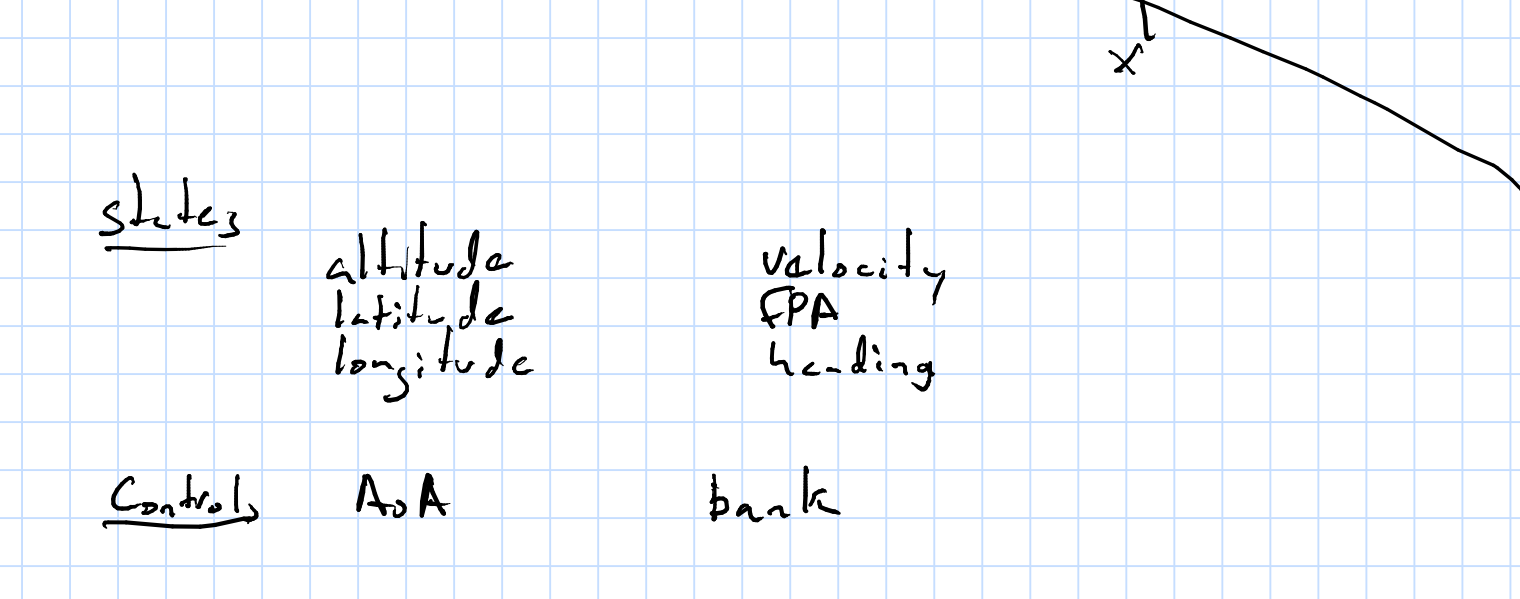


JMU

- Beluza is a general trajectory optimization package written in Python
- Unlike other available trajectory, beluza uses indirect methods
- Indirect methods produce very high quality solutions free from numerical artefact of other methods



States

altitude
latitude
longitude

velocity
EPA
heading

Controls

AOA

bank

Objective (cost)

max terminal \rightarrow min $-V_f$

Parameter Optimization

The most basic class of optimization problem concerns finding \bar{u} of m parameters such that a cost $J(\bar{u})$ is minimized.

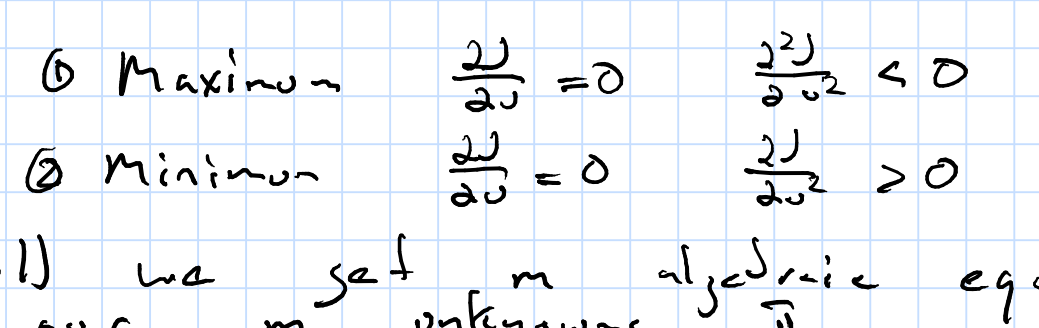
\bar{u} is the control vector

$$\min_{\bar{u}} J(\bar{u})$$

If $J(\bar{u})$ has 1st & 2nd derivatives everywhere, the necessary conditions for a local minimum are

$$\frac{\partial J}{\partial \bar{u}} = 0 \quad (1.1)$$

$$\frac{\partial^2 J}{\partial \bar{u}^2} \geq 0 \quad (1.2) \quad \text{positive semi-definite}$$



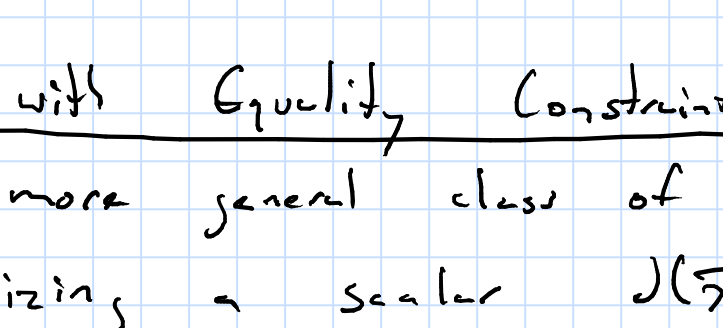
$$\textcircled{1} \text{ Maximum } \frac{\partial J}{\partial \bar{u}} = 0 \quad \frac{\partial^2 J}{\partial \bar{u}^2} < 0$$

$$\textcircled{2} \text{ Minimum } \frac{\partial J}{\partial \bar{u}} = 0 \quad \frac{\partial^2 J}{\partial \bar{u}^2} > 0$$

- (1.1) we set m algebraic equation for our m unknowns, \bar{u}

- Points that satisfy (1.1) are called stationary points

- If $\frac{\partial J}{\partial \bar{u}} = 0$, that is a singular point and more information is needed to determine optimality



- If (1.1) and

$$\frac{\partial^2 J}{\partial \bar{u}^2} > 0 \quad (\text{positive definite}) \quad (1.4)$$

then sufficient conditions of optimality exist.

Problems with Equality Constraints

- A more general class of optimization involves minimizing a scalar $J(\bar{x}, \bar{u})$ where the state vector

$$\bar{x} = [x_1, x_2, \dots, x_n]^T$$

contains n parameters determined by n constraint functions

$$\bar{f}(\bar{x}, \bar{u}) = \begin{bmatrix} f_1(\bar{x}, \bar{u}) \\ \vdots \\ f_n(\bar{x}, \bar{u}) \end{bmatrix} = 0 \quad (1.5)$$

Therefore the resulting problem is

$$\min_{\bar{u}} J(\bar{x}, \bar{u})$$

subject to

$$(1.6)$$

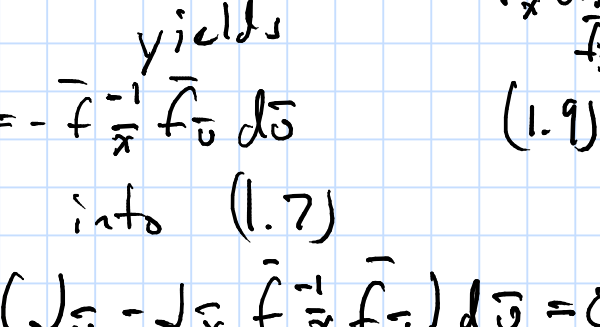
$$\bar{f}(\bar{x}, \bar{u}) = 0$$

Example

$$\min J = \frac{1}{2} \left(\frac{x^2}{a^2} + \frac{u^2}{b^2} \right)$$

$$\text{subject to: } f = x + mu - c = 0$$

Contour Plot



Unadjained Method I

For a stationary, $dJ=0$ for an arbitrary $d\bar{u}$

$$\Rightarrow dJ = \frac{\partial J}{\partial \bar{x}} d\bar{x} + \frac{\partial J}{\partial \bar{u}} d\bar{u} = 0 \quad (1.7) \quad \frac{\partial J}{\partial \bar{x}} = J_{\bar{x}}$$

Also, for $d\bar{u}$ to change without changing \bar{f}

$$\Rightarrow d\bar{f} = \frac{\partial \bar{f}}{\partial \bar{x}} d\bar{x} + \frac{\partial \bar{f}}{\partial \bar{u}} d\bar{u} = 0 \quad (1.8) \quad \bar{f}_{\bar{x}} d\bar{x} + \bar{f}_{\bar{u}} d\bar{u} = 0$$

Rearranging (1.8) yields $d\bar{x} = -\bar{f}_{\bar{x}}^{-1} \bar{f}_{\bar{u}} d\bar{u}$ (1.9)

Substitute (1.9) into (1.7)

$$dJ = (J_{\bar{u}} - J_{\bar{x}} \bar{f}_{\bar{x}}^{-1} \bar{f}_{\bar{u}}) d\bar{u} = 0$$

$$\Rightarrow J_{\bar{u}} - J_{\bar{x}} \bar{f}_{\bar{x}}^{-1} \bar{f}_{\bar{u}} = 0$$

m equations

n equations from $f=0$

$m+n$ equations for $m+n$ unknown (\bar{x}, \bar{u})

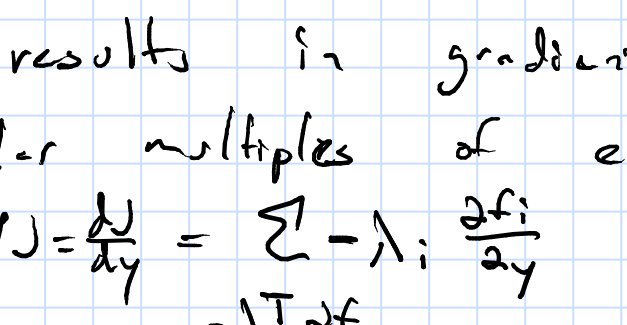
Note:

Symmetry of (1.7) & (1.8) indicates that we interchange the object with a constraint

$$\min f_i(\bar{x}, \bar{u})$$

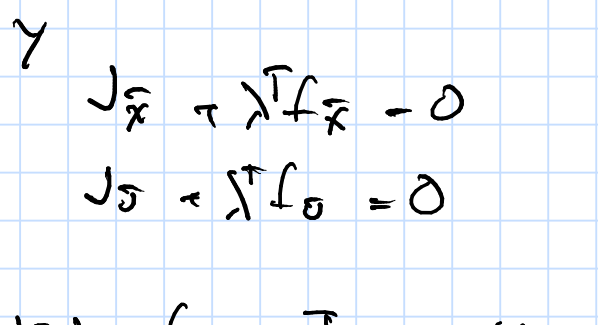
$$\text{s.t. } \begin{bmatrix} x_i, u_i - j^* \\ f_{ji} \\ \text{for } j=1, \dots, m \end{bmatrix} = 0$$

is equivalent to problem in 1.6



Unadjained Method II (Lagrange Multiplier)

$$\bar{y}^T = [\bar{x}^T, \bar{u}^T] \quad (1.11)$$



Tangency results in gradient of J and f , being scalar multiples of each other

$$\nabla J = \frac{\partial J}{\partial \bar{y}} = \sum -\lambda_i \frac{\partial f_i}{\partial \bar{y}}$$

$$= -\lambda^T \frac{\partial \bar{f}}{\partial \bar{y}}$$

$$\Rightarrow J_{\bar{y}} + \lambda^T \bar{f}_{\bar{y}} = 0$$

λ are Lagrange Multipliers

Separate y

$$J_{\bar{x}} + \lambda^T \bar{f}_{\bar{x}} = 0 \quad n \text{ equations} \quad (1.12)$$

$$J_{\bar{u}} + \lambda^T \bar{f}_{\bar{u}} = 0 \quad m \text{ equations} \quad (1.13)$$

Solving (1.12) for $\bar{\lambda}$ yields

$$\bar{\lambda}^T = -J_{\bar{x}} \bar{f}_{\bar{x}}^{-1} \quad (1.14)$$

Note

From 1.14, we can infer an interpretation of λ by setting $du=0$

$$\bar{\lambda}^T = -J_{\bar{x}} \bar{f}_{\bar{x}}^{-1} = -\frac{\partial J}{\partial \bar{x}} \frac{\partial \bar{x}}{\partial \bar{f}} \bigg|_{\bar{u}} = -\frac{\partial J}{\partial \bar{f}} \bigg|_{\bar{u}} \quad 1.15$$

Lagrange indicate how much the constraint affects the cost

Advanced Method

Because $\bar{f}=0$ at the solution we can adjoin \bar{f} to the cost without changing its optimal value

Here $\lambda_1, \dots, \lambda_m$ are not yet specified

$$J'(\bar{x}, \bar{u}) = J(\bar{x}, \bar{u}) + \sum_{i=1}^m \lambda_i f_i(\bar{x}, \bar{u})$$

$$\text{augmented cost} = J(\bar{x}, \bar{u}) + \lambda^T \bar{f} \quad (1.16)$$

Differential change in J' is

$$dJ' = \frac{\partial J'}{\partial \bar{x}} d\bar{x} + \frac{\partial J'}{\partial \bar{u}} d\bar{u} \quad (1.17)$$

We are about how du affect dJ' , so we pick λ to eliminate dx

$$\frac{\partial J'}{\partial \bar{x}} = \frac{\partial J}{\partial \bar{x}} + \lambda^T \frac{\partial \bar{f}}{\partial \bar{x}} = 0$$

$$\Rightarrow \bar{\lambda}^T = -\frac{\partial J}{\partial \bar{x}} \left(\frac{\partial \bar{f}}{\partial \bar{x}} \right)^{-1} \quad (1.18)$$

Same as (1.15)

Therefore

$$dJ' = \frac{\partial J'}{\partial \bar{u}} d\bar{u} = \frac{\partial J'}{\partial \bar{u}} d\bar{u} = 0$$

the resulting new conditions are

$$f(\bar{x}, \bar{u}) = 0 \quad (n \text{ eqns})$$

$$\frac{\partial J'}{\partial \bar{x}} = 0 \quad (n \text{ eqns}) \quad \text{2n eqns}$$

$$\frac{\partial J'}{\partial \bar{u}} = 0 \quad (m \text{ eqns}) \quad \text{2n+m eqns}$$

$(\bar{x}, \bar{\lambda}, \bar{u})$

Problems with Inequality Constraints

Problems may also contain constraints of the form

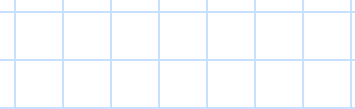
$$\bar{g}(\bar{x}, \bar{u}) \leq 0$$

There are 2 cases that can exist

① Inactive: - solution on the constraint boundary

- choice of \bar{x} does not depend on \bar{u}

- constraint can be neglected



② Active: - solution is on constraint boundary

- constraint can be treated as equality constraint

By Example

$$\min J = \frac{1}{2} \left(\frac{x^2}{a^2} + \frac{u^2}{b^2} \right)$$

$$\text{s.t. } f(\bar{x}, \bar{u}) = x + mu - c = 0$$

where a, b, c are constant

Contour plot

$$\text{Augmented cost } J' = J + \lambda^T f = \frac{1}{2} \left(\frac{x^2}{a^2} + \frac{u^2}{b^2} \right) + \lambda(x + mu - c)$$

$$\frac{\partial J'}{\partial u} = \frac{u}{b^2} + \lambda m = 0$$

$$\frac{\partial J'}{\partial x} = \frac{x}{a^2} + \lambda = 0$$

$$f = x + mu - c = 0$$

3 eqns for 3 unknowns

Solve linear equations

$$\Rightarrow x = \frac{a^2 c}{a^2 + m^2 b^2}, \quad u = \frac{b^2 m c}{a^2 + m^2 b^2}, \quad \lambda = -\frac{c}{a^2 + m^2 b^2}$$

$$J = \frac{c^2}{2(a^2 + m^2 b^2)} \quad (\text{is min})$$