

# The Turnpike Phenomenon for Problems of Optimal Boundary Control

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# Outline

The Turnpike Phenomenon: What is it?

 $L^1$  optimal Dirichlet control of the **wave** equation

L<sup>2</sup> optimal Neumann control of the **wave** equation

Turnpike for linear  $2 \times 2$  systems: Problem definition

A turnpike property relates the dynamic and the static problem

Conclusion open problems



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- Consider a dynamic optimal control problem with a time interval [0, T].
- If all the time-derivatives are set to zero and initial conditions and terminal conditions are canceled, this yields a **static optimal control problem**.
- Turnpike results give relations between the *static optimal control* and the *dynamic optimal control*.
- They state that for sufficiently large *T*, some distance between the static optimal point and the dynamic optimal point becomes small.



For  $T \ge 1$  we consider the problem

$$(\mathbf{OC})_{T} \begin{cases} \min_{u \in L^{2}(0,T), u(t) \ge 0, y(t) \le 0} \int_{0}^{T} \frac{1}{2} |u(t)|^{2} + |u(t)| + |y(t)| \, dt \text{ subject to} \\ y(0) = -1, \ y'(t) = y(t) + \exp(t) \, u(t) \\ y(T) = 0. \end{cases}$$



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$$\hat{y}(t) = e^t \left[ -1 + \int_0^t u(\tau) \, d\tau \right] = t e^{t+1} - e^{2t} \le 0$$

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and for  $t \ge 1$  we have  $\hat{y}(t) = 0$ . The feasible controls are characterized by the *moment equation*  $\int_0^T u(\tau) d\tau = 1$ .



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- In the problem we have terminal conditions.
   What happens, if we cancel y(T) = 0?
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• In fact, if  $\exp T \ge 1 + e$ ,  $\hat{u}$  is again the optimal control!



For sufficiently large *T*, due to the  $L^1$ -norm of *y* that appears in the objective function, the solution has a **finite–time turnpike structure** where the system is steered to zero in the *finite time*  $t_0 = 1$  that is independent of *T* and remains there for all  $t \in (t_0, T)$ .

You can think of the turnpike as a point that does not move.



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Norm minimal exact control, finite horizon  $T \ge 1$ 

Let  $y_0 \in L^1(0, 1)$ ,  $y_1 \in W^{-1,1}(0, 1)$  and  $T \ge 1$  be given. Define (**EC**) :

 $\begin{cases} \min \int_0^T |u_0(t)| + |u_1(t)| \, dt \text{ subject to} \\ y(0,x) = y_0(x), \, y_t(0,x) = y_1(x), \, x \in (0,1) \\ y(t,0) = u_0(t), \ y(t,1) = u_1(t), \ t \in (0,T) \\ y_{tt}(t,x) = y_{xx}(t,x), \ (t,x) \in (0,T) \times (0,1) \\ y(T,x) = 0, \ y_t(T,x) = 0, \ x \in (0,1). \end{cases}$ 



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$$y(t,0) = u_0(t), y(t,1) = u_1(t), t \in (0, T)$$

$$y_{tt}(t,x) = y_{xx}(t,x), \ (t,x) \in (0,T) \times (0,1)$$

$$y(T, x) = 0, y_t(T, x) = 0, x \in (0, 1).$$

### In general the optimal controls are not unique!

$$y_t(0, x) = y_1(x), x \in (0, 1)$$
  
 $y(t, 1) = u_1(t), t \in (0, T)$   
 $x), (t, x) \in (0, T) \times (0, 1)$   
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Let $y_0 \in L^1(0, 1), y_1 \in W^{-1,1}(0, 1)$ and $T \ge 1$ be given. Define ( <b>EC</b> ) : $\begin{cases} \min \int_0^T  u_0(t)  +  u_1(t)   dt \text{ subject to} \\ y(0, x) = y_0(x),  y_t(0, x) = y_1(x),  x \in (0, 1) \\ y(t, 0) = u_0(t),  y(t, 1) = u_1(t),  t \in (0, T) \\ y_{tt}(t, x) = y_{xx}(t, x),  (t, x) \in (0, T) \times (0, 1) \end{cases}$	Let $T \ge 1$ . There exist solutions of (EC) that are 2-periodic i.e. for $k \in \{1, 2,, \}, t \in (0, 2),$ $t + 2k \le T, l \in \{1, 2\}$ we have $u_l(t + 2k) = u_l(t).$
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Let $y_0 \in L^1(0, 1)$ , $y_1 \in W^{-1,1}(0, 1)$ and $T \ge 1$ be given. Define ( <b>EC</b> ) : $\left( \begin{array}{c} \min \int_0^T  u_0(t)  +  u_1(t)   dt \text{ subject to} \\ y(0, x) = y_0(x),  y_t(0, x) = y_1(x),  x \in (0, 1) \end{array} \right)$	Let $T \ge 1$ . There exist solutions of (EC) that are 2-periodic i.e. for $k \in \{1, 2,, \}, t \in (0, 2),$ $t + 2k \le T, l \in \{1, 2\}$ we have $u_l(t + 2k) = u_l(t).$ The set of all solutions is
$\begin{cases} y(t,0) = u_0(t),  y(t,1) = u_1(t), \ t \in (0,T) \\ y_0(t,x) = y_0(t,x), \ (t,x) \in (0,T) \times (0,1) \end{cases}$	measurable convex combinations $(t \in (0, 1))$
$y_{tt}(t,x) = y_{xx}(t,x), \ (t,x) \in (0, T) \times (0, T)$ $y(T,x) = 0, \ y_{t}(T,x) = 0, \ x \in (0, 1).$	$\lambda_{j}^{(\prime)}(t) \geq 0, \ \sum_{j:t+2j \leq T} \lambda_{j}^{(\prime)}(t) = 1.$

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Example: Let  $y_1 = 0$ .

Let

$$k = \max\{j \in \{1, 2, 3, ...\} : j \le T\}$$

and

 $\Delta = T - k \ge 0.$ 

 $(0, T) = (0, \Delta) \cup (\Delta, 1) \cup (1, 1 + \Delta) \cup (2 + \Delta, 2) \cup (2, 2 + \Delta) \dots \cup ((k - 1) + \Delta, k) \cup (k, k + \Delta)$ 

There are k + 1 red intervals and k black intervals!



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$$u_0(t+j) = u_1(t+j) = (-1)^j \frac{y_0(t)}{2(k+1)}.$$



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Example:  $y_1 = 0$ . The control action can be shifted between the different time periods!

Let again

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$$d(t) = \begin{cases} (k+1) & \text{if } t \in (0, \Delta), \\ k & \text{if } t \in (\Delta, 1). \end{cases}$$



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The set of **all optimal controls** has the following structure: For pairs of measurable convex combinations  $I \in \{1, 2\}, t \in (0, 1)$ 

$$\lambda_j^{(l)}(t) \geq 0, \ \sum_{j:t+2j \leq T} \lambda_j^{(l)}(t) = 1$$

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$$u_0(t+2j) = \lambda_{2j}^{(1)}(t) \frac{y_0(t)}{2 d(t)}, \quad u_0(t+2j+1) = -\lambda_{2j+1}^{(2)}(t) \frac{y_0(t)}{2 d(t)},$$
  
$$u_1(t+2j) = \lambda_{2j}^{(2)}(t) \frac{y_0(t)}{2 d(t)}, \quad u_1(t+2j+1) = -\lambda_{2j+1}^{(1)}(t) \frac{y_0(t)}{2 d(t)}.$$



• The convex combinations ( $t \in (0, 1), l \in \{1, 2\}$ )

$$\lambda_j^{(l)}(t) \geq \mathbf{0}, \ \sum_{j:t+2j \leq T} \lambda_j^{(l)}(t) = \mathbf{1}$$

determine the *support* of the corresponding optimal control (together with the support of  $y_0$ ,  $y_1$ ). It can be the whole interval [0, T].



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 If for all t ∈ (0, 1), the λ<sub>j</sub><sup>(l)</sup>(t) are equal for all j, we obtain *periodic* controls. In fact, this yields λ<sub>j</sub><sup>(l)</sup>(t) = 1/d(t). These are the optimal controls with minimal L<sup>2</sup>-norm.



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   In this case, the *support* of the corresponding optimal control can be constrained to a subinterval of [0, T] of minimal length 1.



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   In this case, the *support* of the corresponding optimal control can be constrained to a subinterval of [0, T] of minimal length 1.
- Thus we have optimal controls with support (0, 1). These controls steer the system to rest at the time t = 1.



Adding a tracking-term in the goal function Finite horizon  $T \ge 1$ ,  $\gamma > 0$ . Define (**P**):  $\min_{\substack{u_0, u_1 \in L^1(0, T) \\ + \gamma \int_1^T \int_0^1 |y(t, x)| \, dx \, dt}} \int_0^T |u_0(t)| + |u_1(t)| \, dt$ subject to  $y(0, x) = y_0(x), \ y_t(0, x) = y_1(x), \ x \in (0, 1)$  $y(t,0) = u_0(t), y(t,1) = u_1(t), t \in (0,T)$  $y_{tt}(t,x) = y_{xx}(t,x), (t,x) \in (0,T) \times (0,1),$  $y(T, x) = 0, \ y_t(T, x) = 0, \ x \in (0, 1).$ 



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### Solution of (P)

The nonsmooth problem (**P**) has a *unique* solution. The unique solution of (**P**) steers the state to rest at time t = 1. Then the control is switched off.



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## L<sup>1</sup>-Optimal Dirichlet control of the wave equation

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 $u_0(t)=u_1(t)=0,$ 

 $y(t, x) = 0, x \in (0, 1)$ and  $y_t(t, x) = 0$ .

This is possible due to exact controllability!



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### Proof.

Problem (**EC**) has a unique solution where the support of the controls is in (0, 1). This optimal control  $(u_0^*, u_1^*)$  steers the state to rest at time t = 1. Let  $\nu$ (**EC**) denote the optimal value of (**EC**). Let  $\nu$ (**P**) denote the optimal value of (**P**).



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We can replace (**P**) with the problem without terminal constraints. If T is sufficiently large, the solution should stay the same - however, the proof is not written.



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 $L^1$  optimal Dirichlet control of the **wave** equation

 $L^2$  optimal Neumann control of the **wave** equation

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# L<sup>2</sup>-optimal Neumann control of the wave equation

Norm minimal exact control, finite horizon  $T \ge 2$ 

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Gugat, Arab. J. Math. 2015 Open Access

Let  $T \in \mathbb{N}$  be even. The **unique** solution of (**EC**) is *4–periodic* and

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For  $t_0 = 0$  (*Moving horizon*) this yields the well-known feedback law

$$y_x(t_0, 1) = -\frac{1}{T-1} y_t(t_0, 1).$$



### Adding a tracking-term in the goal function

Finite horizon  $T \ge 2$ ,  $\gamma > 0$ . Define (**P**):

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#### This is an *exponential turnpike* structure!

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For  $d_{-} < 0 < d_{+}$ , define the 2  $\times$  2 matrix

$$D(x) = \left( egin{array}{cc} d_+ & 0 \ 0 & d_- \end{array} 
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Let a 2 × 2 matrix M(x) and  $\eta_0 \in (-\infty, 0]$  be given.



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#### Consider the system equation

(*pde*):  $r_t + D r_x = \eta_0 M r$ , where for  $t \in (0, T)$  and  $x \in (0, L)$  the state is given by  $r(t, x) = \begin{pmatrix} r_+(t, x) \\ r_-(t, x) \end{pmatrix}$ .



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#### Initial conditions (t = 0)

For 
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For  $d_{-} < 0 < d_{+}$ , define the 2 × 2 matrix Dirichlet boundary control ( $x \in \{0, L\}$ )

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#### **Objective Function**

$$J(u, r) = \int_0^T f_0(u_+(t), r_-(t, 0)) dt$$

$$+\int_0^T f_L(u_-(t), r_+(t, L)) dt$$

with strictly convex quadratic functions  $f_0$ ,  $f_L$ .



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Dynamic optimal control problem

Optimal boundary control problem  $\begin{cases} \min_{u \in (L^2(0, T))^2} J(u, r) \\ \text{subject to (pde), initial c. and b.c.} \end{cases}$ 



For real numbers  $\mu_+$ ,  $\mu_-$  define

$$E(x) = \left( egin{array}{cc} \exp(-\mu_+ x) & \mathbf{0} \ \mathbf{0} & \exp(\mu_- x) \end{array} 
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Assume that there exist  $\mu_+ > 0$ ,  $\mu_- > 0$  and  $\nu_a < 0$ ( $\mu_+ < 0$ ,  $\mu_- < 0$  and  $\nu_0 > 0$  respectively) such that for all  $x \in [0, L]$ 

 $\sup_{v: v^{\top} E(x) v = 1} v^{\top} \left[ E'(x) D(x) + E(x) D'(x) - 2 |\eta_0| E(x) M(x) \right] v \le \nu_a < 0 \text{ and respectively}$ 



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1. If M(x) is a diagonal matrix or if  $|\eta_0|$  is sufficiently small or if L > 0 is **sufficiently small**, both conditions hold.



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Can be checked with a 1-d optimization problem!

- 1. If M(x) is a diagonal matrix or if  $|\eta_0|$  is sufficiently small or if L > 0 is **sufficiently small**, both conditions hold.
- 2. If  $M^{\top} = M$ , both conditions equivalent with  $\nu_0 = -\nu_a$ .



## **Problem definition: The static problem**

The static state is denoted by

$$oldsymbol{R}^{(\sigma)}(x)=\left(egin{array}{c} R^{(\sigma)}_+(x)\ R^{(\sigma)}_-(x) \end{array}
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### SICON 2019 (with F. HANTE)

For a finite time horizon T > 0, let the **optimal dynamic control** be

 $u^{(\delta, T)} \in L^2(0, T) \times L^2(0, T)$ 

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• For increasing time horizon  $T \to \infty$ , the average quadratic mean distance between the optimal dynamic and the optimal static control converges to zero with the rate  $O(\frac{1}{T})$ .

#### For the state we have

$$\int_{0}^{T}\int_{0}^{L}\left\|\boldsymbol{r}^{(\delta,\,T)}(\tau,\,\boldsymbol{x})-\boldsymbol{r}^{(\sigma)}\right\|_{\mathbb{R}^{2}}^{2}\,d\boldsymbol{x}\,d\tau\leq\bar{\boldsymbol{D}}.$$



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- The *L*<sup>1</sup>–cost leads to non-smooth optimal control problems. In the applications, the turnpike results allow to obtain
  - *L*<sup>1</sup>*-optimal controls* by *finite time stabilizing feedback controlers* (LIONEL ROSIER) that can be applied independent of the initial state.



Suspension Bridge



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Suspension Bridg

In the L<sup>2</sup>-case, the finite-time turnpike does not occur.
 However, there is exponential turnpike.





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## **Open Problems**

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- Thank you for your attention!