

Trajectory Optimization

Core Idea

Trajectory Optimization problems are often not convex

TrajOpt breaks the problem into a set of convex approximations

Example

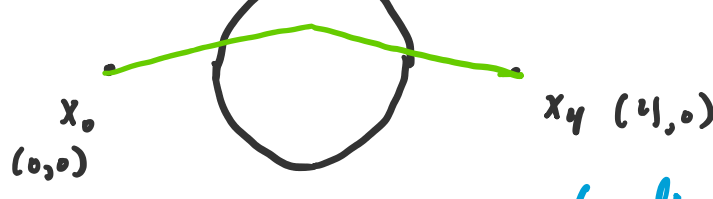


Optimize x_1, x_2, x_3

Objective: $\min \sum \|x_{i+1} - x_i\|^2$ } smoothness

Constraint: $\forall x_i, \|x_i - (2,0)\| \geq 1$ } Do Not intersect with circle

Step 1: Initialization



$$x_1 = (1, 2)$$

$$x_2 = (2, 3)$$

$$x_3 = (3, 2)$$

$$\begin{aligned} \text{Gradients} \\ \nabla_x d &= \frac{(x-2)}{\sqrt{(x-2)^2 + y^2}} \\ &= \frac{(x-2)}{d} \\ \nabla_y d &= \frac{y}{\sqrt{(x-2)^2 + y^2}} \\ &= \frac{y}{d} \end{aligned}$$

Step 2: Iteration 1

A. Check constraint violations:

$$x_1 = (1, 2)$$

- Distance from (2,0)

$$d_1 = \sqrt{(1-2)^2 + (2-0)^2} = 1.02$$

- constraint

$$d_1 - 1 = 1.02 - 1 = .02 \geq 0 \checkmark$$

$$x_2 = (2, 3)$$

- Distance from (2,0)

$$d_2 = .3$$

- constraint

$$.3 - 1 = -.7 < 0 \text{ X violated}$$

$$x_3 = (3, 2)$$

- Distance from (2,0)

$$d_3 = 1.02$$

- constraint

$$.02 \geq 0 \checkmark$$

B. Calculate gradient for Violation

$$\text{Constraint: } g(x,y) = \sqrt{(x-2)^2 + y^2} - 1$$

$$\text{Gradient: } \nabla g = \left(\frac{x-2}{\sqrt{(x-2)^2 + y^2}}, \frac{y}{\sqrt{(x-2)^2 + y^2}} \right)$$

$$\nabla g = \left(\frac{x-2}{d}, \frac{y}{d} \right)$$

$$x_2 = (2, 3)$$

$$d_2 = .3$$

$$\nabla g(2,3) = \left(\frac{2-2}{.3}, \frac{3}{.3} \right)$$

$$\nabla g(2,3) = (0, 1)$$

C. Create Linear Approximation

The linear approximation of g around $(2,3)$ is

$$g(x,y) \approx g(2,3) + \nabla g(2,3) \cdot [x-2, y-3] \text{ } 1^{\text{st}} \text{ order Taylor expansion}$$

$$\text{Given: (1) } g(x,y) = \sqrt{(x-2)^2 + y^2} - 1 \Rightarrow g(2,3) = -.7$$

$$(2) \nabla g = \left(\frac{x-2}{\sqrt{(x-2)^2 + y^2}}, \frac{y}{\sqrt{(x-2)^2 + y^2}} \right) \Rightarrow \nabla g(2,3) = [0, 1]$$

Plugging in...

$$g(x,y) \approx -.7 + 0 \cdot (x-2) + 1 \cdot (y-3)$$

$$\approx -.7 + y - 3$$

$$g(x,y) \approx y - 1$$

From constraint eq

$$g(x,y) = d - 1 \geq 0$$

$$g(x,y) \geq 0$$

substituting

$$y - 1 \geq 0$$

$$y \geq 1$$

D. Convex Optimization

Objective

$$\begin{aligned} \min \quad & \|x_1 - x_0\|^2 + \underbrace{\|x_2 - x_1\|^2}_{\text{Solve for this}} + \|x_2 - x_2\|^2 + \|x_4 - x_2\|^2 \\ & = \|(1,2) - (0,0)\|^2 + \|(x_2) - (1,2)\|^2 + \|(3,2) - (x_2)\|^2 + \|(4,0) - (3,2)\|^2 \\ & = (1^2 + 2^2) + (x_{2x} - 1)^2 + (x_{2y} - 2)^2 + (3 - x_{2x})^2 + (2 - x_{2y})^2 + (1^2 + 2^2) \end{aligned}$$

...

...

$$f(x_{2x}, x_{2y}) = 2x_{2x}^2 - 8x_{2x} + 2x_{2y}^2 - 8x_{2y} + 10.08$$

$$\frac{\partial f}{\partial x_{2x}} = 4x_{2x} - 8, \quad \frac{\partial f}{\partial x_{2y}} = 4x_{2y} - 8$$

Set partial derivatives = 0

$$x_{2x} = 2$$

$$x_{2y} = 2$$

So, the optimal min is $(2, 2)$ BUT $.2 \geq 1$ x, so the constraint is violated

Accounting for constraints, x_{2x} is the same $\hat{?}$ x_{2y} is forced to 1

Quick gut check...

$$f(x_{2x}, 1) = 2x_{2x}^2 - 8x_{2x} + 11.28$$

$$\frac{\partial f}{\partial x_{2x}} = 4x_{2x} - 8 \Rightarrow x_{2x} = 2 \checkmark$$

True optimal point = $(2, 1)$



New path

Step 3: Continue until convergence or # iterations reached

will stop here b/c algorithm converges to $(2, 1)$