

Control with QP and SOCP

Contact-aware Control, Lecture 8

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Quadratic programming

There are special cases of optimization problems that can be solved numerically very efficiently. One of those is *quadratic programming*.

General form of a quadratic program (QP) is given below:

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \mathbf{x}^\top \mathbf{H} \mathbf{x} + \mathbf{f}^\top \mathbf{x}, \\ & \text{subject to} && \begin{cases} \mathbf{A} \mathbf{x} \leq \mathbf{b}, \\ \mathbf{C} \mathbf{x} = \mathbf{d}. \end{cases} \end{aligned} \tag{1}$$

where \mathbf{H} is positive-definite and $\mathbf{A} \mathbf{x} \leq \mathbf{b}$ describe a *convex region*.

Quadratic programming

Special case with an analytic solution

If there are no inequality constraints:

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \mathbf{x}^\top \mathbf{H} \mathbf{x} + \mathbf{c}^\top \mathbf{x}, \\ & \text{subject to} && \mathbf{A} \mathbf{x} = \mathbf{b}. \end{aligned} \tag{2}$$

a quadratic program can be solved analytically.

Quadratic programming

Example

We have the following problem: find such \mathbf{x} that minimizes $\mathbf{x}^\top \mathbf{M} \mathbf{x}$, while $\mathbf{C} \mathbf{x} = \mathbf{y}$. In other words:

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \mathbf{x}^\top \mathbf{M} \mathbf{x}, \\ & \text{subject to} && \mathbf{C} \mathbf{x} = \mathbf{y}. \end{aligned} \tag{3}$$

More concrete:

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 5 & 0 \\ 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, \\ & \text{subject to} && \begin{bmatrix} 1 & 7 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 1. \end{aligned} \tag{4}$$

Quadratic programming

Example - solution using quadprog

We can use a dedicated solver for this class of problems - **quadprog** provided by MATLAB. Here is the solution:

```
0 M = [1 0 1; 0 5 0; 1 0 3];  
   C = [1 7 2];  
2 y = 1;  
  
4 x = quadprog(M, [], [], [], C, y)
```

Average solution time is **0.56** ms

Quadratic programming

Example - solution using CVX

Alternatively we can invoke one of the most powerful convex optimization tools with a user-friendly coding style - CVX:

```
0 M = [1 0 1; 0 5 0; 1 0 3];  
  C = [1 7 2];  
2 y = 1;  
  
4 cvx_begin  
  variables x(3);  
6 minimize( x' * M * x );  
  subject to  
8      C*x == y;  
  cvx_end
```

[Official web page of the CVX package](#)

Quadratic programming

Application

Remember CTC control from the lecture on error dynamics:

$$\begin{cases} \mathbf{v} = \mathbf{H}(\ddot{\mathbf{q}}^* + \mathbf{K}_d \dot{\mathbf{e}} + \mathbf{K}_p \mathbf{e}) + \mathbf{c} \\ \mathbf{T}\mathbf{u} + \mathbf{F}^\top \lambda = \mathbf{v} \end{cases} \quad (5)$$

Assume that $\lambda = [\lambda_1^\top, \dots, \lambda_m^\top]^\top \in \mathbb{R}^{3m}$, $\lambda_i \in \mathbb{R}^3$ and represents a m individual unilateral contact with normal unit-vectors to the contact surface \mathbf{n}_i , and no friction. Then we can write the control law as a solution to the quadratic program:

$$\begin{aligned} & \underset{\mathbf{u}, \mathbf{v}, \lambda_i}{\text{minimize}} && \mathbf{u}^\top \mathbf{u}, \\ & \text{subject to} && \begin{cases} \mathbf{H}(\ddot{\mathbf{q}}^* + \mathbf{K}_d \dot{\mathbf{e}} + \mathbf{K}_p \mathbf{e}) + \mathbf{c} = \mathbf{v} \\ \mathbf{T}\mathbf{u} + \sum_{i=1}^m \mathbf{F}_i^\top \lambda_i = \mathbf{v} \\ -\mathbf{n}_i^\top \lambda_i \leq 0 \end{cases} \end{aligned} \quad (6)$$

Quadratic programming

What it can't do

Remember friction cone constraints:

$$||[\tau_1 \ \tau_2]^\top \mathbf{f}|| \leq \mu \mathbf{n}^\top \mathbf{f} \quad (7)$$

These can't be directly represented in a quadratic program.

Second-order cone programming

General form

The general form of a Second-order cone program (SOCP) is:

$$\begin{array}{ll} \underset{\mathbf{x}}{\text{minimize}} & \mathbf{f}^\top \mathbf{x}, \\ \text{subject to} & \begin{cases} \|\mathbf{A}_i \mathbf{x} + \mathbf{b}_i\|_2 \leq \mathbf{c}_i^\top \mathbf{x} + d_i, \\ \mathbf{F} \mathbf{x} = g. \end{cases} \end{array} \quad (8)$$

QP are a subset of SOCP.

Friction cone and SOCP

We already saw that a no-friction contact can be represented as a QP. Next we will try to show that friction cone (a more general model) can be represented as a SOPC (a more general optimization problem type).

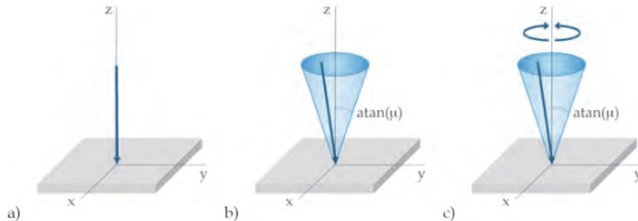


Fig. 2. Contact models commonly used in robotics: a) Point contact without friction; b) Point contact with friction; c) Soft-finger contact

Picture is from *Sancho-Bru, J.L., Pérez-González, A., Mora, M.C., Lón, B.E., Vergara, M., Iserte, J.L., Rodríguez-Cervantes, P.J. and Morales, A., 2011. Towards a realistic and self-contained biomechanical model of the hand. In Theoretical biomechanics. IntechOpen.*

Friction cone and SOCP

Formulation

Coming back to the the friction cone general view:

$$||[\tau_1 \ \tau_2]^\top \mathbf{f}|| \leq \mu \mathbf{n}^\top \mathbf{f} \quad (9)$$

This is has the form of a conic constraint, which is admissible for SOCP:

$$||\mathbf{A}_i \mathbf{x} + \mathbf{b}_i||_2 \leq \mathbf{c}_i^\top \mathbf{x} + d_i \quad (10)$$

Friction cone and SOCP

Application

Consider this problem again:

$$\begin{cases} \mathbf{v} = \mathbf{H}(\ddot{\mathbf{q}}^* + \mathbf{K}_d \dot{\mathbf{e}} + \mathbf{K}_p \mathbf{e}) + \mathbf{c} \\ \mathbf{T}\mathbf{u} + \mathbf{F}^\top \lambda = \mathbf{v} \end{cases} \quad (11)$$

But this time we require that:

$$||[\tau_{i,1} \ \tau_{i,2}]^\top \lambda_i|| \leq \mu \mathbf{n}_i^\top \lambda_i \quad (12)$$

Friction cone and SOCP

Application, part 2

Solution can take the form of a second-order cone program:

$$\begin{array}{ll} \underset{\mathbf{u}, \mathbf{v}, \lambda_i}{\text{minimize}} & \mathbf{u}^\top \mathbf{u}, \\ \text{subject to} & \begin{cases} \mathbf{H}(\ddot{\mathbf{q}}^* + \mathbf{K}_d \dot{\mathbf{e}} + \mathbf{K}_p \mathbf{e}) + \mathbf{c} = \mathbf{v} \\ \mathbf{T}\mathbf{u} + \sum_{i=1}^m \mathbf{F}_i^\top \lambda_i = \mathbf{v} \\ ||[\tau_{i,1} \ \tau_{i,2}]^\top \lambda_i|| \leq \mu \mathbf{n}_i^\top \lambda_i \end{cases} \end{array} \quad (13)$$

You can read and watch more at:

- [CVX user guide](#)
- [quadprog - MATLAB documentation](#)
- [Computational Intelligence 2020, Lecture 5 \(Least Squares and Quadratic Programming\)](#)
- [Computational Intelligence 2020, Lecture 6 \(Domain, Convex Domains\)](#)
- [Computational Intelligence 2020, Lecture 7 \(Linear inequality representation of convex domains\)](#)
- [Computational Intelligence 2020, Lecture 8 \(QCQP, SOCP\)](#)

Lecture slides are available via Moodle.

You can help improve these slides at:

github.com/SergeiSa/Contact-Aware-Control-Slides-Fall-2020

Check Moodle for additional links, videos, textbook suggestions.