

Analysis of the GOP Border security systems of the ROK Army by Using ABMS and NOLH design

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ABMS와 NOLH을 이용한 한국군 GOP 경계시스템에 관한 분석

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ABSTRACT

In this study, the border security problem of the ROK Army is examined by applying the agent-based modeling and simulation (ABMS) concept as well as its platform, MANA. Based on the approximately optimized behavior of the infiltrator obtained using genetic algorithm (GA), we evaluate the GOP border security system which consists of human resources, surveillance, as well as command and control (C2) systems. We use four measures of effectiveness (MOEs) to evaluate its performance, and we apply a near optimal latin hypercube (NOLH) design to deal with the large number of factors of interest in our model. By using a NOLH design, our simulation runs are implemented efficiently. We hope the results of this study provide valuable data for deciding the configuration of the border security system structure and the number of soldiers assigned in the platoon.

Key words : ABMS, border security system, NOLH design, MOEs

요약

본 논문은 에이전트 기반 모델링 및 시뮬레이션 개념과 그 플랫폼인 MANA를 적용한 연구이다. 유전학적 알고리즘을 이용하여 획득한 근사 최적화된 침투하는 적의 행동을 바탕으로 병력, 감시, 통신 시스템들로 이루어진 GOP 경계시스템의 효과를 평가하였다. 경계시스템의 효과 분석을 위해 4가지의 효과척도를 이용하였고, 모델을 구성하는 다수의 경계 자산을 고려하기 위해 NOLH 디자인을 적용하였다. NOLH 디자인을 적용함으로써, 시뮬레이션을 효과적으로 수행할 수 있었다. 이 연구의 결과는 경계시스템을 구성하는 경계자산의 효과적인 구성과 소초에 적절한 병력배치 등을 결정하는데 좋은 자료를 제공할 수 있을 것으로 기대한다.

주요어 : 에이전트 기반 모델링 & 시뮬레이션, 국경경계시스템, NOLH 디자인, 효과척도

1. Introduction

This study develops and analyzes a model of the ROK Army's border security system, which aims to prevent the infiltration of the enemy in the DMZ. The analysis uses a 10×7km section of the DMZ, representing the area of responsibility of a battalion in the 5th infantry division.

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1.1 Scope of this study

Our model in MANA is formulated with Fixed Guard Posts (FGPs), Moving Guard Posts (MGPs), TODs, platoon Command and Control Centers (CCCs), and battalion CCC as available agents. We analyze the border security system in terms of an overall system instead of considering these assets individually.

Our model exploration analyses an efficient experimental design methodology to capture a large number of interactions between the agents that may potentially affect the scenario outcomes. The controllable factors mainly include guard post parameters (detection range), equipment parameter

(for the TODs), network parameters (latency and reliability of network), and reinforcement troop parameters (response time, maneuver speed). The only type of uncontrollable parameters are the enemy parameters (detection range) that characterize infiltration behavior.

The model is run multiple times, varying a large number of design points, i.e., values for the controllable and uncontrollable factors. Then regression analysis, a statistical technique, is used to provide insights to the following question:

- Given the limited resources, what are the optimal combinations of sensors, surveillance and command and control (C2) systems to detect, classify and prevent enemy infiltrations into the DMZ?

1.2 Literature review

Since the 9/11 terrorist attack, there have been numerous types of studies on border security systems and other security systems in the U.S., but few studies have been conducted in South Korea. Berner^[1] analyzes the best combination of broad area maritime surveillance, UAV, and Vertical Take-off UAVs (VTUAVs) through ABMS. Pulat^[2] develops two-side optimization model using a mixed integer linear program to minimize the maximum achievable probability of infiltrator escape. Sung^[3] develops a security guard model to calculate the enemy detection rate and develops a TOD model to optimize the location of new equipment for the DMZ border security system. Based on these two models, Sung suggests an integrated model which incorporates the security guard model and the TOD model. While Sung obtains the detection rate of the TOD model, he also comments that a simulation model would have given better results for this problem because his model does not consider the enemy.

1.3 GOP border security concept

The overall shape of the Korean DMZ is depicted in Fig. 1. The length of the DMZ is approximately 248km (155 miles) and its maximum width is 4 km.

As Fig. 2 depicts, GOP lines are formed on each side of the DMZ from the central line, namely the military demarcation line (MDL). A guard post (GP) depicted in Fig. 2 is a type of security element located between the



Fig. 1. DMZ between South and North Korea

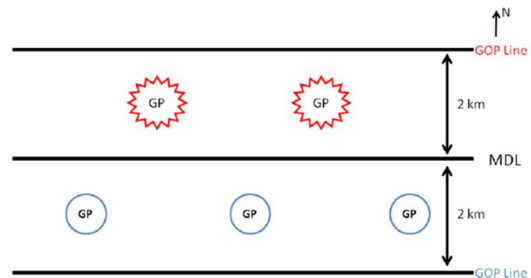


Fig. 2. Configuration of the DMZ

MDL and GOP lines. We do not consider GPs in our study due to their sensitive and classified nature.

2. Model development

Based on the description given in Section 1.3, and information on how the security system is implemented in reality, we develop the GOP border security system model using MANA^[4]. We aim to formulate a model in which we can flexibly develop scenarios so that insights may be obtained as to how different scenarios behave. In order to change the variables specifying these scenarios we use data mining tools as described in Section 3.2.

2.1 GOP border security model with MANA

Fig. 3 shows a simple scenario of the border security model when the number of FGP = 3, MGP = 3 and Reinforcement = 8 per platoon and the number of TOD = 2 per battalion with terrain map. In this simple scenario, the TOD detects the infiltrator, sends the information to the battalion CCC, then the battalion CCC makes decision and directs the platoon CCC to dispatch the reinforcement



Fig. 3. Screenshot of the GOP border security system model

along with the GOP line.

As MANA 4 has its own unit, we define that the MANA distance unit 1 = 10 m and time unit 1 = 5 sec and transform input parameters throughout the model.

When we compare the result of the human resource model of Sung^[3] with the parameters revealed in his paper, we obtain comparable results to his work. Compared to his model, our model's new contribution is to consider the infiltrator, and to jointly combine the human resource model and equipment model, which is separately implemented in his work.

2.2 Infiltrator

The behavior of the infiltrator is a significant variable that affects the results. We only assume that the enemy has a longer detection range than the FGP and MGP when these defensive human resources are in their default state. Furthermore, the objective of the infiltrators is to reach the waypoint without being killed. We exclude the case when the infiltrators attack the border security.

In the previous work^[5], we formulated an optimization model and approximately solved it using a GA in order to capture near optimal behavior of an infiltrating force. we used five simulation model constructs as decision variables to control that behavior, as enumerated below:

- Next waypoint ;
- Avoid FGPs ;
- Avoid Reinforcements ;
- Avoid MGPs ;
- Speed of movement ;

In our experiments, the GA obtains its best solution. The best behavior of the infiltrator is used to evaluate the effectiveness of security assets in the following section.

3. Analysis of the GOP border security system

3.1 MOEs

In the context of the overall security system, it is useful to choose representative MOEs to evaluate and compare the results of different configurations and the values of numerous factors including number of assets, detection range, shooting distance, etc. Although multiple MOEs can result in more sophisticated measures, limitations of the capabilities of MANA must also be considered. Based on these considerations, four MOEs are identified:

- Probability of enemy detection (MOE 1);
- Probability of enemy kill (MOE 2);
- Average time to reach enemy waypoint (MOE 3);
- Probability of enemy mission failure (MOE 4).

Each experiment starts with single infiltrator whose initial location is uniformly distributed along the 2km segment of interest on the northern edge of the GOP line, as described earlier. The infiltrator then proceeds south, towards his final waypoint. One of four events then occurs: He is first detected by a TOD, he is first detected by FGP, he is first detected by a MGP, or he reaches the final waypoint undetected. We let TODD, FGPD, and MGPD denote the first three of these events. Then, we can represent MOE 1 as:

$$MOE 1 = P(TODD) + P(FGPD) + P(MGPD),$$

Where P(.) is the probability associated with the corresponding event. If the enemy is first detected by a TOD then subsequently detected by an FGP or MGP then event TODD occurs but events FGPD and MGPD do not occur, i.e., the detection of the enemy is only counted once. We can define FGPK, MGPK, and Reink in a similar manner, more specifically the event that the infiltrator is killed by a FGP, MGP, or Reinforcement, respectively. In our scenario, Reinforcements are deployed conditioning

Table 1. Important factors and range of the GOP border security system.

Class.	Factor number	Asset	Name	Low Level (Real World)	High Level (Real World)
Human Resources	1	FGP	Num of FGPs per platoon	0	6
	2		Turret height	2 m	10 m
	3	MGP	Num of MGPs per platoon	0	4
	4		Speed of movement	1 km/h	2 km/h
	5		Detection range	20 m	50 m
	6	ReinF	Num of reinF per platoon	5	10
	7		Speed of movement	4 km/h	7 km/h
	8	Common for FGP and ReinF	Detection range/Default state	50 m	200 m
	9		Detection range/Contact state	200 m	300 m
	10		Shooting range/contact	50 m	150 m
	11		Communication latency	25 sec	60 sec
	12		Communication reliability	80 %	100 %
Equipment	13	TOD	Number of TODs per battalion	0	3
	14		Turret height	40 m	60 m
	15		Communication latency	25 sec	60 sec
	16		Communication reliability	25 sec	60 sec
Command and Control Center	17	Platoon CCC	Latency	25 sec	60 sec
	18		Reliability	80 %	100 %
	19	Battalion CCC	Latency	25 sec	60 sec
	20		Reliability	80 %	100 %

on the upgrade of border security level when detection is occurred by FGP, MGP, or TOD. We can define MOE 2 as

$$MOE\ 2 = P(FGPK) + P(MGPK) + P(ReinK \mid \text{Security level upgraded}).$$

MOE 3 represents the average time the infiltrator takes to reach their waypoint, given that they reach it within 1500 time steps. Finally, MOE 4 is the sum of the probability that the enemy is killed and the probability he is not killed but still fails to reach the final waypoint within 1500 time steps.

3.2 Design of experiment

3.2.1 Important Factors and Range

Our simulation model has many inputs that can be varied across a range of reasonable values. When this is

the case we must select a means for considering such an enormous number of possible combinations.

Among the numerous factors which comprise our border security model, twenty significant factors are identified in Table 1. The table groups these factors depending on whether they are specific to human resources, equipment(TOD) or the CCCs. Moreover, the table further indicates whether those human resources are FGPs, MGPs, or reinforcements, and whether it involves the platoon CCC or the battalion CCC.

3.2.2 NOLH Design

Latin Hypercube(LH) designs provide a flexible way to construct efficient designs for quantitative factors^[6]. In this study, the NOLH design helps reduce the computational requirements of some classic designs by many orders of magnitude, while still capable of developing a better understanding of a complex simulation model.

Let k denote the number of factors and let $N \geq k$ denote the number of design points. If each factor has 2 or 5 levels we could employ a 2^k or 5^k factorial design but the number of design points increases exponentially with k and quickly becomes impractical. To avoid the danger associated with high pairwise correlations associated with random LH designs, we employ the ideas of Cioppa and Lucas^[7], who develop NOLH design with good space-filling and orthogonality properties. In Table 2, the number of required design points is dramatically reduced from that of a full factorial 5^k design. Such design points are easily generated through the NOLH.xml file, which can be downloaded at the Seed center website^[8].

We use our 20 factors from Table 2 with varying number of levels for each factor. A full factorial design

would require 1.85×10^{19} design points whereas only 129 design points are used in our NOLH design, as indicated in Table 2. Based on the factors and ranges of the GOP border security system from Table 1, a NOLH design is generated, which an excerpt shown in Table 3.

Fig. 4 shows the pairwise plots for each pair of design points. These show that, at least pairwise, the design points are spread out fairly evenly. The pairwise correlations from Fig. 4 range from -0.097 and 0.105 , suggesting pairwise correlations are relatively low.

3.3 Model run

After generating 129 design points based on a NOLH design, numerous replications of the simulation model are formed. From a single run of the simulation model the enemy is either detected or not, the enemy is either killed or not, and the enemy's mission either fails or succeeds. We encode these events in a binary manner, using 0s and 1s. We use 1000 replications of the simulation to form estimators for MOE 1, MOE 2, MOE 3, and MOE 4.

3.3.1 Analysis tools

We use Xstudy and OldMcddata^[9], developed by the SEED Center, to facilitate the iteration through each of the 129 design points and to process output data from MANA.

Table 2. Requirement for NOLH design

No. of factors	No. of Design points
2~7	17
8~11	33
12~16	65
17~22	129
23~29	257

Table 3. NOLH design with 20 factors and 129 runs

low level	0	2	0	14	2	5	56	5	20	5	80	0	40	5	80	5	80	5	80	5
high level	6	10	4	27	5	10	98	20	30	12	100	3	60	12	100	12	100	12	100	15
factor name	Num FGP	FGP Hei	Num MGP	MGP Spd	MGP Det	Rein Num	Rein Spd	Det Def	Det Con	Com Lat	Com Rei	Num TOD	TOD Hei	TOD Lat	TOD Reli	Plt Lat	Plt Rel	Bat Lat	Bat Rei	Sht Ran
1	1	6	2	20	3	8	80	16	26	12	95	2	52	12	99	9	97	9	98	15
2	5	4	2	20	2	7	74	8	25	8	98	2	52	12	95	11	98	9	92	12
3	3	8	0	18	3	6	88	15	26	9	89	0	50	11	100	9	91	6	97	12
...																				
126	4	7	0	19	3	6	83	18	25	10	94	3	58	6	85	7	90	11	84	7
127	3	9	2	20	3	8	64	10	24	9	98	2	56	7	80	5	86	10	89	7
128	4	5	1	15	3	8	84	17	29	11	86	0	49	5	82	8	88	8	83	9
129	1	6	1	16	2	7	64	11	21	7	83	1	44	7	88	8	89	8	86	5

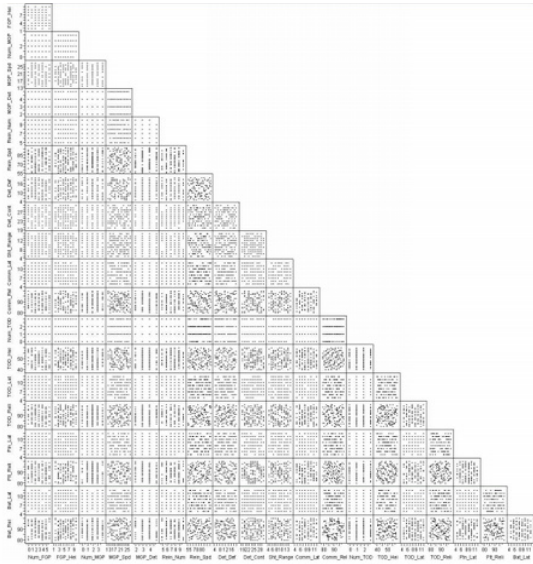


Fig. 4. Pairwise scatter plots for NOLH design with 20 factors in 129 runs

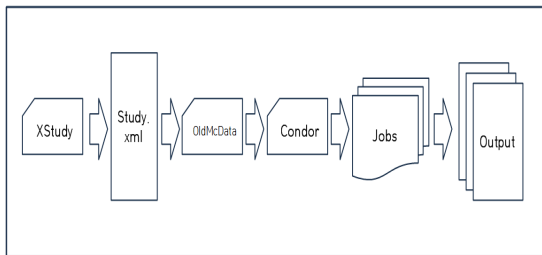


Fig. 5. Data process flow

Xstudy uses information specified within a MANA scenario file to identify the factors that are to be varied. Given a scenario file and the NOLH design file it generates a file called study.xml, which contains the study information about scenario, the number of replications, the exact specification of the algorithm for generating the factor variations, and factors to be used for that variation.

By using the study.xml file from Xstudy, OldMcData generates excursions which are modified from the original scenario file with the factors detailed in the study.xml file. For our GOP border security system model, 129 excursion files are generated and OldMcData conducts experiments using Condor^[10]. Condor is a workload management system developed at the University of Wisconsin-Madison for computationally intensive jobs. When the simulation

run terminates, OldMcData carries out post-processing in which it collects all output data into one file. The overall data process flow is described in Fig. 5.

To run the 129 excursions with 1000 replications for each excursion, the HPC (High Performance Computer) resources at the Naval Postgraduate School were used; the run time was approximately 6.5hours, which corresponds to approximately 240 hours on a typical personal laptop (2.4 GHz, 4.0 GB RAM).

3.4 Results analysis with statistical tools

After running the simulation model for 129 design points and 1000 replications for each design point, a total of 1.29×10^5 data points are analyzed for each MOE. To perform a statistical analysis of the obtained data points and to provide an appropriate visual tool for the decision makers, JMP is primarily used in this study, because its features are dynamic, interactive, visual, and easy to use. For this study, several techniques are used, which are listed below:^[11]

- Variable importance plot : PLS(Partial Least Square) regression in JMP and the variable importance score in TreeNet are used to identify importance factors among the 20 used in our model. Variable importance in the projection (VIP) is obtained by using PLS regression, and it summarizes the contribution a variable makes to the model, i.e., its correlation to the response (MOE). This enables us to choose the number of extracted factors by fitting the model to a subset of these factors and minimizing the prediction error. If a factor has a small VIP, then it means that the factor does not influence much on response. VIP score below one (We use 0.8 instead of 1^[12]) can be considered a small VIP since the average of squared VIP score is one.
- Contour plot : This is a graphical representation of the relationships among three numeric variables in two dimensions. Two important factors are identified from the VIPs, and each MOE provides a third variable for the contour levels.
- Bivariate scatter plot and fitting : This explores how the distribution of one continuous factor is related to another continuous factor. The analysis begins with scatter plot of points, to which we can interactively add other types

of fits, such as simple linear regression and polynomial regressions of selected degree.

3.5 Relation between MOEs

The correlation coefficient(γ) between random variables for the estimators of MOEs (\widehat{MOEs}) can give insights into linear relationship between MOEs. In our model, the enemy detected (X), killed (Y), and mission failed (Z) are the random variables which have binary number ‘0’ or ‘1’, i.e., detected(1) or not detected(0), from which $\widehat{MOE1}$, $\widehat{MOE2}$, and $\widehat{MOE4}$ can be estimated as shown in Section 3.1. Instead of the time length variable, the infiltrator remained (F) variable is added for a better understanding of $\widehat{MOE4}$. This result shows how the mission failure of infiltrator is caused by either a kill or remaining above the GOP line. The multivariate platform of JMP is used to explore the relationship and dependency between random variables. Suppose we compute the sample γ between two random variables, enemy detection (X) and enemy killed (Y), to explore the correlation between $\widehat{MOE1}$ and $\widehat{MOE2}$. We have $n = 1.29 \times 10^5$ with an iid pair of X and Y, written as x_i and y_i where $i = 1, 2, \dots, n$, then we can estimate the γ between X and Y by computing:

$$\gamma_{XY} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(n-1)s_X s_Y},$$

where \bar{x} and \bar{y} are the sample means, s_X and s_Y are the sample standard deviations of X and Y respectively. The matrix in Table 4 summarizes the strength of the linear relationship between each pair of random variables.

Table 4. Pair-wise correlations between \widehat{MOEs}

Variable	Variable	Correlation	Lower 95%	Upper 95%
Detection(X)	Casualty(Y)	0.1326	0.1271	0.1379
Mission(Z)	Detection(X)	0.4182	0.4137	0.4227
Mission(Z)	Casualty(Y)	0.4012	0.3966	0.4058
Remain(F)	Detection(X)	0.3621	0.3573	0.3668
Remain(F)	Casualty(Y)	-0.2065	-0.2117	-0.2013
Remain(F)	Mission(Z)	0.8130	0.8112	0.8149

The dependency between X and Y is only 0.1326 which is lower than one might anticipate. The reason is that most detection of infiltrators was by the TOD. TOD does not have an ability to kill directly from a long distance. From this result, we observe the necessity of enforcing the connection between surveillance systems and weapon systems.

By observing the correlation between F and Z, at around 0.8130, we conclude that the mission failure of the infiltrator is mostly caused by remaining above the GOP line within 1500 time steps.

Considering only one MOE can be relatively simple, but the results above show that it can be insufficient and have limitations in evaluating the overall system. The following section shows the statistical results for each MOE.

3.6 Probability of Enemy Detected(MOE 1)

Simulation results are analyzed through JMP and TreeNet as can be seen in Fig. 6. The VIP plot in the left

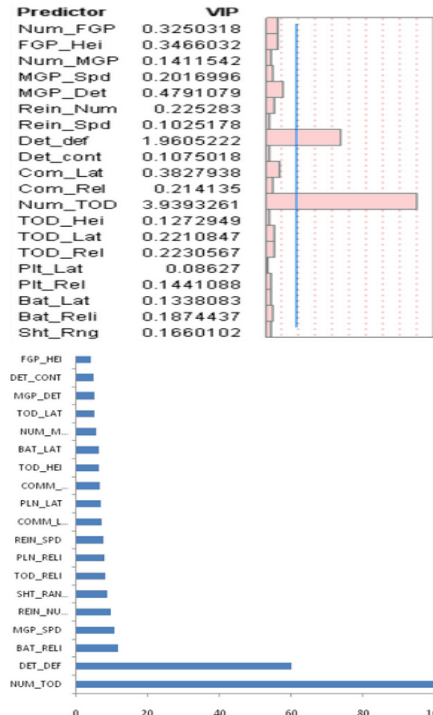


Fig. 6. VIP score(top) and variable importance score(bottom) for MOE 1

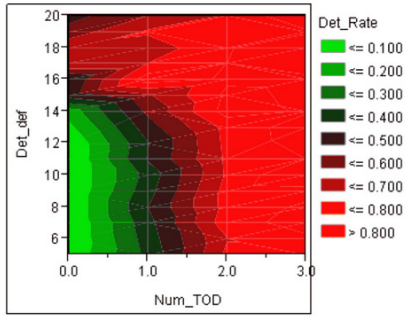


Fig. 7. Contour plot for MOE 1 correspond to the detection range and the number of TOD

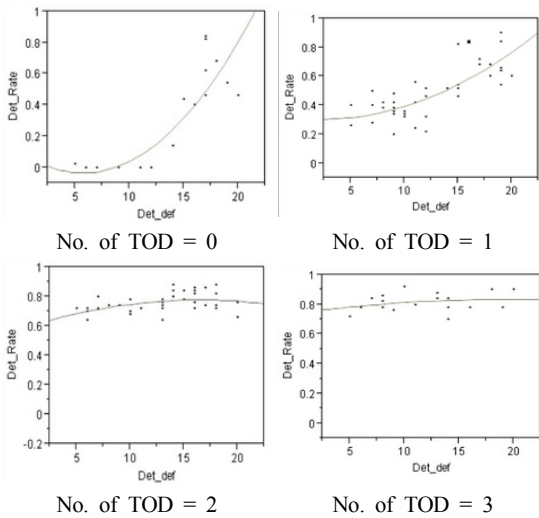


Fig. 8. MOE 1 corresponding to the detection distance and the number of TOD

side of Fig. 6 reports the VIP scores corresponding to the 20 factors. The factors which have VIP score greater than 0.8, denoted by the blue vertical line in the left figure, represent important factors among the twenty. The graph in the right side of Fig. 6 reflects the contribution of each factor in predicting MOE 1 by regression in TreeNet. We identify the number of TODs and detection range (default state) as the important factors with the VIP scores of 3.94 and 1.96, and the variable importance scores of 100% and 60% respectively. These scores are calculated using the results of both methods as shown in Fig. 6.

We can see that the security guards, including FGP and MGP, does not contribute much to the enemy detection

according to the results in Fig. 6.

The contour plot for MOE 1 corresponding to the number of TODs and the detection range (default state) is identified in Fig. 7. In general, as the number of TODs and the detection range increases, MOE 1 tends to increase. Since the infiltrator is assumed to have a detection range of 150m, MOE 1 abruptly increases when the detection range becomes 150m. The detection rate has a relatively high spread from 0 to 0.8 when none of the TODs are used in the operational area, whereas it ranges from 0.6 to 0.9 when two TODs are used. When the number of TODs is greater than or equal to 2, then the MOE 1 is always greater than 0.6. Later, we explore this using a regression.

Since the number of TODs seems to be the most important factor of the surveillance system, i.e., MOE 1 is most sensitive to changes in the number of TODs, this sensitivity is further examined in Fig. 8. The variability chart shows that when the number of TODs is more than one, the value of MOE 1 becomes reasonably stable, whereas it has a huge variability when the number of TOD is strictly fewer than two.

4. Conclusion

In this paper, the border security problem of the ROK Army is examined by applying the ABMS concept and its platform, MANA. Based on the stream of border security using ABMS, this study contributes to the military operations for the following aspects.

First, in a departure from the previous works, this study suggests a way to evaluate the border security system in terms of three main systems, namely surveillance, communication, and weapon system, through the four MOEs. The coefficient correlation matrix of estimators for MOEs demonstrates the dependency of the pair of MOEs and it is expected to help analyze the system in different aspects.

Second, a variety of factors comprising the border security system were considered in order to identify the importance factors and obtain the interaction between these factors. In a contrast to the previous works, the number of assets (TOD, FGP, MGP, and reinforcement

troop) are considered for some of the design point factors. From the results using an enormous number of simulation runs, it was possible to determine the qualitative importance of each factor and to obtain the graphical results for the decision maker.

The results of this study can be a valuable resource in terms of deciding the configuration of the assets, border security system structure, and number of soldiers assigned to either a post or a platoon. The connection between the surveillance and weapon systems needs to be enforced in order to attain an effective border security system, based on the scenario we used.

From the basis of this study, the ROK army can evaluate an ongoing project, more specifically the GOP unmanned border security system, which aims to substitute human resources with cameras and sensors. The ROK Army also has plans to increase the number of TODs to improve the border security system. As we identified in the results of Section 3.5, the effectiveness of TOD is not linearly increased. Thus, the method used in this study can be applied to obtain the optimal number of assets given the limited budget.

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