

# An algebraic approach for behavioral model decomposition

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**Abstract:** This paper presents a constrained decomposition methodology for discrete-event models. Partial models are obtained and used for model-based fault detection and isolation. The constraints of the decomposition ensures that the resulting partial model is decoupled from a given subset of inputs. The constrained decomposition method is formulated using pair algebra concepts as an iterative algorithm for easy implementation.

Keywords: Algebraic approaches, decomposition methods, decoupling, discrete-event systems.

#### 1. INTRODUCTION

This paper presents a decomposition methodology for deterministic behavioral models. The objective is to obtain a partial model decoupled from a given subset of inputs while remaining coupled with respect to another selected inputs subset. The coupling and decoupling properties of the partial model allows us to detect occurrence of only selected unknown inputs or faults and to ignore the rest of them. This scheme is used to produce structured fault indicators to detect and isolate occurring failures based on continuous-time model. The proposed decomposition methodology is based on a particular algebraic formalism, named algebra of functions introduced by Shumsky [1991], Zhirabok and Shumsky [1993] and revisited in Berdjag et al. [2006]. This algebraic formalism remains valid for all types of deterministic models. Therefore, classical fault detection and isolation scheme, described in Patton [1994], may be extended, and specific models like sequential machines may be used for fault detection.

The present paper presents an extension of the decomposition algorithm to propose a constrained decomposition of sequential machines using *pair algebra* Hartmanis and Stearns [1966]. Output data is used to loosen decomposition constraints (output injection technique). Output injection technique is well known and was used to the decoupling continuous-time models.

Fault detection and isolation using discrete event systems was addressed by Sampath et al. [1995] and derived papers. Using output injection improves fault isolation rate. To the best of our knowledge, it was not employed in the past to decouple discrete-event models. This particular point is the major contribution of this paper.

The paper is organized as follows : In section 2, constrained decomposition problem is formulated using set theory framework. In section 3, the decomposition constraints and conditions are detailed. Section 3 contains basic reminders

about *partitions* and *pair algebra* operators. In section 4, constrained decomposition of discrete-event models is presented along with output injection for discrete-event models. An example is given section 5 to illustrate the decomposition and the benefits of output injection. Conclusions and perspectives on future works close the paper.

#### 2. PROBLEM FORMULATION

As a general principle, it is possible to represent deterministic behavioral models, noted  $\Sigma$ , using the following quintuple

$$(\mathcal{X}, \mathcal{U}, \mathcal{Y}, \mathcal{F}, \mathcal{H}) \tag{1}$$

where  $\mathcal{X}$  is the state set,  $\mathcal{U}$  is input set and  $\mathcal{Y}$  is the output set.  $\mathcal{F}$  and  $\mathcal{H}$  are functions defined by :

$$\mathcal{F} : \mathcal{X} \times \mathcal{U} \longrightarrow \mathcal{X} \text{ and } \mathcal{H} : \mathcal{X} \times \mathcal{U} \longrightarrow \mathcal{Y}$$
 (2)

The function  $\mathcal{F}$  is the state function and the function  $\mathcal{H}$  is the output function.

This representation allows to describe continuous-time and discrete-event deterministic models using the same formalism. Indeed, if  $\Sigma$  is a continuous-time model, then the sets  $\mathcal{X}, \mathcal{U}, \mathcal{Y}$  are infinite sets of dimensions n, l, mrespectively, i.e.  $\mathcal{X} \subseteq \mathbb{R}^n, \mathcal{U} \subseteq \mathbb{R}^l, \mathcal{Y} \subseteq \mathbb{R}^m$ , and the state and output functions are defined by :

$$\mathcal{F} : \mathbb{R}^n \times \mathbb{R}^l \longrightarrow \mathbb{R}^n \text{ and } \mathcal{H} : \mathbb{R}^n \times \mathbb{R}^l \longrightarrow \mathbb{R}^m \quad (3)$$

If the model  $\Sigma$  is a discrete-event model with a finite number of states, then the sets are  $\mathcal{X}, \mathcal{U}, \mathcal{Y}$  are finite sets of cardinalities n', l', m':

 $\mathcal{X} = \{x_1, \cdots, x_{n'}\}, \mathcal{U} = \{u_1, \cdots, u_{l'}\}, \mathcal{Y} = \{y_1, \cdots, y_{m'}\}$ If the model  $\Sigma$  is excited by multiple inputs, we assume that  $\Sigma$  contains multiple dynamics. Every single dynamic is affected by a group of inputs, and remains decoupled from the rest. These dynamics are represented by *partial models*.

Definition 1. Let  $\Sigma(\mathcal{X}, \mathcal{U}, \mathcal{Y}, \mathcal{F}, \mathcal{H})$  be a model. The model  $\Sigma_*(\mathcal{X}_*, \mathcal{U}_*, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*)$  is a partial model of  $\Sigma$ , if for a given

relation  $\Theta_{\mathcal{U}} : \mathcal{U} \to \mathcal{U}_*$ , there is two relations  $\Theta_{\mathcal{X}} : \mathcal{X} \to \mathcal{X}_*$ and  $\Theta_{\mathcal{Y}} : \mathcal{Y} \to \mathcal{Y}_*$  satisfying

$$\forall u \in \mathcal{U} , \forall x \in \mathcal{X} , \forall x_* \in \mathcal{X}_* :$$

$$x_* = \Theta_{\mathcal{X}}(x) \Leftrightarrow \begin{cases} \mathcal{F}_*(x_*, \Theta_{\mathcal{U}}(u)) = \Theta_{\mathcal{X}}(\mathcal{F}(x, u)) \\ \mathcal{H}_*(x_*, \Theta_{\mathcal{U}}(u)) = \Theta_{\mathcal{Y}}(\mathcal{H}(x, u)) \end{cases}$$

$$(4)$$

 $(\Theta_{\mathcal{X}}, \Theta_{\mathcal{U}}, \Theta_{\mathcal{Y}})$  is a homomorphism. The partial model  $\Sigma_*$  is coupled with respect to inputs  $u_* \in \mathcal{U}_*$  and decoupled from inputs  $u' \in \mathcal{U}$  if  $u' \notin \mathcal{U}_*$ .

 $\mathcal{U}_*$  is a subset of  $\mathcal{U}$ . If the output set  $\mathcal{Y}_*$  is not empty, then  $\Sigma_*$  provides outputs that are *bisimilar* with  $\Sigma$  outputs. Two outputs are considered as bisimilar, if all of their possible values are taken from the same set, and these values are consistent if the two models are excited by the same input sequence. Thus, if  $\Sigma_*$  and  $\Sigma$  are excited by the same inputs from  $\mathcal{U}$ , then their common outputs will be consistent. Discrepancy appears when the two models are excited by different inputs. Therefore, discrepancy can be used to detect unexpected events in input sequence. Moreover, if the occurring unexpected event belongs to  $\mathcal{U} - \mathcal{U}_*$ , no output discrepancy is observed, since  $\Sigma_*$  is decoupled from  $\mathcal{U} - \mathcal{U}_*$ .

Concretely, decomposition objective is to obtain a decoupled partial model which allows detection of selected faults and ignore the rest of them. The input set  $\mathcal{U}$  is divided in three disjoint subsets :

$$\mathcal{U} = \mathcal{U}_c \cup \mathcal{U}_\rho \cup \mathcal{U}_\gamma \tag{5}$$

where  $\mathcal{U}_{\rho}$  contains inputs to be coupled to,  $\mathcal{U}_{\gamma}$  contains inputs to be decoupled from and  $\mathcal{U}_c$  regroups the rest of inputs.  $\mathcal{U}_{\rho}$  and  $\mathcal{U}_{\gamma}$  forms unknown inputs set. For example, on a real modeled system,  $\mathcal{U}_{\gamma}$  is the set failure events that must be detected,  $\mathcal{U}_{\rho}$  es the set of perturbations or failures that must be ignored and  $\mathcal{U}_c$  is the set of command events. The decoupled partial model  $\Sigma_*$  is described by the quintuple  $(\mathcal{X}_*, \mathcal{U}_*, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*)$  which satisfies :

•  $\mathcal{U}_* \subseteq \mathcal{U}$ 

• 
$$\mathcal{F}(\mathcal{X},\mathcal{U}_*)\subseteq\mathcal{X}_*$$

• 
$$\mathcal{X}_* \cap \mathcal{F}(\mathcal{X}, \mathcal{U}_{\gamma}) = \emptyset$$

• 
$$\mathcal{Y}_* = \mathcal{H}(\mathcal{X}_*, \mathcal{U})$$

The following section details the method to obtain  $\Sigma_*$ , i.e. to determine  $\mathcal{X}_*, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*$ .

#### 3. DECOMPOSITION OF GENERIC BEHAVIORAL MODELS

#### 3.1 Decomposition objective

Consider a partial model  $\Sigma_*$  obtained by decomposition of  $\Sigma$  (1). In order to ensure that  $\Sigma_*$  is decoupled from  $\mathcal{U}_{\gamma}$ and coupled with respect to  $\mathcal{U}_{\rho}$ , decomposition procedure is constrained to *coupling with respect to*  $\mathcal{U}_{\gamma}$  and *decoupling from*  $\mathcal{U}_{\rho}$  properties of the state set  $\mathcal{X}_*$ . Since  $\mathcal{X}_*$  is decoupled, the descriptions

$$(\mathcal{X}_*, \mathcal{U}_*, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*)$$
 and  $(\mathcal{X}_*, \mathcal{U}, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*)$ 

are equivalent for  $\Sigma_*$ .

The functions  $\mathcal{F}_*$  and  $\mathcal{H}_*$  are restrictions of the functions  $\mathcal{F}$  and  $\mathcal{H}$  on the sets  $(\mathcal{X}_*, \mathcal{U}, \mathcal{Y}_*)$ :

$$\mathcal{F}_*: \ \mathcal{X}_* \times \mathcal{U} \to \mathcal{X}_* (x_*, u) \mapsto x_*^+ = \mathcal{F}(x_*, u)$$
(6)

$$\mathcal{H}_*: \ \mathcal{X}_* \times \mathcal{U} \to \mathcal{Y}_* (x_*, u) \mapsto y_* = \mathcal{H}(x_*, u)$$
(7)

The partial model  $\Sigma_*$  is obtained using the *decomposition* function  $\varphi$  defined by :

$$\begin{split} \varphi : \Psi \to \Psi \\ \Sigma \mapsto \Sigma_* \\ (\mathcal{X}, \mathcal{U}, \mathcal{Y}, \mathcal{F}, \mathcal{H}) \mapsto (\mathcal{X}_*, \mathcal{U}, \mathcal{Y}_*, \mathcal{F}_*, \mathcal{H}_*) \end{split} \tag{8}$$

where  $\Psi$  is the set of all the possible quintuples. The function  $\varphi$  determines all the elements of the quintuple that describes  $\Sigma_*$ . Fortunately, the knowledge of the subset  $\mathcal{X}_*$ , if not empty, is sufficient to determine the other remaining elements. Let  $\phi$  be a function that determines  $\mathcal{X}_*$ :

$$\phi: \mathcal{X} \to \mathcal{X}_* \in \Psi_{\mathcal{X}}$$
$$x \mapsto x_* = \phi(x) \tag{9}$$

where  $\Psi_{\mathcal{X}}$  is the set of all the possible subsets generated by elements of  $\mathcal{X}$ .  $\mathcal{Y}_*$  is determined by the relation  $\mathcal{Y}_* = \mathcal{H}(\mathcal{X}_*)$ .  $\mathcal{U}_*$  and  $\mathcal{U}$  are known. If restrictions of  $\mathcal{F}$  and  $\mathcal{H}$ are possible on  $\mathcal{X}_* \times \mathcal{U}$ , then the decoupled partial model of  $\Sigma$  is completely defined.

Hence, the existence of the partial model  $\Sigma_*$  is linked to the existence of a restriction of the state function  $\mathcal{F}$  on  $\mathcal{X}_* \times \mathcal{U}$ . This condition is called *invariance condition*. Output consistency between  $\Sigma_*$  and  $\Sigma$  is linked to the existence of a restriction of the output function  $\mathcal{Y}$  on  $\mathcal{U}_* \times \mathcal{X}_*$ . This condition is called *output condition*.

In the following, the conditions and the constraints are detailed.

#### 3.2 Decomposition constraints

Decoupling constraint  $\Sigma_*$  is decoupled for  $\mathcal{U}_{\gamma}$  if and only if  $\mathcal{U}_{\gamma} \cap \mathcal{U}_* = \emptyset$ . It means that  $\mathcal{X}_*$  do not intersect the state set coupled with respect to the subset  $\mathcal{U}_{\gamma}$ .

$$\mathcal{X}_* \cap \mathcal{X}_\gamma = \emptyset \tag{10}$$

 $\mathcal{X}_*$  and  $\mathcal{X}_{\gamma}$  are given by

$$\mathcal{X}_* = \mathcal{F}(\mathcal{X}, \mathcal{U}_*) \text{ and } \mathcal{X}_{\gamma} = \mathcal{F}(\mathcal{X}, \mathcal{U}_{\gamma})$$
 (11)

Coupling constraint  $\Sigma_*$  is coupled with respect to  $\mathcal{U}_{\rho}$  if  $\mathcal{U}_{\rho} \cap \mathcal{U}_* \neq \emptyset$ . It means that  $\mathcal{X}_*$  intersects the state subset coupled with respect to the subset  $\mathcal{U}_{\rho}$ .

$$\mathcal{X}_* \cap \mathcal{X}_\rho \neq \emptyset \tag{12}$$
$$\mathcal{X}_\rho \text{ is given by } \mathcal{X}_\rho = \mathcal{F}(\mathcal{X}, \mathcal{U}_\rho)$$

## 3.3 Decomposition conditions

Invariance condition Let  $\mathcal{F}_*$  be the restriction of  $\mathcal{F}$  such as

$$\mathcal{F}_* : \mathcal{X}_* \times \mathcal{U} \to \mathcal{X}_*$$
$$(x_*, u) \mapsto x_*^+ = \mathcal{F}(x_*, u) \tag{13}$$

Since  $\Sigma_*$  replicates a part of  $\Sigma$ , the response of the two systems must be equivalent for the same inputs

 $\forall u \in \mathcal{U} , \ \forall x \in \mathcal{X} : \ x_* = \phi(x) \Leftrightarrow \mathcal{F}_*(x_*, u) = \phi\left(\mathcal{F}(x, u)\right)$ (14)

Which implies

$$\mathcal{F}_*(\phi(x), u) = \phi\left(\mathcal{F}(x, u)\right) \tag{15}$$

If this relation is satisfied then the decomposition function  $\phi$  is *invariant* under  $\mathcal{F}$ . By extending the relation (15) to the whole state set, we obtain the invariance condition

$$\mathcal{F}(\phi(\mathcal{X}), \mathcal{U}) \subseteq \phi(\mathcal{X}) \tag{16}$$

Output condition If there is a link between outputs of  $\Sigma_*$  and outputs of  $\Sigma$  for all the states  $x_* \in \mathcal{X}_*$  and  $x \in \mathcal{X}$ , then it is possible to check discrepancy between outputs of  $\Sigma_*$  and  $\Sigma$  and the *output condition* is fulfilled. This condition make sense only if the invariance condition is already fulfilled.

If all the outputs of the two models are bisimilar then following relation is satisfied

$$\forall x \in \mathcal{X} , \forall u \in \mathcal{U} , \exists \zeta, \xi : \xi(\mathcal{H}_*(\phi(x), u)) = \zeta(\mathcal{H}(x, u))$$
(17)

where  $\zeta$ ,  $\xi$  are some functions. If the function  $\mathcal{H}_*$  is a restriction of  $\mathcal{H}$ , the relation (17) is always true. This is the perfect case. Practically, only one single consistent output is required to check consistency of  $\Sigma$  and  $\Sigma_*$ . Relation (17) becomes

$$\exists \zeta, \xi : \xi(\mathcal{H}_*(\mathcal{X}_*, \mathcal{U})) \cap \zeta(\mathcal{H}(\mathcal{X}, \mathcal{U})) \neq \emptyset$$
 (18)

Given that  $\mathcal{H}_*$  is a restriction of  $\mathcal{H}$ , relation (18) can be simplified, giving the final form of output condition :

$$\mathcal{H}(\phi(\mathcal{X}), \mathcal{U})) \cap \mathcal{H}(\mathcal{X}, \mathcal{U}) \neq \emptyset \Leftrightarrow \mathcal{Y}_* \neq \emptyset$$
(19)

#### 3.4 Set delimiters and extension to the finite set case

In order to implement the decomposition conditions and constraints, *set delimiters* are defined. In the case of infinite sets, set delimiters are defined using functions Shumsky [1991]. In the case of finite sets, set delimiters are defined using partitions Hartmanis and Stearns [1966].

The set of all delimiters with the corresponding operations forms a mathematical structure called *algebra*. If definition sets of a model are finite, *pair algebra* is involved Hartmanis and Stearns [1966]. The infinite set case is addressed using *algebra of functions* Shumsky [1991], Zhirabok and Shumsky [1993]. Recently, algebra of functions were used in several topics of model-based monitoring Berdjag et al. [2006]. Unfortunately, pair algebra was not addressed recently.

In the next sections, partitions and pair algebra Hartmanis and Stearns [1966] are presented.

### 4. PARTITIONS AND PAIR ALGEBRA

Some notions on partitions are given, and definitions for the main algebraic operators of pair algebra are detailed.

#### 4.1 Mathematical background

Partition Consider some finite set S. A partition  $\pi$  on S is a collection of disjoint subsets of S whose set union is S. These subsets are called blocks and noted  $B_{\alpha}$  where  $\alpha$  is an element of S.

$$\pi = \{B_{\alpha}\} \text{ such that } \begin{cases} B_{\alpha} \cap B_{\beta} = \emptyset \text{ for } \alpha \neq \beta \\ \bigcup\{B_{\alpha}\} = S \end{cases}$$
(20)

Consider a block B from  $\pi$ , and two elements s and t from S. If s and t are contained in the same block B of  $\pi$ , then we note  $s \equiv t(\pi)$ .

Operations on partitions Let S be a set and let  $\pi_1$  and  $\pi_2$  be two partitions on S. s and t are two elements from S. The following operation and relationship are possible on partitions :

•  $\pi_1.\pi_2$  is the partition on S such that :

$$s \equiv t(\pi_1, \pi_2) \text{ iff } s \equiv t(\pi_1) \text{ and } s \equiv t(\pi_2)$$
  
•  $\pi_1 + \pi_2$  is the partition on  $S$  such that :  
 $s \equiv t(\pi_1 + \pi_2)$  iff there is a sequence in  $S$   
 $s = s_0, s_1, \dots, s_n = t$ 

for which either

 $s_i \equiv s_{i+1}(\pi_1) \text{ or } s_i \equiv s_{i+1}(\pi_2) , \ 0 \le i \le n-1$ 

- $\pi_1 \leq \pi_2$  if and only if  $\pi_1 \cdot \pi_2 = \pi_1$  and  $\pi_1 + \pi_2 = \pi_2$ .  $\pi_2$  is said larger than or equal to  $\pi_1$ .
- $\pi_1 \cong \pi_2$  if and only if  $\pi_1 \le \pi_2$  and  $\pi_2 \le \pi_1$ . Partitions  $\pi_1$  and  $\pi_2$  are equivalent.

The set of all the possible partitions on S is ordered by the order relation  $\leq$ . The smallest partition is noted  $\mathbb{O}$  and the greatest partition is noted  $\mathbb{I}$ . For example, let  $S = \{1, 2, 3\}$ . The smallest partition is given by  $\mathbb{O} = \{\{1\}, \{2\}, \{3\}\}$  and the greatest is given by  $\mathbb{I} = \{\{1, 2, 3\}\}$ 

#### 4.2 Substitution property and pair algebra

Let S, I two sets and  $\delta$  a function defined by

$$\delta: S \times I \longrightarrow S$$

Let  $\pi$  be a partition on S. The partition  $\pi$  is said to have the *substitution property* with respect to the function  $\delta$  if and only if

$$s \equiv t(\pi) \Rightarrow \delta(s, a) \equiv \delta(t, a) \forall a \in I$$
(21)

If  $\pi$  has the substitution property then the function  $\delta_{\pi}$ , defined by  $\delta_{\pi} : \pi \times I \longrightarrow \pi$  such as

$$\delta_{\pi}(B_{\pi}, i) = B'_{\pi} \Leftrightarrow \delta_{\pi}(B_{\pi}, i) \subseteq B'_{\pi}$$

with  $i\in I$  , is the image of  $\delta$  by  $\pi.$   $\delta_{\pi}$  is a restriction of  $\delta$  on  $\pi.$ 

The partition pair is an extension of the substitution property to two partitions. A partition pair  $(\pi, \pi')$  is an ordered pair of partitions on S such as

$$s \equiv t(\pi) \Rightarrow \delta(s, a) \equiv \delta(t, a)(\pi')$$
 (22)

The partition pair is anti-symmetric. If  $(\pi, \pi')$  and  $(\pi', \pi)$  are partition pairs then  $\pi \cong \pi'$ . By the way, if  $\pi$  has the substitution property then  $\pi$  satisfies the relation (22), and  $(\pi, \pi)$  is a partition pair.

The partition pair concept is useful to describe the minimal and the maximal operators m and M.

Definition 2. Let  $\pi$  be a partition on S.  $m(\pi)$  is the minimal partition that forms a partition pair on the left with  $\pi$ , i.e.  $(m(\pi), \pi)$  is a partition pair and if  $(\pi', \pi)$  is a partition pair then  $m(\pi) \leq \pi'$ . The result  $m(\pi)$  is also given by the following relation

$$m(\pi) = \prod \{ \pi_i | (\pi, \pi_i) \text{ is a partition pair } \}$$
(23)

Definition 3. Let  $\pi$  be a partition on S.  $M(\pi)$  is the maximal partition that forms a partition pair on the right with  $\pi$ , i.e.  $(\pi, M(\pi))$  is a partition pair and if  $(\pi', \pi)$  is a partition pair then  $m(\pi) \leq \pi'$ . The result  $M(\pi)$  is also given by the following relation

$$M(\pi) = \sum \{ \pi_i | (\pi_i, \pi) \text{ is a partition pair } \}$$
(24)

Consider know some set of partitions L ordered by the ordering relation  $\leq$ , and a function  $\delta$ . The subset  $\Delta \subseteq L \times$ L of all the partitions pairs with respect to  $\delta$ , with the partition operations "." and "+" forms an algebra called pair algebra. If the pair  $(\pi_1, \pi_2)$  is a partition pair, then we note  $(\pi_1, \pi_2) \in \Delta_{\delta}$ .

#### 5. DECOMPOSITION OF SEQUENTIAL MACHINES

In this section, a constrained decomposition methodology is proposed for sequential machines, which are a common type of deterministic discrete-event models. Sequential machines are noted  $(S, I, O, \delta, \lambda)$  to make distinction with the general case. S, I, O are respectively the state set, the input set and the output set of the model.  $\delta$  is the state function and  $\lambda$  is the output function.

The decomposition problem is formulated as follows : Consider a sequential machine  $\Sigma(S, I, O, \delta, \lambda)$  whose input set is given by

$$I = I_c \cup I_\gamma \cup I_\rho$$

A partial sequential machine  $\Sigma_*$  decoupled from  $I_{\gamma}$  and coupled with respect to  $I_{\rho}$  is searched. The machine  $\Sigma_*$  is defined by the quintuple  $(S_*, I_*, O_*, \delta_*, \lambda_*)$  with

- $S_* = \pi$ , where  $\pi$  is a partition of S.
- $I_* = \pi_I$ ,  $\pi_I$  is a partition of I or of  $I \times O$ .
- $O_* = \pi_O$ , where  $\pi_O$  is a partition of O.  $\delta_* : \pi \times I_* \to \pi$  is a restriction of  $\delta$ .  $\lambda_* : \pi \times I_* \to \pi_O$  is a restriction of  $\lambda$

#### 5.1 Decomposition constraints

In order to express coupling and decoupling constrains using partitions, a neutral element  $i_0$  is added to I.

$$\forall s \in S : \delta(s, i_0) = s \tag{25}$$

Hence,  $\Sigma$  is decoupled from the element  $i_0$  by definition. Consider now a block of the partition  $\pi_I$  that contains  $i_0$ . The corresponding partial model of  $\Sigma$  will be also decoupled from all the elements of this block.

Let 
$$I_{\gamma} = \{a_1, a_2, \ldots\}$$
 and  $I_{\rho} = \{b_1, b_2, \ldots\}.$ 

Decoupling constraint Consider the following partition

 $\pi_{\gamma} = \{\{i_0, a_1, a_2, \ldots\}, \{i_1\}, \ldots, \{i_l\}, \{b_1\}, \{b_2\}, \ldots\}$ (26) where  $i_j$ , with  $j = 1, \ldots, l$ , are elements of  $I_c$ . The partition  $\pi_{\gamma}$  decouples  $I_{\gamma}$ . Using the operator m, the corresponding state set partition is determined.

$$\pi^0_* = m_\delta(\pi_\gamma) \tag{27}$$

 $\pi^0_*$  is the state set partition that is decoupled from  $I_{\gamma}$ . If the machine  $\Sigma_*$  is decoupled from  $I_{\gamma}$  then its state set is a partition of  $\pi^0_*$ , i.e.

$$\pi^0_* \le \pi_* \tag{28}$$

Coupling constraint Consider the partition  $\pi_{\rho}$  that decouples  $I_{\rho}$ 

$$\pi_{\rho} = \{\{i_0, b_1, b_2, \ldots\}, \{i_1\}, \cdots, \{i_l\}, \{a_1\}, \{a_2\}, \ldots\}$$
(29)  
and the corresponding state set partition

$$\bar{\pi}^0_* = m_\delta(\pi_\rho) \tag{30}$$

We saw previously that if the machine  $\Sigma_*$  is decoupled from  $I_{\rho}$  then its state set is a partition of  $\bar{\pi}^0_*$ . Accordingly, if  $\Sigma_*$  is coupled to  $I_{\rho}$  then

$$\bar{\pi}^0_* \nleq \pi_* \tag{31}$$

#### 5.2 Decomposition conditions

Invariance condition Consider the sequential machine  $\Sigma$  and a partition  $\pi$  which has the substitution property with respect to  $\delta$ . It means that the restriction of  $\delta$  on  $\pi$  exists and the quintuple  $(\pi, I, \pi_O, \delta_*, \lambda_*)$  describes a partial model of  $\Sigma$ , for some partition  $\pi_O$  and function  $\lambda_*$ .

It means that if  $\pi$  has the substitution property, i.e.  $(\pi,\pi) \in \Delta_{\delta}$ , then the discrete-event model described by  $(\pi, I, \pi_O, \delta_*, \lambda_*)$  is a partial model of  $\Sigma$ .

From the definitions 2 and 3, if  $(\pi,\pi) \in \Delta_{\delta}$ , then the following relations are satisfied

$$\pi \le M_{\delta}(\pi) \text{ and } \pi \ge m_{\delta}(\pi)$$
 (32)

Given that for the sequential machine case, the relation  $\pi \leq M(\pi)$  implies  $\pi \geq m(\pi)$  and vice versa. Thus, either relation of (32) describes the invariance condition.

#### 5.3 Output condition

Consider a partition  $\pi$  of the state set S. By analogy with the invariance condition, if  $\pi$  has the substitution property, then there is a restriction of  $\lambda$  on  $S_* = \pi$ ,  $I_* = I$  and a partition  $\pi_O$  of O defined by

$$\lambda_{\pi} : S_* \times I_* \longrightarrow \pi_O$$
$$(s_*, i_*) \mapsto \lambda(s_*, i_*)$$

Let  $\pi_{\lambda} = M_{\lambda}(\mathbb{O})$  be a partition induced by the output function  $\lambda$  and the output set O on the state set. If  $\pi \geq \pi_{\lambda}$ is verified,  $(\pi, \pi)$  is a partition pair, all outputs of  $\Sigma$  and outputs of the partial model determined by  $\pi$  are bisimilar, which obviously is the best case.

Practically, to fulfill the output condition, it is sufficient to have one single bisimilar output. It means that partitions  $\pi$  and  $\pi'$  must share the same block and  $\pi' + \pi \neq \mathbb{I}$ . The output condition is given by

$$\pi + M_{\lambda}(\mathbb{O}) \neq \mathbb{I} \tag{33}$$

#### 5.4 Output injection for discrete-event models

In some cases, the constraints of the decomposition are too strong resulting in unsatisfied decomposition conditions. A special technique called output injection may be used to relax the invariance condition. Output injection is a well known technique for continuous-time model decoupling. The main idea is to replace the information loss caused due to the decreased state set  $S_*$  by extending the input set of  $\Sigma_*$  with selected outputs of  $\Sigma$ .

Consider the invariance condition, i.e.  $(\pi,\pi) \in \Delta_{\delta}$ . An injection of the outputs O is equivalent to rectifying the state set partition  $\pi$  with a partition  $\pi_{inj}$  which is determined according to the injected outputs such that

$$\pi^0_* M_\lambda(\pi_{inj}) = \mathbb{O} \tag{34}$$

The extended invariance condition becomes

$$(\pi.\pi_{inj},\pi) \in \Delta_{\delta} \tag{35}$$

The relation (35) is satisfied if the following statements are true.

$$M(\pi) \ge (\pi . \pi_{inj}) \text{ and } \pi \ge m(\pi . \pi_{inj})$$
(36)

If the relations (36) are satisfied, then the decomposition partition  $\pi$  determines a partial sequential machine with an extended input set :

$$\Sigma_*(\pi, I \times \pi_{inj}, \pi_O, \delta_*, \lambda_*)$$

from some partition  $\pi_O$  and function  $\lambda_*$ .

#### 5.5 Decomposition algorithm

Partial model determination problem is solved as a constrained optimization problem. The principle is to determine an initial set of partitions satisfying the decoupling constraint, and to determine  $S_*$  and  $\pi$  using an iterative loop, based on a scheme proposed in Shumsky [1991]. When the extended invariance condition is fulfilled, the loop ends and the output condition and coupling constraint are checked. Finally, the quintuple describing  $\Sigma_*$ is determined (algorithm 5.5).

Algorithm 1. Decomposition algorithm for discrete-event models Require:  $\Sigma(S, I, O, \delta, \lambda)$  {Complete system}

**Require:**  $\pi_{\gamma}, \pi_{\rho}$  { Decomposition constraints}  $\pi_{*}^{0} = m_{\delta}(\pi_{\gamma})$  { Decoupled state set partition }  $\bar{\pi}_{*}^{0} = m_{\delta}(\pi_{\rho})$  { Coupled state set partition }  $\pi_{\lambda} = M_{\lambda}(\mathbb{O})$  { State set partition induced by O } Determine  $\pi_{inj}$  such that  $\pi_{*}^{0}.M_{\lambda}(\pi_{inj}) = \mathbb{O}$  $\xi^0 = \pi^0_*$ , i = 1 {Initialization of the iterative loop} while  $\xi^i \ncong \xi^{i-1}$  do  $\xi^{i+1} = m(\xi^i . \pi_{inj}) + \xi^i$ Increment iend while  $\pi = \xi^i$ if  $\pi = \mathbb{I}$  then return Decoupling impossible else if  $\pi + \pi_{\lambda} = \mathbb{I}$  then Output condition not satisfied by  $\pi$ else Output condition satisfied by  $\pi$ end if if  $\pi \geq \bar{\pi}^0_*$  then Coupling constraint not satisfied by  $\pi$ else Coupling constraint satisfied by  $\pi$ end if return  $\Sigma_*(S_* = \pi, I_* = (I_c \times \pi_{inj}, O_* = \pi_O, \delta_*, \lambda_*)$ end if

#### 6. ILLUSTRATION

Consider a sequential machine  $\Sigma$  described by the table 1.  $\Sigma$  is a five state model, with two known inputs, two

	a	b	f	g	0
1	2	4	5	1	0
2	2	4	2	2	0
3	3	5	3	3	Q
4	3	4	4	3	Q
5	3	1	5	5	Ν

Table 1. Transition table of the model  $\Sigma$ 

unknown inputs f and g and three outputs  $\{O, Q, N\}$ . g

represents the fault we want to detect and occurrence of the event f must be ignored. Therefore,  $\Sigma$  is going to be decomposed in order to obtain the partial model decoupled from  $I_{\gamma} = \{f\}$  and coupled to  $I_{\rho} = \{g\}$ .

The input set partition decoupled from  $I_{\gamma}$  is given by

The corr

$$\pi_{\gamma} = \{\{i_0, f\}, \{a\}, \{b\}, \{g\}\}$$
  
esponding state set partition is given by

 $\pi^0_* = m_\delta(\pi_\gamma) = \{\{1,5\},\{2\},\{3\},\{4\}\}\$ 

The greatest partition  $\pi$  which fulfills the invariance condition is obtained by iteration :

$$\begin{split} \xi^0 &= \pi^0_* \\ \xi^1 &= \xi^0 + m(\xi^0) = \{\{1,4,5\},\{2,3\}\} \not\cong \xi^0 \\ \xi^2 &= \xi^1 + m(\xi^1) = \{\{1,2,3,4,5\}\} \not\cong \xi^1 \\ \xi^3 &= \xi^2 + m(\xi^2) = \{\{1,2,3,4,5\}\} \cong \xi^2 \end{split}$$

Since  $\xi^3 \cong \xi^2$  then  $\pi = \{\{1, 2, 3, 4, 5\}\} = \mathbb{I}$ . In this case, the decomposition is impossible with the decoupling constraints from  $\pi_{\gamma}$ .

A solution may be obtained using output injection. There is three outputs generated by  $\Sigma$  :  $\{O, Q, N\}$ . We don't need to inject all the outputs. The outputs to be injected are a partition  $\pi_{inj}$  of  $O = \{O, Q, N\}$  such that

$$\pi^0_*.M_\lambda(\pi_{inj}) = \mathbb{O}$$

Two partitions of outputs are possible :  $\{\{O, Q\}, \{N\}\}\)$ and  $\{\{O\}, \{Q, N\}\}\)$ . We choose the first one  $\pi_{inj} = \{\{O, Q\}, \{N\}\}\)$ . Choosing the second partition will lead to a similar result. The decomposition algorithm is resumed at the  $\pi$  determination step. The greatest partition  $\pi$ which fulfills the invariance condition with output injection is obtained by the following iteration :

$$\xi^{0} = \pi^{0}_{*}$$
  

$$\xi^{1} = \xi^{0} + m(\xi^{0}.\pi_{inj}) = \{\{1,5\},\{2\},\{3\},\{4\}\} \cong \xi^{0}$$
  
as  $\xi^{1} \simeq \xi^{0}$  then  $\pi = \xi^{1}$ . The decomposition with

Since  $\xi^1 \cong \xi^0$  then  $\pi = \xi^1$ . The decomposition with a decoupling constraint is possible using  $\pi$ .

The verification step consists in the test of the output condition and the coupling to  $I_{\rho} = \{g\}$ .

State set partition induced by the output is obtained by

$$\pi_{\lambda} = \{\{1, 2\}, \{3, 4\}, \{5\}\}$$

The output condition is fulfilled by  $\pi$  since

$$\pi + \pi_{\lambda} = \{\{1, 2, 5\}, \{3, 4\}\} \neq \mathbb{I}$$

To test the coupling constraint, the partition  $\pi_{\rho}$  which decouples  $I_{\rho}$  is calculated :

$$\pi_{\rho} = \{\{i_0, g\}, \{a\}, \{b\}, \{f\}\}\}$$

The corresponding state set partition is given by

$$\bar{\pi}^0_* = m_\delta(\pi_\rho) = \{\{1\}, \{2\}, \{3, 4\}, \{5\}\}\$$

Coupling constraint is fulfilled because

$$\bar{\pi}^0_* \nleq$$

Finally, partial sequential machine  $\Sigma_g$  is determined using the decomposition partition  $\pi$ . The input set of the partial machine is given by

$$I' = \{aO = \{a, \{O, Q\}\}, bO = \{b, \{O, Q\}\}, \dots$$
$$\dots, aN = \{a, \{N\}\}, bN = \{b, \{N\}\}\}$$

The state set is given by

$$S' = \{1' = \{1, 5\}, 2' = \{2\}, 3' = \{3\}, 4' = \{4\}\}$$

$$O' = \{\{O, Q\}, \{N\}\}\$$

The state function  $\delta'$  and the output function  $\lambda'$  are shown through the transition table 2. Output value O' of  $\Sigma_*$  dis

	aO	aN	bO	bN	g	0'			
1'	2'	3'	4'	1'	1'	O'			
2'	2'	2'	4'	4'	2'	O'			
3'	3'	3'	1'	1'	3'	Q'			
4'	3'	3'	4'	4'	3'	Q'			
Та	Table 2. Transition table of $\Sigma_{\alpha}$								

Table 2. Transition table of  $\Sigma_g$ 

equivalent to both output values O or N for  $\Sigma$  and output value Q' is equivalent to Q. For example, if the current output of  $\Sigma$  is O or Q and the output of  $\Sigma_*$  is O' then outputs are consistent.

#### 6.1 Simulations

Simulation results are provided here. The model  $\Sigma$  is excited by two sequences of known and unknown inputs, the first one contains several occurrences unknown input f and the second one contains occurrences of the unknown input g. Sequences composed of known inputs (a, b) of  $seq_1$ and  $seq_2$  combined with outputs from  $\Sigma$  are injected into the decoupled partial model  $\Sigma_*$ . Outputs are compared and a discrepancy indicator sequence is computed. The analysis of the discrepancy indicator sequence permits to detect the event g.

Input sequence containing f The first injected sequence is given by

 $seq_1 = [a, b, a, b, b, f, a, b, a, b, b, f]$ Outputs of  $\Sigma$  and  $\Sigma_*$  are shown figure 1.



Fig. 1. Simulations for the first sequence

An unknown input f appears at the 6th iteration (Figure 3.a). Outputs of  $\Sigma$  and  $\Sigma_*$  remain coherent after f occurs, so the discrepancy indicator (Figure 3.b) remains nil.

Input sequence containing g The second injected sequence is given by

$$seq_2 = [a, b, g, a, b, b, a, b, g, a, b, b]$$

Outputs of  $\Sigma$  and  $\Sigma_*$  are shown figure 2. The unknown input g appears at the 3rd and the 9th iterations (Figure 4.a). Outputs of  $\Sigma$  and  $\Sigma_*$  remain consistent until the first occurrence of event g after what they become inconsistent. The discrepancy can be seen through the indicator (Figure4.b).



Fig. 2. Simulations for the second sequence

#### 7. CONCLUSION

In this paper, constrained decomposition of sequential machines is addressed. An algebraic formulation of the problem and of the solution is presented, based on previous work on continuous-time model decoupling Berdjag et al. [2006]. It is important to notice that the algebraic formalism used to implement the decomposition remain the same in the case of continuous-time models Berdjag et al. [2006].

While the plain decomposition of sequential machines cannot be considered as a contribution, the introduction of a decoupling constraint is new to the best our knowledge. The resulting decoupled partial model can be used to detect unexpected events in a process using a discreteevent model. Another contribution is the use of the *output injection* technique to extend invariance condition in the decomposition methodology.

Future works of authors addresses the decomposition of mixed dynamic models known as hybrid models.

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