

# Time-optimal control of cranes subject to container height constraints

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## Summary

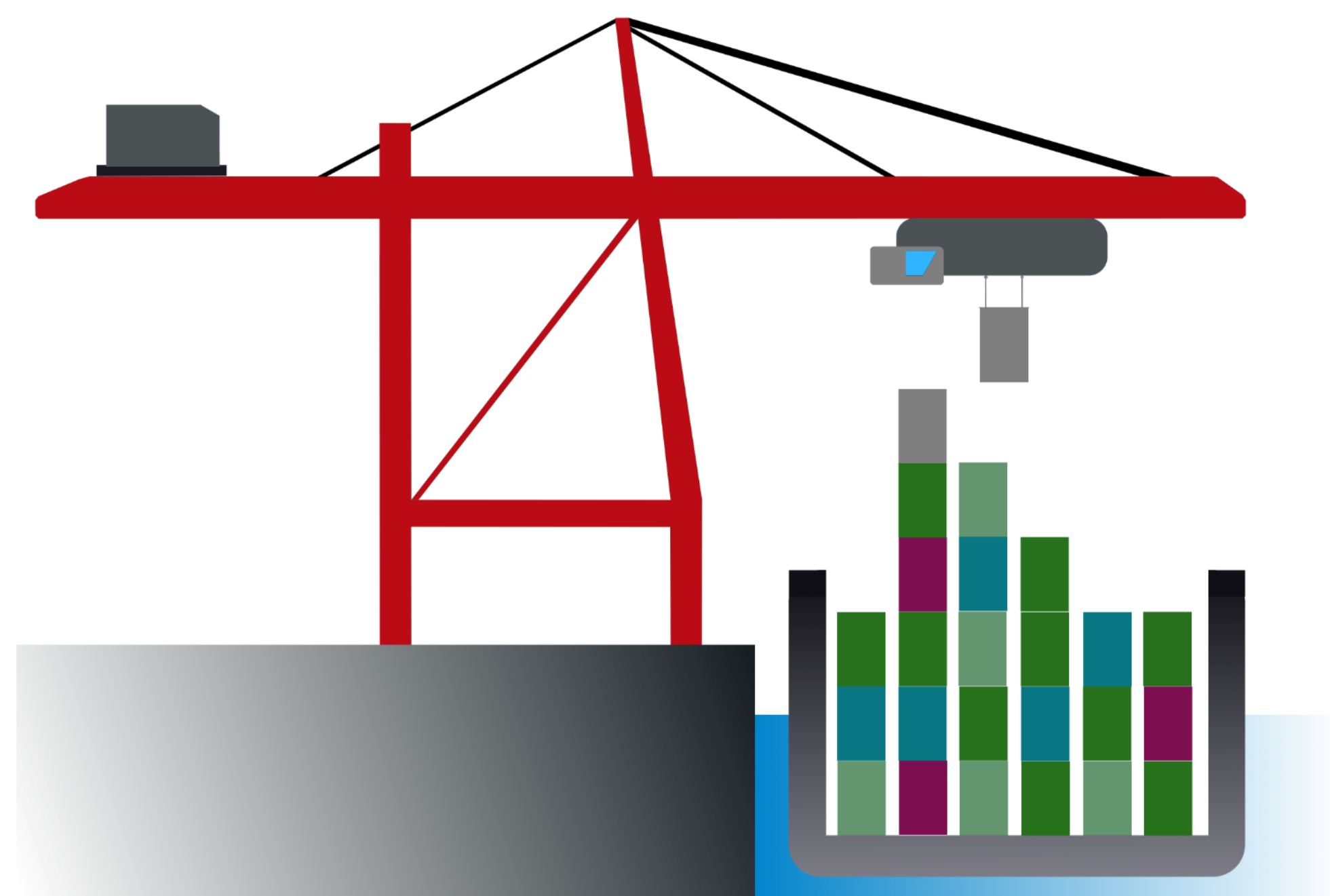
**Objective:** Move the payload into and onto a container ship in minimum time.

**Challenge:** Avoid collision with container stacks.

**Trick:** Variable change in an optimal control problem.

**Outcome:** Non-convex container avoidance constraints become linear bound constraints.

**Benefit:** No functional representation of the container stack heights is required.



## Original problem formulation

The nonlinear state-space representation in the original form is

$$\dot{x}(t) = f(t, x(t), u(t)), \quad (1)$$

where the state variables are

$$\begin{aligned} x_1 &= x_p, & x_3 &= y_p, & x_5 &= l, & x_7 &= \theta, \\ x_2 &= \dot{x}_p, & x_4 &= \dot{y}_p, & x_6 &= \dot{l}, & x_8 &= \dot{\theta}, \end{aligned} \quad (2)$$

and the original time-optimal control formulation is written as

$$\begin{aligned} &\text{minimize } T = \int_0^{t_f} 1 \, dt \\ &\text{subject to } \dot{x}(t) = f(t, x(t), u(t)) \\ &\quad 0 \leq y_p(t) \leq h - s(x_p(t)) \leftarrow \text{avoidance constraints} \\ &\quad \vdots \\ &\quad \text{other constraints} \end{aligned} \quad (3)$$

## Problem reformulation

1. Use spatial derivatives instead of temporal derivatives

$$\dot{x}_1 = \frac{dx_1}{dt} = x_2 \implies \frac{dt}{dx_1} = \frac{1}{x_2}, \quad \frac{dx_2}{dx_1} = \dots \quad (4)$$

2. New state vector is  $x = [t, \dot{x}_p, y_p, \dot{y}_p, l, \dot{l}, \theta, \dot{\theta}]^T$ .

3. With  $x_1(x_p) = t(x_p)$  as the cost function

$$\begin{aligned} &\text{minimize } J = t(x_{p_f}) \\ &\text{subject to } x_2 \dot{x}(x_p) = f(x_p, x(x_p), u(x_p)) \\ &\quad 0 \leq y_p(x_p) \leq h - s(x_p) \leftarrow \text{container constraints} \\ &\quad \vdots \\ &\quad \text{other constraints} \end{aligned} \quad (5)$$

## Geometric constraints

1. Time discretization of the container avoidance constraint in (3) leads to

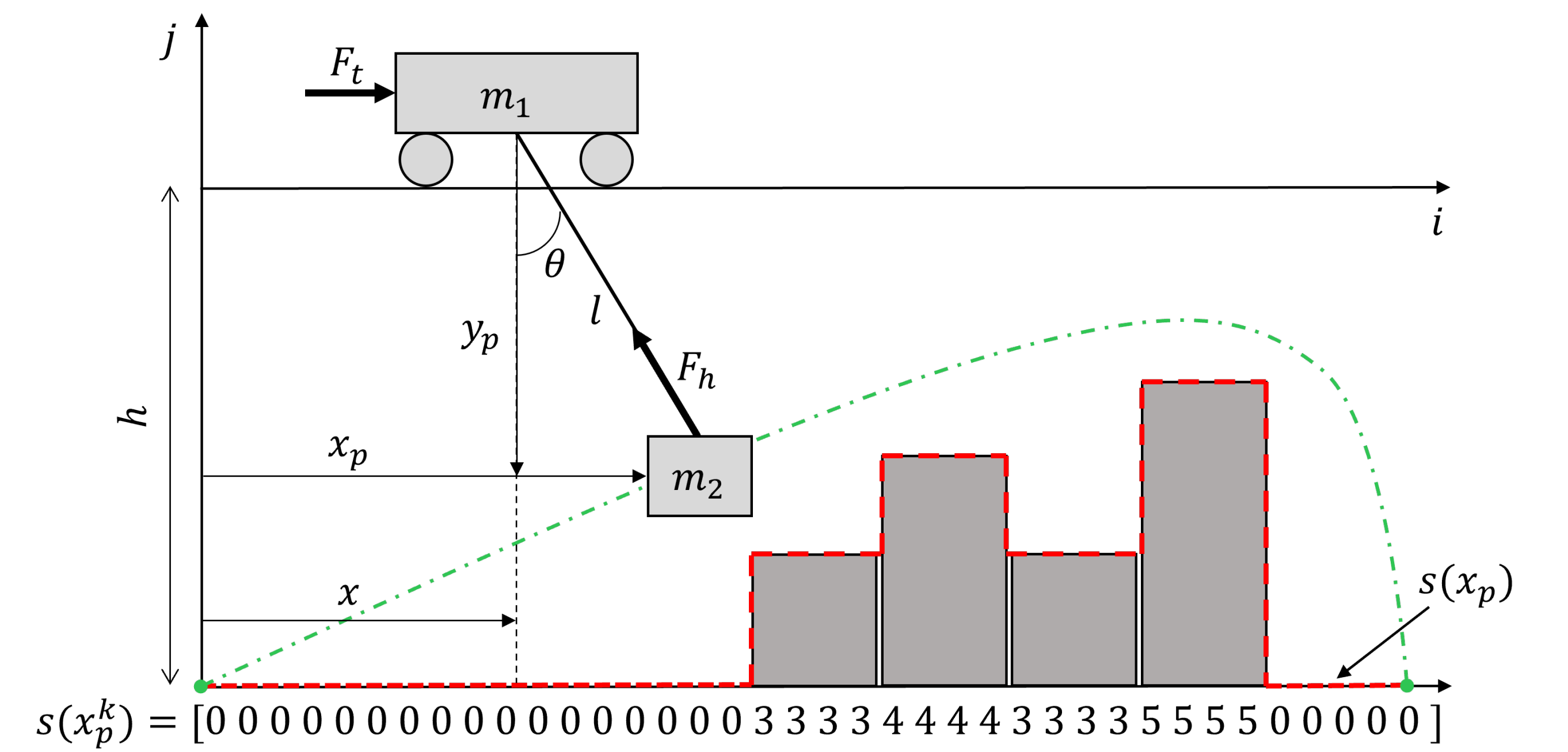
$$0 \leq y_p(t^k) \leq h - s(x_p(t^k)), \quad \text{✗}$$

where the container profile  $s(x_p)$  is generally discontinuous, nonlinear and non-convex.

2. Spatial discretization of the container avoidance constraint in (5) leads to upper bound constraints for  $y_p(x_p)$ .

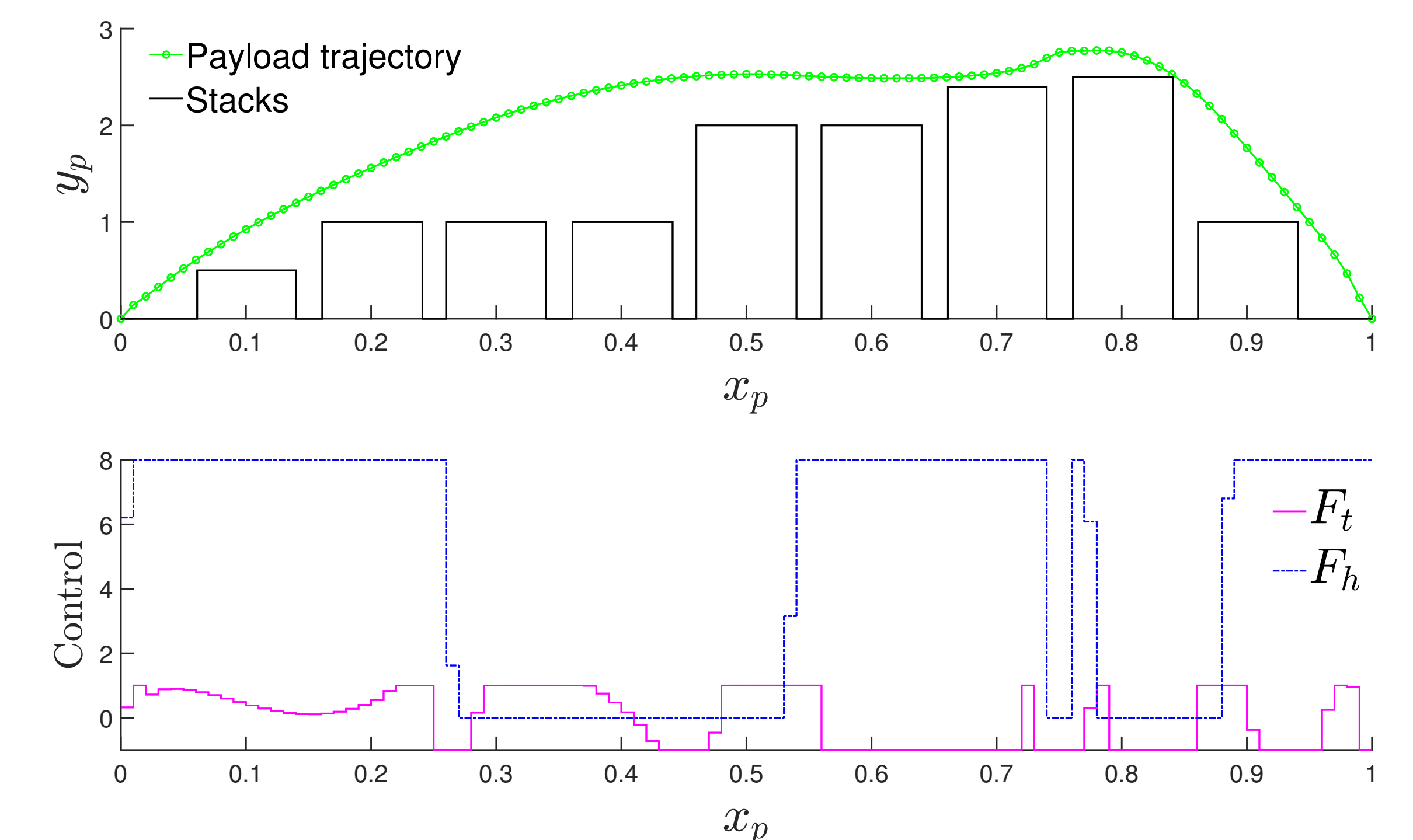
$$0 \leq y_p(x_p^k) \leq h - s(x_p^k). \quad \text{✓}$$

Note that we no longer need an explicit function  $s(x_p)$ , but simply function values which can be computed when setting up the numerical model.



## Simulation example

To illustrate and validate the idea, a small scale scenario of stack configuration was simulated.



## Future work

- Investigate energy consumption and energy optimal solutions.
- Apply the method in closed loop.
- Go beyond the point-mass assumption.
- More physical and geometric constraints to the setup.