

# Multi-Period Optimal Power Flow: LinDistFlow & Copper Plate tADMM Formulations

## 1 LinDistFlow MPOPF with tADMM

### 1.1 Problem Overview

The Temporal ADMM (tADMM) algorithm decomposes the multi-period optimal power flow problem for distribution networks into  $T$  subproblems, each corresponding to one time period. This formulation uses the linearized DistFlow model to capture network physics including voltage drops and reactive power flows.

### 1.2 Variable Color Coding

- $\mathbf{B}_j^{t_0}[\mathbf{t}]$ : Local SOC variables for battery  $j$  in subproblem  $t_0$ , evaluated at time  $t$  (blue)
- $\hat{\mathbf{B}}_j[\mathbf{t}]$ : Global consensus SOC for battery  $j$  at time  $t$  (red)
- $\mathbf{u}_j^{t_0}[\mathbf{t}]$ : Local scaled dual variables for battery  $j$  in subproblem  $t_0$ , for time  $t$  (green)

### 1.3 Sets and Indices

- $\mathcal{N}$ : Set of all nodes (buses)
- $\mathcal{L}$ : Set of all branches (lines)
- $\mathcal{L}_1$ : Set of branches connected to substation (node 1)
- $\mathcal{B}$ : Set of nodes with batteries
- $\mathcal{D}$ : Set of nodes with PV (DER)
- $\mathcal{T} = \{1, 2, \dots, T\}$ : Set of time periods
- $t_0 \in \mathcal{T}$ : Index for a specific time period in tADMM decomposition
- $j \in \mathcal{N}$ : Node index
- $(i, j) \in \mathcal{L}$ : Branch from node  $i$  to node  $j$

### 1.4 tADMM Algorithm Structure

The algorithm alternates between three update steps:

## 1.5 Step 1: Subproblem Update (Blue Variables)

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

$$\begin{aligned} \min_{\substack{P_{\text{Subs}}^{t_0}, Q_{\text{Subs}}^{t_0}, \\ P_{ij}^{t_0}, Q_{ij}^{t_0}, v_j^{t_0}, q_{D,j}^{t_0}, \\ P_{B,j}^t, \mathbf{B}_j^{t_0}[\mathbf{t}] \\ \forall j \in \mathcal{B}, t \in \mathcal{T}}} \quad & c^{t_0} \cdot P_{\text{Subs}}^{t_0} \cdot P_{\text{BASE}} \cdot \Delta t + C_B \sum_{j \in \mathcal{B}} \left( P_{B,j}^{t_0} \right)^2 \cdot P_{\text{BASE}}^2 \cdot \Delta t \\ & + \frac{\rho}{2} \sum_{j \in \mathcal{B}} \sum_{t=1}^T \left( \mathbf{B}_j^{t_0}[\mathbf{t}] - \hat{\mathbf{B}}_j[\mathbf{t}] + \mathbf{u}_j^{t_0}[\mathbf{t}] \right)^2 \end{aligned} \quad (1)$$

**Subject to:**

**Spatial Network Constraints (only for time  $t_0$ ):**

$$\text{Real power balance (substation): } P_{\text{Subs}}^{t_0} - \sum_{(1,j) \in \mathcal{L}_1} P_{1j}^{t_0} = 0 \quad (2)$$

$$\begin{aligned} \text{Real power balance (nodes): } P_{ij}^{t_0} - \sum_{(j,k) \in \mathcal{L}} P_{jk}^{t_0} &= P_{B,j}^{t_0} + p_{D,j}^{t_0} - p_{L,j}^{t_0}, \\ \forall (i,j) \in \mathcal{L}, \end{aligned} \quad (3)$$

$$\text{Reactive power balance (substation): } Q_{\text{Subs}}^{t_0} - \sum_{(1,j) \in \mathcal{L}_1} Q_{1j}^{t_0} = 0 \quad (4)$$

$$\begin{aligned} \text{Reactive power balance (nodes): } Q_{ij}^{t_0} - \sum_{(j,k) \in \mathcal{L}} Q_{jk}^{t_0} &= q_{D,j}^{t_0} - q_{L,j}^{t_0}, \\ \forall (i,j) \in \mathcal{L}, \end{aligned} \quad (5)$$

$$\text{KVL constraints: } v_i^{t_0} - v_j^{t_0} = 2(r_{ij}P_{ij}^{t_0} + x_{ij}Q_{ij}^{t_0}), \quad \forall (i,j) \in \mathcal{L} \quad (6)$$

$$\text{Voltage limits: } (V_{\min,j})^2 \leq v_j^{t_0} \leq (V_{\max,j})^2, \quad \forall j \in \mathcal{N} \quad (7)$$

$$\begin{aligned} \text{PV reactive limits: } -\sqrt{(S_{D,j})^2 - (p_{D,j}^{t_0})^2} &\leq q_{D,j}^{t_0} \leq \sqrt{(S_{D,j})^2 - (p_{D,j}^{t_0})^2}, \\ \forall j \in \mathcal{D} \end{aligned} \quad (8)$$

**Temporal Battery Constraints (entire horizon  $t \in \{1, \dots, T\}$ ):**

$$\text{Initial SOC: } \mathbf{B}_j^{t_0}[\mathbf{1}] = B_{0,j} - P_{B,j}^1 \cdot \Delta t, \quad \forall j \in \mathcal{B} \quad (9)$$

$$\text{SOC trajectory: } \mathbf{B}_j^{t_0}[\mathbf{t}] = \mathbf{B}_j^{t_0}[\mathbf{t} - \mathbf{1}] - P_{B,j}^t \cdot \Delta t, \quad \forall t \in \{2, \dots, T\}, j \in \mathcal{B} \quad (10)$$

$$\begin{aligned} \text{SOC limits: } \text{SOC}_{\min,j} \cdot B_{\text{rated},j} &\leq \mathbf{B}_j^{t_0}[\mathbf{t}] \leq \text{SOC}_{\max,j} \cdot B_{\text{rated},j}, \\ \forall t \in \mathcal{T}, j \in \mathcal{B} \end{aligned} \quad (11)$$

$$\text{Power limits: } -P_{B,\text{rated},j} \leq P_{B,j}^t \leq P_{B,\text{rated},j}, \quad \forall t \in \mathcal{T}, j \in \mathcal{B} \quad (12)$$

**Key Formulation Notes:**

- **Network variables** ( $P_{\text{Subs}}^{t_0}, Q_{\text{Subs}}^{t_0}, P_{ij}^{t_0}, Q_{ij}^{t_0}, v_j^{t_0}, q_{D,j}^{t_0}$ ) are optimized *only* for time step  $t_0$
- **Battery power**  $P_{B,j}^t$  is optimized for the *entire* horizon  $t \in \{1, \dots, T\}$
- **Local SOC trajectory**  $\mathbf{B}_j^{t_0}[\mathbf{t}]$  (blue) is computed for *all* time steps  $t \in \{1, \dots, T\}$

- The ADMM consensus penalty compares the full local trajectory  $\mathbf{B}_j^{t_0}[\mathbf{t}]$  with the global master copy  $\hat{\mathbf{B}}_j[\mathbf{t}]$  (red)
- Each battery  $j \in \mathcal{B}$  has its own local/global SOC variables and dual variables

### 1.6 Step 2: Consensus Update (Red Variables)

For each battery  $j \in \mathcal{B}$  and each time period  $t \in \mathcal{T}$ :

$$\hat{\mathbf{B}}_j[\mathbf{t}] = \text{clamp} \left( \frac{1}{T} \sum_{t_0=1}^T \left( \mathbf{B}_j^{t_0}[\mathbf{t}] + \mathbf{u}_j^{t_0}[\mathbf{t}] \right), \underline{B}_j, \overline{B}_j \right) \quad (13)$$

where  $\underline{B}_j = \text{SOC}_{\min,j} \cdot B_{\text{rated},j}$  and  $\overline{B}_j = \text{SOC}_{\max,j} \cdot B_{\text{rated},j}$ .

### 1.7 Step 3: Dual Update (Green Variables)

For each battery  $j \in \mathcal{B}$ , each subproblem  $t_0 \in \mathcal{T}$ , and each time period  $t \in \mathcal{T}$ :

$$\mathbf{u}_j^{t_0}[\mathbf{t}] := \mathbf{u}_j^{t_0}[\mathbf{t}] + \left( \mathbf{B}_j^{t_0}[\mathbf{t}] - \hat{\mathbf{B}}_j[\mathbf{t}] \right) \quad (14)$$

### 1.8 Convergence Criteria

**Primal Residual (Consensus Violation):**

$$\|r^k\|_2 = \sqrt{\sum_{j \in \mathcal{B}} \sum_{t=1}^T \left( \frac{1}{T} \sum_{t_0=1}^T \mathbf{B}_j^{t_0}[\mathbf{t}] - \hat{\mathbf{B}}_j[\mathbf{t}] \right)^2} \leq \epsilon_{\text{pri}} \quad (15)$$

**Dual Residual (Consensus Change):**

$$\|s^k\|_2 = \rho \sqrt{\sum_{j \in \mathcal{B}} \sum_{t=1}^T \left( \hat{\mathbf{B}}_j^k[\mathbf{t}] - \hat{\mathbf{B}}_j^{k-1}[\mathbf{t}] \right)^2} \leq \epsilon_{\text{dual}} \quad (16)$$

## 2 Copper Plate MPOPF with tADMM (Simplified Case)

### 2.1 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. The hope is to enable parallel computation and improved scalability while still retaining solution optimality.

### 2.2 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)

### 2.3 tADMM Algorithm Structure

The algorithm alternates between three update steps:

### 2.4 Step 1: Primal Update (Blue Variables) tADMM Optimization Model

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

$$\min_{P_{\text{subs}}^{t_0}, P_B^{t_0}, \mathbf{B}^{t_0}} C^{t_0} \cdot P_{\text{subs}}^{t_0} + C_B \cdot (P_B^{t_0})^2 + \frac{\rho}{2} \left\| \mathbf{B}^{t_0} - \hat{\mathbf{B}} + \mathbf{u}^{t_0} \right\|_2^2 \quad (17)$$

Subject to SOC Dynamics for Entire Trajectory:

$$\mathbf{B}^{t_0}[\mathbf{1}] = B_0 - P_B^{t_0} \cdot \Delta t \quad (18)$$

$$\mathbf{B}^{t_0}[\mathbf{t}] = \mathbf{B}^{t_0}[\mathbf{t} - \mathbf{1}] - P_B^{t_0} \cdot \Delta t, \quad \forall t \in \{2, \dots, T\} \quad (19)$$

$$P_{\text{subs}}^{t_0} + P_B^{t_0} = P_L[t_0] \quad (20)$$

$$-P_{B,R} \leq P_B^{t_0} \leq P_{B,R} \quad (21)$$

$$\underline{B} \leq \mathbf{B}^{t_0}[\mathbf{t}] \leq \overline{B}, \quad \forall t \in \{1, \dots, T\} \quad (22)$$

**Key Formulation Notes:**

- Each subproblem  $t_0$  optimizes the battery power  $P_B^{t_0}$  for *only* time step  $t_0$
- However, the SOC trajectory  $\mathbf{B}^{t_0}[\mathbf{t}]$  is computed for *all* time steps  $t \in \{1, \dots, T\}$
- This ensures that the ADMM penalty term can compare the full trajectory  $\mathbf{B}^{t_0}$  with the consensus  $\hat{\mathbf{B}}$
- The power balance constraint is enforced only for the specific time  $t_0$

## 2.5 Step 2: Consensus Update (Red Variables)

$$\hat{\mathbf{B}}[\mathbf{t}] = \text{clamp} \left( \frac{1}{T} \sum_{t_0=1}^T (\mathbf{B}^{t_0}[\mathbf{t}] + \mathbf{u}^{t_0}[\mathbf{t}]), \underline{B}, \overline{B} \right) \quad (23)$$

$$\forall t \in \{1, 2, \dots, T-1\} \quad (24)$$

$$\hat{\mathbf{B}}[\mathbf{T}] = B_{T, \text{target}} \quad (\text{if terminal constraint exists}) \quad (25)$$

## 2.6 Step 3: Dual Update (Green Variables)

$$\mathbf{u}^{t_0}[\mathbf{t}] := \mathbf{u}^{t_0}[\mathbf{t}] + (\mathbf{B}^{t_0}[\mathbf{t}] - \hat{\mathbf{B}}[\mathbf{t}]) \quad (26)$$

$$\forall t_0 \in \{1, \dots, T\}, \forall t \in \{1, \dots, T\} \quad (27)$$

# 3 Convergence Criteria

The algorithm terminates when both residuals fall below specified thresholds:

## 3.1 Primal Residual (Consensus Violation)

$$\|r^k\|_2 = \left\| \text{vec} \left( \left\{ \mathbf{B}^{t_0} - \hat{\mathbf{B}} \right\}_{t_0=1}^T \right) \right\|_2 \leq \epsilon_{\text{pri}} \quad (28)$$

## 3.2 Dual Residual (Consensus Change)

$$\|s^k\|_2 = \rho \left\| \hat{\mathbf{B}}^k - \hat{\mathbf{B}}^{k-1} \right\|_2 \leq \epsilon_{\text{dual}} \quad (29)$$

# 4 Final Solution Recovery

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

$$P_B^t = \begin{cases} \frac{-(\hat{\mathbf{B}}[\mathbf{t}] - B_0)}{\Delta t} & \text{if } t = 1 \\ \frac{-(\hat{\mathbf{B}}[\mathbf{t}] - \hat{\mathbf{B}}[\mathbf{t}-1])}{\Delta t} & \text{if } t > 1 \end{cases} \quad (30)$$

$$P_{\text{Subs}}^t = P_L[t] - P_B^t \quad (31)$$

The final objective value includes both energy and battery costs:

$$J_{\text{total}} = \sum_{t=1}^T \left[ C^t \cdot P_{\text{Subs}}^t \cdot \Delta t + C_B \cdot (P_B^t)^2 \cdot \Delta t \right] \quad (32)$$

## 5 Algorithm Parameters

### 5.1 Objective Function Components

The tADMM objective function for each subproblem  $t_0$  consists of three terms:

$$\text{Energy Cost: } C^{t_0} \cdot P_{\text{subs}}^{t_0} \cdot \Delta t \quad (33)$$

$$\text{Battery Quadratic Cost: } C_B \cdot (P_B^{t_0})^2 \cdot \Delta t \quad (34)$$

$$\text{ADMM Penalty: } \frac{\rho}{2} \left\| \mathbf{B}^{t_0} - \hat{\mathbf{B}} + \mathbf{u}^{t_0} \right\|_2^2 \quad (35)$$

Where:

- $C^{t_0}$ : Energy price at time  $t_0$  [\$/kWh]
- $C_B$ : Battery quadratic cost coefficient [\$/kW<sup>2</sup>/h] (typically  $10^{-6} \times \min(C^t)$ )
- $\rho$ : ADMM penalty parameter

The battery quadratic cost term  $C_B \cdot (P_B^{t_0})^2$  serves as a regularization to:

1. Prevent excessive battery cycling
2. Encourage smoother power trajectories
3. Improve numerical conditioning of the optimization problem

### 5.2 Algorithmic Parameters

- **Penalty Parameter:**  $\rho$  (typically 0.1 to 10.0)
- **Primal Tolerance:**  $\epsilon_{\text{pri}} = 10^{-3}$
- **Dual Tolerance:**  $\epsilon_{\text{dual}} = 10^{-3}$
- **Maximum Iterations:** 1000

## 6 Appendix: Full Variable and Parameter Definitions

### 6.1 System Bases

$$\text{kV}_B = \frac{4.16}{\sqrt{3}} \text{ kV (phase-to-neutral)} \quad (36)$$

$$\text{kVA}_B = 1000 \text{ kVA} \quad (37)$$

$$P_{\text{BASE}} = 1000 \text{ kW} \quad (38)$$

$$E_{\text{BASE}} = 1000 \text{ kWh per hour} \quad (39)$$

### 6.2 SOC Bound Definitions

$$\underline{B} = \text{SOC}_{\min} \cdot E_{\text{Rated}} \quad (40)$$

$$\overline{B} = \text{SOC}_{\max} \cdot E_{\text{Rated}} \quad (41)$$

### 6.3 Physical Interpretation

- $P_B[t] > 0$ : Battery discharging (providing power to the system)
- $P_B[t] < 0$ : Battery charging (consuming power from the system)
- $B[t]$ : Battery state of charge at the end of period  $t$
- $\underline{B} = \text{SOC}_{\min} \cdot E_{\text{Rated}}$ : Lower SOC bound
- $\overline{B} = \text{SOC}_{\max} \cdot E_{\text{Rated}}$ : Upper SOC bound