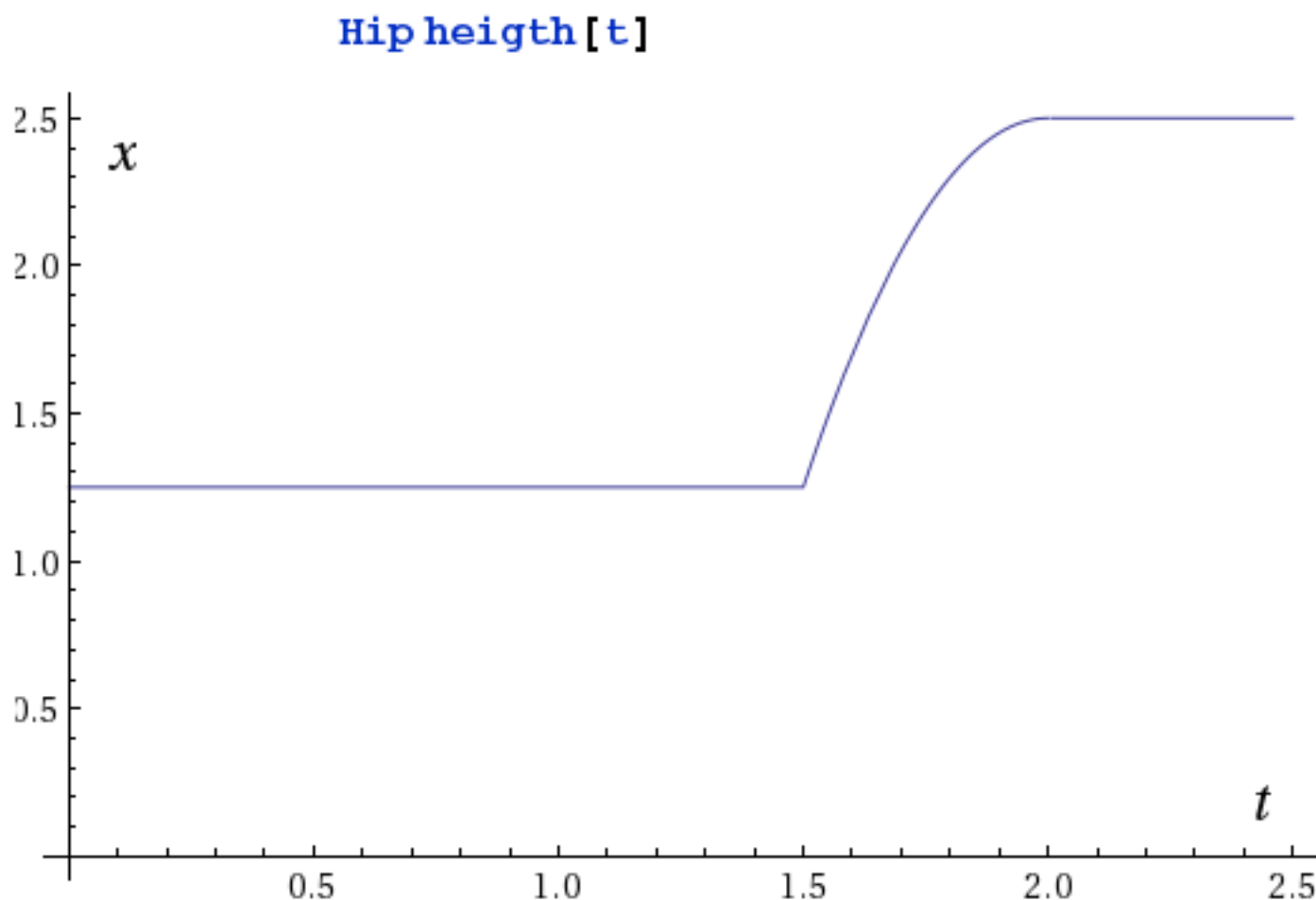


# Exercise : Impulse optimization

This exercise deals with a simple optimization problem :

the preparation and optimal impulse in a one - dimensional vertical movement , corresponding to a vertical jump from a standing position , where the aim is to reach a fixed bar some distance above .

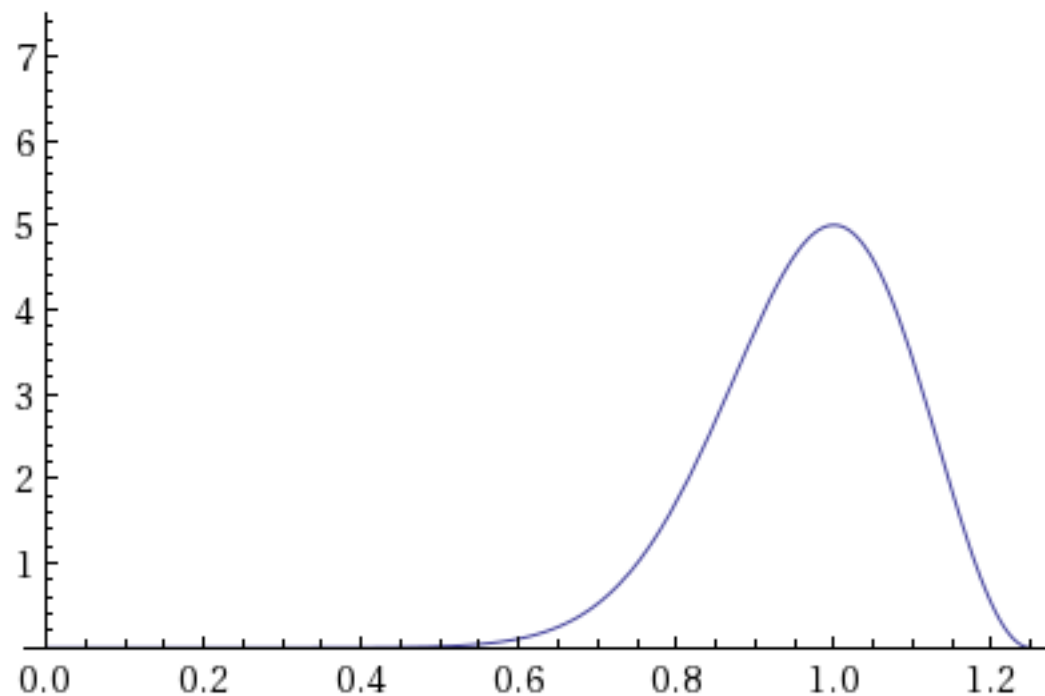
The movement must follow the guiding movement indicated in the figure as closely as possible . The vertical axis  $x$  represents the hip height of person executing the jump , the horizontal axis representing time .



**Fig 1**

We will use a model for the maximum vertical impulse force, which states that when the person is in contact with the ground the maximum vertical force is given by

$$f_{\max}[x] = F_0 \left( \frac{L - x}{x_c} \right)^2 \text{Exp} \left[ 1 - ((L - x) / x_c)^2 \right]$$



The maximum corresponds to  $x = L - x_c$  ( $f_{\max}[x = L - x_c] = F_0$ )

and diminishes when the legs are progressively extended  $x \rightarrow L = 1.25$

or totally flexed  $x \rightarrow 0.5$

Data :

The model data is  $\{L = 1.25, x_c = 0.25, F_0 = 75, g = 10\}$ .

The impulse force can take any value  $f \in [0, f_{\max}[x]]$  and the resulting acceleration is  $\alpha = f - g$  :

(we use as unit of mass the jumping person mass, so that  $M = 1$ )

If the jumping player tries to apply a vertical force  $f$ , the physical model will filter out unphysical values. Large positive values are limited by the mechanical limits modeled via  $f_{\max}[x]$ , whilst negative values are simply not possible. The resulting vertical acceleration  $\alpha$  in response to a desired force  $f$  is then  $\alpha_{\text{Phys}}[f, x] = f_{\text{Phys}} - g$ , given by the constraints

$$\alpha_{\text{Phys}}[f, x] = \begin{cases} -g & \text{if } f \leq 0 \\ f_{\max}[x] - g & \text{if } x < L \ \&\& \ f > f_{\max}[x] \\ -g & \text{if } x > L \\ f - g & \text{Otherwise (unconstrained case)} \end{cases}$$

Given the above restrictions, we want to follow as closely as possible the guiding movement represented in fig. 1

and described by Guiding Movement :

$$xg[t] = \begin{cases} L = 1.25 & t < 1.5 \\ 2.5 - 5(2 - t)^2 & 1.5 \leq t \leq 2 \\ h = 2.5 & t > T = 2 \end{cases}$$

$$vg[t] = \begin{cases} 10(2 - t) & 1.5 < t < 2 \\ 0 & t > T = 2 \end{cases}$$

We want to find out the optimal physical trajectory which :

a) respects the physical limitations for  $\alpha = \alpha_{\text{Phys}}[f, x]$ . Starting from  $x[t = 0] = L$ , it is clear that the physical trajectory will need first to lower  $x$ , then take impulse and finally execute the jump.

b) Minimizes the quantity

$$\begin{aligned} \text{Penalty} = \int_0^T \left( \frac{1}{2} (x[t] - xg[t])^2 + \frac{1}{2} (v[t] - vg[t])^2 + \frac{\gamma}{2} (\alpha[t])^2 \right) dt + \\ + \frac{T}{2} \left( (x[T = 2] - h)^2 + v[T = 2]^2 \right) \end{aligned}$$

and therefore maximizes the payoff

$$\text{Payoff} = - \int_0^T \left( \frac{1}{2} (\mathbf{x}[\mathbf{t}] - \mathbf{x}_g[\mathbf{t}])^2 + \frac{1}{2} (\mathbf{v}[\mathbf{t}] - \mathbf{v}_g[\mathbf{t}])^2 + \frac{\gamma}{2} (\alpha[\mathbf{t}])^2 \right) d\mathbf{t} - \\ - \frac{\Gamma}{2} \left( (\mathbf{x}[\mathbf{T} = 2] - \mathbf{h})^2 + \mathbf{v}[\mathbf{T} = 2]^2 \right)$$

Take $\gamma = 1$ $\Gamma = 250\,000$
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The system is described as follows :

$$\mathbf{x} \equiv \begin{pmatrix} \mathbf{x} \\ \mathbf{v} \end{pmatrix} \quad \dot{\mathbf{x}} = \begin{pmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{v}} \end{pmatrix} = \begin{pmatrix} \mathbf{v} \\ \mathbf{a} \end{pmatrix} \equiv \mathbf{F} \quad (\mathbf{a} = \alpha) ;$$

$$\mathbf{r} = -\frac{1}{2} (\mathbf{x} - \mathbf{x}^g)^2 - \frac{1}{2} (\mathbf{v} - \mathbf{v}^g)^2 - \frac{\gamma}{2} \alpha^2$$

$$\mathbf{p} = \begin{pmatrix} \mathbf{p1} \\ \mathbf{p2} \end{pmatrix}$$

The Hamiltonian control is

$$\mathbf{H} = \mathbf{F} \cdot \mathbf{p} + \mathbf{r} \rightarrow \mathbf{v} \mathbf{p1} + \alpha \mathbf{p2} - \frac{1}{2} (\mathbf{x} - \mathbf{x}^g)^2 - \frac{1}{2} (\mathbf{v} - \mathbf{v}^g)^2 - \frac{\gamma}{2} \alpha^2$$

We know that the optimal  $\alpha$  maximizes  $H$ ,  $\partial_{\alpha} H = 0$  :

$$\partial_{\alpha} H = 0 \quad \rightarrow \quad \boxed{\alpha = p_2 / \gamma \quad f_{\text{imp}} = p_2 / \gamma + g}$$

where  $p_1$  and  $p_2$  are functions of  $t$  whose evolution is dictated by optimization control theory :

$$\boxed{\dot{x}[t] = \partial_p H \quad \dot{p}[t] = -\partial_x H \quad \partial_{\alpha} H = 0}$$

so that the optimal evolution is

$$\dot{p}_1 = -\frac{\partial H}{\partial x} = x - x^g \quad \dot{x} = v$$

$$\dot{p}_2 = -\frac{\partial H}{\partial v} = -p_1 + (v - v^g) \quad \dot{v} = \alpha$$

$$\alpha = p_2 / \gamma \quad f = \alpha + g \quad \rightarrow \quad \alpha = \alpha_{\text{Phys}}[f, x] \text{ due to}$$

the model constrains.

MC Procedure : consider as optimization parameters  $\{t_0, \lambda_1, \lambda_2\}$ , with some initial unknown values for  $\lambda_1 = p_1[0]$ ,  $\lambda_2 = p_2[0]$ , and  $t_0$  being the instant when the player starts de movement ( $t_0 \in [0, 1.5]$ ) and  $\lambda_1 = p_1[t_0]$ ,  $\lambda_2 = p_2[t_0]$  the initial values for  $\{p_1, p_2\}$ . The initial values for  $\{x, v\}$  are known to be  $\{L, 0\}$ .

\* Use a time step such that the interval  $t_0 < t < t_f = T = 2$  is divided in (say  $N_b = 1000$ ) bins and use this time step both to solve the differential equations describing the dynamics, and to compute the integral involved in the Penalty function as a simple discrete sum.

\* Find a reasonable set of parameters  $\{t_0, \lambda_1, \lambda_2\}$  such that the trajectory has the three phases (initially at rest, impulsing phase, and jump). In order to do this estimation, consider that the optimal values should provide  $x[t] \sim x^g[t]$  and  $v[t] \sim v^g[t]$  and assuming this is the case,  $\alpha[t] = p_2[t] / \gamma$  depends solely on  $\{\lambda_1, \lambda_2\}$ . For the 3 phases of the movement to exist, the sign and approximate value of  $\{t_0, \lambda_1, \lambda_2\}$  can be guessed.

\* Using the function `Penalty[t0, λ1, λ2]`,  
write a simple Monte Carlo code which generates random trials

$$\{t_0, \lambda_1, \lambda_2\}' \leftarrow \{t_0, \lambda_1, \lambda_2\} + \{\Delta t_0, \Delta \lambda_1, \Delta \lambda_2\}$$

and updates the values only if

$$\text{Penalty}[t_0, \lambda_1, \lambda_2]' < \text{Penalty}[t_0, \lambda_1, \lambda_2]$$

and reject the trial otherwise.

Report the results of the optimal trajectory :

- \* Physical trajectory vs. guiding trajectory
- \* Minimum Penalty value found
- \* Optionally : discuss whether the optimization hamiltonian is constant ( Non constancy being an indication that the constraints are acting, preventing a further maximization of the payoff functional).