



## Tell me your understanding of “What is Guidance”

- Send to [haichao.hong@sjtu.edu.cn](mailto:haichao.hong@sjtu.edu.cn)
- Include “作业” and YOUR NAME in the subject line
- NO COPY PASTE





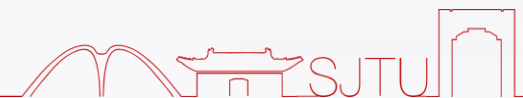
**Haichao Hong / 洪海超**

**Associate Professor**

Email: [haichao.hong@sjtu.edu.cn](mailto:haichao.hong@sjtu.edu.cn)

Office: A227

Homepage: [haichao.de](http://haichao.de)





# Contents

## AE8120

-  What is Guidance
-  **Generic Formulation of Trajectory Optimization**
-  ...



# Keywords:

- Trajectory / Maneuvering
- Motion of Aircraft (Translational Dynamics)
- Select Commands
- Current State to Desired State
- Constraints

# Relationship between

## (Computational) Guidance and Trajectory Optimization

Determine a trajectory

Steer the aircraft

Need to consider constraints

Optimality is not fundamental

Computational guidance mostly onboard

Select a trajectory

Design the maneuver

Can consider constraints

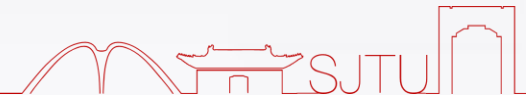
Optimality is fundamental

Not necessarily onboard



## Trajectory Optimization

- Performance index / Objective function / Cost function
  - Equations of motion (often translational dynamics)
  - Initial and terminal conditions
  - Path constraints
- } Constraints





Control history

$$\mathbf{u}(t) \in \mathbb{R}^{m \times 1}$$

Corresponding state history

$$\mathbf{x}(t) \in \mathbb{R}^{n \times 1}$$

Relevant parameters

$$\mathbf{p} \in \mathbb{R}^{s \times 1}$$

Bolza cost function

$$J = \underbrace{\varphi(\mathbf{x}(t_f))}_{\text{Mayer}} + \underbrace{\int_{t_0}^{t_f} L(\mathbf{x}(t), \mathbf{u}(t), t) dt}_{\text{Lagrange}}$$

Aircraft dynamics

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t)$$

Equality and inequality constraints

$$\mathbf{h}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) = 0$$

$$\mathbf{g}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) \leq 0$$





Equality and inequality constraints  $h(x(t), u(t), p, t) = 0$   $g(x(t), u(t), p, t) \leq 0$

- Equality constraints  $\rightarrow$  Inequality constraints
- Not necessarily be the function of state, control, and parameter,

but THINK TWICE!

- $\geq \longleftrightarrow \leq$   $AX \leq b$   $g(x(t), u(t), p, t) \leq 0$

- Main reason causing infeasibility







## A generic formulation:

$$\begin{aligned} & \underset{\mathbf{u}(t), \mathbf{p}}{\text{minimize}} && J = \varphi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}(t), \mathbf{u}(t), t) dt \\ & \text{subject to} && \dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t), \\ & && \mathbf{h}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) = 0, \\ & && \mathbf{g}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) \leq 0 \end{aligned}$$

### Note:

- Constraints are imposed when necessary.
- $t_f$  might or might not be given.





## Before solving it

$$\begin{aligned} & \underset{\mathbf{u}(t), \mathbf{p}}{\text{minimize}} && J = \varphi(\mathbf{x}(t_f)) + \int_{t_0}^{t_f} L(\mathbf{x}(t), \mathbf{u}(t), t) dt \\ & \text{subject to} && \dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t), \\ & && \mathbf{h}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) = 0, \\ & && \mathbf{g}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) \leq 0 \end{aligned}$$

1. Figure out your purpose
2. Know your objective (function)
3. Constraints = Practical concerns





# Contents

- What is Guidance
- Generic Formulation of Trajectory Optimization
- Discretization Methods**
- ...





## Why discretization methods

### Simulation

$$\mathbf{x}(t) = \mathbf{x}(0) + \int_0^{t_f} \dot{\mathbf{x}}(t) dt$$



$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t)$$





## Discretization methods

- Single Shooting
- Multiple Shooting
- Collocation – Full Discretization

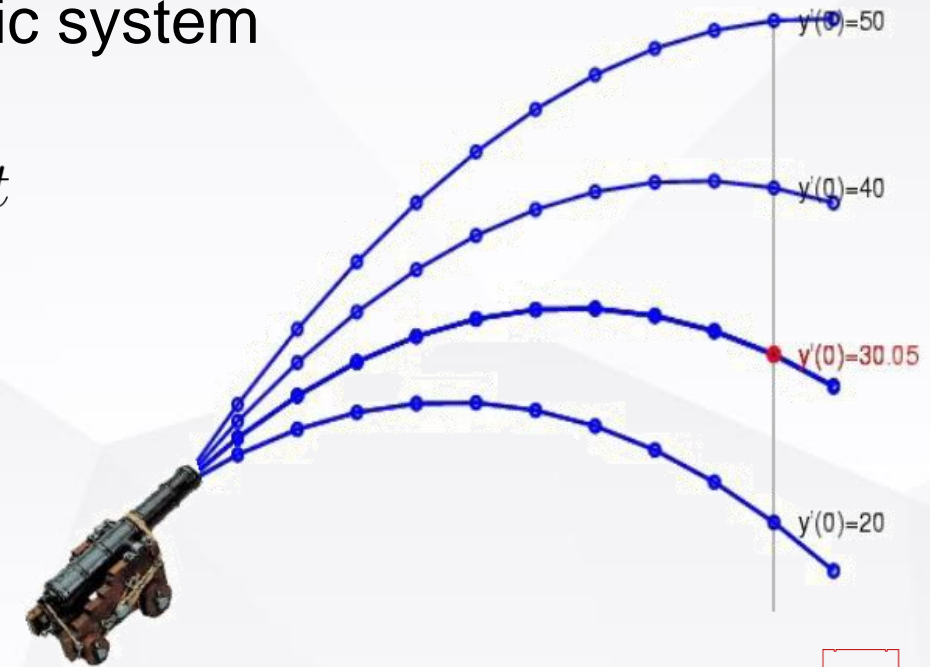


## Single Shooting

- Integration over complete time history in one single sweep
- Restoration of states by integration of dynamic system

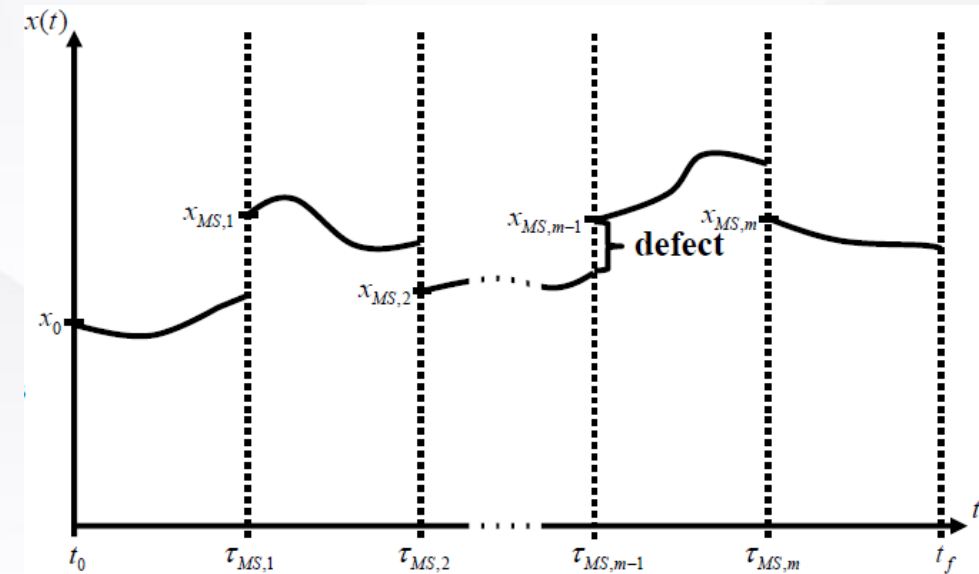
$$\mathbf{x}(t) = \mathbf{x}(t_0) + \int_{t_0}^{t_f} \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}, t) dt$$

- Adjust **only the controls**
- **Least variables**



## Multiple Shooting

- Separation of time history into  $m$  segments (shooting intervals)
- Propagation of differential equations on each interval
- Adjust **the controls and some states**





## Single shooting vs. Multiple shooting:

### Single Shooting

- States are sensitive w.r.t. (i.e. depend on) the initial states
- States are sensitive w.r.t. preceding control variables

**Large, non-linear variations can lead to severe convergence problems**

### Multiple Shooting

- States are only sensitive w.r.t. states at preceding multiple shooting nodes
- States are only sensitive w.r.t. preceding control variables within multiple shooting interval

**Large, non-linear variations can be avoided – improved convergence**

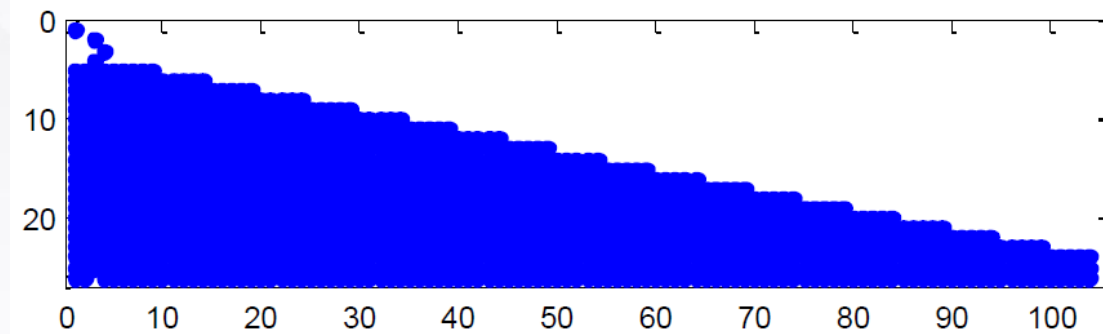




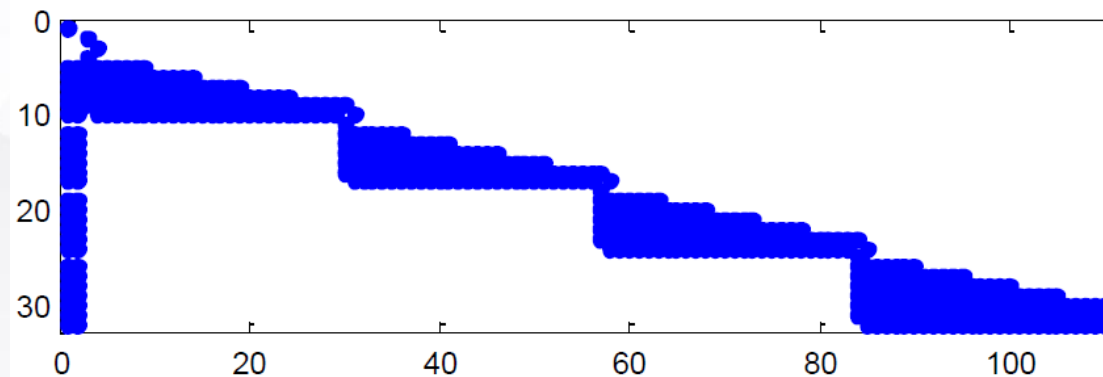


## Single shooting vs. Multiple shooting: Gradient Structure

Single Shooting



Multiple Shooting



## Collocation – Full Discretization

- Discretization of controls and states at time discretization points:
- Propagation of differential equations on EVERY interval
- Adjust **all states and controls**
- **Most variables**
- **Sparse gradient structure**

