Coherence region of the Priority-AND gate: analytical and numerical examples

Ferdinando Chiacchio^{1a}, Jose Ignacio Aizpurua^b, Diego D'Urso^a, Lucio Compagno^a

Abstract

In recent years, the need for a more accurate dependability modelling (encompassing reliability, availability, maintenance, and safety) has favoured the emergence of novel dynamic dependability techniques able to account for temporal and stochastic dependencies of a system. One of the most successful and widely used method is Dynamic Fault Tree that, with the introduction of the dynamic gates, enables the analysis of dynamic failure logic systems such as fault-tolerant or reconfigurable systems. Among the dynamic gates, Priority-AND (PAND) is one of the most frequently used gate for the specification and analysis of event sequences. Despite the numerous modelling contributions addressing the resolution of the PAND gate, its failure logic and the consequences for the coherence behaviour of the system need to be examined to understand its effects for engineering decision-making scenarios including design optimization and sensitivity analysis. Accordingly, the aim of this short communication is to analyse the coherence region of the PAND gate so as to determine the coherence bounds and improve the efficacy of the dynamic dependability modelling process.

Keywords: Markov Chains; Monte Carlo Simulation; Repairable Components; Failure Gates; Dynamic Gates Semantics

1. INTRODUCTION

Fault Tree Analysis is a widely applied technique for the dependability assessment of different applications. Dependability is a term that encompasses a range of attributes which include safety, reliability, availability, maintainability, confidentiality, and integrity. We will not consider confidentiality and integrity attributes because security aspects are outside of the scope of this paper.

For complex systems with dynamic failure logic, the use of Fault Trees is limited because they are unable to accurately capture temporal and stochastic dependencies². To overcome these limitations, Fault Trees have been extended with dynamic gates in the Dynamic Fault Tree (DFT) formalism. Dynamic Fault Trees allow modelling systems with standby logic, sequence of events and conditional triggering dependencies. Table 1 shows the graphical representation of the dynamic gates and their failure behaviour specification.

The design of a Dynamic Fault Tree follows a top-down procedure as with classical (Static) Fault Trees. The top-event represents the system failure condition and this condition is decomposed into a combination of intermediate events defined by Boolean logic and dynamic gates. Basic events are the lowest level events, and generally they represent system component faults. However, the probabilistic analysis of Dynamic Fault Tree models is more intricate than Static Fault Trees. In

Email Address: chiacchio@dmi.unict.it (Ferdinando Chiacchio)

^a Department of Electric, Electronic and Computer Engineering, University of Catania, Italy

^b Department of Electronic and Electrical Engineering, University of Strathclyde, Scotland

¹ Corresponding Author

the last few years, different resolution techniques have been proposed to address the quantification problem. These techniques can be grouped into analytical and simulation approaches. Among the former, Markov chains³ and state space reduction⁴ are suitable when the time to failure/repair follows an exponential distribution, whereas analytical solutions based on structure function⁵ or binary decision diagrams⁶ are normally limited to non-repairable Dynamic Fault Trees. On the other hand, simulation approaches⁷ overcome the limitations of the analytical methods, but they can require long computation times to achieve accurate results.

So as to evaluate the dependability of repairable systems through DFT models, the original formulation of Dynamic Fault Trees² were extended through a formal semantics that defines the behaviour of the DFT gates in case of repairable components (see Manno *et al.*⁸ or Rauzy and Dutuit⁹).

Name	Graphical Representation	Description (N input)
SPARE	primary 1 2 N-1	It triggers only after the primary and all the N-1 spare occur. Spares can be shared with other spare gate. It is possible to identify three different types of spare on the basis of the dormancy factor α which is a multiplicative factor to the active failure rate (when the spare is not in use): $\mathbf{Cold}: \alpha = 0$, spare cannot fail as long is not active. $\mathbf{Hot}: \alpha = 1$, spare can fail at the same rate as when active. $\mathbf{Warm}: 0 < \alpha < 1$, spare can always fail, but at a reduced failure rate when is not in use.
PAND	12 N	It behaves like an AND gate but it triggers only if the input events occur in the order from the leftmost to the rightmost.
SEQ	↑ ↑ ↑ ↑	It forces the input events to occur from the left to the right order. It can model the gradual degradation of a system. It behaves as a Cold Spare Gate.
FDEP	trigger N	This gate models the failure of the dependent input events if the trigger occurs. The output is a dummy or an input for other gates. Note that input events can fail by themselves too.

Table 1. Dynamic gates of Dynamic Fault Tree

The logic of dynamic gates enables the modelling of temporal dependencies of complex systems such as fault-tolerant and reconfigurable systems¹⁰. The semantics of some of these dependencies have been analysed in the literature⁴ including the simultaneity of the events in a PAND gate, the order of components activation in a SPARE gate, the effects of the restoration of the trigger of a FDEP gate, or the treatment of constant probabilities.

Surprisingly, the coherence of the PAND gate¹¹ was never discussed in detail. The coherence is an important property of engineering systems and the most interesting point regarding coherence is that the increase of the reliability of the input components would imply the increase of the reliability of the gate. This is what the designers may be interested in, since they could improve the reliability of the whole system simply by improving the reliability of some basic components. For the PAND gate a detailed analysis of its coherence behaviour may identify useful patterns from the engineering usage perspective. In fact, the increase of the failure probability of the input components of the gate does not always result in an increase of the failure probability of a PAND gate. This is not a

common pattern for other positive static (e.g., AND, OR, XOR, VOTING) gates, as an increase in a component failure probability leads directly to an increase in the failure probability of the gate. This is not typical even for other negative logic gates^{12, 13} (e.g., NOT, NAND, NOR) whose failure probability decreases with the increasing failure probabilities of the components. In fact, as it will be shown, the failure probability of a PAND gate presents a maximum value that depends on the configuration of its input components.

In many industrial applications the identification and quantification of this property is not trivial. However, it may have many implications for engineering design applications. For instance, when engineers need to optimize the system design and this optimization process involves minimising the system failure probability (e.g., strengthen a system section through improved reliability of specific components¹⁷ or maintenance plans which minimize cost and failure probability¹⁴), and the failure model includes time-sequences specified with PAND gates. Therefore, the coherence analysis of the system may be relevant to identify optimal design decisions.

For a system to be coherent it must meet two requirements¹⁵:

- 1) Monotonicity: the reliability improvement of any component will improve the system reliability, and
- 2) Relevance: every component in the system contributes to the system reliability.

For the PAND gate, it will be shown that the monotonicity of the gate depends on its input parameters. Accordingly, the aim of this short communication is to discuss the failure behaviour of the PAND gate and provide a practical way to identify the conditions under which the PAND gate behaves as a coherent gate, within the so-called *coherence region* of the PAND gate. The identification of the coherence region can allow the users to setup a correct intervention for the improvement of a system (e.g., importance measure 14,16, sensitivity analysis 17) and better understand the results of dynamic dependability analyses, particularly for the class of novel techniques belonging to the Dynamic Probabilistic Risk Assessment (DPRA). DPRA techniques make use of non-fixed probability density functions for the failure/repair behaviour of components 18, 19, 20 that, according to the change of operational conditions, need to be interpreted also on the basis of the knowledge of the coherence region of the PAND gate.

The paper is organized as follows. Section 2 describes the failure logic of the PAND gate and frames the design-engineering questions related to its use. In Section 3 authors provide an answer to the questions raised in Section 2 and present a mathematical analysis of a 2-input PAND gate for both repairable and non-repairable components. Finally, Section 4 presents conclusions and discussions.

2. PAND GATE BACKGROUND: FAILURE LOGIC AND ENGINEERING CONSEQUENCES

The original PAND gate failure logic was introduced by Fussell $et\ al.^{21}$ and the definition states that the gate can represent a system that fails if and only if all its inputs fail in left to right order. In any other case the failure does not occur. The graphical representation of the n-input PAND gate is shown in Figure 1.

The PAND gate version with repairable inputs was discussed by Manno et al.⁸

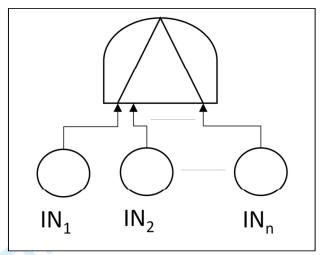


Figure 1: PAND gate with *n* inputs. The gate triggers iff inputs fail in the order from left to right.

Equivalently, the generic *n*-input PAND gate can be modelled with (n-1) interconnected 2-input PAND gates²². Generally, the left side input of the PAND gate corresponds with the monitoring system (sensors or alarms) whereas the right side input corresponds with the actuators^{2,4}. For simplicity, the analysis presented in this paper focuses on the PAND gate with two inputs, but results are directly generalizable to *n*-input PAND gates.

Figure 2 shows the reachability graphs of a 2-input PAND gate for non-repairable² and repairable input events⁸, where the left and right inputs of the PAND gate are components A and B respectively. F_x and R_x denote respectively the failure and the repair transitions of component x associated to the generic failure and repair probability distributions.

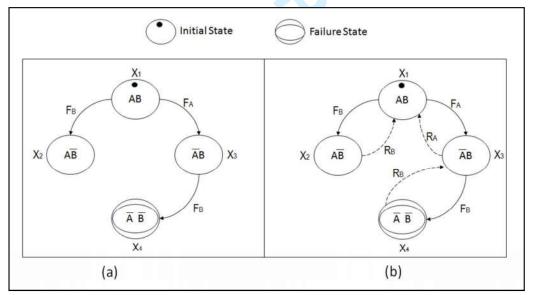
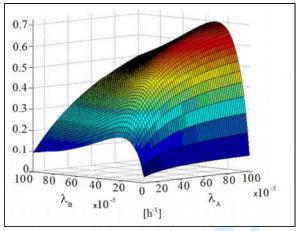


Figure 2: reachability graph for a 2-input PAND gate with (a) non-repairable and (b) repairable components

In the non-repairable model shown in Figure 2.a, if component B fails before A, it is possible to identify an absorbing safe state $A\overline{B}$ in which the system remains permanently. As for the repairable version shown in Figure 2.b, the state $A\overline{B}$ is not absorbing because it is assumed that component B can be repaired and this event brings the system back to the initial state.

Figure 3 shows the sensitivity analysis of the 2-input PAND gate varying the input failure rates in the range of values between $[0, 10^{-3}]$ failures per hour ($[h^{-1}]$) and Figure 4 shows the section of the 3D plot of Figure 3 by fixing the failure rate of input A, $\lambda_A = 10^{-4} h^{-1}$. Observing the two dimensional plot in Figure 4, it is possible to notice that the probability of failure of the PAND gate has an absolute maximum value in λ_B^* and starts decreasing for $\lambda_B > \lambda_B^*$.



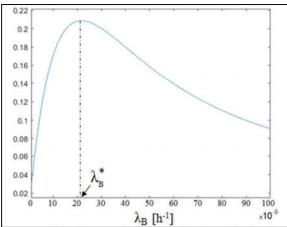


Figure 3: sensitivity analysis for a 2-input PAND gate (with t = 8760h)

Figure 4: section in the plane $\lambda_B (\lambda_A = 10^{-4} \text{ h}^{-1})$

According to the results of this sensitivity analysis, the reliability of the PAND system can sometimes be improved by selecting a component B with a lower reliability. This is the typical characteristic of an incoherent system and, from a design-engineering point of view, raises questions that require some clarifications:

- 1) May the engineer design the system by selecting a component B (the actuator) less reliable than the component A (the monitoring component)?
- 2) How should the engineer select the system components if he/she is allowed to substitute them with other components having different reliability characteristics?

Due to the incoherent behaviour of the PAND gate, the answer to these questions is not straightforward. But, a qualitative analysis of the PAND gate with non-repairable components (Figure 2.a) can reveal some interesting properties that help to answer the first question. Namely, at steady-state the reliability of the system does not tend to zero, as expected in a system with non-repairable components, because it is distributed between the failure state AB and the safe state AB. This observation allows us to answer to the first question and agree with the fact that engineers should design a system in which the actuator component (i.e., input component B) is less reliable than the monitoring component (i.e., input component A). In fact, when the system transits towards the safe state AB (i.e., the actuator component fails), the failure of the system is avoided. This applies for both non-repairable (i.e., the system remains in the state AB) and repairable components.

In the repairable case, we can assume a hypothetical engineering scenario in which component B fails and gets repaired repeatedly such that the system transits from the initial state AB to the safe state $A\overline{B}$ and back continuously. Clearly, from a practical point of view, this scenario is not desirable as it carries the major inconvenience linked with the service unavailability of the system and the economic loss of its maintenance and restoration. In order to be effective, it must remain as much as possible in the initial state AB.

An answer to the second question can offer the opportunity to tackle the paradox of the previous observation. In this case, the problem to solve is how to select the system components so as to

increase the system dependability, but avoiding the issue of the service unavailability. In the next section we will address this issue analysing in deep the incoherent nature of the PAND gate. We will provide some useful information about the PAND gate behaviour and a practical way to recognize the coherence region of the gate.

3. COHERENCE ANALYSIS OF THE PAND GATE

Let us define $Y = \{v_1, v_2, ..., v_n\}$ as the vector of variables that characterize the probability density function (PDF) of the time to failure (or time to occur) of a component (or an event). For instance, the well-known Poisson PDF (modelling random failures) is characterized by a single parameter, called failure rate $(v_1 = \lambda)$, or the Weibull PDF (used to model the bathtub behaviour) has two variables, called shape $(v_1 = \beta)$ and scale $(v_2 = \gamma)$ parameters. Under this setting, it is possible to define the *coherence region*, CR, as:

$$CR: \{ v_i \in \mathbb{R}: \frac{\partial F(Y)}{\partial v_i} \ge 0, i = 1, \dots, n \}$$
 (Eq. 1)

Eq. 1 defines a n-dimensional space in which the probability of failure of a PAND gate F(Y) is monotonically increasing in all its argument v_i belonging to the set of real number \mathbb{R} .

In order to simplify the problem, the behaviour of the gate can be studied using the corresponding Continuous Time Markov Chain model, assuming constant random failures/repairs rate ($F_A = \lambda_A$, $F_B = \lambda_B$, $R_A = R_B = \mu$). Equations in (Eq. 2) show the Kolmogorov differential equations related to the transient analysis of the PAND gate with repairable components (Figure 2.b):

$$\begin{split} \dot{x}_1 &= (x_2 + x_3)\mu - x_1(\lambda_A + \lambda_B) \\ \dot{x}_2 &= x_1\lambda_B - x_2\mu \\ \dot{x}_3 &= x_1\lambda_A - x_3(\mu + \lambda_B) + x_4\mu \\ \dot{x}_4 &= x_3\lambda_B - x_4\mu \end{split}$$
 (Eq. 2)

To obtain the system of equations that defines the scenarios of Figure 2.a, it is possible to simply set $\mu = 0$. The system of differential equations in (Eq. 2) can be written in the matrix form, using the infinitesimal generator matrix (or Generator Matrix) Q:

$$\dot{X} = QX$$

$$\dot{X} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \dot{x}_4 \end{bmatrix} \quad Q = \begin{bmatrix} -(\lambda_A + \lambda_B) & \mu & \mu & 0 \\ \lambda_B & -\mu & 0 & 0 \\ \lambda_A & 0 & -(\mu + \lambda_B) & \mu \\ 0 & 0 & \lambda_B & -\mu \end{bmatrix} X = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$
(Eq. 3)

The probability $P(X_i) = x_i(t)$ of the system to sojourn at time t in the ith state can be found by integrating the system of differential equations (Eq. 3). According to the reachability graph in Figure 2, for the PAND gate, the failure of the system occurs in the state X_4 .

The simplest case to solve analytically is the PAND gate with non-repairable input components (Figure 2.a). This scenario represents the most common example in literature of Dynamic Fault Trees. In the next sections we will examine the steady state and coherence region analysis by starting from the non-repairable case and we will gradually increase the complexity by including

repair rates (Section 3.2) and the restoration of component B in the failure gate⁸ configuration (Section 3.3). The mission time and the failure rate of the components A and B are taken from a previous work²³ and represent the typical scenario of an industrial case of study. The failure and repair rates used in the numerical examples showed in the rest of the paper are measured respectively in number of failures and number of repairs per hours ([h⁻¹]).

3.1 Non-repairable components

According to the reachability graph of Figure 2.a it is possible to identify two absorbing states (X₂ and X₄) that are, by definition of PAND gate², respectively a safe and a failure state. Equations (Eq. 4.a) to (Eq. 4.d) show the solution for the system (Eq. 2) with $\mu = 0$:

$$x_1(t) = e^{-(\lambda_A + \lambda_B)t}$$
 (Eq. 4.a)

$$x_2(t) = \frac{\lambda_B}{\lambda_B + \lambda_A} \left(1 - e^{-(\lambda_A + \lambda_B)t} \right)$$
 (Eq. 4.b)

$$x_3(t) = e^{-\lambda_B t} - e^{-(\lambda_A + \lambda_B)t}$$
 (Eq. 4.c)

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(Eq. 4.b)
$$x_{3}(t) = e^{-\lambda_{B}t} - e^{-(\lambda_{A} + \lambda_{B})t}$$
(Eq. 4.c)
$$x_{4}(t) = \frac{\lambda_{A}}{\lambda_{A} + \lambda_{B}} - e^{-\lambda_{B}t} + \frac{\lambda_{B}e^{-(\lambda_{A} + \lambda_{B})t}}{\lambda_{A} + \lambda_{B}}$$
(Eq. 4.d)

Since the analytical determination of the coherence region is complex, it is possible to evaluate numerically the positivity of the partial derivatives of $x_4(t) = P(X_4)$ by fixing the mission time and the failure rates of one of the two components alternatively. That is, so as to evaluate the partial derivative of the PAND gate with respect to the failure rate of component A, we fix the mission time and the failure rate of component B and vice-versa.

Figure 5 shows the partial derivative of $x_4(t)$ with respect to the failure rate of component A, $\frac{\partial P(X_4)}{\partial \lambda_A}$, in the range of values $\lambda_A \in [0, 10^{-2}] \text{ h}^{-1}$ with mission time $T_m = 8760 \text{h}$, and $\lambda_B = 7.5 \text{x} 10^{-4} \text{ h}^{-1}$.

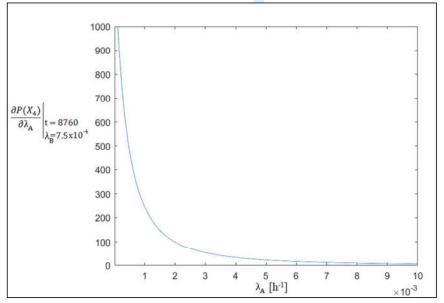


Figure 5: derivative of P(X₄) with respect to the variable λ_A , with t = 8760 and λ_B =0.00075

According to results shown in Figure 5, when we compute the partial derivative of the PAND gate with respect to the failure rate of component A, the PAND gate will always behave as a

coherent system $\left(\frac{\partial P(X_4)}{\partial \lambda_A} > 0, \forall \lambda_A \in [0, 10^{-2}] \ h^{-1}\right)$ and an increase of the failure rate of component A turns in the increase of the PAND gate unreliability.

Assuming the same parameters as in Figure 5, the partial derivative of $x_4(t)$ with respect to the failure rate of component B is shown in Figure 6 (mission time $T_m = 8760h$ and $\lambda_A = 1.7 \times 10^{-4} \, h^{-1}$).

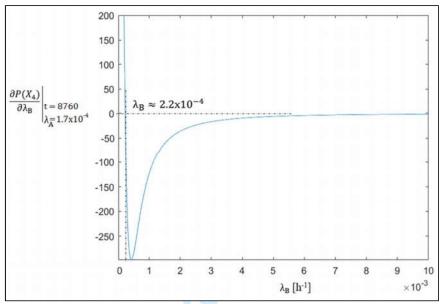


Figure 6: derivative of P(X₄) with respect to the variable λ_B , with t = 8760 and λ_A =0.00017

We can see in Figure 6 that for failure rates higher than $2.2x10^{-4}$ h⁻¹ the system behaves incoherently $(\frac{\partial P(X_4)}{\partial \lambda_B} < 0)$ because the monotonicity condition is not satisfied. Notice also that when the failure rate of component B is much higher than the failure rate of component A $(\lambda_B >> \lambda_A)$, in the long term $P(X_2) = 1$. That is, if $t \to \infty$, $P(X_1) = P(X_3) = 0$, $P(X_2) = \lambda_B/(\lambda_A + \lambda_B)$ and $P(X_4) = \lambda_A/(\lambda_A + \lambda_B)$. Setting $\lambda_B >> \lambda_A$, $P(X_2) = \lambda_B/(\lambda_A + \lambda_B) \approx 1$, while $P(X_4) = \lambda_A/(\lambda_A + \lambda_B) \approx 0$. This is an inherent property of the PAND gate which is in contrast with the definition of the reliability of a system²³. In fact, reliability is a monotonic decreasing probability function that tends to zero for t approaching infinity. In other words, all the absorbing states of the corresponding system reachability graph must be of failure for the system. This shortage has been tackled in Manno⁸ with the introduction of the failure gate and, according to this design, authors suggest to include at least a repair rate from X_2 to X_1 , when interested in the evaluation of the system reliability (see Section 3.3).

3.2 Repairable components

In case of repairable components, it is more appropriate to grasp to the concept of system availability²³ and skip the transient analysis in favour of the steady-state regime. To do that, it is possible to solve the balance equations:

$$\dot{X} = QX = 0$$
 (Eq. 5)

subject to the condition

$$\sum_{i} X_{i} = 1$$
 (Eq. 6)

The solution of the algebraic system (Eq. 5-6) is:

$$P(X_1) = \frac{\mu^2}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.a)

$$P(X_2) = \frac{\lambda_B \mu}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.b)

$$P(X_3) = \frac{\lambda_A \mu}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.c)

$$P(X_4) = \frac{\lambda_A \lambda_B}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.d)

$$P(X_2) = \frac{\lambda_B \mu}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.b)

$$P(X_3) = \frac{\lambda_A \mu}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.c)

$$P(X_4) = \frac{\lambda_A \lambda_B}{(\lambda_A + \mu)(\lambda_B + \mu)}$$
 (Eq. 7.d)

To determine the coherence region, it is sufficient to evaluate the positivity of the partial derivatives of $P(X_4)$. It can be shown that:

$$\frac{\partial P(X_4)}{\partial \lambda_A} = \frac{\lambda_B \mu}{(\lambda_A + \mu)^2 (\lambda_B + \mu)}$$
 (Eq. 8)

and

$$\frac{\partial P(X_4)}{\partial \lambda_B} = \frac{\lambda_A \mu}{(\lambda_B + \mu)^2 (\lambda_A + \mu)}$$
 (Eq. 9)

are always positive for each positive value of failure and repair rate. Since we are not interested in negative failure rates, we can conclude that the repairable PAND gate does not behave incoherently in \mathbb{R}^+ .

3.3 Failure PAND Gate with repairable inputs

The adoption of the failure PAND gate⁸ allows to solve the limit discussed in Section 3.1 concerning the absorbing nature of the safe state X₂. With this configuration we assume that component B can be repaired only when the component B fails before A. Figure 7 shows the reachability graph of this configuration and it is easy to observe that, at steady state, $P(X_4) = 1$ for t $\rightarrow \infty$.

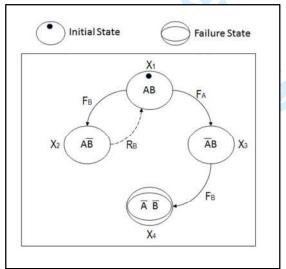


Figure 7: reachability graph for a 2-input failure PAND gate with repairable components

As in Section 3.1, the determination of the coherence region can be performed numerically. In this case, the model presents one more variable, namely the distribution function R_B linked to the repair transition. For the sake of simplicity, in the numerical example used we assume that this transition is exponentially distributed with rate $\mu = 1 \text{ h}^{-1}$. This setting expresses the practical circumstance of a system in which repair transitions are much faster than faults ($\mu >> \lambda_A$, $\mu >> \lambda_B$).

Again, setting the mission time to $T_m=8760h$, Figure 8 and Figure 9 show the positivity of the partial derivative of $x_4(t)$, in the range of values $\lambda \in [0,\ 10^{\text{-1}}]\ h^{\text{-1}}$, respectively for $\frac{\partial \mathit{P}(\mathit{X}_4)}{\partial \lambda_A}$ and $\frac{\partial \mathit{P}(\mathit{X}_4)}{\partial \lambda_B}$, with $\lambda_A=1.7x10^{\text{-4}}\ h^{\text{-1}}$ and $\lambda_B=7.5x10^{\text{-4}}\ h^{\text{-1}}$.

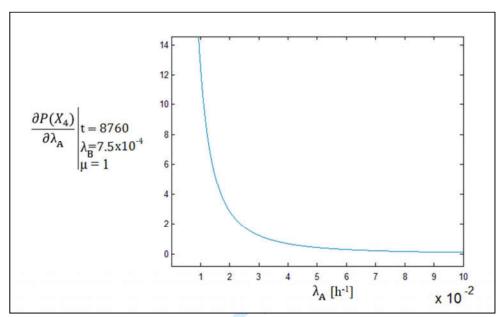


Figure 8: derivative of $P(X_4)$ for the failure gate model, with respect to the variable λ_A (t = 8760 h and λ_B =0.00075 h⁻¹)

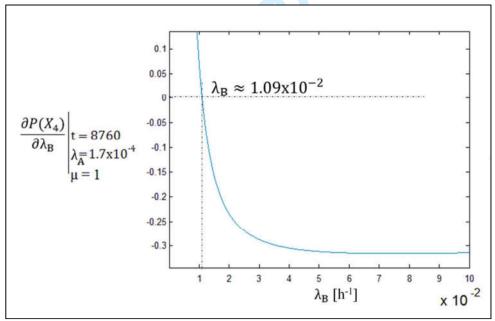


Figure 9: derivative of $P(X_4)$ for the failure gate model, with respect to the variable λ_B (t = 8760 h and λ_A =0.00017 h⁻¹)

Figure 8 demonstrates that, with reference to the failure rate λ_A , the PAND gate keeps behaving as a coherent system $\left(\frac{\partial P(X_4)}{\partial \lambda_A} > 0, \forall \lambda_A \in [0, 10^{-1}] \text{ h}^{-1}\right)$ and an increase of the failure rate of

component A turns in the increase of the PAND gate unreliability. Conversely, Figure 9 shows that values of failure rate higher than $1.09 \times 10^{-2} \, h^{-1}$ make the system behaving incoherently $(\frac{\partial P(X_4)}{\partial \lambda_B} < 0)$.

4. CONCLUSIONS

This short communication discusses the incoherence nature of the PAND gate, one of the most important and utilised gates of Dynamic Fault Tree analysis. The incoherence of the PAND gate emerges due to the existence of the safe state in which the system sojourns in case the events do not trigger in the order specified by the failure logic (from left to right).

In this paper, a 2-input PAND gate for non-repairable and repairable components has been analysed. To simplify the experimental scenario, failure and repair times of occurrence have been set so as to follow the exponential distribution of probability. Results have shown that the failure probability of a PAND gate can present a maximum value; therefore the increase of the reliability of the input components does not always result in an increase of the gate reliability. This behaviour has been revealed for the PAND gate with non-repairable components and for a PAND gate in which the second input component can be repaired only if it fails before the first one. Conversely, the PAND gate with repairable components behaves always as a coherent system.

The coherence analysis may identify a useful region for optimization and sensitivity analysis purposes, and in order to better interpret the results of dynamic dependability models. To be aware of this behaviour, in particular when it is required to perform an improvement of the system reliability, it is suggested to identify the coherence region of operation, studying the positivity of the partial derivative of the failure probability function with respect to the input parameters. This method is not always feasible because the quantification of the closed analytical solution of the PAND gate can be tedious, in particular when systems present non-exponential distributions of probability and repairable components. In this latter case, a well-tuned simulation campaign can help into the identification of approximate boundaries.

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