

Implementation Notes for the Distributed LP Framework

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1 Introduction

This documents describes some implementation details of the distributed LP framework. This document uses the corresponding *IFAC2017* paper as a reference.

2 Dynamics

$$\mathbf{x}^+ = \mathbf{x} + (\mathbf{B}_{\text{in}} - \mathbf{B}_{\text{out}})\mathbf{u} \quad (1)$$

Also,

$$\mathbf{B} = \mathbf{B}_{\text{in}} - \mathbf{B}_{\text{out}} \in \mathcal{R}^{n_s \times n_u}$$

Where \mathbf{u} is

$$\mathbf{u} = [\mathbf{u}_{s_1}^T \ \mathbf{u}_{s_2}^T \ \dots \ \mathbf{u}_{s_{n_s}}^T]^T \in \mathcal{R}^{n_u}.$$

and \mathbf{u}_{s_i} is

$$\mathbf{u}_{s_i} = [u_{s_i \rightarrow s_1} \ \dots \ u_{s_i \rightarrow s_{i-1}} \ \textcolor{red}{u_{s_i \rightarrow s_i}} \ u_{s_i \rightarrow s_{i+1}} \ \dots \ u_{s_i \rightarrow s_{n_s}}]^T. \quad (2)$$

Notice that Eq. 2 is different from the definition in the paper. That is because $\textcolor{red}{u_{s_i \rightarrow s_i}}$ is added to the \mathbf{u}_{s_i} vector. *However, it is always multiplied by zero.* It is added in order to make the implementation easier later on. Therefore, $n_u = n_s^2$ in this implementation.

Dynamics constraints : dynamics over prediction time horizon T_p ,

$$\mathbf{X} = \mathbf{T}_u \mathbf{U} + \mathbf{T}_{x_0} \mathbf{x}[0] \quad (3)$$

where \mathbf{T}_u is,

$$\mathbf{T}_u = \begin{bmatrix} \mathbf{B} & \dots & \mathbf{0} & \\ \mathbf{B} & \mathbf{B} & \dots & \mathbf{0} \\ \dots & \vdots & \dots & \\ \mathbf{B} & \mathbf{B} & \dots & \mathbf{B} \end{bmatrix}_{(n_s T_p) \times (n_u T_p)}$$

and,

$$\mathbf{T}_{x_0} \mathbf{x}[0] = \begin{bmatrix} \mathbf{x}[0] \\ \vdots \\ \mathbf{x}[0] \end{bmatrix}_{n_s T_p \times 1}$$

Assuming that the optimization vector is,

$$\hat{\mathbf{X}} = \begin{bmatrix} \mathbf{X} \\ \mathbf{U} \end{bmatrix}_{(n_s + n_u) T_p \times 1} \quad (4)$$

Eq 3 can be written as,

$$\begin{bmatrix} \mathbf{I}_{n_s T_p} & -\mathbf{T}_u \end{bmatrix} \hat{\mathbf{X}} = \mathbf{T}_{x_0} \mathbf{x}[0] \quad (5)$$

It can be converted to inequalities as follows,

$$\begin{bmatrix} \mathbf{I}_{n_s T_p} & -\mathbf{T}_u \\ -\mathbf{I}_{n_s T_p} & \mathbf{T}_u \end{bmatrix}_{2n_s T_p \times (n_s + n_u) T_p} \hat{\mathbf{X}} \leq \begin{bmatrix} \mathbf{T}_{x_0} \mathbf{x}[0] \\ -\mathbf{T}_{x_0} \mathbf{x}[0] \end{bmatrix}_{2n_s T_p \times 1} \quad (6)$$

Which can be re-written as

$$\mathbf{A}_{\text{dynamics}} \hat{\mathbf{X}} \leq \mathbf{b}_{\text{dynamics}}$$

3 Flow Constraints

Flow constraints over prediction time horizon are described by,

$$\mathbf{T}_{u,c} \mathbf{U} \leq \mathbf{T}_{x_0,c} \mathbf{x}[0] \quad (7)$$

where,

$$\mathbf{T}_{u,c} = \begin{bmatrix} \mathbf{B}_{\text{out}} & \mathbf{0} & \cdots & \cdots & \mathbf{0} \\ -\mathbf{B} & \mathbf{B}_{\text{out}} & \mathbf{0} & \cdots & \mathbf{0} \\ -\mathbf{B} & -\mathbf{B} & \mathbf{B}_{\text{out}} & \cdots & \mathbf{0} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ -\mathbf{B} & -\mathbf{B} & -\mathbf{B} & -\mathbf{B} & \mathbf{B}_{\text{out}} \end{bmatrix}$$

$\mathbf{T}_{u,c} \in \mathcal{R}^{n_s T_p \times n_u T_p}$, and

$$\mathbf{T}_{x_0,c} \mathbf{x}[0] = \begin{bmatrix} \mathbf{x}[0] \\ \vdots \\ \mathbf{x}[0] \end{bmatrix}_{n_s T_p \times 1}$$

Using the same optimization vector in (4), constraints (7) can be written as,

$$\begin{bmatrix} \mathbf{0}_{n_s T_p \times n_s T_p} & \mathbf{T}_{u,c} \end{bmatrix}_{n_s T_p \times (n_s + n_u) T_p} \hat{\mathbf{X}} \leq \mathbf{T}_{x_0,c} \mathbf{x}[0] \quad (8)$$

which can be written as,

$$\mathbf{A}_{\text{flow}} \hat{\mathbf{X}} \leq \mathbf{b}_{\text{flow}}$$

4 Boundary Constraints

Constraints on optimization vector $\hat{\mathbf{X}}$ are,

$$\mathbf{0} \leq \hat{\mathbf{X}} \leq \mathbf{1}$$

or,

$$\begin{aligned}\hat{\mathbf{X}} &\leq \mathbf{1}_{(n_s+n_u)T_p} \\ -\hat{\mathbf{X}} &\leq \mathbf{0}_{(n_s+n_u)T_p}\end{aligned}$$

which can be written as,

$$\begin{bmatrix} \mathbf{I} \\ -\mathbf{I} \end{bmatrix}_{2(n_s+n_u)T_p \times (n_s+n_u)T_p} \hat{\mathbf{X}} \leq \begin{bmatrix} \mathbf{1} \\ \mathbf{0} \end{bmatrix} \quad (9)$$

which can be written as

$$\mathbf{A}_{\text{boundary}} \hat{\mathbf{X}} \leq \mathbf{b}_{\text{boundary}}$$

5 Compact LP Form

Using inequality (6), (8), and (9), the LP problem can be written in compact form as,

$$\begin{aligned} &\min_{\hat{\mathbf{X}}} \mathbf{C}^T \hat{\mathbf{X}} \\ &\text{s.t.} \quad \begin{bmatrix} \mathbf{I}_{n_s T_p} & -\mathbf{T}_u \\ -\mathbf{I}_{n_s T_p} & \mathbf{T}_u \\ \mathbf{0}_{n_s T_p \times n_s T_p} & \mathbf{T}_{u,c} \\ \mathbf{I}_{(n_s+n_u)T_p} \\ -\mathbf{I}_{(n_s+n_u)T_p} \end{bmatrix}_{(5n_s+2n_u)T_p \times (n_s+n_u)T_p} \hat{\mathbf{X}} \leq \begin{bmatrix} \mathbf{T}_{x_0} \mathbf{x}[0] \\ -\mathbf{T}_{x_0} \mathbf{x}[0] \\ \mathbf{T}_{x_0,c} \mathbf{x}[0] \\ \mathbf{1}_{(n_s+n_u)T_p} \\ \mathbf{0}_{(n_s+n_u)T_p} \end{bmatrix}_{(5n_s+2n_u)T_p \times 1} \end{aligned} \quad (10)$$

which can be written as,

$$\begin{bmatrix} \mathbf{A}_{\text{dynamics}} \\ \mathbf{A}_{\text{flow}} \\ \mathbf{A}_{\text{boundary}} \end{bmatrix} \hat{\mathbf{X}} \leq \begin{bmatrix} \mathbf{b}_{\text{dynamics}} \\ \mathbf{b}_{\text{flow}} \\ \mathbf{b}_{\text{boundary}} \end{bmatrix} \quad (11)$$