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A mathematical explanation via “intelligent” PID controllers of the strange ubiquity of PIDs

Brigitte d’ANDRÉA-NOVEL, Michel FLIESS, Cédric JOIN, Hugues MOUNIER, Bruno STEUX

Abstract—The ubiquity of PID controllers in the industry has remained mysterious until now. We provide here a mathematical explanation of this strange phenomenon by comparing their sampling with the the one of “intelligent” PID controllers, which were recently introduced. Some computer simulations nevertheless confirm the superiority of the new intelligent feedback design.

Keywords—PID, model-free control, intelligent PID, sampling.

I. INTRODUCTION

PI and PID controllers (see, e.g., [2], [16]) are still by far the most popular feedback design in industry. To the best of our knowledge, there is no clear-cut explanation of their strange ubiquity for a wide range of systems. Remember that, from a purely mathematical standpoint, they are only fully justified until now for first and second order linear differential equations with constant coefficients! We solve here this long-standing and quite irritating open problem via the newly introduced intelligent PIDs ([6], [7]), which have already been utilized quite successfully in several concrete situations (see, e.g., [1], [4], [10], [11], [13], [14], [18]).

The proof relies on a crude time-sampling of both types of regulators. It shows that the gains in a classic PI or PID take into account, if they are properly tuned, the estimated “structural” part of the intelligent controllers. Thus the efficiency of these intelligent controllers with respect to arbitrary nonlinear plants ([6], [7]) is enough for fulfilling our purpose. Let us nevertheless emphasize that the classic tuning rules are quite intricate whereas their counterparts for intelligent controllers are obvious.

Remark 1.1: Only few references (see, e.g., [3], [12]) in the huge literature on PIDs exhibit some connections with our viewpoint.

Our paper is organized as follows. Section II is devoted to a brief review of model-free control and of the corresponding intelligent PID controllers. Section III, which establishes our results by comparing the sampling of classical and intelligent controllers, gives a table for the connections between classic and intelligent gains. The computer simulations in Section IV confirms the superiority of the intelligent controllers (see, also, [6], [7] for other examples). Some concluding remarks are given in Section V.

II. MODEL-FREE CONTROL

The input-output behavior of the system, which for simplicity’s sake is assumed to be monovariable, is “approximately” governed within its operating range by an unknown finite-dimensional ordinary differential equation, which is not necessarily linear,

\[ E(y, \dot{y}, \ldots, y^{(\nu)}, u, \ddot{u}, \ldots, u^{(\nu)}) = 0 \]  

Replace Equation (1) by the following “phenomenological” model, which is only valid during a very short time interval,

\[ y^{(\nu)} = F + \alpha u \]  

The derivation order \(\nu\), which is in general equal to 1 or 2, and the constant parameter \(\alpha\) are chosen by the practitioner. It implies that \(\nu\) is not necessarily equal to the derivation order \(a\) of \(y\) in Equation (1). The numerical value of \(F\) at any time instant is deduced from those of \(u\) and \(y^{(\nu)}\). Thanks to our quite efficient numerical differentiators, which are moreover real-time, the desired behavior is obtained by implementing, if, for instance, \(\nu = 2\), the intelligent PD controller (i-PD)

\[ u = \frac{1}{\alpha} \left( -F + y^* + K_P e + K_I \int e + K_D \dot{e} \right) \]  

where

- \(y^*\) is the output reference trajectory, which is determined e.g. via the rules of flatness-based control;
- \(e = y - y^*\) is the tracking error;
- \(K_P, K_I, K_D\) are the usual tuning gains.

Let us consider the following special cases:

- If again \(\nu = 2\), we may use an intelligent PD controller (i-PD)

\[ u = \frac{1}{\alpha} \left( -F + y^* + K_P e + K_D \dot{e} \right) \]

- If \(\nu = 1\), we can restrict ourselves to

1See [6], [7] for more details.
2See [8], [15] for details, and also [9].
an intelligent PI controller (i-PI)

\[ u = \frac{1}{\alpha} \left( -F + \dot{y}^* + K_P e + K_I \int e \right) \]  

(5)

or even to an intelligent P controller (i-P)

\[ u = \frac{1}{\alpha} (-F + \dot{y}^* + K_P e) \]  

(6)

**Remark 2.1:** If \( \nu = 2 \) (resp. 1), plugging Equations (3) or (4) (resp. (5) or (6)) in Equation (2) yields the control of a pure double (resp. simple) integrator. This is why tuning the gains of our intelligent controllers is quite straightforward.

**Remark 2.2:** It should be emphasized, if \( \nu = 2 \) (resp. 1), that Equation (3) (resp. (5)) is mathematically sufficient for ensuring stability around the reference trajectory. The integral term \( K_I \int e \) in Equation (3) (resp. (5)) nevertheless adds some well known robustness properties.

### III. Connections between classic and intelligent controllers

#### A. PI and i-P

1) A crude sampling of PIs: Consider the classic continuous-time PI controller

\[ u(t) = k_p e(t) + k_i \int e(\tau) d\tau \]  

(7)

A crude sampling of the integral \( \int e(\tau) d\tau \) through a Riemann sum \( I(t) \) leads to

\[ \int e(\tau) d\tau \approx I(t) = I(t-h) + he(t) \]

where \( h \) is the sampling interval. The corresponding discrete form of Equation (7) reads:

\[ u(t) = k_p e(t) + k_i I(t) = k_p e(t) + k_i I(t-h) + k_i h e(t) \]

Combining the above equation with

\[ u(t-h) = k_p e(t-h) + k_i I(t-h) \]

yields

\[ u(t) = u(t-h) + k_p (e(t) - e(t-h)) + k_i h e(t) \]  

(8)

**Remark 3.1:** A trivial sampling of the “velocity form” of Equation (7)

\[ \dot{u}(t) = k_p \dot{e}(t) + k_i e(t) \]

yields

\[ \frac{u(t) - u(t-h)}{h} = k_p \left( \frac{e(t) - e(t-h)}{h} \right) + k_i e(t) \]

which is equivalent to Equation (8).

2) Sampling i-Ps: Utilize, if \( \nu = 1 \) in Equation (3), the i-P (6), which may be rewritten as

\[ u(t) = \frac{\dot{y}^*(t) - F + K_P e(t)}{\alpha} \]

Replace, according to the computer implementation in [6], [7], \( F \) by \( \dot{y}(t) - \alpha u(t-h) \) and therefore by

\[ \frac{y(t) - y(t-h)}{h} - \alpha u(t-h) \]

It yields

\[ u(t) = u(t-h) - \frac{e(t) - e(t-h) + K_P e(t)}{h \alpha} \]  

(9)

3) Comparison: **FACT.** Equations (8) and (9) become identical if we set

\[ k_p = -\frac{1}{\alpha h}, \quad k_i = \frac{K_P}{\alpha h} \]  

(10)

**Remark 3.2:** It should be emphasized that the above property, defined by Equations (10), does not hold for continuous-time PIs and i-Ps. This equivalence is strictly related to time sampling, i.e., to computer implementation, as demonstrated by taking \( h \to 0 \) in Equations (10).

#### B. PID and i-PD

Extending the calculations of Section III-A is quite obvious. The velocity form of the PID

\[ u(t) = k_p e(t) + k_i \int e(\tau) d\tau + k_d \dot{e} \]

reads \( \dot{u}(t) = k_p \dot{e}(t) + k_i e(t) + k_d \ddot{e}(t) \). It yields the obvious sampling

\[ u(t) = u(t-h) + k_p h \dot{e}(t) + k_i h e(t) + k_d h \ddot{e}(t) \]  

(11)

If \( \nu = 2 \) on the other hand, Equation (4) yields \( u(t) = \frac{1}{\alpha} (\dot{y}^*(t) - F + K_P e(t) + K_D \dot{e}(t)) \). From the computer implementation \( F = \dot{y}(t) - \alpha u(t-h) \), we derive

\[ u(t) = u(t-h) - \frac{1}{\alpha} \dot{e}(t) + \frac{K_P}{\alpha} e(t) + \frac{K_D}{\alpha} \dot{e}(t) \]

(12)

**FACT.** Equations (11) and (12) become identical if we set

\[ k_p = \frac{K_D}{\alpha h}, \quad k_i = \frac{K_P}{\alpha h}, \quad k_d = -\frac{1}{\alpha h} \]  

(13)

#### C. i-PI and i-PID

Equation (12) becomes with the i-PID (5)

\[ u(t) = u(t-h) - \frac{1}{\alpha} \dot{e}(t) + \frac{K_P}{\alpha} e(t) + \frac{K_I}{\alpha} \int e + \frac{K_D}{\alpha} \dot{e}(t) \]  

(14)

Introduce the PII2D controller

\[ u(t) = k_p e(t) + k_i \int e(\tau) d\tau + k_{i2} \int \int e d\tau d\sigma + k_{d2} \dot{e}(t) \]

where a double integral appears. To its velocity form \( \dot{u}(t) = k_p \dot{e}(t) + k_i e(t) + k_{i2} \int \int e d\tau d\sigma + k_{d2} \ddot{e}(t) \) corresponds the sampling

\[ u(t) = u(t-h) + k_p h \dot{e}(t) + k_i h e(t) + k_{i2} h \int \int e d\tau d\sigma + k_{d2} h \ddot{e}(t) \]

Such double integrals do not seem to be common in control engineering.
which is identical to Equation (14) if one sets
\[ k_p = \frac{K_D}{\alpha h}, \quad k_i = \frac{K_D}{\alpha h}, \quad k_{ii} = \frac{K_I}{\alpha h}, \quad k_d = -\frac{1}{\alpha h} \] (15)

The connection between iPIs and PII follows at once.

D. Table of correspondence

The previous calculations yield the following correspondence table between the gains of our various controllers:

**Remark 3.3:** Due to the form of Equation (6), it should be noticed that the tuning gains of the classic regulators ought to be negative.

E. The explanation

The previous calculations and Table 1 explain why sampled classic PI and PID controllers take into account, if their gains are properly tuned, the structural term \(-F/\alpha\), which contains all the structural information of the unknown nonlinear systems, in Equations (3), (4), (5), (6). The superiority of intelligent controllers, which was already noted in [6], [7], is however confirmed:

1) Tuning the gains of intelligent controllers is straightforward whereas it is complex and painful for classic PIDs in spite of all the numerous existing rules in the literature (see, e.g., [2], [16]).

2) Contrarily to intelligent controllers, a correctly tuned classic PI or PID controller is unable to take into account heat effects, ageing processes, characteristic dispersions due to mass production, ….

3) Fault tolerant control is much better handled by intelligent controllers than by classic ones.

IV. CLASSIC VERSUS INTELLIGENT CONTROLLERS

For the nonlinear system
\[ \dot{y} + y^3 = 2u \] (16)
we deduce a classic PI controller thanks to a method due to Broïda and Dindeleux [5] which improves the well-known Ziegler-Nichols rules (see, e.g., [2], [16]). Note however that the open loop response of System (16), with \( y(0) = 0 \), is somehow difficult to exploit as shown by Figure [1]. It yields

- a delay system

\[ \frac{k_e^{-\tau s}}{1 + Ts} \]

where \( k = 1.160, \ T = 0.401, \ \tau = 0.044; \)

- a PI where \( k_p = 6.350, \ k_i = 15.817. \)

Figures [2] and [3], which depict the simulation results for the above PI and an i-PI, do not show any significant difference. Remember however that the i-PI, where \( \alpha = 1, \ K_P = 6, \ K_I = 9, \) does not necessitate any cumbersome identification procedure.

Without any new calibration of the PI for another operating range Figure [4] shows a deterioration of the performances, whereas the performances of the i-PI, which are depicted in Figure [5] remain good.

Introduce now a fault accommodation via a control power loss \( \alpha_{\text{pure}} = 0.996^{t/\alpha} \times u, \ t > 4, \) where the sampling time \( h = 0.01s. \) The i-PI behaves then much better (Figure [2]) than the PI (Figure [3]). Note nevertheless a small deviation of the i-PI controller when the power loss becomes quite important (Figure [3](b)).

V. CONCLUSION

The above numerical simulations as well as many existing experimentations (see [6], [7], and [1], [4], [10], [11], [13], [14], [18]) demonstrate that intelligent PID controllers yield better performances than classic ones. This is achieved moreover thanks to a quite straightforward and natural gain tuning, which contrasts with the numerous complex rules for classic PIDs. Those considerations as well as the results of this communication imply therefore

- that classic PIDs might become obsolete,

- a change of paradigm for control engineering, and for its teaching (see, e.g., [6], [7], and [17]).

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TABLE I: Correspondence between the gains of sampled classic and intelligent controllers.

|       | i-P | i-PD | i-PI | i-PID |
|-------|-----|------|------|-------|
| PI    | $k_p$ | $-1/\alpha h$ | $k_i$ | $K_p/\alpha h$ |
| PID   | $k_p$ | $K_D/\alpha h$ | $k_i$ | $K_P/\alpha h$ | $k_d$ | $-1/\alpha h$ |
| PII   | $k_p$ | $-1/\alpha h$ | $k_i$ | $K_P/\alpha h$ | $k_{ii}$ | $K_I/\alpha h$ |
| PIID  | $k_p$ | $K_D/\alpha h$ | $k_i$ | $K_P/\alpha h$ | $k_{ii}$ | $K_I/\alpha h$ | $k_d$ | $-1/\alpha h$ |

Fig. 1: Open loop case

(a) Input  
(b) Output (--) and denoised output (---)

Fig. 2: PI case

(a) Input  
(b) Output (--) and denoised output (---)

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Fig. 3: i-PI case

Fig. 4: PI in case of large amplitude setpoint change

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Fig. 5: i-PI in case of large amplitude setpoint change

Fig. 6: PI in case of power loss
Fig. 7: i-PI in case of power loss