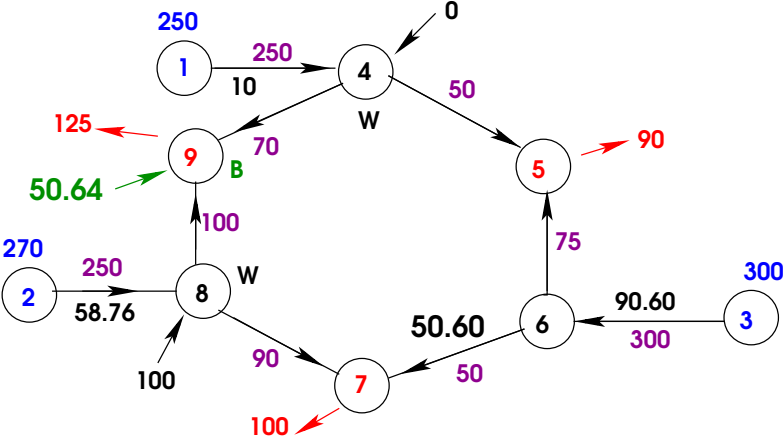


$$\text{cut: } -0.36P_2^g - 0.62P_3^g - 0.125\bar{\lambda}_9 - 12.5\lambda_9 \leq -90.2$$

Updated nominal case



Worst case



A compact formulation (abridged)

A solution $(P^g, \bar{\lambda}, \lambda)$ is **safe** for a line km if:

$$(v_k - v_m) \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right] \leq x_{km} u_{km}$$

for all $\mathbf{w} \in \mathcal{W}$. (Dropped superscript t , also should consider line mk).

This is the same as

$$x_{km} u_{km} \geq \max_{\mathbf{w} \in \mathcal{W}} (v_k - v_m) \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right]$$

In the linear model $\mathcal{W} = \{w : Cw \leq d\}$ this is the same as

$$\begin{aligned} x_{km} u_{km} &\geq \max (v_k - v_m) \left[P^g + \bar{w} + w + \bar{\lambda} - \lambda \left(\sum_i w_i \right) - P^d \right] \\ &\text{s.t. } Cw \leq d \end{aligned}$$

or,

$$\begin{aligned} x_{km} u_{km} &\geq (v_k - v_m) [P^g + \bar{w} + \bar{\lambda} - P^d] + \\ &\max \sum_i [(v_{ki} - v_{mi}) - (v_k - v_m)\lambda] w_i \\ &\text{s.t. } Cw \leq d \end{aligned}$$

A compact formulation (abridged)

A solution $(P^g, \bar{\lambda}, \lambda)$ is **safe** for a line km if:

$$(v_k - v_m) \left[P^g + \bar{w} + \mathbf{w} + \bar{\lambda} - \lambda \left(\sum_i \mathbf{w}_i \right) - P^d \right] \leq x_{km} u_{km}$$

for all $\mathbf{w} \in \mathcal{W}$. (Dropped superscript t). Or,

$$\begin{aligned} x_{km} u_{km} &\geq (v_k - v_m) [P^g + \bar{w} + \bar{\lambda} - P^d] + \\ &\quad \max \sum_i [(v_{ki} - v_{mi}) - (v_k - v_m)\lambda] w_i \\ \text{s.t. } &C\mathbf{w} \leq d \end{aligned}$$

which holds **if and only if** the following system is feasible

$$\pi \geq 0 \tag{1a}$$

$$[\pi^T C]_i = (v_{ki} - v_{mi}) - (v_k - v_m)\lambda \quad \text{for all } i \tag{1b}$$

$$\pi^T d \leq x_{km} u_{km} - (v_k - v_m) (P^g - P^d + \bar{\lambda} + \bar{w}) \tag{1c}$$

In system (1), the variables $\boldsymbol{\pi}$ should be indexed by km and t .

The system must be added to the nominal formulation for each km and t .
Total number of added variables: (**# of lines**) \times (**# of renewables**)