

# Turnpike behaviour for systems that are partially uncontrollable

Martin Lazar  
University of Dubrovnik

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UNIVERSITY OF DUBROVNIK

## The problem framework

We analyse the control system

$$\begin{aligned}x'(t) + Ax(t) &= Bu(t) \\ x(0) &= x_0,\end{aligned}$$

accompanied by the following optimization problem

$$\min_u J_T(u) = \min_u \frac{1}{2} \int_0^T (|u(t) - u_d|_U^2 + |Cx(t) - z_d|_Z^2) dt + p_d \cdot x(T), \quad (\mathcal{P}_T)$$

- $A$  a (possibly) unbounded operator on  $X$
- $u \in L_{\text{loc}}^2([0, \infty); U)$  control
- $C \in L(X, Z)$  is observation operator
- $X, U, Z$  Hilbert spaces
- $B \in \mathcal{L}(U, X)$  control operator
- $u_d$  and  $z_d$  are **time independent** desirable control and observation
- $p_d \in X$  a linear regularization of the final state

## Turnpike

Under suitable conditions:

- observability and controllability (or detectability and stabilizability),
- dissipativity

## Turnpike - another approach

The optimality system for the problem  $(\mathcal{P}_T)$  reads

$$\begin{aligned}x_T'(t) + Ax_T(t) &= -B(B^* p_T(t) - u_d) & -p_T'(t) + A^* p_T(t) &= C^*(Cx_T(t) - z_d) \\x_T(0) &= x_0 & p_T(T) &= p_d,\end{aligned}$$

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and the optimal control is given by

$$u_T = -B^* p_T + u_d.$$

The corresponding stationary problem

$$\min_{u \in U} J_s(u) = \min \left\{ \frac{1}{2} \left( |u - u_d|_U^2 + |Cx - z_d|_Z^2 \right) \mid Ax = Bu \right\}$$

Its optimality system reads

$$A\bar{x} = -B(B^* \bar{p} - u_d) \quad A^* \bar{p} = C^*(C\bar{x} - z_d),$$

and the unique solution is

$$\bar{u} = -B^* \bar{p} + u_d.$$

We want to estimate

$$u_T - \bar{u}.$$

Subtracting two optimality systems

$$\begin{aligned} y_T'(t) + Ay_T(t) &= -BB^* q_T(t) & -q_T'(t) + A^* q_T(t) &= C^* Cy_T(t) \\ y_T(0) &= x_0 - \bar{x} & q_T(T) &= p_d - \bar{p}, \end{aligned} \tag{1}$$

where  $y_T = x_T - \bar{x}$  and  $q_T = p_T - \bar{p}$ .

The obtained system is independent of the target data  $z_d$  and  $u_d$ .

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The obtained system is independent of the target data  $z_d$  and  $u_d$ .

We decompose the solution to (1) into two parts

$$\begin{aligned} y_T &= y_{T,1} + y_{T,2} \\ q_T &= q_{T,1} + q_{T,2}, \end{aligned}$$

First components  $(y_{T,1}, q_{T,1})$  satisfy

$$\begin{aligned} y_{T,1}'(t) + Ay_{T,1}(t) &= -BB^* q_{T,1}(t) & -q_{T,1}'(t) + A^* q_{T,1}(t) &= C^* Cy_{T,1}(t) \\ y_{T,1}(0) &= x_0 - \bar{x} & q_{T,1}(T) &= \mathbf{0}. \end{aligned}$$

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$$\begin{aligned}y_T &= y_{T,1} + y_{T,2} \\ q_T &= q_{T,1} + q_{T,2},\end{aligned}$$

Second components  $(y_{T,2}, q_{T,2})$  satisfy

$$\begin{aligned}y_{T,2}'(t) + Ay_{T,2}(t) &= -BB^* q_{T,2}(t) & -q_{T,2}'(t) + A^* q_{T,2}(t) &= C^* Cy_{T,2}(t) \\ y_{T,2}(0) &= 0 & q_{T,2}(T) &= p_d - \bar{p}.\end{aligned}$$

First components  $(y_{T,1}, q_{T,1})$  satisfy

$$\begin{aligned} y'_{T,1}(t) + Ay_{T,1}(t) &= -BB^* q_{T,1}(t) & -q'_{T,1}(t) + A^* q_{T,1}(t) &= C^* Cy_{T,1}(t) \\ y_{T,1}(0) &= x_0 - \bar{x} & q_{T,1}(T) &= 0. \end{aligned} \quad (2)$$

(2) is the optimality system for the problem

$$\min_{v \in L^2(0,T;U)} J_{T,1}(v) = \min_{v \in L^2(0,T;U)} \frac{1}{2} \int_0^T (|v(t)|_U^2 + |Cy(t)|_Z^2) dt,$$

where  $y$  stands for the solution to

$$\begin{aligned} y'(t) + Ay(t) &= Bv(t) \\ y(0) &= x_0 - \bar{x}. \end{aligned}$$

Consequently,

$$v_{T,1} = -B^* q_{T,1} = \arg \min J_{T,1}$$

and

$$\begin{aligned} \|v_{T,1}\|_{L^2(0,T;U)}^2 + \|Cy_{T,1}\|_{L^2(0,T;Z)}^2 &= 2J_{T,1}(v_{T,1}) \\ &\leq 2J_{T,1}(0) = \|CS_t(x_0 - \bar{x})\|_{L^2(0,T;X)}^2 = Q_T(x_0 - \bar{x}) \cdot (x_0 - \bar{x}), \end{aligned}$$

where

- $S_T$  is the semigroup generated by  $-A$ ,
- $Q_T = \int_0^T S_t^* C^* CS_t dt$  is the **observability Grammian**.



Second components  $(y_{T,2}, q_{T,2})$  satisfy

$$\begin{aligned} y'_{T,2}(t) + Ay_{T,2}(t) &= -BB^* q_{T,2}(t) & -q'_{T,2}(t) + A^* q_{T,2}(t) &= C^* Cy_{T,2}(t) \\ y_{T,2}(0) &= 0 & q_{T,2}(T) &= p_d - \bar{p}. \end{aligned} \quad (3)$$

(3) is the optimality system for the problem (the change of variable  $s = T - t$ )

$$\min_{z \in L^2(0,T;Z)} J_{T,2}(z) = \min_{z \in L^2(0,T;Z)} \frac{1}{2} \int_0^T \left( |z(t)|_Z^2 + |B^* q(t)|_X^2 \right) dt,$$

where  $q$  is the solution to the problem

$$\begin{aligned} q'(t) + A^* q(t) &= C^* z(t) \\ q(0) &= p_d - \bar{p}. \end{aligned}$$

Specially,  $Cy_{T,2} = \arg \min J_{T,2}$ .

Consequently

$$\begin{aligned} \|v_{T,2}\|_{L^2(0,T;U)}^2 + \|Cy_{T,2}\|_{L^2(0,T;Z)}^2 &= 2J_{T,2}(Cy_{T,2}) \\ &\leq 2J_{T,2}(0) = \|B^* S_t^*(p_d - \bar{p})\|_{L^2(0,T;X)}^2 = \Lambda_T(p_d - \bar{p}) \cdot (p_d - \bar{p}), \end{aligned}$$

where  $v_{T,2} = B^* q_{T,2}$ , while  $\Lambda_T$  stands for the **controllability Gramian**

$$\Lambda_T = \int_0^T S_t BB^* S_t^* dt.$$

## Theorem 1. (M. Gugat, M. L, '22.)

The difference of solutions to the original and the stationary optimal control problems  $v_T = u_T - \bar{u}$ , together with the difference of the corresponding optimal states  $y_T = x_T - \bar{x}$ , satisfies the estimate

$$\|u_T - \bar{u}\|_{L^2(0,T;U)}^2 + \|C(x_T - \bar{x})\|_{L^2(0,T;Z)}^2 \leq 2 \left( Q_T(x_0 - \bar{x}) \cdot (x_0 - \bar{x}) + \Lambda_T(p_d - \bar{p}) \cdot (p_d - \bar{p}) \right),$$

where

- $Q_T$  is the observability Grammian for the pair  $(A, C)$ ,
- $\Lambda_T$  is the controllability Grammian corresponding to the pair  $(A, B)$ .

## Theorem 2. (M. Gugat, M. L, '22.)

The difference of solutions to the original and the stationary optimal control problems  $v_T = u_T - \bar{u}$ , together with the difference of the corresponding optimal states  $y_T = x_T - \bar{x}$ , satisfies the estimate

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where

- $Q_T$  is the observability Grammian for the pair  $(A, C)$ ,
- $\Lambda_T$  is the controllability Grammian corresponding to the pair  $(A, B)$ .

No conditions on the linear operators  $A$ ,  $B$  and  $C$ !

## Definition

The operator  $B(C)$  is an **infinite-time admissible control (observation) operator** for the semigroup generated by  $-A$  if the controllability (observability) Gramian  $\Lambda_\infty(Q_\infty)$  is a bounded operator on  $X$ .

In that case

$$\|\Lambda_T\|_{L(X)} \leq \|\Lambda_\infty\|_{L(X)}$$

$$\|Q_T\|_{L(X)} \leq \|Q_\infty\|_{L(X)}$$

for all  $T$ .

## Theorem 3.

Assume that  $B$  and  $C$  are an infinite-time admissible control and observation operator for the semigroup generated by  $-A$ , respectively. Then the following result holds.

a) **(Integral turnpike)** For any target data  $u_d$  and  $z_d$ ,

$$\begin{aligned}\frac{1}{T} \int_0^T u_T &\xrightarrow{T \rightarrow \infty} \bar{u} \quad \text{strongly in } U, \\ \frac{1}{T} \int_0^T Cx_T &\xrightarrow{T \rightarrow \infty} C\bar{x} \quad \text{strongly in } Z.\end{aligned}$$

with the convergence rate of  $O(1/\sqrt{T})$ .

b) **(Measure turnpike)** For every  $\varepsilon > 0$  there exists a constant  $C_\varepsilon > 0$  (that depends on  $x_0 - \bar{x}$  and  $p_d - \bar{p}$ ) such that for every  $T > 0$  we have

$$\mu\{t \in [0, T] \mid |u_T - \bar{u}|^2 + |C(x_T - \bar{x})|^2 \geq \varepsilon\} < C_\varepsilon.$$

c) **(Convergence of the optimal value functions)**

$$\frac{1}{T} \min J_T = \min J_s + \mathcal{O}\left(\frac{1}{\sqrt{T}}\right) \quad \text{as } T \rightarrow \infty.$$

## Parameter dependent problems

A family of the parameter-dependent control systems

$$\begin{aligned}x'(\omega, t) + A(\omega)x(\omega, t) &= B(\omega)u(t) \\ x(\omega, 0) &= x_0(\omega),\end{aligned}$$

accompanied by an optimal control problem of the form

$$\begin{aligned}u_T &:= \min J_T(u) \\ &= \min_u \frac{1}{2} \int_0^T \left( |u(t) - u_d|_U^2 + \left| \int_{\Omega} (C(\omega)x(\omega, t) - z_d(\omega)) d\omega \right|_Z^2 \right) dt + \int_{\Omega} p_d(\omega) \cdot x(\omega, T) d\omega\end{aligned}$$

The corresponding stationary problem:

$$\begin{aligned}\bar{u} &:= \min_{u \in U} J_s(u) \\ &= \min \left\{ \frac{1}{2} \left( |u - u_d|_U^2 + \left| \int_{\Omega} (C(\omega)x(\omega) - z_d(\omega)) d\omega \right|_Z^2 \right) \mid A(\omega)x = B(\omega)u \right\}.\end{aligned}$$

We want to estimate

$$u_T - \bar{u}.$$

The problem is well posed

$$u_T = - \int_{\Omega} B(\omega)^* p_T(\omega) d\omega + u_d$$

where  $p_T$  is the optimal dual state:

$$x'_T(\omega, t) + A(\omega)x_T(\omega, t) = -B(\omega) \int_{\Omega} (B(\nu)^* p_T(\nu, t) - u_d) d\nu$$

$$x_T(\omega, 0) = x_0(\omega)$$

$$-p'_T(\omega, t) + A(\omega)^* p_T(\omega, t) = C(\omega)^* \int_{\Omega} (C(\nu)x_T(\nu, t) - z_d(\nu)) d\nu$$

$$p_T(\omega, T) = p_d(\omega),$$

Similarly

$$\bar{u} = - \int_{\Omega} B(\omega)^* \bar{p}(\omega) d\omega + u_d,$$

where  $\bar{p}$  is the optimal dual variable for the stationary problem.

We repeat the steps from the deterministic case.

- We consider

$$y_T = x_T - \bar{x}$$

$$q_T = p_T - \bar{p}$$

- We decompose the system satisfied by  $(y_T, q_T)$  into two parts.
- Each part is detected as optimality system for a similar kind of problem.

Consequently, we obtain the averaged control analogue of Theorem 7.

#### Theorem 4.

$$\begin{aligned} & \|u_T - \bar{u}\|_{L^2(0,T;U)}^2 + \left\| \int_{\Omega} C(\omega) (x_T(\omega, \cdot) - \bar{x}(\omega)) d\omega \right\|_{L^2(0,T;U)}^2 \\ & \leq \int_{\Omega} \left( Q_T(\omega)(x_0 - \bar{x})(\omega) \cdot (x_0 - \bar{x})(\omega) + \Lambda_T(\omega)(p_T - \bar{p})(\omega) \cdot (p_T - \bar{p})(\omega) \right) d\omega, \end{aligned}$$

where  $\Lambda_T(\omega) \left( Q_T(\omega) \right)$  stands for the controllability (observability) Grammian associated to the parameter value  $\omega$ .

## Theorem 5.

Assume that for almost every  $\omega \in \Omega$  we have that  $B(\omega)$  and  $C(\omega)$  are an infinite-time admissible control and observation operator for the semigroup generated by  $-A(\omega)$ , respectively.

a) **(Integral turnpike)** For any target data  $u_d$  and  $z_d(\omega)$ , we have

$$\frac{1}{T} \int_0^T u_T = \bar{u} + \mathcal{O}\left(\frac{1}{\sqrt{T}}\right) \quad \text{as } T \rightarrow \infty,$$

$$\frac{1}{T} \int_0^T \int_{\Omega} C(\omega)x_T(\omega)d\omega = \int_{\Omega} C(\omega)\bar{x}(\omega)d\omega + \mathcal{O}\left(\frac{1}{\sqrt{T}}\right) \quad \text{as } T \rightarrow \infty.$$

b) **(Measure turnpike)** For every  $\varepsilon > 0$  there exists a constant  $C_\varepsilon > 0$  (that depends on  $\|x_0 - \bar{x}\|_{L^2(\Omega; X)}$  and  $\|p_d - \bar{p}\|_{L^2(\Omega; X)}$ ) such that for every  $T > 0$  we have

$$\mu\{t \in [0, T] \mid |u_T - \bar{u}|^2 + \left| \int_{\Omega} C(\omega)(x_T(\omega) - \bar{x}(\omega))d\omega \right|^2 \geq \varepsilon\} < C_\varepsilon.$$

c) **(Convergence of the optimal value functions)**

$$\frac{1}{T} \min J_T = \min J_s + \mathcal{O}\left(\frac{1}{\sqrt{T}}\right) \quad \text{as } T \rightarrow \infty.$$



## A numerical example

Consider the system for  $\mathbf{x} = (x_1, x_2)$ :

$$x_1' = x_1 + u$$

$$x_2' = x_2 + u.$$

with the optimization problem

$$\min \frac{1}{2} \int_0^T (|u(t)|^2 + |x_1(t) + x_2(t) - 1|^2) dt.$$

$x_1 - x_2$  is independent of the control and uncontrollable.

What can we say about the observation  $C\mathbf{x} = x_1 + x_2$ ?

The corresponding static problem is

$$\min \frac{1}{2} (|u|^2 + |x_1 + x_2 - 1|^2)$$

subject to

$$x_1 = -u$$

$$x_2 = -u.$$

Its solution is  $\bar{u} = -\frac{2}{5}$ ,  $\bar{\mathbf{x}} = (\frac{2}{5}, \frac{2}{5})$ .

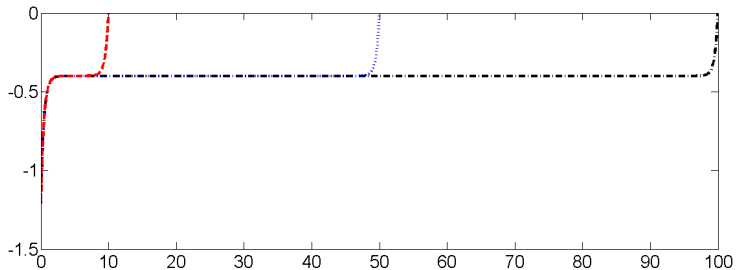


Figure 1: The graph of the optimal control  $u_T(t)$ .

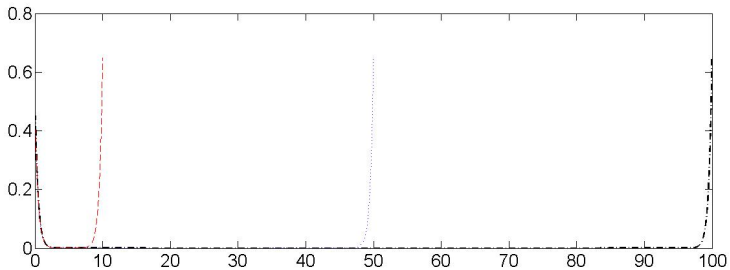


Figure 2: The graph of  $x_{T,1}(t) + x_{T,2}(t) - 0.8$ .

## Conclusion

- Turnpike results for infinite-dimensional, LQ optimal control problems.
- The key estimate on the difference between evolutionary and stationary optimal controls and observations is derived with virtually no assumptions on the operators  $A$ ,  $B$ , and  $C$ .
- The turnpike properties follow directly by assuming infinite-time admissibility of control and observation operator.
- Results obtained both in deterministic and parameter dependent case.
- Exponential turnpike?

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Thanks for your attention!