

# Planning a finite time transition from a non-stationary to a stationary point without overshoot

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**Abstract**—The problem of planning a finite time transition of a trajectory from a non-stationary point to a stationary set-point is addressed. As opposed to standard approaches, where the transition functions are polynomials, the specific choice of non-analytic function proposed may easily be trimmed to show no overshoot and only an adjustable undershoot during transition. The main result is a recursive formula for a simple parametrization of the transition function, as needed in tracking problems. Two examples underscore the ease of the approach.

## I. INTRODUCTION

Consider the problem of planning the transition of a time function  $y(t)$  on a finite time interval  $t \in [t_1, t_2]$ . Let  $t = t_1$  be the time associated with  $r$  left boundary conditions (BC)

$$y^{(i)}(t_1) := \left. \frac{d^i y}{dt^i} \right|_{t=t_1} = \underline{y}_i, \quad i = 0, 1, \dots, r-1. \quad (1)$$

Correspondingly, at  $t = t_2$  let  $y(t)$  satisfy  $r$  right BC

$$y^{(i)}(t_2) := \left. \frac{d^i y}{dt^i} \right|_{t=t_2} = \bar{y}_i, \quad i = 0, 1, \dots, r-1. \quad (2)$$

A straight-forward idea for tackling this problem is to use polynomials for meeting the  $2r$  BC of (1) and (2). The least degree polynomial shows degree  $2r-1$  and is uniquely determined by the BC. Polynomials with degrees larger than  $2r-1$  that meet the BC may be found, as well. Various approaches of polynomial kind are exposed in [8], for example. An approximate optimization-based approach presents [4]. For the planning of transitions between stationary points see [5], where a simple to use formula for the transition polynomial is given. An input shaping approach by means of polynomials with additional exponential decay is derived in [7]. Recently, trajectory generation received attention within the different inversion-based approaches to the control of systems with internal dynamics [1], [2], [6], [3].

In this paper, let the task be confined to the planning of the transition from a non-stationary point of  $y(t)$ , as specified in (1), to a stationary point as given by the right BC

$$y(t_2) = \bar{y}_0, \quad y^{(i)}(t_2) = 0, \quad i = 1, \dots, r-1, \quad (3)$$

being a special case of the right BC (2).

It turns out that in the case of planning from non-stationary to stationary points there are decisive drawbacks when employing polynomials: Primarily, there is no a priori criterion to decide whether the transition polynomial resulting from the BC in (1) and (2) will show an overshoot or undershoot.

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Standard methods as calculating the set of zeroes with respect to the polynomial's first time derivative give a posteriori insight, only. Secondly, it is a well-known fact that large absolute values of  $\underline{y}_i$  and  $\bar{y}_i$  give rise to polynomials with very large degree, accompanied by the problem of a wavy transition in course of time.

Hence, the proposal of this paper is to refrain from polynomials, and rather employ a particular non-analytic function. A formula for the recursive parametrization of this function is provided that may easily be trimmed to show no overshoot and just a reduced undershoot when adjusting one single parameter.

The paper is organized as follows: Section II contains the derivation of the parametrization of a non-analytic function, adequate for solving the above-stated transition problem on a unity time interval. Section III provides the main result that holds for arbitrary time intervals. The paper ends with a discussion and some examples in Section IV.

## II. PARAMETRIZATION OF A NON-ANALYTIC TRANSITION FUNCTION

Consider the transition function

$$y(t) = (c_0 + c_1 t + \dots + c_{r-1} t^{r-1}) e^{\frac{-1}{(t-1)^n}} + \bar{y}_0 \quad (4)$$

with even exponent  $n \in \{2, 4, 6, \dots\}$  and real coefficients  $c_i$ ,  $i = 0, 1, \dots, r-1$ . It is not difficult to show that the ansatz (4) satisfies the stationary right BC (3) in a limit sense

$$\lim_{t \rightarrow 1} y(t) = \bar{y}_0 \quad \text{and} \quad \lim_{t \rightarrow 1} y^{(i)}(t) = 0, \quad i = 1, 2, \dots \quad (5)$$

which implies that  $y(t)$  given in (4) is non-analytic at  $t = 1$ .

The coefficients  $c_i$  serve to satisfy the left BC (1) at the time instant  $t_1 = 0$ . In a next step, the result to be obtained at time instants  $t_1 = 0$  and  $t_2 = 1$  may then be generalized to arbitrary instants of time  $t_1 < t_2$ ,  $t_1, t_2 \in \mathbb{R}$ .

In view of the left BC (1), the coefficients  $c_i$  may be determined by equating  $y^{(i)}(0) = \underline{y}_i$ ,  $i = 0, 1, \dots, r-1$ .

In the first place, observe that

$$y(0) = c_0 e^{\frac{-1}{(-1)^n}} + \bar{y}_0 \stackrel{!}{=} \underline{y}_0 \Rightarrow c_0 = (\underline{y}_0 - \bar{y}_0) e. \quad (6)$$

Thereafter, for  $i = 1, 2, \dots$  determine the  $i$ -th time derivative

$$\begin{aligned} y^{(i)}(t) &= \sum_{\nu=0}^i \binom{i}{\nu} \left( \frac{d^{i-\nu}}{dt^{i-\nu}} \sum_{\mu=0}^{r-1} c_\mu t^\mu \right) \left( \frac{d^\nu}{dt^\nu} e^{\frac{-1}{(t-1)^n}} \right) \\ &= \sum_{\nu=0}^i \binom{i}{\nu} \left( \frac{d^\nu}{dt^\nu} e^{\frac{-1}{(t-1)^n}} \right) \sum_{\mu=i-\nu}^{r-1} c_\mu \frac{d^{i-\nu}}{dt^{i-\nu}} t^\mu \\ &= \sum_{\nu=0}^i \binom{i}{\nu} \left( \frac{d^\nu}{dt^\nu} e^{\frac{-1}{(t-1)^n}} \right) \sum_{\mu=i-\nu}^{r-1} \frac{\mu!}{(\mu + \nu - i)!} c_\mu t^{\mu+\nu-i} \quad (7) \end{aligned}$$

and for adaption to the left BC (1), for any  $i = 1, 2, \dots, r-1$  at  $t_1 = 0$  we have to require that

$$\begin{aligned} \underline{y}_i &\stackrel{!}{=} \lim_{t \rightarrow 0} y^{(i)}(t) = \sum_{\nu=0}^i \binom{i}{\nu} \left( \lim_{t \rightarrow 0} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{(t-1)^n}} \right) \\ &\quad \sum_{\mu=i-\nu}^{r-1} \frac{\mu!}{(\mu+\nu-i)!} c_\mu \left( \lim_{t \rightarrow 0} t^{\mu+\nu-i} \right) \\ &= \sum_{\nu=0}^i \binom{i}{\nu} \left( \lim_{t \rightarrow 0} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{(t-1)^n}} \right) (i-\nu)! c_{i-\nu} \\ &= \sum_{\nu=0}^i \frac{i!}{\nu!} c_{i-\nu} \left( \lim_{t \rightarrow -1} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{t^n}} \right) \end{aligned} \quad (8)$$

which together with equation (6) is a linear system of equations that allows to solve for the  $r$  unknown coefficients  $c_i$  in terms of the BC  $\underline{y}_i$ ,  $i = 0, 1, \dots, r-1$ , in a unique way.

The triangular structure of equation (8) suggests to exploit a simple recurrence scheme. Indeed, rewriting (8) yields

$$\begin{aligned} \underline{y}_i &= i! c_i \left( \lim_{t \rightarrow -1} e^{\frac{-1}{t^n}} \right) + \sum_{\nu=1}^i \frac{i!}{\nu!} c_{i-\nu} \left( \lim_{t \rightarrow -1} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{t^n}} \right) \\ &= \frac{i!}{e} c_i + \sum_{\nu=1}^i \frac{i!}{\nu!} c_{i-\nu} \left( \lim_{t \rightarrow -1} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{t^n}} \right). \end{aligned} \quad (9)$$

Thus, with (6) we derive the recurrence ( $i = 0, 1, \dots, r-1$ )

$$c_i = e \left( \frac{\underline{y}_i}{i!} - \sum_{\nu=1}^i \frac{c_{i-\nu}}{\nu!} \lim_{t \rightarrow -1} \frac{d^\nu}{dt^\nu} e^{\frac{-1}{t^n}} \right), \quad c_0 = (\underline{y}_0 - \bar{y}_0) e. \quad (10)$$

The derivatives on the right hand side of (10) may be evaluated further. To this end, use the chain rule

$$\frac{df(t)}{dt} = f(t) \frac{dg(t)}{dt}, \quad f(t) = e^{g(t)}, \quad g(t) = \frac{-1}{t^n}. \quad (11)$$

In doing so, we may refer to Leibniz' rule for differentiating products again, hence

$$f^{(\nu+1)}(t) = \sum_{i=0}^{\nu} \binom{\nu}{i} \left( \frac{d^{\nu-i}}{dt^{\nu-i}} f(t) \right) \left( \frac{d^{i+1}}{dt^{i+1}} g(t) \right) \quad (12)$$

and shifting  $\nu \rightarrow \nu - 1$  it follows that

$$f^{(\nu)}(t) = \sum_{i=0}^{\nu-1} \binom{\nu-1}{i} f^{(\nu-i-1)}(t) g^{(i+1)}(t), \quad \nu = 1, 2, \dots \quad (13)$$

where the  $\nu$ -th time derivative of  $f(t)$  is expressed in terms of lower order derivatives in form of a recurrence. Finally, recalling (11) it remains to evaluate

$$\begin{aligned} g^{(i+1)}(t) &= \frac{d^{i+1}}{dt^{i+1}} \left( \frac{-1}{t^n} \right) = (-1) \frac{d^{i+1}}{dt^{i+1}} t^{-n} \\ &= (-1)(-n)(-n-1)(-n-2) \dots (-n-i) \frac{1}{t^{n+i+1}} \\ &= (-1)^i \frac{(n+i)!}{(n-1)!} \frac{1}{t^{n+i+1}}. \end{aligned} \quad (14)$$

A consequence is the recurrence

$$\begin{aligned} f^{(\nu)}(t) &= \sum_{i=0}^{\nu-1} \binom{\nu-1}{i} \frac{(n+i)!}{(n-1)!} \frac{(-1)^i}{t^{n+i+1}} f^{(\nu-i-1)}(t) \\ f^{(0)}(t) &= e^{\frac{-1}{t^n}} \end{aligned} \quad (15)$$

which at  $t = -1$  yields

$$\begin{aligned} f^{(\nu)}(-1) &= \sum_{i=0}^{\nu-1} \binom{\nu-1}{i} \frac{(n+i)!}{(n-1)!} (-1)^{n+1} f^{(\nu-i-1)}(-1) \\ f^{(0)}(-1) &= 1/e \end{aligned} \quad (16)$$

to be solved until index  $\nu = r-1$ , as indicated by (10).

### III. MAIN RESULT

Simple steps of manipulation show that a possible transition function, which satisfies the  $2r$  BC of (1) and (3) at arbitrary instants of time  $t_1$  and  $t_2$ , reads

$$y(t) = \underline{y}_0 + (1/e) \left( \frac{t_2 - t_1}{t_2 - t_1} \right)^n \sum_{i=0}^{r-1} c_i \left( \frac{t - t_1}{t_2 - t_1} \right)^i \quad (17)$$

with coefficients  $c_i$  that result from the recurrence

$$c_i = e \left( \frac{\underline{y}_i (t_2 - t_1)^i}{i!} - \sum_{\nu=1}^i \frac{c_{i-\nu}}{\nu!} f^{(\nu)}(-1) \right) \quad (18)$$

$$c_0 = (\underline{y}_0 - \bar{y}_0) e. \quad (19)$$

where the values of  $f^{(\nu)}(-1)$  follow from (16).

### IV. DISCUSSION AND EXAMPLES

In order to find a minimal parameter  $n = n_{\min}$  subject to which no overshoot occurs for  $t \in (t_1, t_2)$ , note that with the coefficients  $c_i$  determined as above, the necessary condition  $\frac{d}{dt} y(t) = 0$  for an extremal point may be written as

$$\begin{aligned} \sum_{i=0}^{r-1} c_i (t - t_1)^i (t_2 - t_1)^{r-1-i} \times \\ (i(t_2 - t)^{n+1} - n(t_2 - t_1)^n (t - t_1)) = 0. \end{aligned} \quad (20)$$

In the main, two cases need to be distinguished:

- 1) When increasing  $n$  starting from 2, given the bottom-up-transition  $\underline{y}_0 < \bar{y}_0$  and  $\underline{y}_1 > 0$  (top-down-transition  $\underline{y}_0 > \bar{y}_0$  and  $\underline{y}_1 < 0$ ), then  $n_{\min}$  is the first number for which the polynomial in (20) shows no zeroes in  $(t_1, t_2)$ . Thus, the overshoot as depicted in the plots of Figure 1 can be avoided by increasing  $n$ .
- 2) When increasing  $n$  starting from 2, given the bottom-up-transition  $\underline{y}_0 < \bar{y}_0$  and  $\underline{y}_1 < 0$  (top-down-transition  $\underline{y}_0 > \bar{y}_0$  and  $\underline{y}_1 > 0$ ), then  $n_{\min}$  is the first number for which the polynomial in (20) shows one single zero in  $(t_1, t_2)$ . In this case, besides avoiding an overshoot, one may additionally reduce the undershoot by a further increase of the parameter  $n$  until the undershoot falls below a specified bound, as shown in the lower plots of Figure 1 (see arrows).

TABLE I

LEFT BC FOR THE PARAMETRIZATION OF THE TRANSITION FUNCTION (17) AS DEPICTED IN FIGURE 2 AND FIGURE 3

left BC	$\underline{y}_0$	$\underline{y}_1$	$\underline{y}_2$	$\underline{y}_3$	$\underline{y}_4$
Figure 1	0	20	30	60	40
Figure 2	0	-15	200	100	40

Either of these cases is demonstrated resorting to an example transition from  $t_1 = 0$  to  $t_2 = 2$  subject to  $r = 5$  non-stationary BC at  $t = t_1$  (see Table I). A stationary value of  $\bar{y}_0 = 10$  shall be reached for both transitions at  $t = t_2$ .

Case 1 is illustrated in Figure 2: A calculation of the corresponding zeroes of (20) for  $n = 2, 4, 6, \dots, 16$  yields that  $n_{\min} = 8$ , where no overshoot takes place, anymore. An increase of  $n$  further accelerates the response.

Case 2 is illustrated in Figure 3: A calculation of the corresponding zeroes of (20) for  $n = 2, 4, 6, \dots, 16$  yields that  $n_{\min} = 6$ , where no overshoot takes place, anymore. A further increase of  $n$  helps accelerate the response and further reduces the undershoot. Such transitions resemble behaviors that are typical within the tracking of non-minimum phase systems.

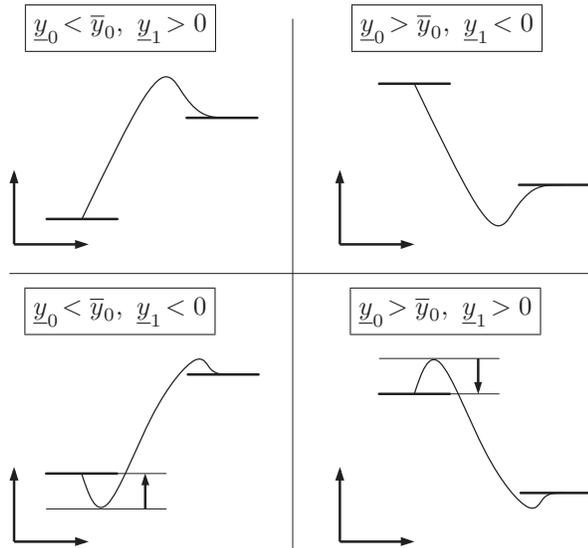


Fig. 1. Dependency of overshoot and undershoot on left side boundary conditions—by increasing  $n$  in the non-analytic transition (17) the depicted overshoots may be avoided and the undershoots (marked with arrow) are reduced

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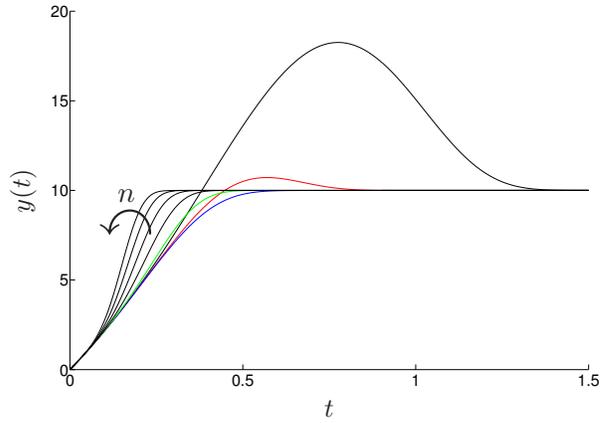


Fig. 2. Case 1: transition function (17) for left BC according to Table I;  $n = 2$  (black line),  $n = 4$  (red line),  $n = 6$  (blue line); for  $n = 8$  (green line) no overshoot occurs anymore; thin black lines show faster response when increasing  $n$  (plotted are  $n = 10, 12, 14, 16$ )

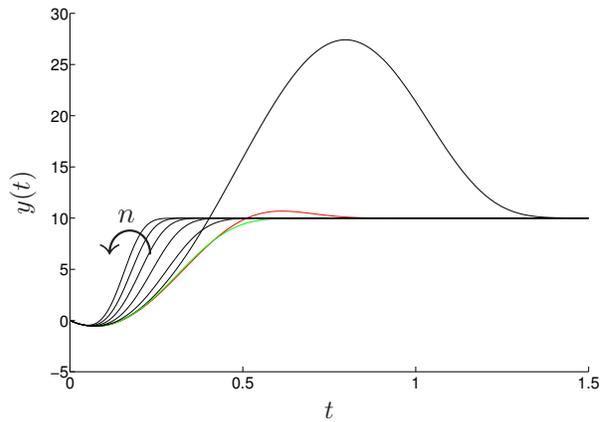


Fig. 3. Case 2: transition function (17) for left BC according to Table I;  $n = 2$  (black line),  $n = 4$  (red line); for  $n = 6$  (green line) no overshoot occurs anymore; increasing  $n$  helps reduce the undershoot further (plotted are  $n = 8, 10, 12, 14, 16$ )

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