



Thesis proposal

From limited model knowledge to careful adaptative model-free control

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Context

In automatic control, the concepts developed mostly postulate the knowledge of an analytical model from which we prove the efficiency of the control technique with respect to a priori defined criteria. It is of course possible to consider that the model is imprecise, but in this case, it is generally necessary to quantify the maximum difference that can exist between the model and the reality. This approach can come up against several pitfalls:

- the dynamics present in the control loop are often ignored; for example, there is little work on
 the attitude stabilization of a VTOL UAV that describes the consideration of the dynamics of
 the thrusters.
- Even if the model is well defined, obtaining the parameters remains complex; still in the field
 of flying objects, the inertias are estimated in an approximate way; even the mass may vary
 according to the load on board; the imprecision of the aerodynamic models for small machines
 poses even more complex problems.
- Once the control algorithm has been established from an inaccurate model, which generally
 needs to be discretized to implement it on a digital system, one rarely obtains a response from
 the closed loop in phase with the theoretical dynamics defined in the specifications. In order
 to approach them, an important work of refinement and adjustment is often essential.

Thus, the idea of doing without all or part of a model to control a process leads to several proposals (data-driven control, [1]). Learning control techniques have been proposed using for example neural networks or kernel approximators [10]. The control value is improved by using the error committed with respect to a reference behavior. Time Delay Control (TDC, [11]) uses the control value at the previous sampling step as an information source to compensate for unknown dynamics and disturbances.

Model-free control ([2], [4], [5], [6], [7]) also offers a strong alternative to dispense with a model as a working basis. The objective is to reduce as much as possible the number of tuning parameters applied to an ultra-local system.

The control by numerical inversion of the behavior model ([3], [8]) focuses on the aspect of respecting the specification of the closed loop dynamics. In a first step, an a priori knowledge of the model was





used to build an algorithm based on a behavior prediction table. According to the output objective fixed by the specification, the optimization algorithm looks for the best control allowing to reach the objective at a sampling step. This control has no parameters to be set, the closed loop dynamics specification is used directly.

Challenge

The objective of the present work is to explore methods to produce a control law:

- 1- having no (or very few) parameters to set, they are extracted from the operating specifications;
- 2- allowing to obtain the desired closed loop dynamics in few sampling periods and without manual adjustment, whatever the neglected dynamics, the parametric variations of the model and the disturbances.

This long-term goal will result in the implementation of learning algorithms of the numerical representation of the system both in an off-line context to build the representation from a model if available, but also on-line to optimize the system response after considering its actual behavior. The fact of having a learning behavior means that, for some areas of the state space domain, the information may be tainted with uncertainty or even no information collected at all. The implementation of a criterion of integrity of the data of the representation (for this purpose we could use the theory of belief functions) will allow a fine management of the model in order not to provoke a loss of control due to a presumptuous use of the model.

Short bibliography

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