

MP-282

Dynamic Modeling and Control of Multicopter Aerial Vehicles

Chapter 7: Control Allocation

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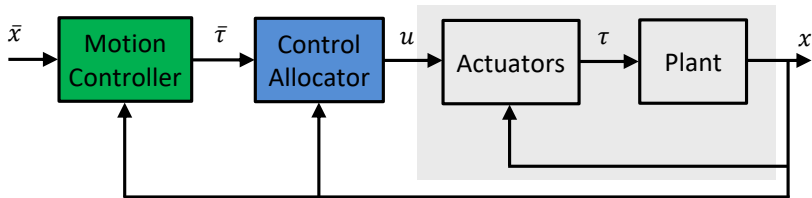
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Introduction . . .

Introduction

Control Structure

A typical control system of an over-actuated mechanical plant can be described by the following block diagram:



Legend:

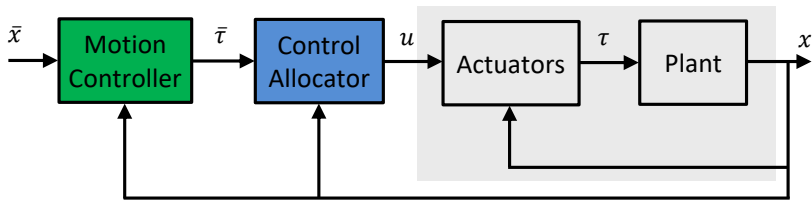
$\mathbf{x} \in \mathbb{R}^n$ - state; $\bar{\mathbf{x}} \in \mathbb{R}^n$ - state command; $\bar{\boldsymbol{\tau}} \in \mathbb{R}^m$ - virtual control¹; $\mathbf{u} \in \mathbb{R}^p$ - actuator commands; $\boldsymbol{\tau} \in \mathbb{R}^m$ - resulting control efforts.

¹The virtual control is typically a command for the resulting control effort.

Introduction

Comments

- The control system design can be divided into the derivation of the **motion controller** and **control allocator**.
- The main benefits of CA is achieved in control systems of over-actuated plants. Its advantages are:
 1. In case of actuator saturation/fault/failure, the control allocator can still produce the actuator commands (sometimes degraded).
 2. The actuator redundancy gives room for optimization (e.g., the minimization of some cost function).



Actuator-Set Model

In general, one can model the resulting control effort $\boldsymbol{\tau} \in \mathbb{R}^m$ as

$$\boldsymbol{\tau} = \mathbf{h}(\mathbf{u}, \mathbf{x}, t)$$

where \mathbf{h} is a known map and t denotes time.

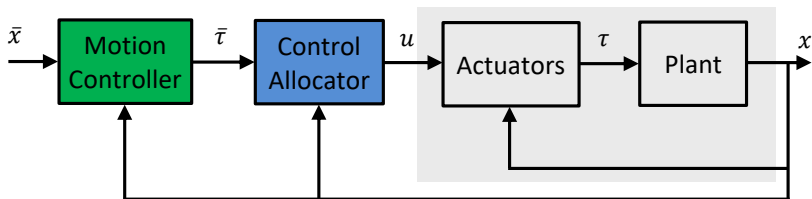
Remark

The time dependence of \mathbf{h} accounts for the actuator dynamics. However, in this chapter, we will assume that the actuator dynamics are very fast and, in this case, it would suffice to write $\boldsymbol{\tau} = \mathbf{h}(\mathbf{u}, \mathbf{x})$.

Introduction

Control Allocation Objective

Consider that a **virtual control** input $\bar{\tau} \in \mathbb{R}^m$ is provided by the **motion controller**. The **control allocation objective** is to **compute the actuator commands** $u \in \mathbb{R}^p$ which ensure that the resulting control effort $\tau \in \mathbb{R}^m$ will be sufficiently close to $\bar{\tau}$.



Introduction

Problem Formulation

It can be done as an optimization problem like

$$\min_{\mathbf{u}, \mathbf{s}} \{ \|\mathbf{Q}\mathbf{s}\| + J(\mathbf{u}, \mathbf{x}, t) \}$$

s.t.

$$\bar{\tau} - \mathbf{h}(\mathbf{u}, \mathbf{x}, t) = \mathbf{s}$$

$$\mathbf{u} \in \mathbb{U}$$

$$\mathbf{u} - \mathbf{u}_{\text{prev}} \in \delta\mathbb{U}$$

where

J is some **cost function**.

$\mathbf{Q} \in \mathbb{R}^{m \times m}$ is a **weighting matrix**.

$\mathbf{s} \in \mathbb{R}^m$ is a **slack variable**.

\mathbf{u}_{prev} is the previous value of \mathbf{u} (previous sampling time).

\mathbb{U} and $\delta\mathbb{U}$ are given compact set.

Remarks

(1) A generic example of J :

$$J(\mathbf{x}, \mathbf{u}, t) = \frac{1}{2}(\mathbf{u} - \mathbf{u}_n)^T \mathbf{W}(\mathbf{u} - \mathbf{u}_n)$$

where $\mathbf{W} \in \mathbb{R}^{p \times p}$ is a weighting matrix and $\mathbf{u}_n \in \mathbb{R}^p$ is the nominal value of \mathbf{u} . Basically, by this J , the optimization problem tries to minimize the deviation of \mathbf{u} w.r.t. \mathbf{u}_n .

(2) Note that the cost term $\|\mathbf{Q}\mathbf{s}\|$ along with the equality constraint $\bar{\tau} - \mathbf{h}(\mathbf{u}, \mathbf{x}, t) = \mathbf{s}$ forces τ towards $\bar{\tau}$, thus contributing with the control allocation objective (see slide 7).

(3) The last two constraints represent the actuator physical bounds.

Fixed-Rotor MAVs . . .

Fixed-Rotor MAVs

Actuator-Set Model

The actuator-set models for the fixed-rotor MAVs have the common form²

$$\begin{bmatrix} F^c \\ \mathbf{T}_B^c \end{bmatrix} = \mathbf{\Gamma} \mathbf{f}$$

where $\mathbf{\Gamma} \in \mathbb{R}^{4 \times n_r}$ is the allocation matrix and n_r is the total number of rotors.

Assumption: The actuator dynamics are very fast, implying that $\bar{\mathbf{f}} \approx \mathbf{f}$.

Define: The virtual control input is

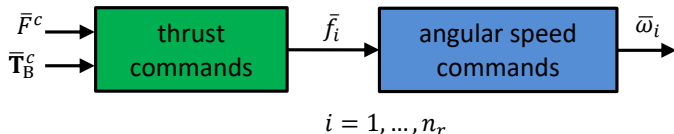
$$\bar{\boldsymbol{\tau}} \triangleq \begin{bmatrix} \bar{F}^c \\ \bar{\mathbf{T}}_B^c \end{bmatrix}$$

²See [Chapter 3](#).

Fixed-Rotor MAVs

Control Structure

We know from [Chapter 6](#) that the control allocation for fixed-rotor MAVs can be divided into two parts:



The blue block is simply realized by the inversion of the thrust model for each individual rotor, *i.e.*,

$$\bar{\omega}_i = \sqrt{\bar{f}_i / k_f}, \quad i = 1, \dots, n_r$$

In the sequel, **we are going to focus on the green block.**

Fixed-Rotor MAVs

Formulation 1

Computation of **thrust commands** considering the **rotor bounds**:

$$\min_{\bar{\mathbf{f}}} \|\bar{\mathbf{f}}\|^2$$

s.t.

$$\bar{\boldsymbol{\tau}} = \mathbf{\Gamma} \bar{\mathbf{f}}$$

$$\mathbf{e}_i^T \bar{\mathbf{f}} \in [f_{\min}, f_{\max}], \quad \forall i = 1, \dots, n_r$$

$$\mathbf{e}_i^T (\bar{\mathbf{f}} - \bar{\mathbf{f}}_{\text{prev}}) \in [\delta f_{\min}, \delta f_{\max}], \quad \forall i = 1, \dots, n_r$$

where

$f_{\min} \in \mathbb{R}$ and $f_{\max} \in \mathbb{R}$ are the thrust bounds.

$\delta f_{\min} \in \mathbb{R}$ and $\delta f_{\max} \in \mathbb{R}$ are the thrust rate bounds.

$\bar{\mathbf{f}}_{\text{prev}}$ is the previous value of $\bar{\mathbf{f}}$ (previous sampling time).

Remarks

(1) Note that this problem is similar to the prototype one given in slide 8, except that here we are not considering the slack variable s . Neglecting s is not an issue since we assure that the optimization input $\bar{\tau}$ is inside its feasible set.

(2) The above problem is a **quadratic program**, for which there exist many efficient (commercial and free) solvers available. For solving it in MATLAB, one can use the **quadprog** command (from the Optimization Toolbox).

Fixed-Rotor MAVs

Formulation 2

One can ignore the inequality constraints in formulation 1 to obtain the following simplified problem:

$$\begin{aligned} \min_{\bar{\mathbf{f}}} \quad & \|\bar{\mathbf{f}}\|^2 \\ \text{s.t.} \quad & \\ & \bar{\boldsymbol{\tau}} = \mathbf{\Gamma}\bar{\mathbf{f}} \end{aligned}$$

The above optimization problem has a unique closed-form solution that can be obtained using Lagrange multiplier, resulting

$$\bar{\mathbf{f}}^* = \mathbf{\Gamma}^\dagger \bar{\boldsymbol{\tau}}$$

where $\mathbf{\Gamma}^\dagger \triangleq \mathbf{\Gamma}^T (\mathbf{\Gamma}\mathbf{\Gamma}^T)^{-1}$ is the Moore-Penrose's pseudo-inverse matrix.

Fixed-Rotor MAVs

After computing the optimal solution $\bar{\mathbf{f}}^*$, it is required to saturate it so as to respect the thrust and thrust-rate bounds ³ (respectively):

$$\bar{\mathbf{f}} \in \mathbb{U} \cap \left(\delta\mathbb{U} \oplus \bar{\mathbf{f}}_{\text{prev}} \right)$$

Remarks

- (1) The above solution can produce a dangerous mismatch between the virtual control $\bar{\boldsymbol{\tau}}$ and the resulting control effort $\boldsymbol{\tau} \triangleq (F^c, \mathbf{T}_B^c)$.
- (2) However, it provides a lighter computational implementation compared with formulation 1.

³The symbol \oplus denotes the set (or Minkowski) sum, defined as $\mathcal{A} \oplus \mathcal{B} \triangleq \{\mathbf{a} + \mathbf{b} : \mathbf{a} \in \mathcal{A}, \mathbf{b} \in \mathcal{B}\}$.

Vectoring-Rotor MAVs ...

Vectoring-Rotor MAVs

Example 1: Quadcopter with Longitudinal-Vectoring Rotors

In [Chapter 3](#), we obtained the actuator-set model:

$$\begin{bmatrix} F_1^c \\ F_3^c \\ \mathbf{T}_B^c \end{bmatrix} = \mathbf{\Gamma}_{LV4} \mathbf{f}$$

where

$$\mathbf{\Gamma}_{LV4} \triangleq \begin{bmatrix} s\beta_1 & s\beta_2 & s\beta_3 & s\beta_4 \\ c\beta_1 & c\beta_2 & c\beta_3 & c\beta_4 \\ lc\beta_1 + ks\beta_1 & -lc\beta_2 - ks\beta_2 & -lc\beta_3 + ks\beta_3 & lc\beta_4 - ks\beta_4 \\ -lc\beta_1 & -lc\beta_2 & lc\beta_3 & lc\beta_4 \\ -ls\beta_1 + kc\beta_1 & ls\beta_2 - kc\beta_2 & ls\beta_3 + kc\beta_3 & -ls\beta_4 - kc\beta_4 \end{bmatrix}$$

Vectoring-Rotor MAVs

Consider $\beta_1 = \beta_2 = \beta_3 = \beta_4 \triangleq \beta$ and neglect the actuator dynamics. Given $\bar{\mathbf{F}}_G^c$, $\bar{\mathbf{T}}_B^c$, and $\mathbf{D}^{B/R}$, the actuator commands $\bar{\beta}$ and \bar{f}_i , $i = 1, \dots, 4$, are obtained by the following procedure:

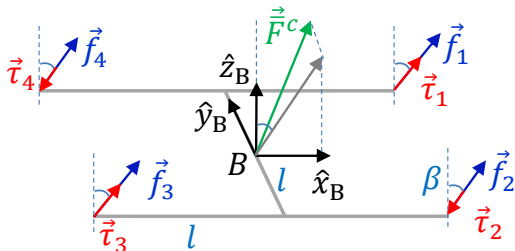
1. compute $\bar{\mathbf{F}}_B^c \triangleq (\bar{F}_1^c, \bar{F}_2^c, \bar{F}_3^c) = \mathbf{D}^{B/R} \bar{\mathbf{F}}_G^c$
2. compute $\bar{\beta} = \text{atan } \bar{F}_1^c / \bar{F}_3^c$
3. compute

$$\bar{\mathbf{f}} = \left(\check{\mathbf{\Gamma}}_{LV4} \right)^{-1} \begin{bmatrix} \bar{F}_3^c \\ \bar{\mathbf{T}}_B^c \end{bmatrix}$$

where $\check{\mathbf{\Gamma}}_{LV4}$ is obtained from $\mathbf{\Gamma}_{LV4}$ by eliminating its first line.

See the figure below.

Vectoring-Rotor MAVs



On the other hand, the attitude command can be set to

$$\bar{\mathbf{D}}^{B/R} = \mathbf{D}_{\bar{\mathbf{n}}_B}(\bar{\psi})\mathbf{D}_1(-\bar{\lambda})$$

where $\bar{\psi}$ and

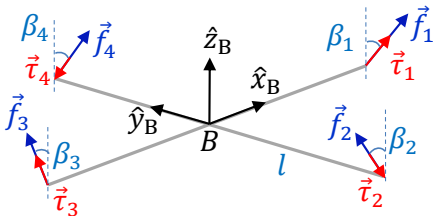
$$\bar{\lambda} \triangleq \arcsin \frac{\bar{F}_2^c}{\|\bar{\mathbf{F}}_B^c\| \cos \beta} \quad \bar{\mathbf{n}}_B \triangleq \frac{\bar{\mathbf{F}}_B^c}{\|\bar{\mathbf{F}}_B^c\|} \quad \square$$

Vectoring-Rotor MAVs

Exercise: Quadcopter with Transversal-Vectoring Rotors

Consider now the MAV illustrated below. Its actuator-set model was obtained in [Chapter 3](#); it has the format:

$$\begin{bmatrix} \mathbf{F}_B^c \\ \mathbf{T}_B^c \end{bmatrix} = \mathbf{\Gamma}_{\text{TV4}}(\beta_1, \dots, \beta_4) \mathbf{f}$$



Design a control allocator for this model. Which is the corresponding attitude command?

Complementary Reading ...

Complementary Reading

We suggest the following complementary texts:

- A survey on CA in general → [Ref \[1\]](#).
- CA for aerospace systems → [Ref \[2\]](#).
- CA for MAVs using pseudo-inverse matrix and saturation → [Ref \[3\]](#).

References . . .

References

-  [1] Johansen, T.A. and Fossen, T.I. Control Allocation - A Survey. Automatica, 49, 2013.
-  [2] Oppenheimer, M., Doman, D., Bolender, M. (2010). Control allocation. In W. S. Levine (Ed.), The control handbook, control system applications (2nd ed.). (Chapter 8)
-  [3] Ducard, G.J.J. and Hua, M.D. Discussion and Practical Aspects on Control Allocation for a Multi-Rotor Helicopter. International Archives of the Photogrammetry, Remote Sensing and Spatial Information Sciences, September 2011, Zurich.
-  [4] Boyd, S. and Vandenberghe, L. Convex Optimization. Cambridge University Press, 2004.

Thanks!