

Multi-Period Optimal Power Flow: Temporal ADMM (tADMM) Formulation

Decomposition with T Single-Step Blocks

1 Problem Overview

The Temporal ADMM (tADMM) algorithm decomposes the multi-period optimal power flow problem into T single-step subproblems, each corresponding to one time period. This approach enables parallel computation and improved scalability.

1.1 Variable Color Coding

Following the handwritten PDF formulation:

- \mathbf{B}_{t_0} : Local SOC variables for subproblem t_0 (blue)
- $\hat{\mathbf{B}}$: Global consensus SOC trajectory (red)
- \mathbf{u}_{t_0} : Local scaled dual variables for subproblem t_0 (green)

2 tADMM Algorithm Structure

The algorithm alternates between three update steps:

2.1 Step 1: Primal Update (Blue Variables)

For each subproblem $t_0 \in \{1, 2, \dots, T\}$:

$$\min_{P_{\text{subs}, t_0}, P_{B, t_0}, \mathbf{B}_{t_0}} C_{t_0} \cdot P_{\text{subs}, t_0} + \frac{\rho}{2} \left\| \mathbf{B}_{t_0} - \hat{\mathbf{B}} + \mathbf{u}_{t_0} \right\|_2^2 \quad (1)$$

$$\text{s.t. } \mathbf{B}_{t_0}[\mathbf{t}_0] - B_0 + P_{B, t_0} \cdot \Delta t = 0 \quad \text{if } t_0 = 1 \quad (2)$$

$$\mathbf{B}_{t_0}[\mathbf{t}_0] - \hat{\mathbf{B}}[\mathbf{t}_0 - \mathbf{1}] + P_{B, t_0} \cdot \Delta t = 0 \quad \text{if } t_0 > 1 \quad (3)$$

$$P_{\text{subs}, t_0} + P_{B, t_0} - P_L[t_0] = 0 \quad (4)$$

$$-P_{B, R} - P_{B, t_0} \leq 0 \quad (5)$$

$$P_{B, t_0} - P_{B, R} \leq 0 \quad (6)$$

$$\text{SOC}_{\min} \cdot E_{\text{Rated}} - \mathbf{B}_{t_0}[\mathbf{t}] \leq 0, \quad \forall t \in \{1, \dots, T\} \quad (7)$$

$$\mathbf{B}_{t_0}[\mathbf{t}] - \text{SOC}_{\max} \cdot E_{\text{Rated}} \leq 0, \quad \forall t \in \{1, \dots, T\} \quad (8)$$

2.2 Step 2: Consensus Update (Red Variables)

$$\hat{\mathbf{B}}[\mathbf{t}] = \text{clamp} \left(\frac{1}{T} \sum_{t_0=1}^T (\mathbf{B}_{t_0}[\mathbf{t}] + \mathbf{u}_{t_0}[\mathbf{t}]), \text{SOC}_{\min} \cdot E_{\text{Rated}}, \text{SOC}_{\max} \cdot E_{\text{Rated}} \right) \quad (9)$$

$$\forall t \in \{1, 2, \dots, T-1\} \quad (10)$$

$$\hat{\mathbf{B}}[\mathbf{T}] = B_{T, \text{target}} \quad (\text{if terminal constraint exists}) \quad (11)$$

2.3 Step 3: Dual Update (Green Variables)

$$\mathbf{u}_{t_0}[\mathbf{t}] := \mathbf{u}_{t_0}[\mathbf{t}] + (\mathbf{B}_{t_0}[\mathbf{t}] - \hat{\mathbf{B}}[\mathbf{t}]) \quad (12)$$

$$\forall t_0 \in \{1, \dots, T\}, \forall t \in \{1, \dots, T\} \quad (13)$$

3 Convergence Criteria

The algorithm terminates when both residuals fall below specified thresholds:

3.1 Primal Residual (Consensus Violation)

$$\|r^k\|_2 = \left\| \text{vec} \left(\left\{ \mathbf{B}_{t_0} - \hat{\mathbf{B}} \right\}_{t_0=1}^T \right) \right\|_2 \leq \epsilon_{\text{pri}} \quad (14)$$

3.2 Dual Residual (Consensus Change)

$$\|s^k\|_2 = \rho \left\| \hat{\mathbf{B}}^k - \hat{\mathbf{B}}^{k-1} \right\|_2 \leq \epsilon_{\text{dual}} \quad (15)$$

4 Final Solution Recovery

After convergence, the final power schedules are reconstructed from the consensus trajectory:

$$P_B^*[t] = \begin{cases} \frac{-(\hat{\mathbf{B}}[\mathbf{t}] - B_0)}{\Delta t} & \text{if } t = 1 \\ \frac{-(\hat{\mathbf{B}}[\mathbf{t}] - \hat{\mathbf{B}}[\mathbf{t}-1])}{\Delta t} & \text{if } t > 1 \end{cases} \quad (16)$$

$$P_{\text{Subs}}^*[t] = P_L[t] - P_B^*[t] \quad (17)$$

5 Algorithm Parameters

- **Penalty Parameter:** ρ (typically 1.0 to 10.0)
- **Primal Tolerance:** $\epsilon_{\text{pri}} = 10^{-3}$
- **Dual Tolerance:** $\epsilon_{\text{dual}} = 10^{-3}$
- **Maximum Iterations:** 1000