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Assessing network-based earthquake early warning systems in low-seismicity areas

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Earthquake early warning (EEW) technology, designed to alert the public of earthquake risks after initial P-wave detection but before the onset of strong tremors, has developed rapidly. Methodologies from various fields are combined in EEW systems to estimate earthquake locations, magnitudes, and expected intensities based on the initial P-wave data. These systems operate automatically because prompt responses are required. However, as no common evaluation framework for EEW system verification exists, potentially divergent evaluations from reviewers or countries could ensue. Moreover, evaluating EEW systems is more complicated when the target area does not experience frequent earthquakes. We aimed to establish a guidance review process for lowseismicity areas to ensure reliable and stable integrated EEW system operation. We incorporated management aspects through actual system operator surveillance and designed an EEW assessment process based on feedback from our surveys. Using this approach, we created a comprehensive and wellinformed evaluation process that considers the diverse perspectives of experts involved in EEWs. Our proposed assessment method allows for a uniform and consistent evaluation process, regardless of changes in the methods or technologies used by EEW systems. The method aims to guide EEW system assessments in low-seismicity areas.

KEYWORDS

warning assessment, analytic hierarchy process, earthquake early warning, lowseismicity area, public services, seismic hazard

1 Introduction

For rapid mitigation against seismic hazards, automated alerting technology based on seismic networks EEWs is used, with automated event estimation employed to facilitate the issuing of timeous public alerts (Allen and Melgar, 2019; Velazquez et al., 2020). Such systems operate at national or local government levels, as their fundamental technology requires 24 h monitoring (Mayer et al., 2008). Several countries have implemented EEW systems as public services, including Japan (Hoshiba and Ozaki, 2014), Republic of Korea (Lee et al., 2022), Taiwan (Hsiao et al., 2009), and the United States (Kohler et al., 2018). Among these countries, Republic of Korea experiences the lowest occurrence frequency of medium-to large-scale earthquakes.

After the introduction of EEW in Republic of Korea in 2015 (KMA, 2018), public awareness remained low until two earthquakes occurred (local magnitude $[M_L] = 5.8$ on

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12 September 2016, and $M_L = 5.4$ on 15 November 2017). The low level of public awareness could be ascribed to the occurrence and recording of only ten earthquakes exceeding $M_L > 5.0$ since instrumental observations started in 1980. However, the two events of 2016–2017 were a timely reminder of the importance of EEW, and EEW technology has since developed rapidly in the country. Initially, a cell broadcasting service (CBS) was intended for governmental use, but the Korea Meteorological Administration (KMA, 2017) has extracted and applied relevant parts of the service for earthquake, tsunami, and volcano warnings (KMA, 2018). The network-based EEW system was optimized further to suit domestic observation environments, estimation of magnitudes (Sheen et al., 2018), and detection of initial P-waves (Cho et al., 2022).

KMA EEW progress relies on warning time reduction, as it is crucial to warn people as quickly as possible in the case of an earthquake (Bostrom et al., 2022; Ahn et al., 2023). Warning time is a critical factor in assessing preparedness for an earthquake and a vital consideration from a risk management perspective (Allen et al., 2009). The evaluation of EEWs is often based on lead time, i.e., the time between an alert is received and the arrival of tremors or shaking (Cremen et al., 2022). A primary method to reduce EEW lead time is decreasing the number of stations (nS) used for detecting earthquakes in the initial analysis version of the event. However, the likelihood of prediction errors increases when fewer stations are involved in earthquake determination and analysis.

Although Republic of Korea has not experienced social disruption induced by false earthquake alarms for KMA, such incidents could cause substantial social issues. On 5 January 2018, a false alarm occurred when an EEW mistakenly identified two small earthquakes as one large earthquake in Japan, causing temporary public panic (BBC News, 2018). The head of the Japan Meteorological Administration was forced to apologize publicly for the error. Owing to a computer error, a warning of a large-scale earthquake off the coast of California (United States) was issued on 22 June 2017, 92 years after it had actually occurred. This warning caused widespread public confusion (BBC News, 2017). Another notable example is a false alert issued in Mexico on 28 July 2014, by Sky Alert, a popular smartphone application, prompting response from numerous people. This incident triggered a social issue in Mexico, invoking Aesop's fable "The Boy Who Cried Wolf" (Reddy, 2020). Although SkyAlert is not a public service, the consequences of the false alarm emphasized the importance of proper EEW management.

While the fact that Republic of Korea experiences fewer earthquakes and alerts may be considered a positive aspect, it has also led to many Koreans being unaware of EEWs. According to a 2020 survey by Ahn (2021), only 84 (approximately 44%) of 192 general public respondents (i.e., non-seismologists and noncivil engineers) were aware of EEWs. As a result of the limited number of alert cases and the lack of public awareness, obtaining public evaluations of these services remains a challenge.

Owing to these limitations, EEWs are generally evaluated by experts rather than the public. Past studies have either assessed the point source or alert accuracy. Point source assessments verify EEW algorithms for magnitude and hypocenter accuracy (e.g., Cua et al., 2009; Chen et al., 2015; Massin et al., 2021). Alert assessments are performed to prevent errors during operations or simulations (e.g., Chen et al., 2019; Chung et al., 2019; Zuccolo et al., 2021). However, management aspects require a comprehensive review of the overall EEW. Cochran et al. (2018) proposed a framework for a Testing and Certification Platform (TCP) that assesses the performance of the overall system alerts issued by the decision module and those generated by individual algorithms. The TCP assessment consists of both point-source alert elements (i.e., magnitude, epicentral location, origin time, and alert latency) and ground-motion prediction accuracy. TCP is evaluated on the basis of alerts, and the more earthquakes available for evaluation, the better. Unfortunately, in countries with low earthquake frequency and relatively few strong motions, the Cochran et al. (2018) platform is limited.

Expert assessments vary based on the knowledge and interests of the individual conducting the evaluation (Binger et al., 2012; May et al., 2016). Additionally, the evaluator must make a minimum number of decisions for the assessment, which could be challenging. Experts tend to be cautious during decisionmaking because their decisions ultimately affect the public services provided to citizens (Meijer and Grimmelikhuijsen, 2020). For KMA EEW, the public is informed of earthquake risks for events with $M_L > 3.5$. However, since 2010, there have been only five events of $M_L \ge 5.0$. As a result, evaluations often include earthquakes with slight magnitudes which was events of $M_L \ge 2.5$ (Cho et al., 2022). These issues raise concerns regarding the appropriate magnitude target for assessment, review area, and acceptable accuracy standard. Answers from the experts could differ, i.e., it is crucial to recognize that even experts could hold different perspectives on related issues.

The fundamental goal of an EEW is achieving promptness and accuracy (Allen et al., 2009; Kamigaichi et al., 2009; Satriano et al., 2011; Finazzi, 2020) However, balancing these two aspects is challenging, and performance scores could vary depending on the criteria and the reviewer. Therefore, as regards the decision-making process in assessing EEW, it is essential to integrate expert opinion with the goal of providing an effective public service, as different people could have different values and priorities, even when working toward a common mission. Therefore, we aimed to establish an EEW review method that incorporates management aspects through actual system operator surveillance. We designed an EEW assessment process based on the feedback received from our surveys, such as criteria for analysis, pairwise comparison, and so on. Our approach was to create a comprehensive and well-informed evaluation process that considers the diverse perspectives of the experts involved in EEWs.

2 Criteria and methods

2.1 Alert criteria and EEW service

Conservative alert standards are in place in the KMA EEW service. The Korean Peninsula is an area of low seismicity; therefore, even a small tremor could unnerve citizens. Furthermore, considering that the country hosts semiconductor factories, high-speed rails, and nuclear facilities, alert services are conducted with sensitivity preparation. In the event of an earthquake of $M_{\rm L} \geq 3.5$, seismic information is disseminated within 5–10 s after the initial detection.



The KMA EEW uses at least four stations for optimal performance (Cho et al., 2022), aiming to make decisions on alerts about 5 s after the first observation when an earthquake occurs on the Korean Peninsula. In EEW, the theoretical minimum number of stations (nS) for hypocenter determination is three (Yamada et al., 2021), but relying on the assessment accuracy can be challenging. Low accuracy is particularly associated with earthquakes occurring outside the observation network. Based on these settings, when an earthquake with $M_L > 3.5$ (or > 4.0, if occurring over the sea) occurs, a warning message is automatically propagated to the public. If an earthquake with an $M_L > 5.0$ occur, warnings with a stronger alert sound are disseminated. These alert criteria were designed as a policy decision in Republic of Korea. Therefore, the EEW alert decisions might differ in other countries.

Additionally, the KMA operates a network-based EEW that incorporates the algorithms ElarmS-3.0 (Chung et al., 2019), RTLoc (Satriano et al., 2008), and Maxwell–Hertz electrostatic potential theory (Sheen, 2016). These algorithms have been optimized for the Korean Peninsula by KMA and are used simultaneously for monitoring earthquakes. The KMA EEW is based on the mutual interaction and automatic collaborative decision making of the algorithms, and produces an alert when two or more algorithms detect an earthquake. In principle, if two logically distinct algorithms working concurrently sound an alert, the likelihood of an actual earthquake occurrence is higher.

By combining three algorithms, we achieve analysis accuracy as well as 24 h service stability. This approach has been instrumental in overcoming occasional system failures experienced by the KMA, including bugs, delays, and network issues. However, owing to the complexity of the algorithm, which involves the integration of multiple systems, it becomes sensitive to minor enhancements.

2.2 Disaster communications to the public

The KMA EEW alerts, determined by a combination of technology and policy, are issued to citizens in the event of an earthquake through the CBS. Using CBS technology, a warning is sent to numerous users in the vicinity of base stations (Doi, 2011; Wu et al., 2022). Alarm is transmitted to mobile phones in approximately 3 s (Minson et al., 2018), and the public receives messages shortly after earthquake detection. Alerts are forcibly sent to individual mobile phones in the form of emergency text messages. Figure 1 shows an example of earthquake warning information delivered to the public. The message provides information on the

epicenter, event time, and magnitude of the earthquake, along with guidelines to "drop, cover, and hold on" (DCHO) (Porter, 2016).

Following alert dissemination, evaluations from citizens regarding the system's effectiveness could be gathered. Speediness, as perceived by the citizens, is the time difference between receiving the alert and feeling the earthquake vibrations. It is challenging for citizens to quantitatively assess promptness, as the warning and the shaking occur within a very short time frame. However, as regards source information (i.e., location, magnitude, and event time), people judge the warning accuracy by comparing it to subsequent earthquake information released through news broadcasts. Collecting feedback is vital for understanding the needs of the public with respect to the DCHO advice, which is a crucial aspect of emergency preparedness. However, intermediate and large earthquakes do not occur frequently in the Korean Peninsula, and assessing KMA EEW might be difficult for the general public. Consequently, only people who manage and develop the system can evaluate the KMA EEW.

2.3 Assessment survey and design

Operator or manager assessments may vary depending on the expertise of the reviewer (Cooke and Goossens, 2004). Here, evaluations conducted from the perspective of a small group of developers could lead to serious issues in the future. To address this concern, we gathered opinions from individuals with relevant experience and field knowledge. We surveyed 18 KMA employees who operated EEW from July 14–16, 2021. The survey research was designed and conducted in the following order 1) assessment parameter selection, 2) relational questionnaire development based on assessment parameters and pairwise analysis questionnaires for the factors, 3) operator interviews, 4) questionnaire criteria determination, analytic hierarchy process (AHP) analysis, and weight function designation, and 5) EEW operational aspect assessment.

2.3.1 Aspect-based assessment parameter selection

To select appropriate assessment parameters for EEW, two aspects were considered, namely, operations and management, which are distinct but interrelated. Operation refers to the dayto-day activities and tasks of a system (Whipple and Frankel, 2000). In EEWs, operations focus on the reliability of alert production (Medina-Cetina and Nadim, 2008). In contrast, management involves EEW planning and control to achieve high performance (Ittner and Lacker, 1997), considering alert reliability and seismic information accuracy (Ruhl et al., 2019; Esposito et al., 2022). Managers are responsible for overseeing objective reviews and implementing improvements, ensuring that a balance is maintained between safety, effectiveness, and sustainability (Too and Weaver, 2014). Figure 2 shows the EEW algorithm analysis process with both aspects. The assessment steps are shown in the EEW flow, considering algorithm analysis based on simulations.

As regards the operational aspects, system behavior was categorized into three steps. Step 1 is the evaluation preparation stage, comprising input of the earthquake scenario and setting the minimum targeted nS. Step 2 is the detection phase of the analysis



process [i.e., short-time-average over long-time-average trigger (Allen, 1982)] or the filter picker (Lomax et al., 2012). If the nS condition is satisfied, it is verified to determine whether it becomes an associated event (Cho et al., 2022). If the alarm fails to generate an event, verification is required whether this event was alarm worthy. Verification allows the determination of the alarm being operationally a false alarm or normal operation. Step 3 comprises predicting the source and assessing the alarm based on associated event information (Weyrich et al., 2021). Alarms can be produced by EEWs based on the predicted source information, which is termed decision making of warning (DMW).

As regards the management aspect, it is essential to verify the success of the alert and the source information that influenced the decision to issue an early warning. By considering all the parameters that impacted the final alert, it could be ensured that the algorithm and system were free of issues. Figure 2 presents representative error sources that could be easily identified from an administrator viewpoint. However, various errors could occur in operations, and this procedure is only applicable to natural earthquakes.

In the review of EEW assessment, we ignored the associate event time, which depends on the nS. Additionally, waveform data used in

KMA seismic stations are applied only to locations that pass strict criteria for latency, background noise, and trigger rate (KMA, 2017; Ahn et al., 2021). Consequently, data lagging or failed equipment rarely occur for KMA EEW in real-time operations. However, from a management perspective, a station detection rate review within a certain time frame could be performed separately, because of trigger checking at the observatory.

We selected four parameters for this study based on aspects. 1) The first parameter is "prevent missed alerts (PM)" in the system, i.e., confirming whether a targeted earthquake alert was missed. Alerts in normal operations are issued when the national observation network detects a warning event using n stations. 2) The second parameter is "prevent false alerts (PF)" in the system, which determines whether a targeted earthquake alert is incorrect. This involves checking alarms that are normal operations, non-earthquakes, or source errors when EEW alerts occur. 3) The third parameter is the accuracy of magnitude (AM), and 4) is the accuracy of location (AL). These four parameters were reviewed from operational and administrative perspectives and represent all the information that the public sees when an alert is triggered.

2.3.2 Assessment criteria preparation

To help respondents better understand the assessment parameters, we provided training before conducting the research. During the training, we introduced various existing analytical examples, explained the unique features of the criteria, and received approval for ethical review surveys involving human participants. Further, we discussed the EEW technical and theoretical limitations. These are 1) Inability to fully understand the underground geological structure, implying that seismic information is subject to uncertainty at source (Iervolino et al., 2009; Tidlund et al., 2022; Murray et al., 2023; Ni et al., 2023). 2) Detection accuracy varies depending on the observation network because of wave propagation attenuation (Jackson and Anderson, 1970; Vipin and Sitharam, 2011; Goda et al., 2023). 3) Accuracy limitations of source detection in coastal earthquakes (Angove et al., 2019; Lim et al., 2020; Takabatake and Kojima, 2023), and the difficulty of installing seismometers in the seabed (Podolskiy et al., 2021). The system operators understood and agreed with these EEW limitations. In addition, we emphasized understanding of the characteristics of earthquakes in the Korean Peninsula, as well as the psychological conditions of the public. Given the infrequent occurrence of seismic events in the Korean Peninsula, even weak shaking could elicit feelings of surprise and unease among the general population (Kwon et al., 2020; Yeon et al., 2020). Therefore, unlike other countries, the KMA issues warning criteria from even slight magnitudes (M_L 3.5 and over).

Considering KMA policy and technology characteristics, we designed a questionnaire for the EEW assessment. In the criteria establishment stage of the survey, we created five response items to establish the assessment criteria. We classified the criteria between inland areas and outside (i.e., ocean, outside the country) of the seismic observation network. Both groups 1) within the seismic observation network (= inside seismic network, ISN) and 2) outside the seismic observation network (= outside seismic network, OSN) were considered. Categorization by observation network was included in all five questions.

The survey questions were:

Q.1: "What is the minimum magnitude of an earthquake to be analyzed for performance assessment?" We needed to establish a baseline progression for our evaluation, considering the margin of error in our prediction scale. However, detecting initial waveforms from slight magnitude earthquakes is complicated owing to their small amplitudes and poor observation environments. The EEW performance results could, therefore, vary depending on the criteria and magnitude, with greater sensitivity for weaker events than larger ones.

Q.2: "Compared with $M_L \geq 5.0$, how important is an earthquake of $M_L < 5.0$?" This item was included in the questionnaire, as the level of ensuing damage could differ according to the magnitude of the event. While seismic intensity is important from a damage perspective, the basic information transmitted is magnitude and location. Therefore, we investigated the relativity of magnitude. The initial standard for early warnings of earthquakes in Republic of Korea was $M_L > 5$.

Q.3: "How important is the comparison of earthquake locations, both inside and outside the observation network?" This question was included because it is associated with the observation network conditions, and the current criteria classify inland and outside events (i.e., teleseismic or occurring outside the country). These criteria represent the KMA strategy for considering the impact of earthquake detection relative to the observation network. Inland earthquakes can be detected rapidly and analyzed as they occur between observation networks, but oceanic analysis could likely be inaccurate in relation to the source owing to biased and sparse observation networks.

Q.4: "What is the maximum allowable error range when evaluating the accuracy of the magnitude?"

Q.5: "What is the maximum allowable error range when evaluating the accuracy of the epicenter?" The fourth and fifth questions are related to accuracy, as accuracy could affect missed and false alarms. Additionally, if the difference between the predicted information and precise information analyzed afterward is significant, an operating system check could be required.

In the criteria establishment stage of our survey, we analyzed the results based on the five questionnaires and collected the majority opinion regarding operational recommendations. Table 1 shows a summary of the established criteria from the questionnaire. The criteria in Q.1 allowed us to define the analysis object range. We established the range for earthquakes to be considered by ISN and OSN, i.e., 0.5 smaller than the magnitude of the alert criteria.

We set the weight for event sources based on Q.2 and Q.3. From an operational side, all earthquakes are valued equally because the evaluation pertains only to whether an alert should be issued. For instance, in a point-source assessment by Cochran et al. (2018), all earthquakes were also valued equally. However, the damage caused by an earthquake depends on the location and magnitude of the event. Therefore, we proposed weighing of the source effects. The score for each earthquake event was:

Event Value
$$(EV) = \delta M \times \delta L$$
 (1)

where EV is event value, which is the judgment of earthquakes from a management perspective. δM is the weight for magnitude, and δL is the weight for the source location (i.e., ISN, OSN) of the earthquake. For example, if an earthquake of M_L 5.5 occurred inland, the event value would be 1. Conversely, if an earthquake with magnitude M_L 3.2 occurred inland near the coast, the event value would be 0.18.

The criteria for an EEW false alarm was selected based on answers to Q.4 and Q.5. The criteria set for a false alarm were assessed based on the estimated source information produced by EEWs. In practice, the results of both source location and magnitude of EEWs depend on the nS (Lim et al., 2020). The more stations are involved in collecting information, the better would be the accuracy of epicenter analysis. From a management perspective, the criteria for a false alarm could be a crucial factor for planning system improvements.

2.3.3 Assessment parameter scoring

Assessments of EEWs must consider both the appropriateness of the alerts and the accuracy of earthquake source estimates. The source estimates in EEWs are updated rapidly as they increase because of a higher number of nS. However, we do not recommend reviewing all the steps that are updated over time, because the public responds to the first warning message (Lassa, 2008; Vihalemm et al., 2012); therefore, the information at the moment the alarm occurs is the only concern. Alert time depends on

	Assessment criteria	Criteria			
		ISN	OSN		
(Q.1)	Minimum magnitude of assessment	3.0	3.5		
(Q.2)	Weight of magnitude (δM)	$1.0 \ (5.0 \le M)$	$1.0 (5.0 \le M)$		
		$0.7 (3.5 \le M < 5.0)$	$0.7 \ (4.0 \le M < 5.0)$		
		$0.3 \ (2.0 \le M < 3.5)$	$0.3 \ (2.0 \le M < 3.5)$		
(Q.3)	Weight of epicenter (δL)	1.0	0.6		
(Q.4)	Maximum allowable magnitude error (ΔM)	0.4	0.6		
(Q.5)	Maximum allowable epicenter error (ΔL)	10 km	30 km		

TABLE 1 Performance assessment criteria based on questionnaire responses.

nS, which is the most important factor in EEW alarm decision making. Our proposed assessment is designed to evaluate performance according to the target nS.

The first step was to score alert appropriateness (AA). We designed the scoring to include only events where the alert was a normal operation under the target nS conditions. If an earthquake required an alert but resulted in a false or missed alarm, we scored it as zero. This allowed us to determine if the system has effective controls for false or non-alarms. Consequently, our assessment could result in a low score if the system failed to control alarms for multiple cases. The formula for this is:

Prevent false alerts
$$(PF) = EV \times AA$$
 (2)

Prevent Miss alerts
$$(PM) = EV \times AA$$
 (3)

where EV is the event value by Eq. 1, AA is scoring of alert appropriateness (only 1 or 0). Therefore, a normal alert is 1, and false and missed alarms are 0.

The second step is to score the accuracy of an estimated source. By default, we designed the accuracy calculation to be based only on positive cases (normal alerts). The basic framework is similar to that of the Cochran et al. (2018) model. The expression is:

Accurate of magnitude
$$(AM) = EV \times \left(1 - \frac{M_{No} - M_{EEW}}{\Delta M}\right) \times AA$$
(4)

Accurate of location (AL) = EV ×
$$\left(1 - \frac{L_{No} - L_{EEW}}{\Delta L}\right) \times AA$$
 (5)

where M_{NO} and L_{NO} are the notified magnitude and epicenter of the earthquake according to the KMA. The notified earthquake information is calculated through manual analysis by seismic analysts at the KMA. The M_{EEW} and L_{EEW} are the calculated source information (i.e., magnitude and epicenter) according to the target nS in the EEW. The maximum allowable error values, ΔM and ΔL , are based on the answers to Q.4 and Q.5. The accuracy value is close to 1 for a small error and close to 0 for a large error.

2.3.4 Pairwise analysis and assessment parameter weight

We assumed that different experts would place different values on the four parameters; therefore, we performed a relative comparison of the four parameters to obtain their values. For the weights of the four assessment parameters, we used the Analytic Hierarchy Process (AHP) that involves pairwise comparison, weight calculation, and consistency verification, as described by Saaty et al. (1990). The AHP process could be used as a rational decision-making tool for risk assessment to stratify, simplify, and systematize multiple criteria (Wen, 2015; Thaker et al., 2018; Providakis et al., 2022). This survey aimed to identify the most important parameters for operators by employing relative comparability.

During the pairwise comparison process, the study assessed the relative importance of the four assessment parameters using a ninepoint scale, as shown in Figure 3. Following this step, the results were subjected to consistency verification to ensure their reliability. The consistency index (CI) used to assess the reliability of the responses was calculated through the maximum Eigenvalue operation of the comparison matrix. A lower CI value indicated higher reliability. For this study, an average weight of 11 questionnaires with a consistency index of 0.2 or less was used for the analysis. A summary of the results of these weights is presented in Table 2.

In pairwise analysis, the most important factor from the perspective of managers and operators is controlling false alarms. This factor is crucial because the issuing of a warning and the level of the warning depend on the accuracy of magnitude and location. Although the algorithm itself does not reveal this, operators consider the activation of a warning without an actual earthquake a significant problem. Therefore, false alarm control was allocated the highest weight, followed by non-alarm control, location accuracy, and magnitude accuracy.

To determine the assessment of the overall score (*OS*), we used the weights and assessment points of the four parameters. The final score was obtained by multiplying the score of each element by its weight, as follows:

Overall Score (OS) =
$$\alpha \cdot PM + \beta \cdot PF + \gamma \cdot AM + \kappa \cdot AL$$
 (6)

where *PM*, *PF*, *AM*, and *AL* refer to parameters such as prevent missed alerts, prevent false alerts, accuracy of magnitude, and accuracy of location, with α , β , γ , and κ as their respective weights (Table 2).

3 Results

3.1 Assessment preparation

We verified the proposed assessment of results in EEWs using past earthquake data. We analyzed 58 domestic earthquakes that

Q. Indicate the factor that you think is relatively more important when evaluating the performance of the EEW system.										
[Factor]	←	← Important			Rating		lm	portant	\rightarrow	[Factor]
Prevent missed alerts	(5)	4	3	2	1	2	3	Ø	(5)	Prevent false alerts
Prevent missed alerts	(5)	4	3	Q	1	2	3	4	(5)	Accurate magnitude estimation
Prevent missed alerts	(5)	4	3	Q	1	2	3	4	(5)	Accurate epicenter estimation
Prevent false alerts	(5)	Ś	3	2	1	2	3	4	(5)	Accurate magnitude estimation
Prevent false alerts	(5)	Ć	3	2	1	2	3	4	(5)	Accurate epicenter estimation
Accurate magnitude estimation	(5)	4	3	2	1	2	Ś	4	(5)	Accurate epicenter estimation

FIGURE 3

Questionnaire example for pairwise assessment parameter comparison. EEW, earthquake early warning.

TABLE 2	Weighting of four	assessment	parameters	based on o	questionnaire	answers	received from	n Korean	Metrological	Administration	(KMA).

Section	ΡΜ (α)	PF (β)	AM (γ)	AL (κ)	Consistency index
SR-1	0.073	0.475	0.125	0.327	0.159
SR-2	0.264	0.527	0.123	0.087	0.05
SR-3	0.237	0.189	0.361	0.212	0.68
SR-4	0.533	0.272	0.13	0.065	0.26
SR-5	0.125	0.625	0.125	0.125	0.00
SR-6	0.351	0.351	0.109	0.189	0.00
SR-7	0.28	0.553	0.083	0.083	0.03
SR-8	0.076	0.484	0.304	0.136	0.11
SR-9	0.523	0.272	0.089	0.117	0.14
SR-10	0.056	0.546	0.159	0.239	0.09
SR-11	0.092	0.495	0.206	0.206	0.05
SR-12	0.248	0.62	0.066	0.066	0.12
SR-13	0.202	0.558	0.087	0.154	0.07
Weight	0.208	0.501	0.134	0.157	

SR, survey respondents.

occurred between January 2015 and August 2021, with $M_L \ge$ 3.0 and \ge 3.5 for the ISN and OSN, respectively. The details of the events are shown in Figure 4 and summarized in Table 3. In this analysis, the operational decisions for checking the appropriateness of the alerts are shown in Table 4 using the criteria of Cho et al. (2022). However, differing from the Cho et al. (2022) study, the current assessment imposed stricter criteria for matching source information based on the survey result.

We assessed the performance of the initial P-wave trigger criteria in the ElarmS-3 algorithm, which was proposed by Chung et al. (2019) and include minPa, Range-Post Trigger (RP), NEtoZ, and Zero-Crossing (ZC). MinPa is the criterion for the initial P-wave, representing the minimum amplitude of acceleration during 4 s. Noise signals (i.e., boxcars, spikes) and block S-waves are removed using RP and NEtoZ, respectively. The validity of the signal is determined by ZC by counting the number of samples based on zero. Ultimately, we designed five simulation matrices based on the study by Cho et al. (2022). The configuration of each module is summarized in Table 5.

We assessed the effectiveness of the EEWs by simulating past earthquake events. Although past earthquake simulations cannot completely replicate real-time data delays, simulation analyses based



on recorded data could produce reliable results similar to those of real-time operations (Lim et al., 2020). The simulation method used in this study does not allow for the delay of telecommunication time, and time-related errors were not surveyed, which is a limitation of the study. We assumed that the processing of observation data was conducted without delay.

Figure 5 shows an instance of simulation results for an earthquake. Here, we observed that the analysis results differed each time the simulation was executed, despite using identical simulation environments. The recording values of several observatories were transmitted simultaneously to the algorithm, and the buffer order changed owing to differences in micro-small system time. These issues were also found in a couple of pairs of a real-time system. However, as they do not significantly affect the overall analysis, we decided to tolerate the problem and considered it in this study. We conducted 20 repetitive simulations for all events to consider any unspecified deviations that could arise during simulation reproduction.

3.2 Performance assessment case

We conducted a performance evaluation of the actual EEW algorithm using the OS. The accuracy of the network-based EEWs increased along with an increase in the nS (Ruhl et al., 2019; Cho et al., 2022), as shown in Figure 5. Accordingly, the OS should receive higher scores as the nS increases. To verify this finding, we examined the evaluation results for condition 1, as shown in

Figure 6. The figure shows that as the nS increased, the OS also increased, and the deviations in the 20 simulation cases declined.

We conducted performance assessment of the actual EEW algorithm by applying the assessment process developed for OS. This was intended to review the suitability of the assessment process and evaluate the algorithmic condition algorithm performance. The simulation analysis results for five algorithmic conditions and 58 earthquakes were obtained for this performance assessment. In this study, the five algorithmic conditions were termed A1–A5.

As four is the minimum nS for the KMA EEW, we compared the OS under five conditions with the evaluation results shown in Figure 7. As the proposed OS could calculate a score based on a single earthquake, the overall results are presented in a box and whisker plot to show the mean and median values. The results in Figure 7 indicate that A1–2 without the RP condition was more stable than A3–5.

We had to determine the reason for the addition of the RP condition resulting in relatively low scores; therefore, we determined the number of points each parameter earned. Figure 8 shows the average scores obtained in the parameter domain. Compared with A3–5, A1–2 exhibited a slight increase in the AM and a significant increase in the PF. We expected that using the RP condition would be effective for controlling false alarms; however, the results were not satisfactory.

In the case of A1–2 without the RP condition, we found that the analysis of A2 was more stable than that of A1. The criterion for minPa in A2 is the optimization proposed by Cho et al. (2022). Ultimately, the optimized amplitude criterion could secure EEW stability. In addition, the optimization criterion proposed in the study by Cho et al. (2022) was A5. However, in the current study, A2 was found to ensure EEW stability. This finding indicates that using minPa alone was more effective than using both conditions.

4 Limitation

To address earthquake hazards and establish strategies in various countries, EEWs are currently subjected to various research initiatives (e.g., Minson et al., 2018; Santos-Reyes, 2019; McBride et al., 2020; Ahn, 2021; Bostrom et al., 2022; Dallo et al., 2022; Sumy et al., 2022; Sutton et al., 2022). Similar strategies for earthquake responses are evolving to suit the characteristics of each country. Accordingly, the KMA needed to develop network-based EEW assessment strategies for Republic of Korea, a low-seismicity area.

The KMA evaluation method proposed in this study could potentially be adopted by other countries to assess stability during minor earthquakes. However, its applicability might vary depending on the disparities in the network-based EEW algorithms employed by each country and the perspectives of the network managers.

The EEWs are not only network based but also include other types, such as performance-based EEW (Convertito et al., 2008), OnSite EEW, and hybrid methods (Iervolino et al., 2006). The performance-based EEWs have focused on Intensity Measures (IM) and damage estimation for specific structures (Iervolino et al., 2009; Iervolino, 2011). The OnSite EEW issues an alert based on IM in a small area. The IM-based EEW could be tied

TABLE 3 Assessment target event list.

YYMMDE) (UTC)	ML	Location	YYMMDD (UTC)		ML	Location
[.] 150108	11:15	3.5	OSN	`161212	08:53	3.3	ISN
[•] 150803	01:11	3.7	OSN	`161214	08:20	3.3	ISN
ʻ151221	19:31	3.9	ISN	`170105	20:31	3.3	ISN
[•] 160106	11:39	3.0	ISN	`170331	04:46	3.3	ISN
[•] 160210	20:57	3.1	ISN	`170415	02:31	3.1	ISN
ʻ160602	19:53	3.0	ISN	`170505	14:16	3.0	ISN
ʻ160705	11:33	5.0	OSN	`171115	05:29	5.4	ISN
¹⁶⁰⁹¹²	10:44	5.1	ISN	`171115	05:32	3.6	ISN
[°] 160912	10:48	3.1	ISN	`171115	06:09	3.5	ISN
¹⁶⁰⁹¹²	11:10	3.1	ISN	`171115	07:49	4.3	ISN
[°] 160912	11:32	5.8	ISN	`171116	00:02	3.6	ISN
[°] 160912	11:34	3.6	ISN	`171119	14:45	3.5	ISN
[°] 160912	11:36	3.4	ISN	`171119	21:05	3.6	ISN
[°] 160912	11:38	3.0	ISN	`171225	07:19	3.5	ISN
[°] 160912	11:39	3.0	ISN	`180210	20:03	4.6	ISN
[°] 160912	11:40	3.0	ISN	`190108	18:10	3.7	OSN
[°] 160912	14:18	3.0	ISN	`190210	03:53	4.1	OSN
[°] 160912	14:52	3.1	ISN	`190419	02:16	4.3	OSN
[°] 160912	15:37	3.1	ISN	`190421	20:45	3.8	OSN
ʻ160912	23:24	3.2	ISN	`190627	06:19	3.9	OSN
[°] 160913	05:31	3.0	ISN	`190721	02:04	3.9	ISN
⁽ 160913	15:48	3.0	ISN	`190921	06:11	3.5	OSN
¹⁶⁰⁹¹⁹	11:33	4.5	ISN	`191027	06:37	3.4	ISN
ʻ160921	02:53	3.5	ISN	`191229	15:32	3.5	ISN
¹⁶⁰⁹²⁸	07:34	3.1	ISN	`200129	15:52	3.2	ISN
[•] 161002	11:53	3.0	ISN	`200503	13:07	3.1	ISN
[•] 161010	13:59	3.3	ISN	`200511	10:45	3.8	OSN
⁽¹⁶¹¹⁰⁵	21:26	3.5	OSN	`210419	05:20	3.7	OSN
ʻ161113	12:52	3.5	ISN	`210821	00:40	4.0	OSN

TABLE 4 Performance assessment criteria comparison.

Section		Survey results			
		ISN	OSN		
Normal operation	$\Delta L \le 100$	$\Delta \ L \leq 10$	Δ L \leq 30		
		$\Delta~M \le 0.4$	Δ M \leq 0.6		
False alert	$\Delta L > 100$	$\Delta L > 10$	$\Delta L > 30$		
		$\Delta M > 0.4$	Δ M > 0.6		
Missed alert	Non-associated event (= missing earthquake)				

TABLE 5 Assessment configurations of five conditions in early earthquake warning (EEW) algorithms.

Algorithmic condition	minPa	RangePostTrig (R	
	log (Pa)	Rv	Ra
A 1	-2.5	OFF	OFF
A 2	-2.7	OFF	OFF
A 3	Off	0.000013	0.00074
A 4	-2.5	0.000013	0.00074
A 5	-2.7	0.000013	0.00074



FIGURE 5

Differences between simulation results performed with the same system environment parameters of a magnitude 3.4 earthquake on 27 October 2019, in the Korean Peninsula.











Mean score of four parameters for algorithmic condition at number of stations (nS) = 4. PM, prevent missed alerts; PF, prevent false alerts; AM, accuracy of magnitude; AL, accuracy of location.

to a strategic alerting strategy to consider damage. We believe that developing a damage-based strategic evaluation technique holds practical significance (Iervolino et al., 2011; Papadopoulos et al., 2023). The KMA is also engaged in the development and experimentation of on-site EEW systems based on IM (Ahn et al., 2023). However, these methods differ from network-based EEWs.

The current study concentrated on alerts disseminated in the public domain. Given the widespread use of smartphones, CBS stands out as the most efficient means of conveying alerts to individuals. However, to guarantee the reliability of alerts in the public sphere, this study defined successful, false, and missed alarms from an administrator standpoint. However, we note that this viewpoint could contrast with the that of the public receiving the service.

5 Conclusion

We developed an assessment for EEWs based on management aspects. While promptness is an integral factor in EEW assessment, we considered that reducing warning time based on nS could result in less accurate information. Therefore, the proposed performance assessment could comprehensively review EEW services from multiple aspects.

Surveys involving experts were conducted to establish standards for assessment criteria consistent with KMA EEW operational management. The proposed weights sufficiently addressed the preferences of operators with EEW technical system knowledge and the managers of the systems. These experts selected operational impact factors according to their operational principles, with most operators expressing a desire to minimize EEW false alerts.

The comprehensive review was scored by assigning weights to each element based on the AHP analysis. Accordingly, a quantitative assessment method was developed that could replace the conventional method, which only considers warnings. The established assessment equation reflected performance improvement based on accuracy. Performance assessments can be conducted with changes in algorithms or parameters.

Our method could be appropriate for regions that are new to EEW in low-seismicity areas. However, our study is limited in that it only surveyed operators who were familiar with EEW technology. They responded with answers characterized by 1) imagining enduser satisfaction, 2) considering only technical aspects, and 3) summarizing all perspectives. Therefore, we note that the criteria could change depending on the survey responses, as individuals have various perspectives, backgrounds, and types of thinking.

The proposed EEW system assessment offers a quantitative approach that takes into account operational aspects. This method would be also expected the quantitative values of factors during actual operational periods, such as delays and anomalies arising from issues at the observation stations. Given the range of potential scenarios, it is imperative that this method undergoes further refinements for future operations.

Data availability statement

The original contributions presented in the study are included in the article, further inquiries can be directed to the corresponding author.

Author contributions

J-KA: Conceptualization, Data curation, Formal Analysis, Methodology, Supervision, Writing-original draft, Writing-review and editing. SC: Formal Analysis, Visualization, Writing-original draft. E-HH: Funding acquisition, Project administration, Supervision, Writing-original draft. W-HB:

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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