

# Optimal Control and Optimization in Robotics

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# Introduction

My internship is co-supervised by Justin Carpentier (Willow, Inria Paris) and Nicolas Mansard (Gepetto, LAAS/CNRS) in the Willow research group at INRIA Paris in France.



**Figure 1:** Justin Carpentier



**Figure 2:** Nicolas Mansard



L'Institut national de recherche en informatique et en automatique (Inria) est un établissement public à caractère scientifique et technologique français spécialisé en mathématiques et informatique, placé sous la double tutelle du ministère de l'Enseignement supérieur, de la Recherche et de l'Innovation et du ministère de l'Économie et des Finances créé le 3 janvier 1967 dans le cadre du « plan calcul ».

**WILLOW**

*Computer vision and machine learning research laboratory*



**WILLOW** is based in the Laboratoire d'Informatique de l'École Normale Supérieure (CNRS/ENS/INRIA UMR 8548) and is a joint research team between INRIA Rocquencourt, École Normale Supérieure de Paris and Centre National de la Recherche Scientifique.

Their research is concerned with representational issues in visual object recognition and scene understanding.

# Motivations and Problems in a general context

I want my robot to move one thing somewhere and pass some point at some moment with a lowest cost.

**Optimization:** lowest cost

**Control:** from some point to another point  
and *Constraints*

## Optimal Control Problem

- Goal 1: Controllability

- Goal 2: Optimal Control

## Transformation of the problem

- Adding penalty

- Discretization

## Differential Dynamic Programming

- Dynamic Programming

- Linear Quadratic Regulator (LQR)

- Backward and Forward pass

- Roll-out

# Optimal Control Problem

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# Optimal Control Problem

## Goal 1: Controllability

$$\begin{array}{ll}\text{find} & u \\ \text{subject to} & x(0) = x_0, \quad x(T) = p, \\ & \dot{x}(t) = f(x(t), u(t)).\end{array}\tag{1}$$

## Goal 2: Optimal Control

$$\begin{array}{ll}\text{minimize}_{u} & \int_0^T l(x(t), u(t)) dt \\ \text{subject to} & x(0) = x_0, \quad x(T) = p, \\ & \dot{x}(t) = f(x(t), u(t)).\end{array}\tag{2}$$



# Transformation of the problem

Adding penalty to the terminal lost:

$$\begin{aligned} & \underset{u}{\text{minimize}} && \int_{[0, T[} l(x(t), u(t)) dt + l_T(x(T)) \\ & \text{subject to} && x(0) = x_0, \\ & && \dot{x}(t) = f(x(t), u(t)). \end{aligned} \tag{3}$$

Discretization of functions and variables:

$$\begin{aligned} & \underset{x \in \ell_{N+1}^{\infty}, u \in \ell_N^{\infty}}{\text{minimize}} && J(x, u) = \sum_{i=0}^{N-1} L(x_i, u_i) + L_T(x_N) \\ & \text{subject to} && x(0) = x_0, \\ & && x_{i+1} = F(x_i, u_i) \quad \forall i \in [0..N-1] \end{aligned} \tag{4}$$

# Differential Dynamic Programming

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# Dynamic Programming

Optimize one by one:

$$\min_U J(U) = \min_{u_0} \min_{u_1} \dots \min_{u_{N-1}} J(U) \quad (5)$$

Definitions of Value Function and Q-functions:

$$V_i(x_i) = \min_{u_i} L(x_i, u_i) + V_{i+1}(x_{i+1}) \quad (6)$$

$$V_N(x_N) = L_T(x_N)$$

$$\begin{aligned} Q_i(x_i, u_i) &= L(x_i, u_i) + V_{i+1}(x_{i+1}) \\ &= L(x_i, u_i) + V_{i+1}(f(x_i, u_i)) \\ &= L(x_i, u_i) + \min_{u_{i+1}} Q_{i+1}(f(x_i, u_i), u_{i+1}), \quad i \leq N-2 \end{aligned} \quad (7)$$

$$V_i(x_i) = \min_{u_i} Q_i(x_i, u_i) \quad (8)$$

# Linear Quadratic Regulator (LQR)

The LQR is an algorithm that solves the problem 4 *in one iteration* in case  $L, L_T$  are **quadratic** and  $F$  is **linear** (or quadratic).

Two phases in the algorithm:

- **Backward and Forward pass:**

Compute  $V_x, V_{xx}$  and  $Q_u, Q_{uu}, Q_{ux}$  alternately

- **Roll-out:**

Compute  $\delta_u^*$  by  $V_x, V_{xx}$  and  $Q_u, Q_{uu}, Q_{ux}$

## Backward and Forward pass

**Backward pass:** from the partial derivatives of  $V_{i+1}$  back to the partial derivatives of  $Q_i$

$$\begin{aligned}Q_u &= L_u + \underline{V'_x} F_u \quad ^1 \\Q_{uu} &= L_{uu} + \underline{V'_x} \cdot F_{uu} + F_u^T \underline{V'_{xx}} F_u \\Q_{ux} &= L_{ux} + \underline{V'_x} \cdot F_{ux} + F_u^T \underline{V'_{xx}} F_x\end{aligned}\tag{9}$$

**Forward pass:** from the partial derivatives of  $Q_i$  back to the partial derivatives of  $V_i$

$$\begin{aligned}V_x &= Q_x - Q_u Q_{uu}^{-1} Q_{ux} = Q_x + Q_u K \\V_{xx} &= Q_{xx} - Q_{xu} Q_{uu}^{-1} Q_{ux} = Q_{xx} + Q_{xu} K\end{aligned}\tag{10}$$

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<sup>1</sup>" $Q, V'$ " mean " $Q_i, V_{i+1}$ ".

**Roll-out:** determine the new trajectory  $x, u$  by computing the best control change  $\delta_u$ .

$$\begin{aligned}\delta_u^*(i) &:= \arg \min_{\delta_u(i)} Q_i(x_i + \delta_x(i), u_i + \delta_u(i)) \\ \delta_u^* &= -Q_{uu}^{-1}(Q_u + Q_{ux}\delta_x) \quad \forall i \in [0..N-1]\end{aligned}\tag{11}$$

$$\begin{aligned}u^* &:= u + \delta_u^*, \\ x_0^* &= x_0, \\ x_{i+1}^* &= F(x_i^*, u_i^*) \quad \forall i \in [0..N-1]\end{aligned}\tag{12}$$

# Sequential Quadratic Programming (SQP)

What if the  $F, L$  are not quadratic? Use the Sequential Quadratic Programming to solve general non-linear problems.

$$\begin{aligned} & \underset{x}{\text{minimize}} && f(x) \\ & \text{subject to} && c(x) = 0 \end{aligned} \tag{13}$$

$$A(x)^T = [\nabla c_1(x), \nabla c_2(x), \dots, \nabla c_m(x)] \tag{14}$$

Goal: KKT necessary conditions

$$F(x, \lambda) = \begin{bmatrix} \nabla f(x) - A(x)^T \lambda \\ c(x) \end{bmatrix} = 0 \tag{15}$$

Resolve the quadraticalized problems sequentially to find an optimal step and update the trajectory with a roll-out policy (like line search).    Partie technique, voir la chapitre 18 de [1]



Jorge Nocedal and Stephen J. Wright.

*Numerical Optimization.*

Springer, 2006.