

# Minimum Time Speed Optimization Along a Fixed Path

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## Motivation

- minimum time trajectories are desired in a wide range of applications:
  - surveillance
  - flight paths
  - machine paths
  - racing
- trajectory generation problems are hard
- currently, trajectory generation is often divided into two parts
  - 1) a feasible path is generated through obstacles
  - 2) the time to traverse that path is optimized
- this research focuses on the second step of trajectory generation
- already studied for robotic manipulators; can be generalized to other vehicles

## Dynamics

- vehicles have a generalized position  $q \in \mathbf{R}^p$
- vehicles take control inputs  $u \in \mathbf{R}^r$
- vehicles have dynamics:

$$R(q)u = M(q)\ddot{q} + C(q, \dot{q})\dot{q} + d(q)$$

- $R: \mathbf{R}^p \rightarrow \mathbf{R}^{p \times r}$ : controls matrix
- $M: \mathbf{R}^p \rightarrow \mathbf{S}_{++}^p$ : mass matrix
- $\mathbf{S}_{++}^p$ : set of positive definite  $p \times p$  matrices
- $C: \mathbf{R}^{2p} \rightarrow \mathbf{R}^{p \times p}$ : centrifugal matrix, linear in  $\dot{q}$
- $d: \mathbf{R}^p \rightarrow \mathbf{R}^p$ : position dependent force

## Control Constraints

- vehicles have control constraints:

$$(\dot{q}^2, \ddot{q}, u) \in \mathcal{C}(q)$$

- $\mathcal{C}(q) \subseteq \mathbf{R}^{p \times p \times m}$ : a set valued mapping to convex sets
- control limits can be dependent on generalized position
- path constraints (i.e. speed limits) can also be represented

## Path

- vehicles travel a path:

$$s: [0, 1] \rightarrow \mathbf{R}^p$$

- $s(\theta(t)) = q(t), \quad t \in [0, T]$
- $\theta: [0, T] \rightarrow [0, 1]$
- $\theta(0) = 0, \quad \theta(T) = 1, \quad \dot{\theta} > 0$

## Problem Statement

$$\begin{aligned} & \text{minimize } T \\ & \text{subject to } R(q(t))u(t) = M(q(t))\ddot{q}(t) + C(q(t), \dot{q}(t))\dot{q}(t) + d(q(t)), \quad t \in [0, T] \\ & \quad s(\theta(t)) = q(t), \quad t \in [0, T] \\ & \quad (\dot{q}(t)^2, \ddot{q}(t), u(t)) \in \mathcal{C}(q(t)), \quad t \in [0, T] \end{aligned}$$

- the functions  $\theta$  and  $u$  are optimization parameters
- $R, M, C, d, s$ , and  $\mathcal{C}$  are problem data
- through a clever change in variables, known since the 1970's, this problem becomes a convex optimization problem
- convexity guarantees global optimality of solution
- discretize  $\theta$  to solve the problem
- the discrete convex optimization problem can be solved very efficiently
- time to solve is linear in the discretization of  $\theta$

## Example: Spacecraft

- a spacecraft modeled as a point mass with a force that can be applied in any direction subjected to gravity

$$q = (x, y, z), \quad u = (f_x, f_y, f_z)$$

$$M = mI, \quad C = 0, \quad d = -gme_3, \quad R = I$$

- $m$ : mass
- $g$ : acceleration due to gravity
- $e_3$ : unit vector in the  $z$  direction

$$\mathcal{C} = \left\{ (\dot{q}^2, \ddot{q}, u) \mid \|u\|_2 \leq u_{\max} \right\}$$

- $u_{\max}$ : maximum force that the thruster can produce

## Example: Car

- a car modeled using the friction circle assumption and treated as front wheel drive (less force available when accelerating)

$$q = (x, y), \quad u = (f_{\text{long}}, f_{\text{lat}})$$

$$M = mI, \quad C = 0, \quad d = 0, \quad R = \begin{bmatrix} \cos(\phi(q)) & -\sin(\phi(q)) \\ \sin(\phi(q)) & \cos(\phi(q)) \end{bmatrix}$$

- $\phi: \mathbf{R} \rightarrow \mathbf{R}$ : orientation of the vehicle

$$\mathcal{C} = \left\{ (\dot{q}^2, \ddot{q}, u) \mid \|u\|_2 \leq \mu_s F_N, \quad |f_{\text{long}}| \leq W_f \mu_s F_N \right\}$$

- $\mu_s$ : friction between the road and wheels
- $W_f$ : percentage of mass supported by the front tires
- $F_N = mg$ : normal force to the road

## Example: Extensions

- more complex models fit in this formulation
- vectored thrust, variable friction, banked turns, aerodynamic drag
- can be applied to planes, underwater vehicles, ducted fans

## Results

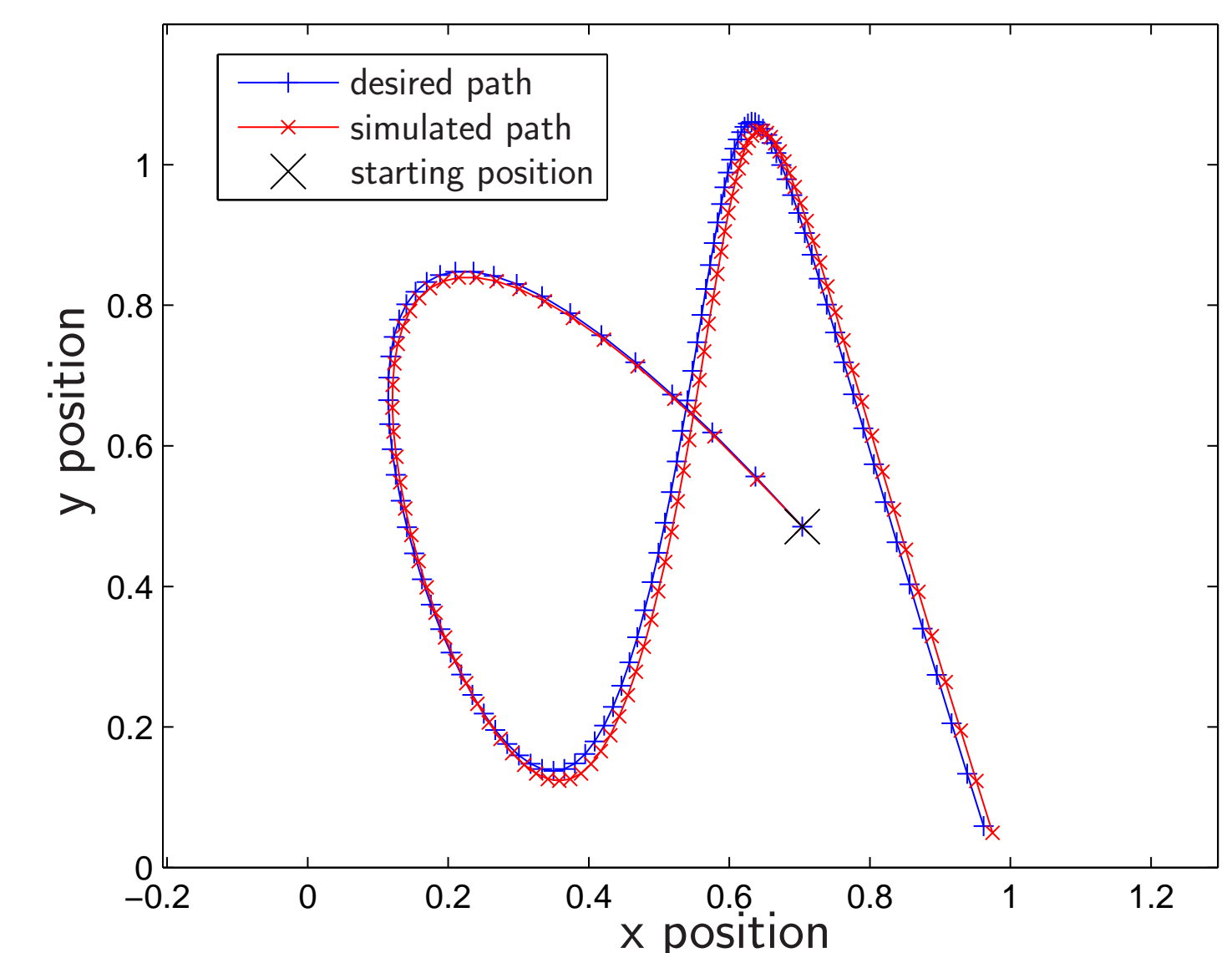


Figure 1: resultant path from applying minimum time control to a car example with  $\theta$  discretized at 100 points

- controls track the desired path
- computation took 16.8 milliseconds
- could be implemented with feedback or recomputation to reduce error

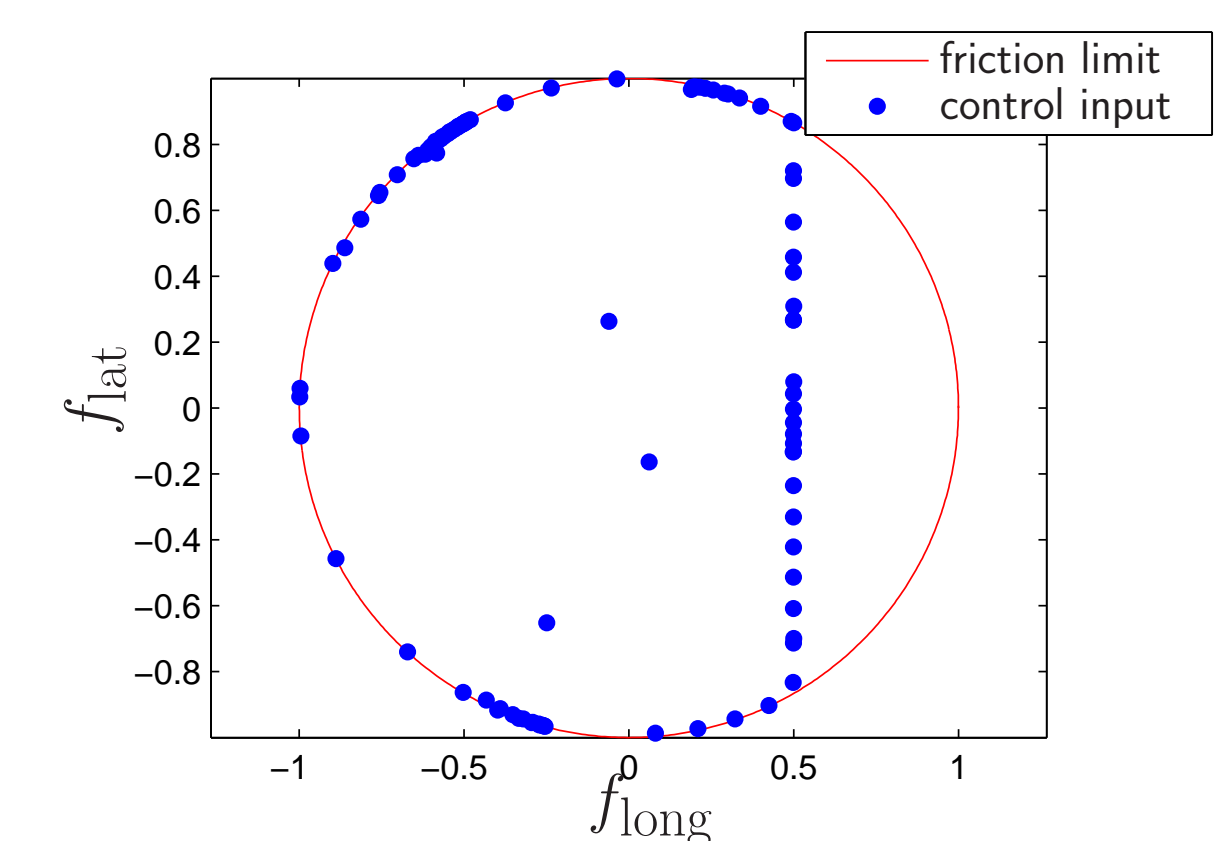


Figure 2: forces applied

- forces obey the force constraints
- forces are at their limits as expected, otherwise faster travel is possible

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