Improving Dependability of Robotics Systems

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INTRODUCTION

The use of robotics systems is widespread and spans a variety of application areas. From healthcare, to manufacturing, to nuclear power plants, to space missions, these systems are typically conceived to perform difficult, dangerous or critical tasks. The nature of such tasks (e.g., surgery operations, radioactive waste clean-up or space mining) places high demands on the dependability of robotics systems.

The preoccupations in the dependability of robotics systems are not new. Fault Tree Analysis (FTA; Vesely, 1981) and Failure Modes and Effects Analysis (FMEA; IEEE Std.352, 1987) are among the most often used techniques in various domains of robotics. For instance, Visinsky, Walker, and Cavallaro (1993) describe the use of FTA for robots operating in remote and hazardous environments. Other fields of application include industrial robots like in Karbasian, Mehr, and Agharajabi (2012), and modular and swarm robots like in Murray, Liu, Winfield, Timmis, and Tyrrell (2012).

The widespread use of FTA in the dependability assessment of complex systems is mainly due to the flexibility and ease of use of the fault trees. These are static (i.e., 'pure' Boolean) models, and therefore enable the use of efficient Boolean calculus in the elimination of component failures that are irrelevant to the total failure of the system. This logical reduction (known as qualitative analysis) simplifies the process to produce overall probabilities of system hazards (i.e., quantitative analysis). Nevertheless, such convenience comes with the loss of the significance of the sequencing of failure events—i.e., the dynamic features often exhibited by modern systems cannot be captured by combinatorial models like this type of fault trees.

Robotics systems are certainly not an exception when it comes to sequence-dependent failures. For example, preclusion of the dynamic aspects due to the use of static fault trees in the analysis of modular robotic systems is clearly noted in Murray et al. (2012). To overcome such drawback, an alternative can be the utilization of fault trees that are extended with capabilities to capture the dynamic features. A well-known example is the Dynamic Fault Tree (DFT) approach (Dugan, Bavuso, & Boyd, 1992). This method was primarily conceived for quantitative analysis, which is often state-based (i.e., Markov analysis which is based on state transition diagrams [Markov models] is the DFT most prominent solving technique). That is, the full power of the Boolean methods was sacrificed here, especially when it comes to analyzing the dynamic parts of the system at the level of the fault tree (i.e., reducing the DFT).

Theoretically, some later research efforts have provided workarounds to the question of FTA with dynamic aspects. To deal with it, a technique which is relevant to this article consists of extending the Boolean methods with temporal logic calculus. In this connection, a set of temporal laws that enable qualitative analysis of fault trees extended with dynamic features can be found in Walker and Papadopoulos (2009). In the same vein, the algebraic formalism in Merle, Roussel, Lesage, and Bobbio (2010) proposes formal descriptions of dynamic behaviors and provides proofs of a number of theorems useful for the qualitative analysis of this type of fault trees. The

latter approach also deals with the corresponding probabilistic algebraic analysis.

In practice, automation of such advanced FTA as part of integrated dependability and systems engineering processes requires an automated generation and synthesis of these fault trees from failure behavioral models that are linked to the system specifications. The work in Mahmud, Walker, and Papadopoulos (2012) describes a suitable approach to generating and synthesizing fault trees that preserve the significance of the event-order from hierarchical models. Application areas for this approach include the automotive domain (Chen, Mahmud, Walker, Feng, Lönn, & Papadopoulos, 2013). More details about integration in an extended FTA through a Model-Based development process can be found in Kolagari et al. (2015).

In this article, emphasis is put on the significance of the sequencing of failure events and its implications in dependability analysis of robotics systems using FTA. Accordingly, a suitable technique for automated generation and synthesis of extended fault trees from system models is presented. Furthermore, we outline a novel approach to automated reduction of these fault trees. The article is structured as follows: the background section provides a literature review and highlights the relevant approaches to generating and synthesizing fault trees from systems models. The next section outlines an algorithm for the logical reduction of fault tree algebraic expressions that are extended with temporal semantics. Then, an advanced fault tree synthesis approach is presented in the following section. Finally, we present some future research directions, then we conclude.

BACKGROUND

A number of approaches to generating fault trees from system models can be found in the literature. However, a special focus is put on automata representations of the dysfunctional behaviors of systems from which the fault trees get generated.

In this regard, the existing approaches fall into two main categories. The first category concerns the generation of static fault trees; it includes approaches that have been used in the context of influential modeling languages like Altarica (Rauzy, 2002) and AADL (Joshi, Vestal, & Binns, 2007). The former has been used in several aerospace projects including Airbus civil aircraft programs. The latter is increasingly being accepted by the aerospace community as a future standard. Moreover, there have been efforts recently to generate static fault trees from modeling languages specific to robotics systems, e.g. RobotML (Yakymets, Dhouib, Jaber, & Lanusse, 2013).

The second category of methods for fault tree generation includes approaches to generating dynamic fault trees (Dehlinger, & Dugan, 2008), and fault trees that are extended with the significant temporal information about the order in which failure events occur (Mahmud, Papadopoulos, & Walker, 2010; Mahmud et al., 2012; Mahmud & Mian, 2013). The DFTs are solved quantitatively, often, by converting them into equivalent Markov models. Although, there have been efforts on the minimization of the obtained Markov models (Crouzen, Hermanns, & Zhang, 2008; Boudali, Crouzen, & Stoelinga, 2010), there is still less focus on qualitative analysis at the level of the fault tree in this method. However, the approach in Mahmud et al. (2010, 2012) and Mahmud and Mian (2013) allows to reduce the extended fault trees into equivalents which can be proven by using the temporal laws and theorems in Walker and Papadopoulos (2009) and Merle et al. (2010). This is a State Automata-based top-down deductive approach to Fault tree synthesis with Order-dependent behaviors, see SAFORA in Mahmud (2012). It has demonstrated its value in the framework of a rich model-based design founded on EAST-ADL (Chen et al., 2013; Kolagari et al., 2015). EAST-ADL is an approach for describing automotive electronic systems through an information model that captures engineering information in a standardized form (EAST-ADL, 2010).

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