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Introduction



Brief Introduction of the Framework to Ass ess Diagnosis Error Probabilities



Insights derived from the Suggested Frame work



Summary





• As the advanced MCR (Main Control Room) is being adopted in NP Ps (Nuclear Power Plants), the operators may obtain the plant dat a via computer-based system.



Introduction



• A new framework to assess diagnosis error probabilities in advanced MCR has been suggested.

	Generally, the HEPs directly extracted from data-source include the huge effect of PSFs!!		Z=Y*X
Data-source	HEP assessment (Z) - Pr. (diagnosis error) - Pr. (execution error)	PSF (<mark>Y</mark>) analysis	Nominal HEP assessment (X) -Nominal Pr. (diagnosis error) -Nominal Pr. (execution error)
Full-scope simulator of the advanced MCR	 Analysis of diagnosis error by using information processing model Estimation of Pr. (diagnosis error) The Pr.(diagnosis error) was fitted to binomial distribution. It was used as likelihood distribution for Bayesian inference. 	 PSF selection by reviewing HRA methods PSF evaluation by using developed framework Estimation of PSF weighting by using the profiling technique 	 Estimation of the nominal Pr.(diagnosis error) by using the updated TRC model Bayesian inference was used for updating the TRC model. Prior distribution: The existing TRC model, distributed as log-normally.

Purpose :

To provide the insights in the process of developing a framework to assess the proba bility of diagnosis error in the advanced MCR

Brief Introduction of the Framework to assess di agnosis error probabilities

Overview of the framework



Brief Introduction of the Framework to assess di agnosis error probabilities

• Overview of the framework

Update of TRC model	<u>Data-source</u>	<u>Result</u>	
Bayesian inference was applied.	1. Domestic full-scope simulator of the advanced MCP	PSFs	Multiplier
		Teamwork	11.00
$p(\theta y) = \frac{p(y \theta)\pi(\theta)}{(\pi(y \theta) - (\theta))}$	Duration: 2009 –2014 Scenario: LOCA, SGTR, and SBO	Both time constraint and Training	5.72
$\int p(y \theta)\pi(\theta)d\theta$	The number of the crews: 9 crews	HSI	1.03
$\pi(\theta)$: prior distribution	The number of tasks corresponding to	Procedure	2.50
$p(y \theta)$: photoi distribution $p(y \theta)$: likelihood distribution	HFES: 7 tasks	Stress level	2.15
$p(\theta y)$: posterior distribution		Experience	1.39
y : a data point	2. HAMMLAB (HAlden huMan-Mach	Time constraint	3.00
TRC model: $\pi(\theta) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left[-\frac{(\ln \theta - \mu)^2}{2\sigma^2}\right]$ Observed data: $p(y \theta) = \frac{n!}{y! (n-y)!} \theta^y (1-\theta)^{n-y}$	ine LABoratory) Duration: 2007 – 2010 Scenario: SGTR (base and complex sce narios), and LOFW (base and complex s cenarios) The number of the crews: 14 crews The number of tasks corresponding to HFEs : 11 tasks	10 [°] 10	Median HEP for diagnosis Upper bound (95%)

• Insights from analyzing diagnosis errors

- 1) Significant relationship between the cognitive activities and speech act codi ng scheme
 - In this study, diagnosis error were analyzed by using ATHEANA's information processing model.
 - In order to properly distinguish the nature of verbal protocol data, speech act coding sc heme was used.

Cognitive step in ATEANA model	Speech act coding scheme	Definition
Monitoring /detection	Announcement	A statement to the public which gives information
	Inquiry	A statement for asking the status of information
Situation assessment	Judgement	A statement identification based on observation and inquiry
Response planning	Suggestion	A statement of recommendation for specific action or an introduction of an id ea



<Relationship between 'M&D' and coding scheme>

<Relationship between 'S/A and coding scheme>

<Relationship between 'R/P' and coding scheme>

Insights from analyzing diagnosis errors

- 2) Patterns of cognitive activities
 - Most crews performed a part of cognitive activities (monitoring/detection an d response planning).
 - Since they performed their tasks using the given procedures, most crews di d not perform situation assessment.
 - Most crews performed monitoring/detection again since they checked the re sult of their diagnosis.

• Insights from calculating diagnosis error probabilities

1) Factors to increase the probabilities of diagnosis errors

- For two tasks, all crews failed to diagnose the situation
- I. Tasks under insufficient procedure

When the given procedure was insufficient, all crews failed to diagnose the situati on. In the situation under that indicators or components were not addressed in th e given procedures, all crews were difficult to correct diagnose the situation.

II. Tasks under broken indicators

When the crucial indicator malfunctioned, all crews failed to diagnose the situatio n. In this situation, operators could not obtain the cue from the related indicators, then all crews were not able to correct diagnose the situation correctly.

• Insights from analyzing PSFs

- 1) Most influential PSF -> 'teamwork'
 - When 'teamwork' PSF was 'poor', most crews failed to correctly diagnose the necessary actions.
 - Even the RO recognized the required cue, the SS neglected RO's opinion and they eventually failed to diagnose the given task.
 - When communication and coordination between the crew members were inappro priate, they eventually failed to diagnose the given task.
- 2) 'Procedure' and 'time constraints' are also highly influential PSF s
 - When operators perform the diagnostic activities, those two were the cr ucial PSFs as addressed in many papers.
- 3) Multipliers of 'Experience' and 'Stress level' PSFs are different to those in THERP
 - It seems that because the advanced MCR is designed to enhance human performance, the effects of those PSFs to the diagnosis error probabilitie s might be reduced.

- Insights from updating TRC model by Bayesian inference
- 1) Because of the limited available data, it is necessary to collect m ore diagnosis error data from the full-scope simulator of the adv anced MCR.
 - Until now, it is difficult to provide the updated TRC model with accurate values.
- 2) Nonetheless, this is a good starting point to suggest the framew ork to estimate diagnosis error probability in the advanced MCR.





- In this study, the insights were derived from the new framework t o assess the probabilities of diagnosis error in the advanced MCR.
- Insights from analyzing diagnosis errors
 - Significant relationship between the cognitive activities and speech act c oding scheme
 - Patterns of cognitive activities
- Insights from calculating diagnosis error probabilities
 - $\boldsymbol{\cdot}$ Factors to increase the probabilities of diagnosis errors were scrutinized.
- Insights from analyzing PSFs
 - 'Teamwork', 'Procedure', and 'Time constraint' are most influential PSFs.
- Insights from updating TRC model by Bayesian inference
 - \cdot More accurate and reliable framework will be suggested when sufficient data are accumulated.

THANK YOU

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