

# Midterm presentation

## The Quadrocopters

Technische Universität München

27. Mai 2015

# Overview

- 1 Motivation
- 2 Model
- 3 Realtime Optimization Approach
- 4 Results

# Optimal Control Problem

$$\min_{x,u} J(x, u) \quad \dot{x} = f(x, u)$$

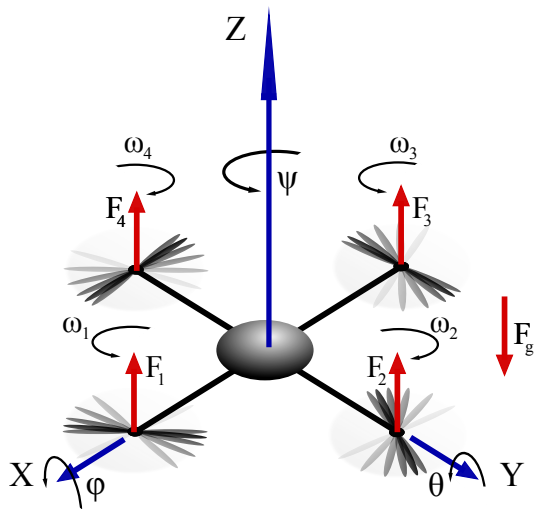
$x$  : state

$u$  : control

→ additional difficulty: realtime approach

# Model

# Forces and Torques



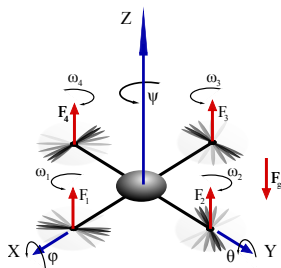
# Newton-Euler Equations

Forces

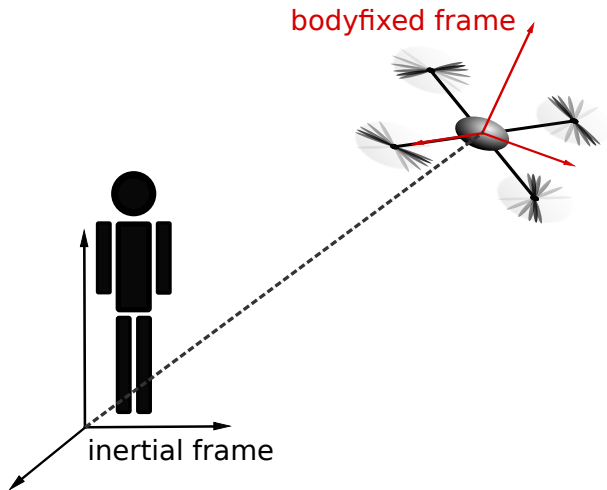
$$F_{\text{ext}} = F_g + \sum_{i=1}^4 F_i$$

Torques

$$\tau_{\text{ext}} = \sum_{i=1}^4 \tau_i + (\tau_\varphi + \tau_\theta)$$



# Coordinate Systems



# Quaternions

$$q = a + ib + jc + kd \quad a, b, c, d \in \mathbb{R}$$



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problem  $\rightarrow \|q\| = 1$  additional constraint

# Dynamics

Equations representing dynamics...

$$T(x, u) = M \cdot \begin{pmatrix} \dot{x}_8 \\ \vdots \\ \dot{x}_{13} \end{pmatrix} + \Theta(x)$$

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$$T(x, u) = M \cdot \begin{pmatrix} \dot{x}_8 \\ \vdots \\ \dot{x}_{13} \end{pmatrix} + \Theta(x)$$

...expressed as system of differential equations:

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ \vdots \\ x_7 \\ x_8 \\ \vdots \\ x_{13} \end{pmatrix} = \begin{pmatrix} \dot{x}_1 \\ \vdots \\ \dot{x}_7 \\ M^{-1}(T(x, u) - \Theta(x)) \end{pmatrix}$$

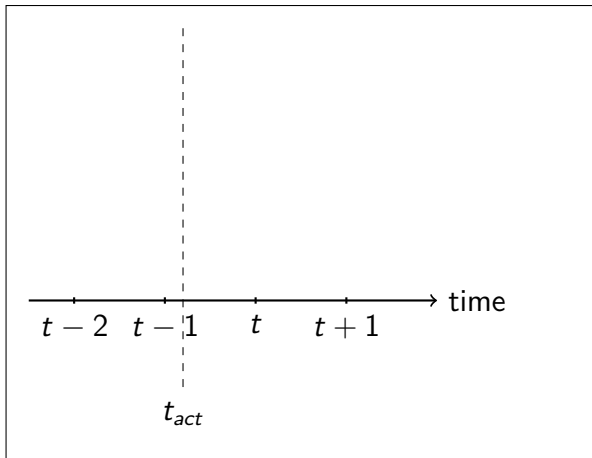
Refinement of the model

→ wind

→ aerodynamical forces

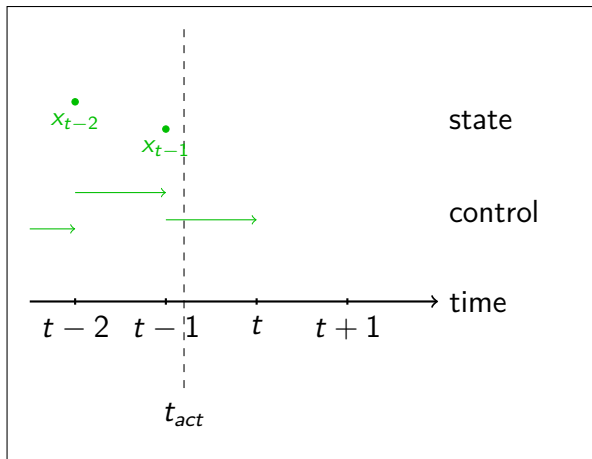
# Realtime Optimization Approach

# Setting

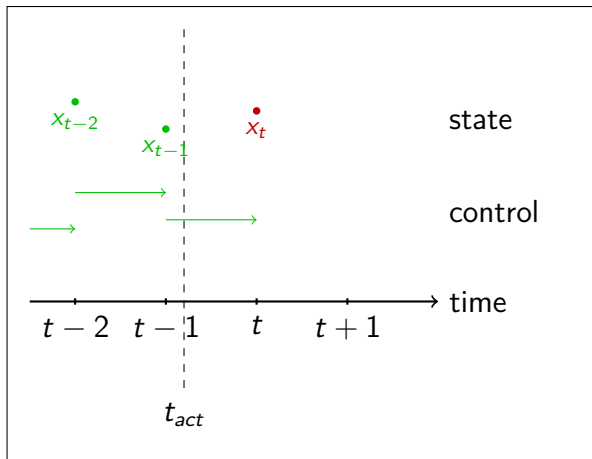




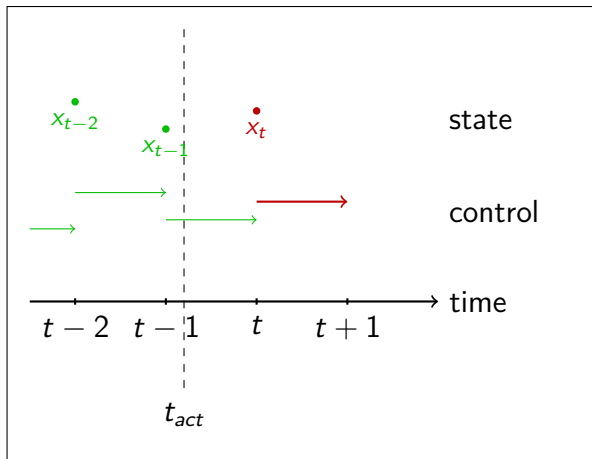
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# Minimization Problem

$$\min_{\substack{s_t, \dots, s_N \\ q_t, \dots, q_{N-1}}} \sum_{i=t}^{N-1} J_i(s_i, q_i) \quad \text{s.t.} \quad \begin{cases} x_t - s_t = 0 \\ h_i(s_i, q_i) - s_{i+1} = 0 \quad \forall i = t, \dots, N-1 \end{cases}$$

$J_i(s_i, q_i)$       discretized goal function

$x_t - s_t = 0$       expected state = real state

$h_i(s_i, q_i)$       solution of the ODE at time  $i$

# The Lagrangian

$$L^t(y) = \sum_{i=t}^{N-1} J_i(s_i, q_i) + \lambda_t^T (x_t - s_t) + \sum_{i=t}^{N-1} \lambda_{i+1}^T (h_i(s_i, q_i) - s_{i+1})$$

$$y := (\lambda_t, s_t, q_t, \lambda_{t+1}, s_{t+1}, q_{t+1}, \dots, \lambda_N, s_N)$$

We are looking for  $y^*$  satisfying the KKT conditions:

$$\Rightarrow \nabla_y L^t(y^*) = 0$$

# The SQP Method

How to find  $y^*$ ?

$$y_{k+1} = y_k + \alpha_k \Delta y_k$$
$$\min_{\Delta y_k} \frac{1}{2} \Delta y_k^T A_k \Delta y_k + \nabla_{y_k} J(y_k)^T \Delta y_k$$

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Special case:

$$A_k = H(y_k) \quad \text{approximated Hessian } \nabla_{y_k}^2 L(y_k)$$

$$\alpha_k = 1$$

# Newton-Raphson

Find  $\Delta y_k$  with:

$$\nabla_{y_k} L(y_k) + H(y_k) \Delta y_k = 0$$



# Newton-Raphson

Find  $\Delta y_k$  with:

$$\nabla_{y_k} L(y_k) + H(y_k) \Delta y_k = 0$$

There is only time for 1 SQP step:

$$y_{t+1} = y_1 = y_0 + \Delta y_0$$

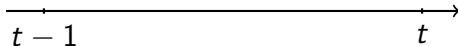
# Riccati Recursion

Approximated Hessian:

$$H^t(y^t) = \begin{pmatrix} -E & & & & & & & & & \\ & -E & Q_t^H & M_t^H & A_t^T & & & & & \\ & (M_t^T)^H & R_t^H & B_t^T & & & & & & \\ & A_t & B_t & & -E & & & & & \\ & & & -E & Q_{t+1}^H & M_{t+1}^H & A_{t+1}^T & & & \\ & & & (M_{t+1}^T)^H & R_{t+1}^H & B_{t+1}^T & & & & \\ & & & A_{t+1} & B_{t+1} & & & & & \\ & & & & & & \ddots & & & \\ & & & & & & & \ddots & & \\ & & & & & & & & Q_{N-1}^H & M_{N-1}^H & A_{N-1}^T \\ & & & & & & & & (M_{N-1}^T)^H & R_{N-1}^H & B_{N-1}^T \\ & & & & & & & & A_{N-1} & B_{N-1} & -E \\ & & & & & & & & & -E & Q_N^H \end{pmatrix}$$

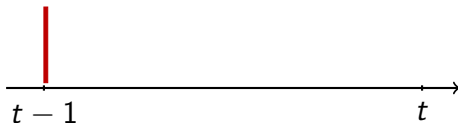
# Summary

What happens in interval  $[t - 1, t]$  ?



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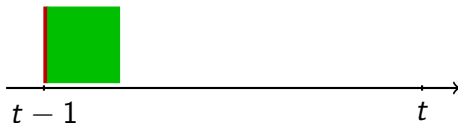
What happens in interval  $[t - 1, t]$  ?



- 1 calculate control  $u_{t-1}$  (Riccati Part II)

# Summary

What happens in interval  $[t - 1, t]$  ?



- 1 calculate control  $u_{t-1}$  (Riccati Part II)
- 2 calculate  $y$  (Riccati Part II)

# Summary

What happens in interval  $[t-1, t]$  ?



- ① calculate control  $u_{t-1}$  (Riccati Part II)
- ② calculate  $y$  (Riccati Part II)
- ③ prepare  $u_t$  (Newton & Riccati Part I)

# Finite Horizon

$$\min_{\substack{s_t, \dots, s_N \\ q_t, \dots, q_{N-1}}} \sum_{i=t}^{N-1} J_i(s_i, q_i) \quad s.t. \quad \begin{cases} x_t - s_t = 0 \\ h_i(s_i, q_i) - s_{i+1} = 0 \quad \forall i = t, \dots, N-1 \end{cases}$$

How to choose  $N$ ?

$N = t_{end} \rightarrow$  problem size decreasing

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How to choose  $N$ ?

$N = t_{end} \quad \rightarrow$  problem size decreasing

$N = t + n \quad \rightarrow$  problem size constant



# Results

Questions?