# Investigation of Insider Identification by Using Human Bio-Monitoring

Nuclear Security and Emerging Disruptive Technologies: The Impact of Cyber and Artificial Intelligence 22 March, 2018

Hoam Faculty House, Seoul National University

### Man-Sung Yim

Researchers: Young-A Suh, Jung-hwan Kim





Department of Nuclear and Quantum Engineering

# **Research Motivation (Problem Statement)**



**Fig.1.** Distribution for 119 cases of serious nuclear and radiological incidents worldwide, from 1960-2013, by motivation, attack type, and insider involvement. Source: WMD incident data (www.hegghammer.com)

## INTRODUCTION

**Problem Statement (1) Insider Threat Problem** 

- The **insider threat** refers to harmful acts that a trusted insider might carry out (Schultz, 2002).
- The importance of the insider threat
  - Nuclear and International communities acknowledge demand
  - Indeed, disgruntled insiders at nuclear facilities have perpetrated many of the known acts of nuclear sabotage (Mott, 2007).
- **Current Status:** There are administrative and technical measures to prevent an insider threat.
- Technical measure limitations
  - The IAEA report (IAEA, 2008) mentioned traditionally lots of focus on physical security against outsiders, which are a less difficult menace to detect than insiders (Wolkov & Balatsky, 2013).
  - Insiders can bypass many security safeguards by nature of their access authorization.
- Limitations of current administrative programs
  - 1) Subjective and potentially biased.
  - 2) Infrequently employed.
  - 3) Reactive Approach.

# Research Hypotheses: EEG indicators may be useful to measure insider traits

	Specific features	Possible EEG Indicators	
Emotional changes	Negative Emotion (fear, Sadness)	Frontal asymmetry Coherence Gamma	
	Anger	F3-F4 Asymmetry	
	Anxious	A selective increase in right parietal activity (P3-P4 or T5-T6).	
	Worry	Alpha Increased gamma in left temporal and posterior area	
Cognitive Change	High Stress	Alpha Theta	
	High cognitive workload/ High engagement Poor decision-making Poor rational decision Poor emotional Understanding Poor social skill	Beta/(Alpha+Theta) Beta/Alpha Coherence Asymmetry	
Malicious intention	Bad thinking/Willingness Impulse control capacity	Gamma/Alpha P300 N200	

At-risk workers' psychological features and possible EEG based measurements

# **Overview of KAIST Research related to Human Bio-Monitoring**



5

- To utilize human bio-monitoring to minimize the risk of nuclear insider threats and human error.
  - To examine the feasibility of using Electroencephalogram(EEG) signal indicators to identify potentially-at-risk workers, especially those with malicious intentions in Nuclear Power Plants (NPPs).

# **Research Method**

**Experimental Set-up** 

#### \_Task 1. Decision-making Tasks when a subject is placed in 140 situations similar to an insider



#### **Casual Questions (20 questions)**

#### Insider Situation-related Questions (140 questions)

 To verify the utility of the developed experimental tasks, the subjects were asked to perform two different sets of tasks.

#### Task 2. Professor's Email Intrusion

- Eye Closed (EC): Normal Resting states.
- Eye Open (EO): Normal Resting states.
  - EC and EO: Check the health of subjects/ develop a baseline.
- Internet Surfing: Focus on the tasks but without malicious intent.
  - i.e., Open their email to check their work.
- Professor's email Intrusion (malicious access, forbidden activity) to find the answer sheet to a test.
  - If the subject gets a score of 100 on the test, they will get more prizes.

According to literature reviews (Almehmadi, A., & El-Khatib, K., 2014 and Hashem, Y., et al, 2015), these experimental tasks are good trials for measuring an individual's malicious intentions.

### **Research Method**

**Experimental Set-up** 

- Subjects and sample size: 87 healthy subjects
  - **99.1% power (**the ability of a trial to detect a difference between two different groups) based on Post-Hoc Power analysis (Levine M & Ensom, MH. , 2001)
  - Age: mostly ~25 years old (except two people (60 years old))
    - The literature in 2012 (Taylor, G. S., & Schmidt, C., 2012), correlation were computed to determine the relationship between system performance and participant age, handedness and gender. No significant correlations were found (P>0.05 in each case).
  - Sex: 60 men and 27 women
  - Handed: 20 left-handed and 67 right-handed
  - Nationality: 10 from foreign countries and the others from Korea.
- Stimulus was mostly in visual form (80% getting data from vision senses) (Zeng, et al, 2009).
- **Controls:** Turn off the light, same room, silent environment, etc.



# **Research Method**

#### **Data processing**





	Band	Frequency (Hz)	Normally	Brain state		
Unconscious	Delta	1~4	Sleep, Continuous-attention tasks	Instinct		
	Theta	4~7 Hz	Drowsiness, idling	Emotion, stress		
Conscious	Alpha	7-13Hz	Relaxed/reflecting	Consciousness Vision function		
	Beta	13-30Hz	Active thinking, focus, hi alert, anxious	Thought Busy, active concentration		
	Gamma	30-50Hz	Short-term memory matching of recognized objects	Will Emotion		
Characteristics of EEG rhythms						

Task 1. Decision-making Tasks when a subject is placed in a similar insider situation



Reading (Thinking about problem)

- An increase in the **"Theta" indicator** is associated with high mental stress (cognitive workload).
- The type of tasks, influence the theta indicator.
- Thus, for similar insider situations, our simulation (Task 1) can represent when a subject is placed in a high mental stress environment.

Task 1. Decision-making Tasks when a subject is placed in a similar insider situation



Comparison between (a) Casual Reading (Thinking) and (b) Insider situation Reading (Thinking)

- In decisions regarding casual questions, there were **NO significant differences** between YES and NO. However, in decisions related to insider situation, there were **significant differences** between YES and NO.
- Whether the outcome of a decision is good or bad, the cognitive thinking process of a potential insider (before committing to a decision) evoked significantly different patterns of changes in the EEG indicators.
  - The power of "Alpha" decreased significantly for the insider situation but not for casual reading.
    - When they thought their bad decision (YES), they felt more worry and fear.
  - The power of "Gamma" slightly decreased when they thought bad decision (YES).
    - This can be interpreted as a subject's willingness to decide YES (Bad decision) is weaker than NO.
    - Or, it can represents a subject's greater worry when he/she considers the NO button.
  - The power of "Beta" decreased significantly when they thought bad decisions (YES).
    - Usually, increasing "Beta" is related to conscious focus; high attentional level and problem solving (Gola, M., et al, 2013).
    - When a subject thought about committing a crime(YES), their cognitive processing was very high because of the volume of information to consider or their imagining they are a real NPP worker, can result in a decrease in Beta (Yuan, H., et al., 2010).

Task 1. Decision-making Tasks when a subject is placed in a similar insider situation



Comparison between (a) BRAIN MASTER and (b)EMOTIV EPOC EEG devices.

For all 140 scenario tasks, the significant indicator(P<0.01) differences between the two groups were marked as\*\*.

BRAIN MASTER	ΕΜΟΤΙΥ ΕΡΟΟ	
Non-wearable device (19channels)	Wearable device (14channels)	
50 Subjects (Man:36, Woman:14)	37 Subjects (Man:24, Woman:13)	

- Fig 6. allows a quick identification of bad actor, supports current reactive program.
- Depending upon whether the outcome of a decision is good or bad, the corresponding decision making process of a potential insider (clicking YES or NO button) evoked significantly different patterns of changes in the EEG indicators.
- Alpha and Theta and Gamma values were decreased, and "Activity" (Beta/Alpha) and "Willingness" (Gamma/Alpha) values were increased when the subjects made a decision to become an insider after a period of thinking about the actions required to becoming an insider. 12
- Two indicators (Beta/Alpha and Gamma/Alpha) can be used to detect the insider right after committing a crime.

Task 1. Decision-making Tasks when a subject is placed in a similar insider situation



# Variation in Difficulty of the Tasks reflect strength of two indicators for YES Response

- Expected difficulty of the tasks involved (i.e., outsiders' demands) is directly related to the type of an insider.
  - Simple (easy) task (i.e., Delivering a message or lending a security card): This could be carried by a non-violent passive insider.
  - **Difficult task** (i.e., Planting a bomb in a facility or releasing a biochemical weapon in an NPP): These types of tasks belong to the actions of a violent-active insider.
- When the subject's deep level thinking is in serious conflict with their highest life value priorities, he/she will hesitate in making the decision. Thus, the Gamma/Alpha between YES-difficult and YES-easy was decreased due to their weak willingness.
- The thinking process, when a subject was placed in a difficult task situation, was very busy and positively activated (i.e., he/she was really contemplating on carrying out the action). Thus, the Beta/Alpha between YES-difficult and YES-easy was increased due to their highly cognitive workload and engagements.
- These results indicate that the details of the variations of the scale of Gamma/Alpha and Beta/Alpha indicators may help to identify an insider's type.

#### Task 2. Professor's Email Intrusion



Absolute Power Difference between non-malicious (surfing) and malicious activity (intrusion)

- When subject did **malicious action** (professor's email intrusion), the absolute power of:
  - Alpha indicator decreases (Fear or worry emotion, also true for an increase in the cognitive workload) ;
  - Beta indicator decrease (poor cognition);
  - Gamma indicator increases (worry and anxious feelings/ willingness); and
  - Gamma/Alpha indicator increases (big worry and strong willingness).

#### Predicting the High Potential Insider (At-Risk Insider)

Machine learning (Training Set)						
Support Vector Machine (SVM) Fine Gaussian SVM	Classification Rate	85-96 %				
<b>K-Nearest Neighbor (KNN)</b> Fine KNN	Classification Rate	91-98.8%				
Prediction test						
Ability to Predict (SVM)	Validation Accuracy	0.9252- <mark>0.9861</mark>				
Ability to Predict (KNN)	Validation Accuracy	0.9391- <mark>1</mark>				

**Classification Rate and Prediction Results** 

#### • Use of machine learning

- For more accurate classification between YES and NO
- To show the feasibility of predicting an At-risk insider based on "Activity" (Beta/Alpha) and "Willingness" (Gamma/Alpha) EEG indicators



#### **Malicious Intention Mapping-Coherence (brain function connectivity)**



#### **Malicious Intention Mapping-Coherence (brain function connectivity)**

# **Summary of the Results**

- The EEG frequency domain analysis showed that the selected indicators were useful as a quantitative measure to identify an "At-Risk (High Potential Risk)" insider.
- With the use of machine learning classification and these EEG indicators, "identifying" a high potential risk insider appears feasible.
- Possible explanations of the observed results from the channel analysis (malicious intention mapping) were proposed.
- The issue of information security needs to be addressed to implement the proposed approach to various security sensitive organizations.