Collision hazard Modeling and Analysis in a multi-mobile robots system transportation task with STPA and SPN

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Abstract

This paper investigates the ability to use complex multi mobile robotic systems in risky and dynamic environments, such as industrial plants and laboratories, with the presence of human factor. More specifically, it presents an approach to analyze the dominant risk, extract, model and quantify the hazard scenarios, then propose requirements using the combination between System Theoretic Process Analysis (STPA) and Stochastic Petri Nets (SPN). This approach is demonstrated with a case study related to a chemical transportation task within a miniature analysis laboratory in oil and gas industry. The main purposes of this article are: to investigate how risk and safety of these systems should be managed, to create a framework for modeling and quantifying collision hazard, and to generate the needed constraints and requirements to improve the operation of robots and preserve safety in the laboratory.

The results obtained reveal that the main problems that may hinder the safe operation of robots are those of control, communication, power and retrieving sensors information. The proposed safety measures contribute to improve autonomy features in mobile robots. The STPA-SPN combination offers a better modeling and assessment of robots performance as well as their collision hazard frequency. *Key words*

Hazard Analysis, STAMP, System Theoretic Process Analysis (STPA), Autonomous multi-mobile robots, Stochastic Petri Nets (SPN), Collision Hazard Frequency.

1. Introduction

The current progress in mobile robots technology pushes the world to revolutionize several industry sectors (automotive, nuclear, petrochemical, ...); mobile robots will facilitate human's labor by helping him/her to perform different tasks that are very often difficult and tiring for

workers like cleaning, environmental surveillance, objects transport,... especially with the advent of so-called autonomous control which allows the robot self-navigating without any human intervention. Recently, a lot of researchers care about robotics and deal in their works with safe navigation issues (Liu, 2017),(Fan et al., 2018),(Tang, Thomas and Kumar, 2016),(Szatmary and Richert, 2017),(Liu et al., 2017), (Pandey and Parhi, 2016), (Li and Savkin, 2018); coordination between multi robots(Li et al., 2012),(Li, Kong and Guo, 2014),(Yoo and Kim, 2015), (Mendiburu, Morais and Lima, 2016), (Jin et al., 2019) and with Human-Robot Interaction (Lasota, Fong and Shah, 2017). Moreover, as safety in every institution and for every application is of prime importance, it is a barrier against the widespread of robotic applications(Saenz et al., 2018). Therefore, a number of researchers directed toward managing risks and safety design sectors in which a set of methods or combinations are employed to dealing with the issue of safety

assessment in robotic applications.(Kazanzides, 2009)has combined between two conventional methods Failure Mode and Effects Analysis (FMEA) and Fault Tree Analysis (FTA), the aim is to analyze hazards and implement a safety control loop for a surgical robot. A reverse of the previous approach has been proposed by (Lee and Yam, 2012) where FTA and FMEA are used to examine the potential failures of Human-cooperative robots (HCRs) which lead to a hazardous movement of the robot and provide the suitable safety functions against failures with the required safety level. A novel Hazard and Operability studies (HAZOP) has been proposed by (Böhm and Gruber, 2010) applied for a therapeutic mobile robot. The authors tried to coverage by their approach all kinds of hazards associated with components and related to operations. (Alexander, Herbert and Kelly, 2009) have combined a set of techniques to identify hazards and derive safety requirements for autonomous systems; Energy Trace and Barrier Analysis (ETBA) and checklists firstly used to provide a basic hazard identification then a detailed analysis provided using Functional Failure Analysis (FFA) and HAZOP method. Another approach based on SHARD analysis (Software Hazard Analysis and Resolution in Design) along with a hazard check list is implemented for a personal autonomous mobile robot by (Woodman et al., 2012).(Dogramadzi et al., 2014) have developed an environmental approach called Environmental Survey Hazard Analysis (ESHA) for hazard identification of autonomous mobile robot applications.(Martin-Guillerez et al., 2010), (Machin, 2015) and (Guiochet, 2016) have discussed in their studies the issue of interaction between human and robot by developing the HAZOP technique basing on a unified modeling language (UML) which have been implemented together for a mobile manipulator robot. Nevertheless, the increased complexity of new robotic systems composition, due to the intensive use of software, level of autonomy which characterize their control systems and the number of interaction among components, lead to the appearance of new hazard type which caused by new causal factors rather than failure of components. Traditional risk analysis techniques such as FMEA, HAZOP, FTA..., based on reliability theory and functional decomposition, that describe the system as a number of almost independent subsystems(Scarinci et al., 2019), are no longer adequate to represent indirect interactions among system components and to consider conditions which cause inappropriate behaviors.

Last decades, Leveson developed a method called System Theoretic Process Analysis STPA(Leveson, 2011),(Leveson and Thomas, 2018) based on system theory which considers safety as a system's control problem rather than a component failure problem. STPA is a hazard analysis technique dedicated to identify hazards of automated complex systems and their control/ command architectures which include interactions between their components; that interests to handle hardware, software, human operators and integrate them in a unified process (Scarinci et al., 2019). The STPA method has proven its strength and capability over last few years through its analytical capacity of many varieties of automated and high complex systems including: driving systems(Abdulkhaleq et al., 2018), aircraft Systems(Plioutsias and Karanikas, 2015), (Scarinci et al., 2019), aerial transportation systems(Fleming et al., 2013), aerospace exploration systems (Ishimatsu et al., 2014), robotic surgical systems(Alemzadeh et al., 2014); even autonomous ones such as: autonomous vehicles, (Wróbel, Montewka and Kujala, 2018), autonomous vessels(Banda and Kannos, 2017), autonomous ships (Rokseth, Haugen and Utne, 2019). In addition, STPA has been applied for several industries such as power industry(Khan, Madnick and Moulton, 2018); Subsea (Rachman and Ratnayake, 2015),(Kim et al., 2018), (Zhang et al., 2019b); process industry (Hardy and Guarnieri, 2011), (Rodríguez and Díaz, 2016) and Nuclear industry (Song, 2012), (Thomas, Lemos and Leveson, 2012), (Lee et al., 2013), (Uesako, 2016). Furthermore, a number of comparison studies of STPA hazard analysis with different conventional risk analysis methods have reported positive evaluation about applying STPA on various complex systems, such as(Ishimatsu et al., 2010)who have compared STPA analysis results with FTA results. This comparison showed that STPA identifies additional causal factors than the ones identified by FTA. (Nakao et al., 2011)conclude that STPA makes the generation of safety requirements and constraints easily and flexibly. Also, (Fleming et al., 2013) argue that STPA allows the identification of faults related to software and dynamic behavior of systems, unlike the classical methods such as FMEA, FTA, etc.

It is true that STPA has attracted a big attention from a large number of researchers in a short period of time due to its suitability for complex systems and due to its benefits mentioned above. However, the original STPA is still fully qualitative technique. It does not focus on modeling and assessing hazard scenarios which is considered as a significant step in safety analysis. It is incapable yet to quantify risks. In this context, we aim in this study to deal with this issue by combining STPA with other method allowing to obtain a quantification of risk scenarios. In this article, we adopt as a quantitative method Stochastic Petri Net (SPN), which is best suited in modeling of complex and dynamic systems behavior. Petri Nets (PN)are a state-transition approach that has been proposed to generate large-scale Markov processes (SIGNORET, 2008). It has shown excellent modeling and calculation skills especially when it is coupled with Monte-Carlo simulation in the field of dependability and safety(Malhotra and Trivedi, 1995; Dutuit et al., 1997; Signoret, 2009).

This combination intends to make the analysis more effective and to better assess hazard scenarios. From literature, some research studies have proposed to combine STPA with other methods that offer an evaluation to the risk such as FMEA(La and Kwon, 2018) and Bowtie (Bensaci *et al.*, 2020) for semi-quantitative analysis. Further, there exist a proposal to combine STPA with SPN for process industry by (Zhang *et al.*, 2019b).Their aim was to conduct quantitative models based on STPA for only feedback control loops behavior, not for system behavior as a whole.

In this study, the approach is conducted for a multi-robots system by modeling the whole system behavior during its operation. The main purposes of this article are therefore to investigate how risk and safety of autonomous multi-robots systems should be managed, to create a framework for modeling and quantifying the collision hazard, and to generate the needed constraints and requirements to ensure the safe operation of robots. The case study used in this article investigates the use of a set of mobile robots for chemical transportation tasks within high-risk environments such as industrial laboratories in oil and gas industries.

The remainder of the paper is organized as follows: Section 2 presents the methodology proposed in this article, introduces the System-Theoretic Process Analysis (STPA) and a background on Petri nets for risk and safety uses is also presented. The application of the proposed methodology to an autonomous multi-robots work in chemical laboratory is given in section 3. Section 4 is devoted to results discussion. Finally, conclusion is made in Section 5.

2. Methodology Description

The proposed methodology is a combination of two methods, namely:

- System-Theoretic Process Analysis (STPA)that is used to identify and analyze hazard scenarios
- Petri net method (PN) is performed to model the functional and dysfunctional behaviors of robots and their coordinating
 actions, in order to assess the occurrence frequency of collision scenarios and robots performance.

The proposed methodology follows the steps illustrated in figure 1.



2.1.STPA Hazard Analysis

System Theoretic Process Analysis (STPA) is a purely qualitative method for hazard analysis based on Systems-Theoretic Accident Model and Processes (STAMP) which has been developed by (Leveson, 2004). It has been specially created for complex automated systems as long as today the automation invaded multiple domains and applied first to identify undesired/unsafe system behaviors through a structured, top-down approach. Requirements are subsequently generated from the results of STPA in order to handle these unsafe behaviors. STPA treats safety as a control problem rather than a failure prevention problem(Khan, Madnick and Moulton, 2018). In this context, the STPA method is suited to distinguish potentially dangerous design imperfections, including software and hardware errors, external disturbances and improper interactions among various control structure components, which result mainly from lacking control or violation of safety constraints related to the development, design, and operation of the system (Ishimatsu et al, 2014).

The STPA analysis is carried out in five main steps as shown in the following figure 2 (Leveson and Thomas, 2018).



Fig. 2.STPA methodology. (UCA: Unsafe Control Action)

- The first step in any analytical method is to define its purpose, mainly by identifying the types of losses, accidents and high level hazards of the system under study.
- The second step is to model the control structure of the system. The control structure constructs as a set of feedback control loops, which defines functional relationships and interactions between its components.
- The third step is to identify unsafe control actions and connect every unsafe control action with their hazards defined in the first step. Control actions are analyzed using predefined guide words by verifying:
 - The control action causes a hazard if it "is provided".
 - The control action or the preventive measure causes a hazard if it is "not provided".
 - The control action causes a hazard if it "is provided too early or too late".
 - The control action causes a hazard if it is "applied for a long time or lose it too soon".
 - The fourth step is to extract the causal factors for which unsafe control action could occur at the system level.
- Finally, the fifth step consists of proposing safety constraints and safety requirements to generate safety policies for the development, design, control algorithm and operating part of the system.

2.2.Petri Net (PN)

Petri nets (PNs) were first introduced by Carl Adam Petri in his doctoral thesis presented in 1962. Its initial goal was a graphic representation of the behavior of automata. At the beginning of the 1980s, they were proposed for the first time to be used in dependability in order to generate Markov processes on a large scale, then they were considered as a model suitable for Monte-Carlo simulation(SIGNORET, 2008). The Petri net is a powerful modeling formalism suitable for dynamic discrete event systems, combining a well-defined mathematical theory with a graphical representation of the behavior of the systems. Its theoretical aspect provides a modeling and behavioral analysis of the dynamics of the system, while its graphical representation allows to visualize the changes in the modeled state of the system (Wang, 2012). Therefore, PNs have been used to model various types of complex systems (Wang, 2007). Thus, they have shown excellent modeling and calculation capacities in the field of dependability and safety(Malhotra and Trivedi, 1995; Dutuit *et al.*, 1997; Signoret, 2009). They are very useful for modeling the functional and dysfunctional behavior of systems and for evaluating several performances(SIGNORET, 2008). For example, many existing works have used PNs to assess reliability (Kumar and Aggarwal, 1993; Yang *et al.*, 2011) and availability (Jian, Shaoping and Yaoxing, 2008; Kumar, Jain and Gandhi, 2012; Zhang *et al.*, 2019a).

As described by Petri, the basic PN is a special type of directed bipartite graph, mainly composed of three complementary parts:

- A static graph, which does not change over time. It is made up of three basic elements:
 - Places, represented as circles, each place represents a state or condition.
 - Transitions, represented by bars, each transition represents an event, a transformation or a change of state.

• Directed arcs, represented by arrows, connect squares to transitions or transitions to squares. Each arc can be assigned with a natural number, named a weight (normally assumed to be 1).

- **Dynamic elements**, called tokens and represented by small balls which indicate the state of the system at a given moment. That makes it possible to describe the behavior of the modeled system. Tokens can represent objects, machines, humans, information, conditions, etc. The distribution of tokens in the squares is called network marking. The presence or absence of a token in a place can indicate whether a condition associated with that place is true or false. If each entry place "p" of the transition "t" contains at least the number of tokens equal to the weight of the directed arc connecting "p" to "t", then this transition "t" is activated.
- Predicates and assertions, which can be introduced in the PN by means of variables. The predicate (often preceded by two question marks "??") is a condition to enable or disable transitions when variables are checked or not checked, and the assertion (often preceded by two exclamation marks "!!") is a formula for updating variables after crossing the associated transition (Signoret, 2008; Wang, 2012). The use of these two elements can improve the readability of complex PN models, thus improving their comprehensibility(Signoret *et al.*, 2013).

In the real world, almost all events are related to time. For this, the need to include temporal variables in models of dynamical systems is obvious. When a Petri net integrates the time variable into the model of the system, it becomes a timed Petri net. Timed PNs are an extension of basic Petri Nets for which a time variable called delay with its probability density function is associated with each transition. Tokens can move between squares when the activated transition is fired. The transition is fired when the associated delay elapses (since the transition remains on during the delays). The delay between the activated transition and the crossing can be characterized as fixed (deterministic) or random (stochastic). Once the transitions are characterized by deterministic and stochastic delays in a model, Petri nets will become difficult to solve analytically. In this case, a Monte Carlo simulation is generally coupled with the SPN model in order to facilitate the evaluation of system's performance (SIGNORET, 2008).

3. Application to Autonomous Multi-mobile robots system work in petrochemical plants

In this section, STPA is applied to analyze the functioning of a multi-mobile robot moving products in a chemical laboratory, and then SPN is used to model their behavior as well to assess the collision hazard. A brief description of the object of study is presented first, followed by the presentation of the approach application with the derived results.

3.1. System description

Our system is composed of two wheeled mobile robots that cooperate together in order to transport dangerous chemicals (toxic, flammable, explosive ...) within a chemical analysis laboratory, in the presence of analysis machines and human workers. The laboratory has five main rooms represented in Figure 3, a big room for analysis, rooms for chemicals storage and one other room for providing analysis results. The multi-robot system coordinates their motion according to a distributed architecture. The presence of these robots in such hazardous plants should carry a high level of dependability and safety, which inevitably demands the need for systems that are increasingly available, reliable and the need for optimal control and good communication between robotic entities. The objective of our study is therefore to collect the constraints and safety requirements necessary to ensure the good operating conditions of these robots in such environment, to assess the system performance and the occurrence frequency of collision hazard.



Fig. 3.Model of the working environment of robots.

3.2. Hazard analysis with STPA

We present here the application of the STPA analysis step by step as well as its obtained results.

- Step1: defining the purpose of the analysis (define system boundary). This step contians the following substeps :

• Identify losses:

Seven types of losses are identified and classified in decreasing way in figure 4:

- L1: Loss of human life (operators and workers), human injuries, illnesses, asphyxia, poisoning...
- L2: Toxic or polluted environment (toxic gas emission, dust, fume of chemicals emission, chemicals leakage...)
- L3: Loss or damage to analysis machines or other materials in the laboratory
- L4: Loss of chemical products (spill of chemicals)
- L5: Loss of mission (no Transportation of products \rightarrow no analysis)
- L6: Loss or damage to robots
- L7: Loss of reputation of the company producing the robots.



Fig. 4.Classification of losses levels.

• Identify system-level Accidents, Hazards and safety constraints:

The table 1 below clearly identifies the accidents, hazards and high level constraints of the system. For each hazard, a constraint must be formulated.

System- level Accidents (S-lA)	System-level Hazards (S-lH)	System-level constraints (S-IC)
System- level Accidents (S-IA) S-IA1: Collision between robots loaded with chemicals (two or more) S-IA2: Collision between robot (loaded with chemicals) and human worker S-IA3: Collision between robots (without chemicals) S-IA4: Collision between robot and human worker (without chemicals) S-IA5: Robot crashes to the wall/ or other static obstacles (analysis machines) or falls down S-IA6: Chemicals spill	 System-level Hazards (S-IH) S-IH1: Robots (loaded with dangerous chemicals or not) enter together in a narrow area (Robots doesn't respect the safe distance between them or to humans) (L1- L7) S-IH2: Robot (loaded with dangerous chemicals or not) doesn't respect the safe distance to other static obstacles (L2- L7) S-IH3: Robot (loaded with dangerous chemicals or not)doesn't respect the safe velocity in its navigation (L1- L7) S-IH4: Robot (loaded with dangerous chemicals or not) enters into a slippery area (products spill) (L1- L7) S-IH5: Robot (loaded with dangerous chemicals or not) enters in an uncontrolled state (L1- L7) 	 System-level constraints (S-IC) S-IC1: The robots must not enter in the same area together especially if the area is narrow (the robot must detect if there is a place to enter) (one by one) → S-IH1 S-IC2: The robots have to respect the safe distance and doesn't violate the minimum distance of separation and if it violated, then the violation must be detected to avoid any collision → S-IH2 S-IC3: The robots must not exceed certain limit speed, especially in the case of carrying the products and if the limit speed exceeded, then the robot must be stopped → S-IH3 S-IC4: Must maintain permanent cleanliness of the floor and not throw slippery materials on the floor → (safety constraint to S-IH4 outside the boundary of the robot control)
		S-IC5: The robots must detect slippery to avoid and if it not the case they must adjust their velocity and slow down \rightarrow (safety constraint to S-IH4 Relating to robot control)
		S-IC6: The control parameters and algorithms of robots must be monitored and if there is any issue in the control (or indications that the robots are not well controlled) the robot must be stopped \rightarrow S-IH5

Table 1. Table lists system-level accidents, hazards and constraints.

Step2: model the control structure for autonomous control

The goal of the STAMP approach is to increase our understanding of systems and processes control structure, and to model the control structure based on functional diagram. The studied system is composed of two autonomous controllers. The two controllers communicate together to coordinate the motion of the robots.

- High level Functional control block diagram of one robot:

Before moving on to the control architecture of an autonomous mobile robot we thought to start by defining its functional model to identify the main components and the function of each component. As the following figure 5 is shown, an autonomous robot's navigation revolves around three main tasks: perception and location (sensors), planning (controller), and control (motors).



Fig. 5. Functional control block for an autonomous mobile robot.

Detailed functional control block diagram for one robot:

For more details, detailed functional control architecture is proposed according to the STAMP model and illustrated in figure 6. This architecture defines the main parts of an autonomous mobile robot identified by the function of each part; these parts interact with each other by control actions (arrows in red) and its feedback (arrows in green).



Fig. 6.Part 1- detailed functional control architecture for one autonomous mobile robot.

- Detailed Functional control block diagram for two robots:

In the case of an autonomous multi-robots (case of two robots) operating in a common indoor environment, coordination of actions is of prime importance between robots, in particular, for distributed control. This coordination is carried out through communication protocols (messages, sharing information, etc.). The functional control architecture for two robots coordination is shown in figure 7.



Fig. 7.Part 2-functional control architecture for multi- robots coordination (case of two robots).

- Step3&4: identify unsafe control actions and their causal factors

Table 2 presents the set of hazard scenarios identified for each functionality of autonomous mobile robot controller and its causal factors.

Control action/ communication	Guideword	Scenario (UCA)	S-IH	Causal factors
	Not providing causes hazard	UCA1: The controller doesn't calculate and generate the path	Н5	 Planning and calculation units fail to perform their function (component failure) Insufficient information to support calculation (unclear map or sensors problem) High calculation number → the calculation is blocked because of the huge number of equations Memory card saturation
Plan and	Providing causes hazard	UCA2: The controller generates the wrong path which lead to the wrong destination	Н5	 Insufficient sensor information Failure of sensors False algorithm programmer error Wrong order of operator
generate the path		UCA3: The controller generates the long path (without optimization)	Н5	- Problem in the optimization algorithm , ineffective approach, inappropriate optimization criteria
(From path planning controller)		UCA4 : The controller doesn't update the path in a dynamic environment	H1	 Repeated sensors data , Failure of controller, sensors Inadequate Algorithm
	Too early, too late, out of order	UCA5: Take a long time to calculate and generate the path	H1,H2, H5	 High Number of calculation Lock of controller Delay in sensor information Problems in data fusion Inadvertent lock of software, calculation
	Stopped too soon, applied too long	UCA6: The planning of path stopped too soon	H1, H2,H5	Instant cut of dataDischarged battery
	Not providing causes hazard	UCA7: Collision avoidance feature is not provided during the robot navigation	H1, H2, H4, H5	 Failure of sensors (doesn't indicate if there is an obstacle) Failure of collision avoidance controller

Fable 2.	Possible	scenarios	obtained	from	STPA	application.
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			-	-Inadequate control algorithm (which relate between features)- No dependency between sub-controllers
	Providing causes hazard	UCA8: Providing the wrong command to avoid collision (speed up / speed down, stop, turn left, turn right)	H1, H2, H5	 Inadequate calibration of sensors/ data fusion Missing sensors /communication data
Avoid collision (from collision avoidance controller)	Too early, too late, out of order	UCA9: Collision avoidance feature provided too late (static obstacle, other robots or human)	H1, H2	 Delay of sensors information Missing data from sensors Missing Data from communication (too late or too early) Loose dependency between controllers
	Stopped too soon, applied too long	UCA10: The collision avoidance feature stopped too soon during the robot navigation	H1,H2, H5	 Inadequate collision avoidance Algorithm (non robust) Failure of controller, the sensor which detects the distance between the obstacles and the robot Inadequate sensors calibration
Establish and	Not providing causes hazard	UCA11: Doesn't provide the dependency between sub-controllers (path tracking or path planning controller and collision avoidance controller)	H1,H2, H5	 No data input from the sensor which detect obstacles (failure of sensors) or insufficient. Programmer error Inadequate control algorithm Failure of software Failure of node connecting between controllers
maintain dependency between sub-	Too early, too late, out of order	UCA12: Providing the dependency too late	H1,H2, H5	Input data sent from sensors too lateTake more time in data fusionBug of software
controllers (Software)	Stopped too soon	UCA13: Loose dependency between sub-controllers	H1,H2, H5	 Lock of software Program corruption Failure of node connecting between controllers / or lose the node. Lose connection with master
	Not providing causes hazard	UCA14: Doesn't provide communication between robots	H1	No Connection between robotsFailure of communication components
	Providing causes hazard	UCA15: Provide insufficient information UCA16: Provide huge information even unnecessary ones	H1,H5	- Programmer error - Cyber-security issues
Establish and maintain communication	Too early, too late, out of order	UCA17: Provide communication with delay (too late)	H1	Weak flow of networkStrong flow of dataLoss of power
between robots	Stopped too soon, applied too long	UCA18: Loose communication between robots	H1	 Lock of communication software Failure of communication components Received information higher than processor capacity Weak flow of network
		UCA19: The communication stopped too soon	H1	- Network interruption
	Not providing causes hazard	UCA20: The controller doesn't track the path	H2,H5	 Inadequate path tracking control algorithm Failure of motion controller Failure of position sensors, Inadequate sensors fusion Inadequate sensors calibration Failure of actuators(blockage of wheels)
Track the path (from Path tracking controller)	Providing causes hazard	UCA21: Track the path with no precision	H5	 Inadequate tracking control algorithm Failure of sensors, actuators Inadequate sensors calibration Provide sensors data with delay Inadequate data fusion

	Stopped too soon, applied too long	UCA22: Stooped to track too soon (interruption)	Н5	 Interruption in sensors information No Power (battery failure or Discharge) Failure of controllers, sensors or actuators
	Not providing causes hazard	UCA23: The velocity command is not provided to the wheels when the robot is near to a dynamic obstacle (human, other robots)	H1	 Problem in communication block (command publisher) Inadequate aggregation Algorithm (decision problem) Failures of components of the main controller Lock of software (of controllers and sensors) Sensors information doesn't received by controller (problem in subscriber blocks) Sensors information badly fused Failure of sensors Failure of actuators controller (no signal provided to the wheels)
Send the velocity		UCA24: The controller does not provide the velocity command to the right/left wheel during the robot navigation	Н5	 Inadequate sensors information Failure of actuators controller (no signal provided to the wheel)
command to the wheels (from motion controller after aggregation of	Providing causes hazard	UCA25: The controller provides high velocity in inappropriate situations (Robot carry chemicals, slippery floor, smoke, human work with robots in the same room)	H3,H4	 Inadequate control algorithm (the controller not adapt to these situations) Sensors doesn't detect slippery/ the load Inadequate sensors calibration Lack of sensors
the set of features)		UCA26: The controller provides the same value of velocity even if the robot in front of an obstacle	H1,H2, H3	 Missing sensors indication/ or wrong fusion Lock of software/ controller
	Too early, too late, out of order	UCA27: Send the command too late after a delay time	Н1,Н2, Н5	Send sensors information too late (delay)Problem controller softwareMemory saturation
	Stopped too soon, applied too long	UCA28: The sending command stopped too soon	H1,H2	 Failure of main controller/ controller of actuators, sensors. Inadequate sensors information (doesn't update information about the environment) Loose communication with master
		UCA 29: The same value of command applied too long even in new situations	H1, H2,H3	 Failure of sensors. Inadequate sensors information The sensor indicates the same environment state (don't update new data about the environment)

- Step5: generate the important safety constraints and safety requirements

The important safety constraints and safety requirements proposed to ensure the good control of autonomous multi-robot in hazardous environment are detailed in Table 3.

Table 3.	Important safety	constraints and	safety re	equirements	proposed f	from STP.	A results.
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Control Action/ communication	UCA	Safety constraints	Suggested safety requirements
	UCA1/ UCA 3	The controller must calculate and generate the correct and the optimized path.	The controller must choose the safe trajectory which takes the shortest time and that requires to choose the good optimization criterion
Plan and generate the path	UCA4	Planning should always be adapted to changes in the environment	The path shall be updated after each action of the robot or make the combination with the obstacle avoidance controller
	UCA5	The controller must not exceed the calculation time specified by the programmer	If the controller exceeds a certain time predefined during the planning, it is

			necessary to stop manually the operation of the controller
	UCA 7/ UCA 10	The controller should not let the robot gets in contact with any of other dynamic or static obstacles (Objects, Humans, robots) during its navigation.	The robot shall respect an adequate minimum safety distance MSD (not less than 0.5 m) between robots and any other obstacles during its navigation to avoid any contact. It must not be violated under any condition.
Avoid collision	UCA 8	The controller should not specify the wrong command to avoid collision	The controller shall check all the directions around the robot before specifying the correct avoidance command
	UCA 9	The collision avoidance controller must not let the robot get in touch with any static or dynamic obstacles especially human or a robot carrying a product	The robot shall respect a specified safety distances between other static or dynamic obstacles to avoid any collision and three layers of safety distances (zones) surrounding the robot should be made. If there is an obstacle detected on the third one, the controller begin to reduce the speed of the robot and if it isn't detected until the first zone, the controller must stop the motion immediately
	UCA 10	The collision avoidance feature must operating all the navigation time	A contact sensor must be in working order to stop the robot, in case of a contact or a collision.
Establish and maintain	UCA 14	The robots should communicate and negotiate with each other before doing action.	Tasks, motion and actions of robots must be organized between them by ensuring a continuous exchange of information inter- robot using messages to avoid entanglements and unwanted action in order to ensure an effective cooperation
communication between robots	UCA 15/ UCA 16	Sufficient information (to be exchanged) provided by communication should be specified	the type of information should be specified (robots positions, velocities, type of tasks, information about the rooms state if they are occupied or empty)
	UCA 17	Communication should be maintained	It is necessary to use a multilayer software like ROS which supports the huge number of information diffusion and facilitates its transmission and if there is a loose of connection between the nodes it is necessary to indicate to the master
	UCA 20/ UCA 23	Robots should not deviate from the specified path during their navigation	- The controller shall detect any deviation from the specified path using multi-sensors which detects the robots position (odometer, laser telemeter)
Track the path	UCA 21	The precision should be considered in the tracking feature	 Use sensors redundancy (multi-sensors) to minimize the risk of failure It is necessary to choose the good filter for sensors information to increase the precision of tracking The risk of sensor defect must be minimized by the use of the fault tolerant approach To minimize the risk of non-observability, an estimator must be used
Send the velocity command	UCA 23, 24, 25, 26, 27, 28, 29	The velocity command must be always available	 The programmer shall specify multi-level of safety distances: SD 1 for humans / SD 2 for robots / SD 3 for other objects Detect slippery (using Camera) and reduce the speed of the robot in that case Detect the load using sensor for weight Speed down the robot when it is carrying products or in front of human, or other robots

- Step 6:modeling and quantfyingcollisionhazard using SPN

This section presents the integration part of the SPN model for modeling and assessing collision scenarios of two mobile robots on the basis of their normal operating scenario. The operating state of these multi-robot is decribed using petri net modeling by illustrating coordination between their tasks.

1) Normal operation modeling of robots with SPN:

Figure 8 illustrates a model represented by SPN, drawn using the GRIF software (Grif, 2020). This model describes the normal operating scenario of two robots (R1 and R2) in an analysis laboratory. The laboratory consists mainly of an analysis room, two chemical storage rooms, and a battery charging room (see figure 3).

Part 1, shown in figure 8, presents the phase of sending orders and preparing the robots for an operation. The two robots represented by two places (P2 and P3), the two markings indicate that the robots are ready, and that they are waiting for the command issued by the person in the analysis room (Pa free) to move. The "demands_ S1" and "demands_ S2" transitions represent the product demands from storage room 1 (roorm1) and storage room 2 (room 2) respectively. The order is made according to the needs of the analyzes (every 4 hours a demand is launched for each room). One list for product requests for storage room 1 and another for storage room 2. Places P5 andP6 indicate the number of demands!D1,!D2 respectively. Once a demand is initiated, the command will be sent to the robots depending on the availability of each of the two robots (!Robot 1 = true or! Robot 2 = true). If both robots are available, there will be a priority for robot 1. Two counters are provided to count the number of cycles per robot (Nc1 = Nc1 + 1 and Nc2 = Nc2 + 1). If there are no robots in room 1, the value of the room1 variable is false (! Room1 = false). Once one of the robots enters room 1, the room 1 variable will return true and the same with room 2.

To know which robot enters each room, 4 variables are defined as follows:

- (! R11 = true) if robot 1 goes to room 1;

- (! R21 = true) if robot 2 goes to room 1;

- (! R22 = true) if robot 2 goes to room 2;

- (! R12 = true) if robot 1 goes to room 2.



Fig. 8.Normal operating model of the two robots.

Parts 2 and 3, shown in figure 8, present the transportation of products with robots. The robots move from the analysis room to storage rooms 1 and 2 respectively. Persons in rooms 1 and 2 (P1 and P2 respectively) place the products on the robots. Once the robots leave room 1 or room 2, the room1 or room2 variables return to their free states (! Room 1 = false,! Room 2 = false), i.e. there is no robot in storage rooms. The robots return to the analysis room with products. The transitions (Tr7, Tr8, Tr9, Tr10, Tr11, Tr12, Tr13, Tr14) are defined by a dirac law with deterministic delays, which represent transportation periods for the two robots.

Then, in parts 4 and 5, shown in figure 8, a recognition of robots (the two robots define by two tokens) is performed to identify them and each robot can return to its initial place. If robots makes a number of turns (Nc) greater than Ncmax, they go to the battery loading room before being available again and the Nc counter returns to the value 0.

Finally, part 6, shown in figure 8, represents the phase of preparing machines by the person in the analysis room (Pa) to perform the analysis.

2) Modeling of collision scenarios

Figure 9 illustrates the SPN model for the main loss scenarios that lead to a collision state. We identified the six main causes from the results obtained by the STPA method (failure of the motor of the left wheel or that of the right wheel, failure of the laser

scanner sensor, failure of the communication system, failure of the control card and battery problem). Note that the failure of the two motors of each robot can occur independently or be caused by a common cause failure (CCF) event.

Based on STPA results, the SPN model is used to evaluate the collision frequency according to the characteristics determined by the studied system.



Fig. 9. Main causal scenarios leading to collision for the two robots.

The tokens initially remain in places (P27, P29, P31, P33, P35, P37) for robot 1, and in places (P39, P41, P43, P45, P47, P51) for robot 2, indicating the good functioning of the two robots and all their components. In addition, the tokens in P55 and P57 indicating the safety state of the two robots, which means that there is no collision (see figure 10).



Fig. 10. Collision scenarios.

When tokens fire failure detection transitions and reach places (P28, P30, P32, P34, P36, P38) for robot 1 or places (P40, P42, P44, P46, P48, P50) for robot 2, it means that robots enter in an unsafe state, which is caused by a failure of one or more of its

components. Those failures are the main factors of collision event (the tokens fire the Tr 61 or Tr 63 transitions according to the predicate "? # 32 == 1 | # 34 == 1 | # 36 == 1 | # 38 == 1 | @ (1,1) (# 28 == 1, # 30 == 1) "or"? # 44 == 1 | # 46 == 1 | # 48 == 1 | # 50 == 1 | @ (1,1) (# 40 == 1, # 42 == 1)", where the palces 28, 30, 32, 34, 36, 38 represent the failure state of the components of the first robot and places 40, 42, 44, 46, 48, 50 represent the failure state of the components of the second robot, and reach places P56 or P58 from P55 or P57, which realizes the assertion "! R1_FD = true " if robot 1 fails to operate or"! R2_FD = true " if robot 2 fails to operate.

The defective robots will be transferred from places (P7, P8, P9, P10, P11, P12, P13, P14) to the maintenance service (P59 or P60), shown in figure 11, and the tokens remain in these places until the repair of the problem is completed. When the robots return to their initial states through the transitions Tr62 and Tr64, the variables "R1_FD and R2_FD" will be assigned as false, the transitions Tr81 and Tr82 will be crossed to reach the places P19 and P21 and the robots continue their missions.



Fig. 11. Model of robots in maintenance service.

3) Simulation results

Failure probability data retrieved from references (IEC 61508:2010; Fazlollahtabar and Niaki, 2017) are re-evaluated based on new operating conditions. The table 4 below shows the estimated failure data of robots components.

Components	Probability distribution
Electric motor	$\mathrm{Exp}~(9\times10^{-6})$
Laser scanner	Exp (5×10^{-6})
Communicating system	Exp (9×10^{-4})
Control card	Web (2, 9.5× 10 ⁴)
Battery	Web (4.7, 4.9× 10 ⁴)

Table 4. Failure pro	bability data.
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Failure events are assumed to be distributed according to two probability laws: the exponential law and the Weibull law. The duration of the experiment for the simulation is four years (34,560 hours), the number of histories is 10⁶ history, and the confidence interval is estimated at 90%. Figures 12, 13 and14 below illustrate The instant unavailability of the two robots, the mean time to collision for the two robots and the frequency of collision respectively for an experiment period of 4 years with the two lower and upper confidence interval (CI) bound.



Fig. 12. Instant unavailability of the two robots.

The robots unavailability, the mean time to collision and the collision frequency were directly calculated by Monte-Carlo simulation using Grif software (Grif, 2020).

As shown in figures 12 and 13, the unavailability and the mean time of collision of the two robots are the 0.014 and 542 hours respectively. Which means that the first collision happenes after 542 hours of the two robots continuous operation.



Fig. 13. Mean time to collision of the two robots.



Fig. 14. Collision frequency related to the two robots.

According to the graph infigure 14, we observe that the value of frequency is between $1.82 \times 10^{-3} h^{-1}$ and $1.86 \times 10^{-3} h^{-1}$. It means that every 22 to 23 days there is a risk that one of the two robots will collide in the lab.

It should be noted that the degradation of performance begins after 23 days of continuous operation, and therefore the collision frequency is considerable.

The critical parts of the system represented in comunicating systems and batteries.

4. Discussion

The approach used in this article is a combination between the STPA method and the Stochastic Petri Net (SPN) model. This approach provides a detailed analysis of the various functionalities of autonomous mobile robot controllers using more detailed control architecture by STPA analysis. The result of this analysis identifies the necessary safety constraints and also proposes requirements to optimize the control of these robots in complex high-risk environments. In addition, the SPN helps us to model the behavior of robots during their normal operation and also to model hazard scenarios as well as to quantitatively assess the risk of collision.

The results reveal that the main problems that may hinder the safe operation of robots are the problems of control, communication, power and retrieving environment information from sensors.

The most marked property of STPA analysis is the ability to extract many sets of hazardous events, including those caused by failures of system elements, such as risks resulting from unintentional interactions between elements. In addition, the STPA analysis provides a large number of potential scenarios, which does not only deal with the system in operating modes / malfunctions but also takes into account the time factor and its influence (action executed too late or too early).

Among limitations that we faced are:

The complexity of the whole behavior modeling for a multi-robots system using SPN, in particular, when dealing with a large number of robots (more than two). Increasing the number of robots makes a large scale model, which causes a problem of readability, understandability and evaluability.

In our study, we select only the collision scenarios caused by robot components failure for the numerical experiment, while ignoring other factors such as human errors, degraded component operation...

5. Conclusion

In this article, we investigate how risk and safety of autonomous multi-robots systems should be managed during a chemical transportation task in high risk environment using the combination STPA and SPN, The STPA method is applied to identify and analyze unsafe robot behavior, also to generate the needed constraints and requirements to ensure the safe operation of these robots

in such environment. Whereas the SPN is used mainly to complement the STPA in order to create a framework for modeling and quantifying the collision hazard.

The proposed methodology allowed us to provide a detailed analysis of complex industrial systems equipped with autonomous mobile robots, to also collect a large number of potential risk scenarios, to model the behavior of the system during its normal operation and its dysfunctional and unsafe behavior. Further, to assess the robots performance and the occurrence frequency of collision hazard. Thus, safety requirements and recommendations have been suggested to resolve the navigation and control problems of differential wheel mobile robots, while maintaining the safety and security of the entire laboratory. In our case study, we limited the number of robots to two robots because of the high complexity of the Petri net model. The goal is to facilitate the comprehensibility of the model and to minimize the simulation time. In addition, we considered the collision scenarios caused by robot component failures, while ignoring other factors, such as human errors, degraded component operation ... In future work; to have more rigorous quantification of loss scenarios it is preferable to take into account each unsafe control action.

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