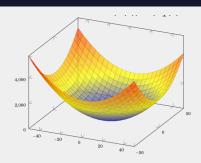
## MOTION PLANNING FOR AUTONOMOUS VEHICLES

HAMILTONIAN (OPTIMAL CONTROL THEORY)

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# HAMILTONIAN (OPTIMAL CONTROL

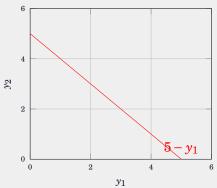
THEORY)

#### **CONTENTS**

- Constrained Minimization of functions
  - Elimination method (direct method)
  - The Lagrange multiplier method: examples, general formulation
- Constrained Minimization of functional: Point constraints, differential equation constraints
- Hamiltonian
- The necessary condition for optimal control
- Boundary conditions for optimal control: with the fixed final time and the final state specified or free
- Boundary conditions for optimal control: with the free final time and the final state specified, free, lies on the moving point  $x_f = \theta(t_f)$ , or lies on a moving surface m(x(t)))

#### **CONSTRAINED MINIMIZATION OF FUNCTIONS**

Find the point on the line  $y_1 + y_2 = 5$  that is nearest the origin.



$$\label{eq:subjection} \begin{array}{ll} \underset{y_1,y_2\in\mathbb{R}}{\text{minimize}} & f(y_1,y_2)=y_1^2+y_2^2, \quad \text{square distance} \\ \text{subject to} & y_1+y_2=5 \end{array}$$

3 4

 $\label{eq:force_point} \begin{array}{ll} \underset{y_1,y_2\in\mathbb{R}}{\text{minimize}} & f(y_1,y_2)=y_1^2+y_2^2, \quad \text{square distance} \\ \text{subject to} & y_1+y_2=5 \end{array}$ 

■ The differential

$$df(y_1, y_2) = \left(\frac{\partial f(\cdot)}{\partial y_1}\right) \Delta y_1 + \left(\frac{\partial f(\cdot)}{\partial y_2}\right) \Delta y_2 \tag{1}$$

where  $f(\cdot) = f(y_1, y_2)$ .

3

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where  $f(\cdot) = f(y_1, y_2)$ .

■ If  $f(y_1^*, y_2^*)$  is the extreme point,

$$df(y_1^*, y_2^*) = \left(\frac{\partial f(y_1^*, y_2^*)}{\partial y_1}\right) \Delta y_1 + \left(\frac{\partial f(y_1^*, y_2^*)}{\partial y_2}\right) \Delta y_2 \tag{2}$$

3

■ If and only if  $y_1$  and  $y_2$  are **independent**  $\Delta y_1$  and  $\Delta y_2$  can be selected arbitrarily.

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- $\blacksquare \text{ That result in } \Big( \frac{\partial f(y_1^*, y_2^*))}{\partial y_1} \Big) \Delta y_1 = 0 \text{ and } \Big( \frac{\partial f(y_1^*, y_2^*))}{\partial y_2} \Big) \Delta y_2 = 0.$
- However, in this example,  $y_1$  and  $y_2$  are dependent.

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- However, in this example,  $y_1$  and  $y_2$  are dependent.
- Hence, considering  $f(y_1, y_2)$  only function of  $y_2$

$$df(y_2^*) = \left(-10 + 4y_2^*\right) \Delta y_2 = 0$$
  

$$\Rightarrow y_2^* = 2.5, \ y_1^* = 2.5$$
(3)