

given  $\min_u c(x, u) \text{ s.t. } x(s+1) = F(s, x(s), u(s))$   
 (DT-OCF)  $c(x, u) = l(t, x(t)) + \sum_{s=0}^{t-1} \mathcal{L}(s, x(s), u(s))$

there are 2 main interpretations / strategies

→ show that (DT-OCF) can be rewritten as

$$\min_{\bar{u} \in \mathbb{R}^m} \bar{c}(\bar{x}, \bar{u}) \text{ s.t. } \bar{f}(\bar{x}, \bar{u}) = 0$$

(i.e. determine  $\bar{u}, \bar{x}, \bar{c}, \bar{f}$ )

— recalling that  $(u: [0, t] \rightarrow \mathbb{R}^m) \in (\mathbb{R}^m)^{[0, t-1]}$ ,

let  $\bar{u} \in \mathbb{R}^{t \cdot m}$  be  $[\bar{u}]_s = u(s)$ ,

and similarly  $[\bar{x}]_s = x^u(s)$

where  $x^u$  is the traj generated by  $u$

— then  $\bar{c}(\bar{x}, \bar{u}) = c(x, u)$ ,

$$\bar{f}(\bar{x}, \bar{u}) = 0 \Leftrightarrow x(s+1) = F(s, x(s), u(s))$$

$$\bar{c}(\bar{x}, \bar{u}) = c(\psi_x^{-1}(\bar{x}), \psi_u^{-1}(\bar{u}))$$

there is a "natural" correspondence between discrete-time signals

$$z: [0, t) \rightarrow \mathbb{R}^d$$

and "stacked" / block vectors

$$\bar{z} = \begin{bmatrix} z(0) \\ z(1) \\ z(2) \\ \vdots \\ z(t-1) \end{bmatrix} \in \mathbb{R}^{t \cdot d}$$

$$\bar{z} = [z(0), z(1), \dots, z(t-1)] \in \mathbb{R}^{d \times t}$$

$$\{z: [0, t) \rightarrow \mathbb{R}^d\} = \mathcal{Z} \xrightleftharpoons[\psi^{-1}]{\psi} \bar{\mathcal{Z}} = \{\bar{z} \in \mathbb{R}^{t \cdot d}\}$$

1°. dynamic programming  $\xleftrightarrow[\text{problems / strategies}]{\text{spectrum of}}$  2°. trajectory optimization

interp: output will be an optimal control policy  $u^*: [0, t) \times X \rightarrow \mathcal{U}$  that tells us the optimal action  $u^*(\tau, x(\tau))$  for every time  $\tau$  and state  $x(\tau)$

• output will be an optimal control signal  $u_{x_0}^*: [0, t) \rightarrow \mathcal{U}$  that tells us the optimal action  $u_{x_0}^*(\tau)$  for every time  $\tau$  from initial state  $x(0) = x_0$

$u(t, x(t))$  for every  
time  $t$  and state  $x(t)$

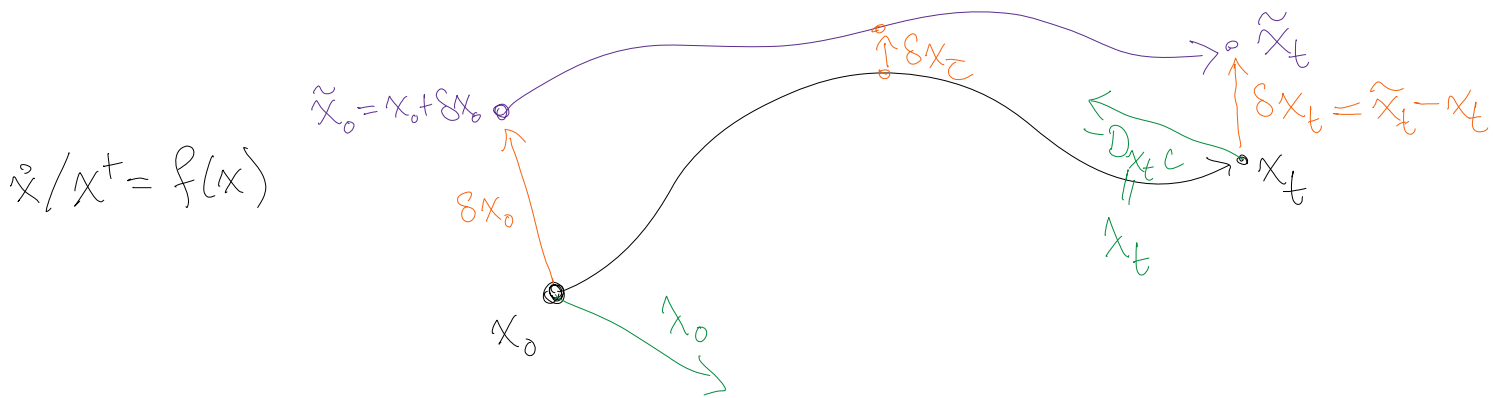
stat: solve a sequence of parametrized  
NLP backward in time

time  $\tau$  from initial state  $x(0) = x_0$

• solve one NLP "forward"  
in time

\* since these are working from same  
underlying problem, they're consistent.

$u^*(\tau, x^*(\tau)) = u_{x_0}^*(\tau)$   
assuming  $x^*(0) = x_0$  and  
 $u_{x_0}^*$  is applied until  $\tau$



$$\dot{x}/x^+ = f(x)$$

$$* \quad \delta \dot{x}_\tau / \delta x_\tau^+ \simeq A(\tau) \cdot \delta x_\tau, \quad A(\tau) = D_x f(x(\tau))$$

$$\Rightarrow \delta x_\tau = \Phi(\tau, 0) \delta x_0$$

$$\hookrightarrow \dot{\lambda}_\tau / \lambda_\tau = \lambda_\tau \cdot A(\tau)$$