HRA Data Collection from the Simulations of Abnormal Situations

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1. Introduction

It is widely recognized that reliability of operators are critical to complex socio-technical systems [1]. For this reason, human reliability analysis (HRA), which aims to identify unsafe actions (UAs) that contribute to risks of the systems and assess the failure rates of the actions, has been conducted [2, 3].

Although many techniques of HRA have been developed and used in many years, many reports indicated lack of database for supporting empirical bases of HRA methods [1, 4]. Thus, there have been recent efforts to collect data about human reliability from plant experience, simulator experiment or qualification, and laboratory experiments [3]. As one of these efforts, KAERI also established a guideline to collect information about human reliability and performance shaping factors from simulators [5]. This guideline particularly presented a set of worksheets that allows comprehensively gathering objectively observable information in simulations of emergency situations.

This paper reports the process and preliminary results of the data collection from the simulations of abnormal situations based on the developed worksheets in KAERI database guideline. We analyzed operator behaviors of the sixteen experiments for the two kinds of abnormal situations: RCP (reactor coolant pump) cyclone filter blockage and CDP (condensate pump) valve stuck. The UAs of operators were identified and quantified.

2. Data Collection Process

2.1 Simulation Data

We analyzed the behaviors of operators in the main control room of an OPR 1000 under abnormal situations. The operation team of OPR 1000 consists of five operators: a shift supervisor (SS), reactor operator (RO), turbine operator (TO), electric operator (EO), and shift technical assistant (STA). When cues of abnormal situations were recognized, the operators should find out related abnormal operating procedures (AOPs) and follow the procedures to cope with the situations. If the operators aware urgency of the situations, they can also take rapid actions before following a procedure.

The eight operation teams of OPR 1000 participated in the experiments for each scenario. For the scenario of CDP valve stuck, a situation of simultaneous closes of the valve (CDV201) before the condensate polishing system and the bypass valve (CDV200) of the condensate polishing system. The operators should find out the cause of problem and request a local operator to open CDV200 within 10 minute. If necessary, the operators can cut reactor power back. The AOP for low flow of condensate tanks describes all actions to be conducted. For the scenario of RCP cyclone filter blockage, the cyclone filter after the high pressure cooler in RCP 1A seal injection flow was blocked. The operators should find out the cause of problem and trip the RCP 1A when the inlet temperature of seal coolers reaches the trip level. The AOP for loss of RCP seal describes various actions to be conducted for similar situations, but it does not describe any symptom and measure related to the blockage of the cyclone filter.

From the sixteen simulations, the behaviors and communications of control room operators were recorded. In addition, the associated events such as manipulations or alarms annunciated and the process parameter such as pressure or temperature of important component were obtained.



Fig. 1. The overall process to collect information for HRA data [5].

2.2 Worksheets and Analysis Process

To accumulate qualitative HRA data, the obtained records were analyzed and converted to worksheets designed in the KAERI guideline. Fig. 1 shows the process to develop the HRA data worksheets. The worksheets were developed by following phase one to five of the process for ISLOCA (Interfacing System Loss of coolant accident) scenario, an emergency situation.

The worksheets include plant level information worksheet, timeline worksheet, and task level worksheet. The contents of the worksheets can be summarized by Table I.

Work	Category	Data item				
sheet						
Plant	Plant &	Plant/simulator name				
level	simulation	Plant type				
	overview	Operating mode				
		Simulation date				
		Ingress/injection time of initiating				
		event				
		Simulation completion time				
		Crew/shift/team name				
	Crew ()	Age				
		Work experience of plant				
		operation (yr)				
		Work experience in current				
		position/role (yr)				
		Certified License				
		Work experience in current team				
		(yr)				
	Training &	Simulator training frequency				
	education	Training experience on the				
		scenario				
	Environmen t	Simulation environment				
	Observed	Observed procedural path				
	response	1 I				
	Scenario	Simulation mode				
		Initiating event				
		Multiple initiating events				
		Failed system or component				
		Failed/masked alarm or indicator				
		Scenario/event summary				
		Expected Procedural path				
		Allowable time				
	Crew	Leadership of SS				
	characteristi	Cooperative attitude				
	cs and	Supervising level of STA				
	dynamics	Independent checker				
	5	Procedure compliance				
		Communication level				
Time	Contents of	Start time of utterance				
line	communicat	End time of utterance				
	ion	Operating personnel				
		Utterance				
	Action log	Related action log				
	Procedure &	Related procedure				
	Task	Related sten				
	Tubic	Responsible person of step				
		Task type				
		Component type				
	ΙΙΑ	UA candidata				
	UA	Code of UA or				
		Related UA/recovery				
		(Situation) Description				
Teals	ITA	UA codo				
1 ask	UA	UA COUE				

level	information	ion UA overall description			
		UA effect			
		Related UA/Recovery Code			
		Recovered UA			
		Recovery timing			
	UA type	Reporting UA Instruction UA			
	51				
		Manipulation UA			
	UA Initiator	UA performer			
		Related operator (causality)			
	Plant/	Failed system/ component			
	system state	Failed alarm/ indicator/display			
	5	Failed switch/controller			
	Time	Time pressure			
	pressure	rine pressure			
	Task	Task familiarity			
	familiarity				
	Task	UA occurred during the			
	complexity	performance of a contingency			
	(diagnosis)	action part			
		The type of state identification			
		Note or caution			
		Change of procedure strategy			
		Procedure conformity			
	Task	Number of detailed instructions			
	complexity	Number of manipulations			
	(execution)	Component manipulation mode			
		Continuous action step			
	Procedure	Confusing statement			
	quality	Multiple constraints			
	(clarity)	Clarity of decision-making criteria			
	Procedure	Description of object			
	quality	Specification of means			
	(description)	-			
	HMI &	Information clarity			
	Information	Feedback information			
	quality				
	Communicat	Procedure compliance			
	ion quality	Precise instruction			
		Controversial expression			
		Standard terminology			
		Communication level			
		Reporting omission			
	Recovery	Recovery code			
	information	Recovery description			
		Recovery worker			
		Time to recovery			
		Recovery cue			
		Recovery Initiator			

Using the worksheets, the data items of the plant level worksheet and the data items in communication contents, action log, and procedure & task of the timeline worksheet were obtained, first. All kinds of deviations were then extracted by comparing communication contents, actions log, process parameter data and procedures. The UAs were determined among the deviations by considering rationales of human behaviors such as procedures, the expert judgment, the thermalhydraulic analyses, success criteria used in the PSA, and the status of CSFs. The identified UAs were then investigated in detail and the task level worksheet was written for each UA.

Some instructions in AOPs were not explicitly represented; hence, to analyze the simulations of abnormal situation based on the worksheets, which were developed for proceduralized processes, the related AOPs were formalized to be analyzed. The instructions of AOPs were distinctly numbered and the recognition of significant alarms and search of proper procedures were supplemented to the early tasks of the instructions.

2.3 Taxonomy of UAs

The taxonomy of UAs was defined to quantify the frequency of UAs. Because the purpose of data collection is to support HRA, some HRA techniques such as THERP [6], ASEP [7], K-HRA [8], SPAR-H [9], HEART [10], HCR [11], Phoenix [12] and CBDT [13] and existing HRA databases such as operating experience data by GRS[4] and CORE-DATA [14] were reviewed. By considering the taxonomy of human error in these references, the taxonomy that is based on the structure of crew failure modes in the Phoenix method [12] and reflects generic task characteristics like THERP and ASEP method [6, 7]. Table II shows taxonomy of UAs.

Table II: Taxonomy of UAs

UA types	Task type			
Failure of information gathering	Alarm			
and reporting – checking	Indicator			
discrete state (omission error,	Synthetically evolution			
commission error)	Synthetically evaluation			
Failure of information gathering	Reading simple value			
and reporting – measuring	Comparison of			
parameter (omission error,	parameter			
commission error)	Evaluating trend			
Failure of situation assessment	Transferring procedure			
and instruction (omission error,	Operating control			
commission error)	Communicating			
Failure of diagnosis (omission	-			
error, commission error)				
Failure of manipulation –	Two-position control			
omission error	Rotary control			
	A set of sequential			
	operation			
	Dynamic			
Failure of manipulation – wrong	Two-position control			
device	Rotary control			
	A set of sequential			
	operation			
	Dynamic			
Failure of manipulation – wrong	Two-position control			
direction	Rotary control			
	A set of sequential			
	operation			
	Dynamic			
Failure of communicating to	-			
external agent (omission error,				
commission error)				
Unauthorized control	-			

3. Probability of UAs

Based on the taxonomy of UAs, the probabilities of UAs were estimated by the following equation [2].

$$P_{UA} = \frac{n_i}{m_i} = \frac{n_i}{n_i + 0_i}$$

Here, n_i is the frequency of UAs observed in situations of UA type *i*, m_i is the number of the total possible situations of type *i*, and 0_i is the frequency of situations of type *i* where no UA is observed.

The examples of obtained probabilities are shown in Table III. The fourth column shows the probabilities presented in the human error probability in THERP database in similar situations or contexts in the given UA type. To easily compare the probabilities in this study with in THERP [6], total 0_i and n_i for task types in some UA types were calculated as well.

Table III: Examples of Obtained Probabilities

UA		n_i	\mathbf{P}_{UA}	THERP
Failure of information gathering and reporting(commission) – (total: checking & measuring)		1	0.006	~0.004
Failure of information gathering and reporting(commission) – checking discrete state (total)	112	0	0	~0.003
Failure of information gathering and reporting(commission) – checking discrete state(alarm)	72	0	0	-
Failure of information gathering and reporting(commission) – checking discrete state(indicator)	40	0	0	_
Failure of information gathering and reporting(commission) – measuring parameter (total)	63	1	0.016	~0.005
Failure of information gathering and reporting(commission) – measuring parameter (comparison)	63	1	0.016	_
Failure of information gathering and reporting(omission) –(total: checking & measuring)	175	0	0	-
Failure of manipulation – wrong device (total)		0	0	~0.0005

4. Conclusion

Because the number of simulations was limited and data of various situations will be obtained continuously, it is uncertain to conclude the resulted probabilities. However, in this study, it was revealed that the designed worksheets were feasible to collect HRA data, especially in abnormal situations. The defined taxonomy of UAs was unambiguous to distinguish actions of operators and quantify the probabilities. Based on the worksheet, operator behaviors in many different kinds of scenarios will be analyzed and the relations between human reliabilities and the observed factors will be also investigated.

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