



Developments of the Nationwide Earthquake Early Warning System in Japan After the 2011 *M*_w9.0 Tohoku-Oki Earthquake

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OPEN ACCESS

Edited by:

Katsuichiro Goda, Western University , Canada

Reviewed by:

Jindong Song, Institute of Engineering Mechanics, China Earthquake Administration, China Gemma Cremen, University College London, United Kingdom

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Specialty section:

This article was submitted to Solid Earth Geophysics, a section of the journal Frontiers in Earth Science

Received: 16 June 2021 Accepted: 17 September 2021 Published: 04 October 2021

Citation:

Kodera Y, Hayashimoto N, Tamaribuchi K, Noguchi K, Moriwaki K, Takahashi R, Morimoto M, Okamoto K and Hoshiba M (2021) Developments of the Nationwide Earthquake Early Warning System in Japan After the 2011 M_w9.0 Tohoku-Oki Earthquake. Front. Earth Sci. 9:726045. doi: 10.3389/feart.2021.726045 In Japan, the nationwide earthquake early warning (EEW) system has been being operated by the Japan Meteorological Agency (JMA) since 2007, disseminating information on imminent strong ground motion to the general public and advanced technical users. In the beginning of the operation, the system ran based mainly on standard source-based algorithms with a pointsource location estimate and ground motion prediction equation. The point-source algorithms successfully provided ground motion predictions with high accuracy during the initial operation; however, the 2011 M_{W} 9.0 Tohoku-Oki earthquake and the subsequent intense aftershock and triggered earthquake activities underscored the weaknesses of the source-based approach. In this paper, we summarize major system developments after the Tohoku-Oki event to overcome the limits of the standard point-source algorithms and to enhance the EEW performance further. In addition, we evaluate how the system performance was influenced by the updates. One of significant improvements in the JMA EEW system was the implementation of two new ground motion prediction methods: the integrated particle filter (IPF) and propagation of local undamped motion (PLUM) algorithms. IPF is a robust point-source algorithm based on the Bayesian inference, and PLUM is a wavefield-based algorithm that predicts ground motions directly from observed shakings. Another notable update was the incorporation of new observation facilities including S-net, a large-scale ocean bottom seismometer network deployed along the Japan and Kuril trenches. The prediction accuracy and warning issuance performance analysis for the updated JMA EEW system showed that IPF improved the source-based ground motion prediction accuracy and reduced the risk of issuing overpredicted warnings. PLUM made the system less likely to underpredict strong ground motions and improved the warning issuance timeliness. The detection time analysis for the S-net incorporation suggested that S-net enabled the system to issue the first EEW report earlier than before the S-net incorporation for earthquakes around the Japan and Kuril trenches. Those findings indicate that the JMA EEW system has made substantial progress both on software and hardware aspects over the 10 years after the Tohoku-Oki earthquake.

Keywords: earthquake early warning, ground motion prediction, real-time analysis, hypocenter determination, ocean bottom seismometer, 2011 Tohoku-Oki earthquake

INTRODUCTION

It has been 10 years since the 2011 off the Pacific coast of Tohoku earthquake (Tohoku-Oki earthquake), a M_w 9.0 megathrust event in the Japan Trench, occurred on March 11, 2011. The resulting strong ground shaking and large tsunami caused severe damage in a large part of eastern Japan. The Tohoku-Oki earthquake and subsequent intense seismic activity also had shed light on technical limitations of the Japanese nationwide earthquake early warning (EEW) system, which afterwards led to further developments of the system.

To mitigate earthquake damage, EEW systems have been developed and/or operated in many earthquake-prone regions around the world such as Japan (e.g., Nakamura, 1988; Hoshiba et al., 2008), Mexico (Espinosa-Aranda et al., 2009), the west coast of the United States (Böse et al., 2013; Kilb et al., 2021), Taiwan (Chen et al., 2015), Europe (Clinton et al., 2016), China (Zhang et al., 2016), Korea (Sheen et al., 2017), and Israel (Nof and Kurzon, 2021). In Japan, the Japan Meteorological Agency (JMA) has been operating the nationwide EEW system since 2007 to disseminate information on imminent strong ground motion to the general public and advanced technical users (Hoshiba et al., 2008).

In the beginning of the operation, the system ran based mainly on standard source-based algorithms that used a point-source model (PSM) estimate and ground motion prediction equation (GMPE) (Kamigaichi, 2004). The PSM algorithms successfully provided ground motion predictions with high accuracy during the initial operation (Doi, 2011; JMA, 2014). However, the prediction performance was poor for the Tohoku-Oki event and subsequent intense earthquake sequences in 2011 (Hoshiba et al., 2011; Kodera et al., 2018). For the mainshock, the system successfully issued a public warning for the Tohoku region (the nearest region from the epicenter) before the S-wave hit; on the other hand, the system underpredicted ground motions for the Kanto region, ~300 km to the southwest of the epicenter, due to the fault finiteness (i.e., the PSM algorithms were unable to capture the spatial extent of the fault rupture). Just after the mainshock, the system missed several large aftershocks; the P-wave picking algorithm did not work because the P-wave onsets overlapped with large coda waves of the mainshock or previous aftershocks. In addition, the system overpredicted ground motions for multiple simultaneous earthquakes, failing to associate P-wave travel times and mislocating the hypocenters.

Over the past 10 years, JMA has taken various measures to overcome those technical limitations and to enhance the system performance further. One of significant updates of the JMA EEW system was the implementation of two novel approaches: the integrated particle filter (IPF) and propagation of local undamped motion (PLUM) algorithms (Tamaribuchi et al., 2014; Kodera et al., 2018). Another notable development was the incorporation of new observation facilities. In this article, we report the system improvements on prediction algorithms and seismic networks from 2011 to 2020 and investigate how those updates influenced the system performance using actual EEW reports issued from April 2016 to December 2020.

The Initial Japan Meteorological Agency Earthquake Early Warning System

JMA launched EEW service for the general public in October 2007 (Hoshiba et al., 2008). The initial EEW system was operated by incorporating two different seismic networks (Kamigaichi et al., 2009; Doi, 2011), versatile seismometers of JMA (Harada, 2007; ~220 accelerometers) and Hi-net of the National Research Institute for Earth Science and Disaster Resilience (NIED; Okada et al., 2004; Aoi et al., 2020; ~800 high-sensitivity velocity meters) (Figure 1A). Versatile means that the seismometers can calculate and transmit various observation quantities related to EEW such as P-wave arrival, displacement amplitude, and epicentral distance given by the B-A algorithm (Odaka et al., 2003; Tsukada et al., 2004; Harada, 2007). The system provided ground motion predictions based on the PSM estimation. Source parameter estimates were given by several algorithms running in parallel. The $B-\Delta$, territory, and grid-search algorithms (Kamigaichi, 2004) were employed as the main source estimation methods. Real-time data from the JMA versatile seismometers were fed into these algorithms. The not-yet-arrived-data (NYAD) algorithm (Horiuchi et al., 2005), which provided source parameters from the Hi-net stations, was also implemented as an independent subsystem. In addition, the system was receiving source estimates from an external subsystem based on a classical hypocenter determination approach using automatic P- and S-wave picks at the JMA and Hi-net stations (we refer to this algorithm as Hypoon). The system received hypocenter estimates from NYAD and Hypoon only if the location errors were smaller than rejection thresholds. Source estimates given by different algorithms were combined into a single event in the event identification process if the source parameters were similar to each other. After that, ground motion predictions were calculated from the estimated source parameters and GMPE. The GMPE employed in the JMA EEW system was one proposed by Si and Midorikawa (1999), which determines a peak ground velocity (PGV) from a point-source location, magnitude, and hypocentral distance. The PGV was converted into a JMA seismic intensity (JMA, 1996) with the empirical equation proposed by Matsuoka and Midorikawa (1994). Finally, the system issued EEW reports including predicted JMA intensities and estimated source parameters.

In the JMA EEW system, JMA seismic intensity is used as the ground motion metric. The JMA intensity is represented in two different ways: the 10-degree discrete representation I_{JMA} (i.e., $I_{JMA} = 0, 1, 2, 3, 4$, five lower (5L), five upper (5U), 6L, 6U, 7) and the continuous representation I_{inst} (e.g., $I_{inst} = 2.5, 4.8$). I_{inst} can be converted into I_{JMA} by rounding off I_{inst} to the nearest I_{JMA} value (e.g., $2.5 \le I_{inst} < 3.5$ corresponds to $I_{JMA} = 3$, and $4.5 \le I_{inst} < 5.0$ is equivalent to $I_{JMA} = 5L$). The JMA EEW system issues a public warning if the maximum predicted intensity is 5L or more on I_{JMA} (4.5 or more on I_{inst} ; Hoshiba et al., 2008). $I_{JMA} = 5L$ corresponds to the intensity threshold at or above which severe earthquake damage is likely to occur.

JMA evaluated the system prediction accuracy by calculating a prediction score, defined as the ratio of sub-prefectural areas



whose intensity prediction error is within one unit on I_{JMA} among all areas with predicted or observed $I_{JMA} \ge 4$ (Doi, 2011). From October 2007 to March 2010, the system successfully provided accurate ground motion predictions, and the prediction score was as high as ~80%. However, the prediction score decreased to 28% in Japanese fiscal year 2010 (from April 2010 to March 2011) due to the occurrence of the Tohoku-Oki earthquake (JMA, 2014).

The Japan Meteorological Agency Earthquake Early Warning System After the 2011 Tohoku-Oki Earthquake

Over the 10 years after the Tohoku-Oki earthquake, JMA has upgraded the EEW system by introducing new ground motion prediction algorithms and seismic observation facilities (Figure 1B).

New Ground Motion Prediction Algorithms The Integrated Particle Filter Algorithm

IPF (Tamaribuchi et al., 2014) is a PSM algorithm that has been employed in the JMA EEW system since December 14, 2016 (JMA, 2016) to reduce the risk of overprediction due to the misassociation of P-wave picks for multiple simultaneous earthquakes. IPF estimates source parameters based on the Bayesian inference (Liu and Yamada, 2014; Wu et al., 2015); hypocenter locations are determined using not only P-wave travel times but also various observed quantities such as velocity amplitude, untriggered station distribution, and epicentral distance estimated from $B-\Delta$. For the association of P-wave picking data, IPF also takes into account velocity amplitudes in addition to travel time differences, to discriminate the occurrence of multiple simultaneous earthquakes from a single large event. Retrospective simulations with the intense earthquake sequences caused by the 2011 Tohoku-Oki and 2016 Kumamoto earthquakes showed that IPF could reduce the substantial number of overpredicted warnings compared to the PSM algorithms in the initial JMA EEW system (Tamaribuchi et al., 2014; Kodera et al., 2016).

In the updated EEW system, IPF is implemented as the main PSM algorithm, replacing the $B-\Delta$, territory, and grid-search algorithms. The NYAD and Hypoon algorithms are still working in the updated system; JMA has decided to leave the two algorithms because the two algorithms use denser seismic networks than IPF (the difference comes from whether Hi-net is incorporated or not) and therefore could sometimes provide source parameter estimates timelier than IPF.

IPF is being operated with JMA versatile seismometers. Additionally, IPF uses three seismic networks maintained by NIED: DONET, S-net, and a part of KiK-net (Kaneda et al., 2015; Kawaguchi et al., 2015; Kanazawa et al., 2016; Uehira et al., 2016; Mochizuki et al., 2016; Okada et al., 2004; Aoi et al., 2020; **Figures 1B**, **2A**,**B**; as of December 2020).



The Propagation of Local Undamped Motion Algorithm

PLUM (Kodera et al., 2018) is a wavefield-based algorithm (Hoshiba, 2013; Hoshiba and Aoki, 2015) that has been implemented since March 22, 2018 (JMA, 2018) to provide accurate ground motions for large earthquakes with nonnegligible finite faults and for intense earthquake sequences for which PSM algorithms could fail to estimate correct source parameters. PLUM provides ground motion predictions without assuming specific source models; instead, PLUM predicts future ground motions directly from ground shakings observed near the target sites, assuming unattenuated plane wave incident. In the JMA EEW system, a predicted intensity at target site k is given by $I_{pred}^{(k)} = \max\{Ir_{obs}^{(1)} - F_0^{(1)}, \cdots, Ir_{obs}^{(N)} - F_0^{(N)}\} + F_0^{(k)}$, where $Ir_{obs}^{(i)}$ ($i = 1, \dots, N$) are observed real-time seismic intensities (Kunugi et al., 2013) at seismometers located within 30 km from target site k. $F_0^{(i)}$ and $F_0^{(k)}$ are scalar site amplification factors at

individual sites (Iwakiri et al., 2011) converted into equivalent seismic intensity differences. Kodera et al. (2018) showed that the JMA EEW system could predict ground motions without underprediction for the Tohoku-Oki mainshock if PLUM was implemented. In addition, other offline simulations indicated that PLUM reduced the number of missed earthquakes for intense earthquake sequences during the 2011 Tohoku-Oki and 2016 Kumamoto events (Kodera et al., 2016; 2018). In the first year after the PLUM implementation, the JMA EEW system issued warnings with a better detection rate, especially for earthquakes whose observed ground motions were near the warning threshold (Kodera et al., 2020).

As of December 2020, PLUM is being operated with two JMA seismic networks, versatile seismometers and intensity meters (Figure 2C).

The Hybrid Algorithm

In the updated EEW system, two different ground motion predictions are given by the PSM and PLUM algorithms. The

two ground motion predictions are combined in the groundmotion-based event identification process to obtain the final ground motion prediction result (**Figure 1B**). The event identification process assumes that the two ground motion predictions are from the same event if there are one or more overlapped sub-prefectural areas with predicted intensities of 3 or more. The final ground motion predictions are given by taking the maximum of predicted intensities for each area. We refer to this procedure as the hybrid algorithm (Kodera et al., 2018).

New Seismic Observation Facilities

After the Tohoku-Oki earthquake, in addition to the introduction of the new ground motion prediction algorithms, JMA has incorporated new inland and offshore seismic observation facilities for more robust and timelier EEW issuances (**Figure 2**).

Inland Networks

The initial JMA EEW system was being operated with ~220 JMA versatile seismometers and ~800 Hi-net stations (Figure 1A). On March 31, 2015, JMA installed 50 new versatile seismometers mainly on the Pacific side to enhance the detection capability (JMA, 2015; Figure 1B). Deep borehole seismometers of KiK-net (15 stations in the southern Kanto region) were also added for the main PSM algorithms to obtain additional lead times for earthquakes in the Kanto region (JMA, 2015; Figures 1B, 2B). On March 22, 2018, the start date for the PLUM operation, the system incorporated JMA intensity meters (seismometers that can transmit seismic intensities only) to perform the PLUM algorithm with a denser seismic network (Kodera et al., 2018). As of December 2020, real-time seismic intensities from ~660 JMA intensity meters are fed into PLUM (Figures 1B, 2C). The Hi-net high-sensitivity velocity meters (Figure 2D) are used only for the NYAD and Hypoon algorithms.

Offshore Networks

To enhance the detection capability for offshore earthquakes, JMA also incorporated two ocean-bottom seismometer (OBS) networks maintained by NIED (**Figures 1B, 2B**). One of the OBS networks is DONET (Kaneda et al., 2015; Kawaguchi et al., 2015; Aoi et al., 2020), which consists of 22 OBSs across the Kumano-Nada (DONET1) and 29 OBSs off the Kii channel (DONET2). JMA incorporated the DONET stations partially on March 31, 2015 (JMA, 2015) and fully on June 27, 2019 (JMA, 2019b; **Figure 1B**). The other network is S-net (Kanazawa et al., 2016; Mochizuki et al., 2016; Uehira et al., 2016; Aoi et al., 2020), 150 OBSs deployed along the Japan and Kuril trenches. 125 and 25 OBSs of S-net were introduced in the system on June 27, 2019 (JMA, 2019b) and on March 24, 2020 (JMA, 2020a), respectively (**Figure 1B**).

For robust magnitude estimates with OBSs, a new magnitude estimation algorithm specialized for OBSs was developed and introduced (Hayashimoto et al., 2019; Hayashimoto et al., 2021; submitted to Quarterly Journal of Seismology). For inland seismometers, the JMA EEW system calculates magnitudes from the vector sum of three-component displacements. However, for OBSs, three-component displacements could overestimate magnitudes because of horizontal site amplification thick sedimentary caused by lavers (Hayashimoto and Hoshiba, 2013; Nakamura et al., 2014) and acceleration offset signal contamination by device orientation changes when strong motions hit (Hayashimoto et al., 2016; Nakamura and Hayashimoto, 2019; Takagi et al., 2019). Hayashimoto et al. (2019) found that those overestimation effects could be mitigated by using vertical-component instead of three-component displacements. JMA therefore implemented a new magnitude estimation equation based solely on verticalcomponent displacements for OBSs. In addition, to eliminate possible amplitude anomalies at OBSs near hypocenters, the new magnitude estimation algorithm requires three or more OBSs to calculate magnitudes (i.e., magnitude estimates are unavailable until the system has received displacement amplitudes from three or more OBSs; Havashimoto et al., 2021, submitted).

Performance Evaluation for the Updated Japan Meteorological Agency Earthquake Early Warning System

To evaluate the performance of the updated JMA EEW system, we assessed how the IPF and PLUM implementation influenced the prediction accuracy and warning issuance performance. Also, we analyzed how the S-net incorporation contributed to the detection timeliness for earthquakes around the Japan and Kuril trenches. A part of the prediction accuracy and warning issuance performance analysis in this study is the same as what Kodera et al. (2020) conducted but with an extended analysis period.

Prediction Accuracy Changes With the Integrated Particle Filter and Propagation of Local Undamped Motion Implementation

We investigated how the prediction accuracy of the JMA EEW system changed with the IPF and PLUM implementation, using the maximum and final ground motion prediction results. We focused on the maximum ground motion predictions for the accuracy evaluation because the system issues warnings once predicted ground motions exceed the warning threshold. The final ground motion predictions were also used because their prediction errors would indicate the upper limit of prediction accuracy for target algorithms. For PLUM, the final ground motion predictions are the same as the maximum ground motion predictions. In this analysis, the prediction accuracy was evaluated for three different periods: 1) April 1, 2016 to December 14, 2016, 2) December 14, 2016 to March 22, 2018, and 3) March 22, 2018 to December 31, 2020. December 14, 2016 and March 22, 2018 are the start dates for the IPF and PLUM operation, respectively. To evaluate the system performance before the IPF and PLUM implementation, we took the analysis period from April 1, 2016. We focused only on recent EEW reports for the evaluation of the previous PSM algorithms to exclude the possible performance change due to different software versions. The analysis period was relatively short compared to the entire operation period of the previous PSM algorithms but was enough long to capture the algorithms' characteristics because during the period the system processed



FIGURE 3 | Comparisons of observed and predicted intensities using the maximum ground motion prediction results for the three periods: (A) April 1, 2016 to December 14, 2016 (the start date for the IPF operation), (B) December 14, 2016 to March 22, 2018 (the start date for the PLUM operation), and (C) March 22, 2018 to December 31, 2020. Intensities were plotted on an area-by-area basis; that is, each intensity was obtained by taking the maximum among those at all sites in a sub-prefectural area. Circles and bars in magenta indicate the mean and standard deviation of predicted intensities for each observed I_{JMA} (data points with a predicted intensity equal to 0 were eliminated for the calculation). Green zone is one in which prediction errors were within one I_{JMA} unit (we assumed that those errors were acceptable). Orange and blue zones are ones in which ground motions were overpredicted and underpredicted with intensity errors of two or more units, respectively. Data points with predicted and observed $I_{JMA} \leq 3$ (i.e., $I_{inst} < 3.5$) are in the gray zone. The prediction score (PS) was calculated, defined as the ratio of areas whose intensity prediction error was within one I_{JMA} unit among areas with an observed or predicted intensity of four or more (data points in the gray zone were eliminated for the calculation). For (C), comparison plots for the PLUM and hybrid algorithms are shown in addition to that for PSM because the period was after the PLUM operation.

many earthquakes including the 2016 Kumamoto earthquake sequence.

For each period, we calculated the prediction score. For the score calculation, predicted intensities were compared to observed intensities on an area-by-area basis (i.e., each intensity was obtained by taking the maximum of predicted or observed intensities in a sub-prefectural area; for the definition of the sub-prefectural areas, see **Figure 1C** in Kodera et al. (2020)). The prediction score was defined as the ratio of areas in which the difference between predicted and observed I_{JMA} was equal to or less than one I_{JMA} unit among areas with predicted or observed $I_{JMA} \ge 4$ (Doi, 2011). A predicted/observed I_{inst} was set to 0.0 if there was no available predicted/observed intensity for a target area. Intensities predicted by the PSM algorithms were obtained from hypocenter estimates using two or more stations. Intensities predicted by PLUM were what satisfied the EEW issuance condition of PLUM (i.e., predicted $I_{inst} \ge 2.5$; Kodera et al.,

2020). The prediction accuracy analysis in this study was done with earthquakes for which the JMA EEW system issued EEW reports.

During the first period, which is before the IPF and PLUM operation, the prediction score was 69.7% with the maximum ground motion predictions (**Figure 3A**) and 86.1% with the final predictions (**Figure 4A**). The prediction score with the maximum predictions was low mainly due to overprediction for events of the 2016 Kumamoto earthquake sequence (Kodera et al., 2016). The system still had difficulty in distinguishing multiple simultaneous events from a large earthquake because IPF had not been implemented yet. The system overpredicted ground motions for several concurrent aftershocks, mislocating the hypocenters. The prediction score with the final predictions was higher than that with the maximum predictions, although the prediction accuracy was still affected by the overprediction for the Kumamoto earthquake sequence. The underprediction case



with predicted $I_{JMA} = 2$ and observed $I_{JMA} = 6$ L in **Figure 4A** was due to duplicated event declarations; the system declared two events, one of which had a poor source location estimate because the location was determined only with seismometers far from the actual epicenter.

For the second period, just after the IPF implementation, the prediction score was 83.6% with the maximum ground motion predictions and 92.5% with the final predictions, higher than those during the first period (Figures 3B, 4B). IPF did not provide mislocated hypocenters for multiple simultaneous events, and therefore the prediction accuracy was improved compared to that before the IPF implementation. However, there was still an overprediction case in which ground motions were overpredicted with errors of $\geq 2 I_{IMA}$ units. The overprediction was caused by M4.5 and M4.0 earthquakes in January 2018 that occurred simultaneously, ~400 km away from each other (JMA, 2019a). The magnitude of the M4.5 earthquake was overestimated because the system calculated the magnitude with displacement amplitudes of the M4.0 earthquake. Although the IPF and NYAD algorithms provided correct hypocenter location estimates for the two events, the subsequent event identification process associated the two earthquakes incorrectly. In 2019, JMA updated the event identification and

magnitude calculation criteria to address the overprediction issue (JMA, 2019a). Without ground motion predictions for the M4.5 and M4.0 earthquakes, the prediction score would have been 89.8% with the maximum predictions and 97.4% with the final predictions (**Supplementary Figures S1B, S2B**).

During the third period, after the PLUM implementation, the prediction score of PSM was 81.5 and 88.5% with the maximum and final ground motion predictions, respectively (Figures 3C, 4C). As during the second period, IPF provided accurate source parameter estimates without hypocenter mislocation for multiple simultaneous earthquakes. However, an overprediction issue occurred again on July 30, 2020 (JMA, 2020b). For a M6.0 earthquake that occurred near the Torishima island (the epicenter was ~500 km southward of the Japan mainland), the system issued an overpredicted warning, estimating a M7.3 earthquake whose location was ~440 km northward of the actual epicenter. The overprediction was significant because no felt ground shaking was observed although EEW users received the warning. The Hypoon algorithm provided a mislocated hypocenter estimate, and the system used the source estimate because the location estimation errors (calculated by Hypoon) were lower than the rejection criteria. The magnitude was calculated with a displacement amplitude observed at a seismometer in the Hahajima island, ~800 km southward of the estimated source location. After the overprediction issue, as a tentative measure, JMA modified the magnitude calculation logic, imposing the condition that seismometers used for the magnitude estimation must be located within 700 km from the estimated epicenter. Without the M6.0 near-Torishima event, the prediction score of the PSM algorithms would have been 85.5 and 94.3% with the maximum and final ground motion predictions, respectively (**Supplementary Figures S1C, S2C**). For the maximum ground motion predictions, there were several overprediction cases in which $I_{JMA} = 4$ was predicted but $I_{JMA} \leq 2$ was observed. Those overpredictions occurred mainly because of magnitude overestimation in an early stage with a small number of seismometers and unstable hypocenter location estimates using seismometers at the network boundary.

During the second and third periods, the PSM algorithms tended to underpredict ground motions especially for strong shakings of $I_{JMA} \ge 5L$ in the final EEW reports, with prediction errors of one to three I_{JMA} units (Figures 4B,C). Most of those strong shakings were caused by large inland earthquakes and were observed near the epicenters. The PSM algorithms underpredicted the strong ground motions because of GMPE errors and/or minor magnitude underestimation, although accurate source locations were estimated. The underprediction case with predicted $I_{IMA} = 2$ and observed $I_{IMA} = 5L$ during the third period was due to a hypocenter estimate with a large location error caused by duplicated event declarations (two events were declared for this earthquake, and the PSM algorithms provided an accurate source location estimate in the other declared event; therefore, EEW users also received more accurate EEW reports for this earthquake).

During the third period, the PLUM prediction score was 84.6% (Figures 3C, 4C; the maximum and final ground motion predictions are the same for PLUM). The prediction score for PLUM was influenced by prediction errors for inland earthquakes. The assumption of unattenuated wave incidence did not hold for sites near epicenters of shallow inland earthquakes; PLUM therefore overpredicted ground motions for several inland events. In addition, PLUM missed ground motions for three inland earthquakes with observed $I_{IMA} \ge 5L$. This was because the strong motions were so localized that seismometers used for PLUM did not observe $I_{IMA} \ge 3$, the EEW issuance threshold of PLUM. On the other hand, PLUM did not cause significant overpredictions, compared to PSM; the maximum overprediction error of PLUM was 3 on the I_{IMA} units (i.e., predicted $I_{IMA} = 5U$ but observed I_{IMA} = 3; the intensity difference between 5L and 5U is counted as one I_{JMA} unit although the interval is 0.5 I_{inst}) while that of PSM was 6 (predicted $I_{IMA} = 5U$ but observed $I_{IMA} = 0$) with the final ground motion predictions during the third period (Figure 4C). In addition, PLUM predicted strong shakings caused by large inland earthquakes with smaller underprediction errors than PSM.

The prediction score for the hybrid algorithm was 81.0 and 87.3% with the maximum and final predictions, respectively (**Figures 3C, 4C**). Without the M6.0 near-Torishima earthquake, the score would be 84.7 and 92.2% (**Supplementary Figures S1C, S2C**). The prediction score for the hybrid algorithm

was similar to but slightly smaller than that for PSM, influenced by PLUM ground motion predictions. This indicates that the hybrid algorithm inherited the characteristics of both PSM and PLUM.

Warning Issuance Performance Changes With the Integrated Particle Filter and Propagation of Local Undamped Motion Implementation

We also evaluated how the warning issuance performance was changed by the IPF and PLUM implementation. We investigated which earthquakes with $I_{JMA} \ge 1$ ($I_{inst} \ge 0.5$; felt earthquakes) were warned/missed by the JMA EEW system, taking the same analysis period as in *Prediction Accuracy Changes With the Integrated Particle Filter and Propagation of Local Undamped Motion Implementation*. In addition, the warning timeliness was assessed by calculating warning issuance time differences between PSM and PLUM. From April 2016 to December 2020, there was no overprediction case in which the system issued a warning for an earthquake with observed $I_{JMA} = 0$, except for the M6.0 near-Torishima earthquake in July 2020.

During the first period, the JMA EEW system issued warnings for large earthquakes with a high detection rate (**Figure 5A**). Especially, warnings were issued for all earthquakes with observed $I_{JMA} \ge 5$ U, except for two inland earthquakes of the 2016 Kumamoto earthquake sequence. However, the system issued overpredicted warnings for M2–3-class (i.e., $2.0 \le M <$ 4.0) earthquakes of the Kumamoto earthquake sequence with observed $I_{JMA} \le 3$, providing mislocated hypocenter estimates.

During the second period, the system issued warnings with a similar detection rate to that during the first period for earthquakes with observed $I_{JMA} \ge 5L$ (Figure 5B). An overpredicted warning was issued for the simultaneous occurrence of the M4.5 and M4.0 earthquakes in January 2018, which was not due to hypocenter mislocation but due to incorrect event association in the event identification process (in Figure 5B; only the M4.5 earthquake is plotted). IPF did not cause overpredicted warnings for earthquakes with observed $I_{JMA} \le 3$.

During the third period, PSM and PLUM detected warning events with a low missing rate for observed $I_{IMA} \geq 5U$ (Figure 5C). For earthquakes with $4.5 \leq I_{inst} < 5.0$ (earthquakes near the warning threshold), PLUM had a higher detection rate than PSM. This was because PLUM was less likely to underpredict strong motions for inland earthquakes than PSM (see Prediction Accuracy Changes With the Integrated Particle Filter and Propagation of Local Undamped Motion Implementation). There were two earthquakes with observed $I_{IMA} = 3$ for which the system issued overpredicted warnings, owing to source parameter estimates by IPF. Those were due to magnitude overestimation in an early stage and unstable hypocenter location estimates at the seismometer network boundary. Although the system also issued overpredicted warnings after the IPF implementation, the overpredicted earthquakes with observed $I_{JMA} \leq 3$ had larger magnitudes (M5-6-class; 5.0 \leq M < 7.0) than those during the first period (M2-3-class).



To assess the warning timeliness, we evaluated warning issuance time differences between PSM and PLUM, using the 13 earthquakes warned by both of PSM and PLUM during the third period (**Figure 5D**). For five out of the 13 earthquakes, the warning issuance times of PLUM were earlier than those of PSM. The median time difference was \sim 3.4 s. The warning timeliness was improved for those events because PLUM was less likely to underpredict strong ground motions than PSM, and PLUM used the denser JMA seismic network than IPF.

Detection Capability Changes With the S-Net Incorporation

Focusing on the S-net incorporation, we investigated how the new observation facilities changed the detection capability of the JMA EEW system. The timeliness of the first EEW report was evaluated for earthquakes around the Japan and Kuril trenches, where S-net has been deployed. Here, we use the term "detection time" as the time when the system issued the first EEW report.

The target area for the detection time analysis is shown in **Figure 6A**. For earthquakes inside the area, we calculated the cumulative distribution functions (CDFs) of detection times before and after June 27, 2019, the start date for the S-net

incorporation (**Figure 6B**). The CDF after the S-net incorporation took smaller values for all detection times than the CDF before the incorporation, indicating that S-net successfully improved the detection timeliness for earthquakes around the Japan and Kuril trenches. Especially, after the S-net incorporation, the system issued the first EEW report within 21.78 and 30.14 s from the origin time for 50 and 75% of earthquakes inside the area, respectively, which were 3.90 and 8.83 s earlier than before the incorporation.

In addition, we also calculated the CDFs for earthquakes outside the target area (**Figure 6C**). The CDF values did not change drastically before and after the S-net introduction. The detection time difference was less than ~ 2.0 s, implying that the detection time variability due to temporal seismicity change was within ~ 2.0 s. The detection time reduction for earthquakes inside the area was larger than ~ 2.0 s; therefore, the timeliness improvement by S-net was significant.

DISCUSSION

IPF has been implemented to overcome the overprediction issue caused by the mislocation of multiple simultaneous earthquakes. Before the IPF operation, the prediction score for the PSM



algorithms was 69.7% with the maximum ground motion predictions; after the IPF implementation, the prediction score increased to more than 80% (83.6% from December 14, 2016 to March 22, 2018 and 81.6% from March 22, 2018 to December 31, 2020). After the IPF operation, the PSM algorithms caused significant ground motion overpredictions for the M4.5 and M4.0 earthquakes in January 2018 and for the M6.0 near-Torishima earthquake in July 2020, which were not due to IPF but due to the event identification process and the Hypoon algorithm. Without the two overprediction cases, the prediction score would have been 89.8% from December 14, 2016 to March 22, 2018 and 85.5% from March 22, 2018 to December 31, 2020. Those results indicate that IPF provided source parameter estimates with a higher accuracy than the previous PSM algorithms. The warning performance analysis showed that, before the IPF implementation, the system issued overpredicted warnings for M2-3-class earthquakes because of mislocated source estimates provided by the previous PSM algorithms. On the other hand, IPF did not cause overpredicted warnings for earthquakes with observed $I_{JMA} \leq$ 3, except for two M5-6-class events. There were several earthquakes with observed $I_{IMA} = 4$ for which IPF satisfied the warning condition; however, their magnitudes were M4class or more. These imply that IPF reduced the risk of issuing overpredicted warnings with small (M2-3-class) earthquakes.

The prediction score for PSM using the final EEW reports was ~90% after the IPF operation; however, even with the final ground motion predictions, the PSM algorithms were likely to underpredict strong shakings caused by large inland earthquakes, because of GMPE errors and/or minor magnitude underestimation. Especially, the underprediction due to GMPE errors underscored the prediction accuracy limitation of the source-based approach; that is, correct hypocenter location

and magnitude estimates do not always provide accurate ground motions because of GMPE errors (Hoshiba et al., 2010).

After the IPF operation, significant ground motion overpredictions occurred for the M4.5 and M4.0 earthquakes in January 2018 and for the M6.0 near-Torishima earthquake in July 2020. The first case was due to incorrect event association in the event identification process. The other case was because of a mislocated hypocenter estimate from the Hypoon algorithm. Those overprediction cases indicate that there are still technical challenges that need to be addressed to attain more robust source-based ground motion predictions. A possible solution for the technical issues is to skip the event identification process by integrating the different PSM algorithms into a single algorithm. The integration may be possible by feeding all available seismic data into IPF, although the current IPF algorithm does not incorporate Hinet high-sensitivity velocity meters. Several previous studies suggested that the Hi-net velocity sensors could be used for a PSM algorithm in combination with accelerometers if the instrumental response was corrected (Yamada et al., 2014; Noguchi et al., 2020; Yamada et al., 2021).

PLUM has been introduced to provide robust ground motion predictions for complex earthquake scenarios such as a large earthquake with a finite fault and an intense earthquake sequence. The prediction score for PLUM was 84.6%, which was affected by prediction errors for shallow inland earthquakes. On the other hand, PLUM predicted strong ground motions caused by large inland earthquakes with smaller underprediction errors than PSM. The warning performance analysis showed that, for earthquakes near the warning threshold (i.e., the peak observed intensity was $4.5 \leq I_{inst} < 5.0$), PLUM detected the strong motions and satisfied the warning condition with a lower missing rate than PSM. In addition, there were five events for which the warning issuance time of PLUM was earlier than that of

PSM. Those results indicate that PLUM attains the robustness and timeliness of the ground motion prediction for strong shakings, at the expense of the prediction accuracy for shallow inland events (robustness here means that the capability of predicting ground motions without missing or underprediction). The PLUM's characteristics discussed here were already mentioned in Kodera et al. (2020) but were validated with a longer analysis period in this study. Reducing prediction errors for inland earthquakes is one of technical issues for PLUM. A more complicated wave propagation model may be required to improve the PLUM prediction performance for inland earthquakes. Another technical challenge for PLUM is that lead times given by PLUM are limited because ground motions are predicted based on direct observations at nearby seismometers. Introducing P-wave information and/or an attenuating wave propagation model may lengthen available lead times (Kodera, 2018; 2019).

Owing to the PLUM implementation, the JMA EEW system attained robust and timely ground motion predictions for earthquakes with high observed intensities. At the same time, the system had a higher possibility of overprediction for shallow inland earthquakes than before. Although the PLUM implementation caused an adverse effect on the system prediction accuracy, the prediction accuracy analysis for the hybrid algorithm indicates that the adverse impact was not very large. The prediction score for the hybrid algorithm was 81.0% with the maximum ground motion predictions and 87.3% with the final predictions. Those percentages were lower than those for PSM; however, the differences were limited to within ~1%, implying that the prediction errors due to PLUM were acceptable.

Two large-scale OBS networks, DONET and S-net, have been incorporated into the JMA EEW system to improve the detection capability for offshore earthquakes. The detection capability analysis for S-net showed that S-net shortened detection times for earthquakes around the Japan and Kuril trenches substantially. This indicates that the current JMA EEW system would issue public warnings with longer lead times than the system in 2011 if large earthquakes occur again around the Japan trench. In the western region of the Nankai trough, along which a M9-class megathrust earthquake is anticipated, NIED is planning to construct a new large-scale OBS network, N-net (Aoi et al., 2020). The N-net observation data will be transmitted to the JMA EEW system in real time. N-net would improve the detection timeliness for earthquakes around the network as well as S-net.

CONCLUSION

The 2011 M_w 9.0 Tohoku-Oki earthquake and subsequent intense earthquake sequence underscored technical limitations of the source-based algorithms employed in the initial JMA EEW system. To overcome the technical issues and to enhance the system performance further, JMA has implemented the IPF and PLUM algorithms and incorporated new observation facilities including S-net.

To evaluate the prediction accuracy changes with the IPF and PLUM implementation, we calculated a prediction score, defined as the ratio of sub-prefectural areas for which ground motions are predicted within intensity errors of one I_{JMA} unit among all areas with predicted or observed $I_{JMA} \ge 4$. Before the IPF implementation, the prediction score based on the maximum prediction results was as low as 69.7% because of mislocated hypocenter estimates provided by the previous PSM algorithms for multiple simultaneous earthquakes. IPF increased the prediction score to more than 80%, providing accurate source parameter estimates based on the Bayesian inference. In addition, IPF reduced the number of overpredicted warnings for earthquakes with a small magnitude (M2–3-class) and low peak intensity ($I_{IMA} \le 3$).

The PLUM implementation made the system less likely to underpredict strong ground motions caused by large inland earthquakes, which was due to PLUM's ground motion prediction approach using direct observation of ongoing ground shakings. PLUM also enabled the system to issue warnings with a low missing rate for earthquakes whose observed intensities were just above the warning threshold. The warning timeliness was improved by PLUM for five out of the 13 earthquakes warned by both of PSM and PLUM. The median of the improved warning times was ~3.4 s.

The detection time analysis for S-net showed that the incorporation of S-net shortened 50th and 75th percentile detection times (times required to issue the first EEW report) by 3.90 and 8.83 s, respectively, for offshore earthquakes around the Japan and Kuril trenches.

Those findings indicate that, over the 10 years after the Tohoku-Oki earthquake, the JMA EEW system has made substantial progress both on software and hardware aspects. At the same time, there is still room for further improvements of the system on the ground motion prediction algorithms and observation facilities. The JMA EEW system in the next 10 years will achieve more accurate, timely, and robust EEW issuance for possible future large earthquakes.

DATA AVAILABILITY STATEMENT

EEW reports issued by the JMA EEW system were obtained from an internal database of JMA. Earthquake locations and observed JMA intensities can be obtained from the JMA website at https:// www.data.jma.go.jp/svd/eqev/data/bulletin/index_e.html (last accessed June 2021).

AUTHOR CONTRIBUTIONS

YK designed the study, performed the analyses, and drafted the article. NH, KN, KM, RT, MM, and KO maintained the JMA EEW system and made the database of EEW reports. KT developed the IPF algorithm. YK and MH proposed the PLUM algorithm. NH, KN, and MH developed the algorithms

to process OBS data. All authors read and approved the final article.

ACKNOWLEDGMENTS

The authors would like to thank the handling editor and two reviewers for their useful comments and suggestions that improved the article. The JMA EEW system has been developed based on joint research of JMA and the Railway Technical Research Institute, technological achievements by NIED, and research led by Kyoto University that was supported by funding for Next

REFERENCES

- Aoi, S., Asano, Y., Kunugi, T., Kimura, T., Uehira, K., Takahashi, N., et al. (2020).
 MOWLAS: NIED Observation Network for Earthquake, Tsunami and Volcano. *Earth Planets Space* 72, 126. doi:10.1186/s40623-020-01250-x
- Böse, M., Allen, R., Brown, H., Gua, G., Fischer, M., Hauksson, E., et al. (2014). "CISN ShakeAlert: An Earthquake Early Warning Demonstration System for California," in *Early Warning for Geological Disasters—Scientific Methods and Current Practice*. Editors J. Zschau and F. Wenzel (Berlin, Germany: Springer), 49–69. doi:10.1007/978-3-642-12233-0_3
- Chen, D. Y., Hsiao, N. C., and Wu, Y. M. (2015). The Earthworm Based Earthquake Alarm Reporting System in Taiwan. *Bull. Seismological Soc. America* 105 (2A), 568–579. doi:10.1785/0120140147
- Clinton, J., Zollo, A., Marmureanu, A., Zulfikar, C., and Parolai, S. (2016). State-ofthe Art and Future of Earthquake Early Warning in the European Region. *Bull. Earthquake Eng.* 14 (9), 2441–2458. doi:10.1007/s10518-016-9922-7
- Doi, K. (2011). The Operation and Performance of Earthquake Early Warnings by the Japan Meteorological Agency. *Soil Dyn. Earthquake Eng.* 31 (2), 119–126. doi:10.1016/j.soildyn.2010.06.009
- Espinosa-Aranda, J. M., Cuellar, A., Garcia, A., Ibarrola, G., Islas, R., Maldonado, S., et al. (2009). Evolution of the Mexican Seismic Alert System (SASMEX). *Seismological Res. Lett.* 80 (5), 694–706. doi:10.1785/gssrl.80.5.694
- Harada, S. (2007). Earthquake Versatile Observation System. Q. J. Seismology 70, 73–81.
- Hayashimoto, N., and Hoshiba, M. (2013). Examination of Travel Time Correction and Magnitude Correction of Tonankai Ocean Bottom Seismographs for Earthquake Early Warning. *Q. J. Seismology* 76, 69–81.
- Hayashimoto, N., Nakamura, T., and Hoshiba, M. (2019). A Technique for Estimating the UD-Component Displacement Magnitude for Earthquake Early Warnings that Can Be Applied to Various Seismic Networks Including Ocean Bottom Seismographs. Q. J. Seismology 83, 1.
- Hayashimoto, N., Nakamura, T., and Hoshiba, M. (2016). Stability of Ocean Bottom Seismograph Data Exposed to strong Shaking: Efforts for Utilizing OBS for Earthquake Early Warning. ECGS ESC/EAEE Jt. Workshop Proc. 31, 41–49.
- Horiuchi, S., Negishi, H., Abe, K., Kamimura, A., and Fujinawa, Y. (2005). An Automatic Processing System for Broadcasting Earthquake Alarms. Bull. Seismological Soc. America 95 (2), 708–718. doi:10.1785/0120030133
- Hoshiba, M., and Aoki, S. (2015). Numerical Shake Prediction for Earthquake Early Warning: Data Assimilation, Real-Time Shake Mapping, and Simulation of Wave Propagation. *Bull. Seismological Soc. America* 105 (3), 1324–1338. doi:10.1785/0120140280
- Hoshiba, M., Iwakiri, K., Hayashimoto, N., Shimoyama, T., Hirano, K., Yamada, Y., et al. (2011). Outline of the 2011 off the Pacific Coast of Tohoku Earthquake (M W 9.0) -Earthquake Early Warning and Observed Seismic Intensity-. *Earth Planet. Sp* 63, 547–551. doi:10.5047/eps.2011.05.031
- Hoshiba, M., Kamigaichi, O., Saito, M., Tsukada, S. y., and Hamada, N. (2008). Earthquake Early Warning Starts Nationwide in Japan. *Eos Trans. AGU* 89, 73–74. doi:10.1029/2008EO080001
- Hoshiba, M., Ohtake, K., Iwakiri, K., Aketagawa, T., Nakamura, H., and Yamamoto, S. (2010). How Precisely Can We Anticipate Seismic Intensities? A Study of Uncertainty of Anticipated Seismic Intensities for the Earthquake

Generation World-Leading Researchers from the Cabinet Office of Japan. The JMA EEW system uses seismic data obtained from observational facilities of NIED, in addition to those of JMA (as of December 2020). Figures were drawn using Generic Mapping Tools (Wessel and Smith, 1998).

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/feart.2021.726045/full#supplementary-material

Early Warning Method in Japan. *Earth Planet. Sp* 62 (8), 611–620. doi:10.5047/eps.2010.07.013

- Hoshiba, M. (2013). Real-time Prediction of Ground Motion by Kirchhoff-Fresnel Boundary Integral Equation Method: Extended Front Detection Method for Earthquake Early Warning. J. Geophys. Res. Solid Earth 118, 1038–1050. doi:10.1002/jgrb.50119
- Iwakiri, K., Hoshiba, M., Nakamura, K., and Morikawa, N. (2011). Improvement in the Accuracy of Expected Seismic Intensities for Earthquake Early Warning in Japan Using Empirically Estimated Site Amplification Factors. *Earth Planet. Sp* 63 (2), 57–69. doi:10.5047/eps.2010.12.002
- Japan Meteorological Agency JMA (2020b). Press release: On the implementation of an urgent improvement of earthquake early warning. Retrieved from: https:// www.jma.go.jp/jma/press/2008/07a/kaizen_20200807.pdf (last accessed June, , 2021).
- Japan Meteorological Agency JMA (2014). Document No. 1: On the Dissemination History and Accuracy Evaluation of Recent Earthquake Early Waning reportsThe Fifth Meeting of the Committee of Evaluation and Improvement of Earthquake Early Warning. Retrieved from: https://www.data.jma.go.jp/svd/ eqev/data/study-panel/eew-hyoka/05/shiryou1.pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2019a). Document No. 2: Document of the Ninth Meeting of the Technical Panel of the Committee of Evaluation and Improvement of Earthquake Early Warning. The Ninth Meeting of the Technical Panel of the Committee of Evaluation and Improvement of Earthquake Early Warning. Retrieved from: http://www.data.jma.go.jp/svd/ eqev/data/study-panel/eew-hyoka/t09/20190305_siryou2.pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2016). Press Release: On a Technical Improvement of Earthquake Early Warning (The IPF Algorithm) and Measure for the False Earthquake Early Warning Report on August 1, 2016. Retrieved from: https://www.jma.go.jp/jma/press/1612/13a/EEW_kaizen_ 201612.pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2018). Press Release: On an Improvement of Earthquake Early Warning—The Prediction Accuracy of Seismic Intensity when Large Earthquakes Occur Is Improved. Retrieved from: https://www. jma.go.jp/jma/press/1803/08c/EEW_kaizen_201803.pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2020a). Press Release: On the Additional Use of New Ocean Bottom Seismometers for Earthquake Early Warning. Retrieved from: https://www.jma.go.jp/jma/press/2003/19a/20200319_eewkatsuyouS6. pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2015). Press Release: On the Start Date for Using New Observation Data for Earthquake Early Warning. Retrieved from: https://www.jma.go.jp/jma/press/1503/24a/eewkatsuyou20150324.pdf (last accessed June, 2021).
- Japan Meteorological Agency JMA (2019b). Press Release: On the Use of New Ocean-Bottom Seismic Observation Data for Earthquake Early Warning. Retrieved from: https://www.jma.go.jp/jma/press/1906/21a/ eewkatsuyou20190621.pdf (last accessed June 2021).
- Japan Meteorological Agency JMA (1996). Seismic Intensity. (Tokyo, Japan: Gyosei).
- Kamigaichi, O. (2004). JMA Earthquake Early Warning. J. JAEE 4 (3), 134–137. doi:10.5610/jaee.4.3_134

- Kamigaichi, O., Saito, M., Doi, K., Matsumori, T., Tsukada, S., Takeda, K., et al. (2009). Earthquake Early Warning in Japan: Warning the General Public and Future Prospects. *Seismological Res. Lett.* 80 (5), 717–726. doi:10.1785/ gssrl.80.5.717
- Kanazawa, T., Uehira, K., Mochizuki, M., Shinbo, T., Fujimoto, H., Noguchi, S., et al. (2016). S-net Project, Cabled Observation Network for Earthquakes and Tsunamis. SubOptic. WE2B–3.
- Kaneda, Y., Kawaguchi, K., Araki, E., Matsumoto, H., Nakamura, T., Kamiya, S., et al. (2015). "Development and Application of an Advanced Ocean Floor Network System for Megathrust Earthquakes and Tsunamis," in *Seafloor Observatories*. (Berlin, Germany: Springer), 643–662. doi:10.1007/978-3-642-11374-1_25
- Kawaguchi, K., Kaneko, S., Nishida, T., and Komine, T. (2015). "Construction of the DONET Real-Time Seafloor Observatory for Earthquakes and Tsunami Monitoring," in *Seafloor Observatories*. (Berlin, Germany: Springer), 211–228. doi:10.1007/978-3-642-11374-1_10
- Kilb, D., Bunn, J. J., Saunders, J. K., Cochran, E. S., Minson, S. E., Baltay, A., et al. (2021). The PLUM Earthquake Early Warning Algorithm: A Retrospective Case Study of West Coast, USA, Data. *J. Geophys. Res. Solid Earth* 126, e2020JB021053. doi:10.1029/2020JB021053
- Kodera, Y. (2019). An Earthquake Early Warning Method Based on Huygens Principle: Robust Ground Motion Prediction Using Various Localized Distance-Attenuation Models. J. Geophys. Res. Solid Earth 124 (12), 12981–12996. doi:10.1029/2019JB017862
- Kodera, Y., Hayashimoto, N., Moriwaki, K., Noguchi, K., Saito, J., Akutagawa, J., et al. (2020). First-year Performance of a Nationwide Earthquake Early Warning System Using a Wavefield-Based Ground-Motion Prediction Algorithm in Japan. Seismological Res. Lett. 91 (2A), 826–834. doi:10.1785/ 0220190263
- Kodera, Y. (2018). Real-time Detection of Rupture Development: Earthquake Early Warning Using P Waves from Growing Ruptures. *Geophys. Res. Lett.* 45, 156–165. doi:10.1002/2017GL076118
- Kodera, Y., Saitou, J., Hayashimoto, N., Adachi, S., Morimoto, M., Nishimae, Y., et al. (2016). Earthquake Early Warning for the 2016 Kumamoto Earthquake: Performance Evaluation of the Current System and the Next-Generation Methods of the Japan Meteorological Agency. *Earth Planets Space* 68 (1), 202. doi:10.1186/s40623-016-0567-1
- Kodera, Y., Yamada, Y., Hirano, K., Tamaribuchi, K., Adachi, S., Hayashimoto, N., et al. (2018). The Propagation of Local Undamped Motion (PLUM) Method: A Simple and Robust Seismic Wavefield Estimation Approach for Earthquake Early Warning. *Bull. Seismological Soc. America* 108 (2), 983–1003. doi:10.1785/ 0120170085
- Kunugi, T., Aoi, S., Nakamura, H., Suzuki, W., Morikawa, N., and Fujiwara, H. (2013). An Improved Approximating Filter for Real-Time Calculation of Seismic Intensity. Jssj 65 (3), 223–230. doi:10.4294/zisin.65.223
- Liu, A., and Yamada, M. (2014). Bayesian Approach for Identification of Multiple Events in an Early Warning System. *Bull. Seismological Soc. America* 104 (3), 1111–1121. doi:10.1785/0120130208
- Matsuoka, M., and Midorikawa, S. (1994). The Digital National Land Information and Seismic Microzoning. in Proceeding of the 22nd Symposium of Earthquake Ground Motion, AIJ, Tokyo, Japan, 23–34.
- Mochizuki, M., Kanazawa, T., Uehira, K., Shimbo, T., Shiomi, K., Kunugi, T., et al. (2016). S-net Project: Construction of Large Scale Seafloor Observatory Network for Tsunamis and Earthquakes in Japan. AGU Fall Meeting, San Francisco, United States. NH43B–1840.
- Nakamura, T., and Hayashimoto, N. (2019). Rotation Motions of Cabled Ocean-Bottom Seismic Stations during the 2011 Tohoku Earthquake and Their Effects on Magnitude Estimation for Early Warnings. *Geophys. J. Int.* 216 (2), 1413–1427. doi:10.1093/gji/ggy502
- Nakamura, T., Nakano, M., Hayashimoto, N., Takahashi, N., Takenaka, H., Okamoto, T., et al. (2014). Anomalously Large Seismic Amplifications in the Seafloor Area off the Kii peninsula. *Mar. Geophys. Res.* 35, 255–270. doi:10.1007/s11001-014-9211-2
- Nakamura, Y. (1988). On the Urgent Earthquake Detection and Alarm System (UrEDAS). in Proceedings of Ninth World Conference on Earthquake Engineering, Tokyo and Kyoto, Japan, August 2–9, 1988, 7, 673–678.
- Nof, R. N., and Kurzon, I. (2021). TRUAA-earthquake Early Warning System for Israel: Implementation and Current Status. Seismological Res. Lett. 92 (1), 325–341. doi:10.1785/0220200176

- Noguchi, K., Hayashimoto, N., Tamaribuchi, K., and Kodera, Y. (2020). Improvement of IPF Method with Utilization of Hi-Net. *SSJ Fall Meet*. 2020, S15–S06.
- Odaka, T., Ashiya, K., Tsukada, S., Sato, S., Ohtake, K., and Nozaka, D. (2003). A New Method of Quickly Estimating Epicentral Distance and Magnitude from a Single Seismic Record. *Bull. Seismological Soc. America* 93 (1), 526–532. doi:10.1785/0120020008
- Okada, Y., Kasahara, K., Hori, S., Obara, K., Sekiguchi, S., Fujiwara, H., et al. (2004). Recent Progress of Seismic Observation Networks in Japan -Hi-Net, F-Net, K-NET and KiK-Net-. *Earth Planet. Sp* 56, xv–xxviii. doi:10.1186/ BF03353076
- Sheen, D. H., Park, J. H., Chi, H. C., Hwang, E. H., Lim, I. S., Seong, Y. J., et al. (2017). The First Stage of an Earthquake Early Warning System in South Korea. Seismological Res. Lett. 88 (6), 1491–1498. doi:10.1785/ 0220170062
- Si, H., and Midorikawa, S. (1999). New Attenuation Relationships for Peak Ground Acceleration and Velocity Considering Effects of Fault Type and Site Condition. *Nihon Kenchiku Gakkai Kozokei Ronbunshu* 64 (523), 63–70. doi:10.3130/ aijs.64.63_2
- Takagi, R., Uchida, N., Nakayama, T., Azuma, R., Ishigami, A., Okada, T., et al. (2019). Estimation of the Orientations of the S-net Cabled Ocean-Bottom Sensors. Seismological Res. Lett. 90 (6), 2175–2187. doi:10.1785/ 0220190093
- Tamaribuchi, K., Yamada, M., and Wu, S. (2014). A New Approach to Identify Multiple Concurrent Events for Improvement of Earthquake Early Warning. *Jssj* 67 (67), 41–55. doi:10.4294/zisin.67.41
- Tsukada, S., Odaka, T., Ashiya, K., Ohtake, K., and Nozaka, D. (2004). Analysis of the Envelope Waveform of the Initial Part of P-Waves and its Application to Quickly Estimating the Epicentral Distance and Magnitude. *Jssj* 56 (56), 351–361. doi:10.4294/zisin1948.56.4_351
- Uehira, K., Kanazawa, T., Mochizuki, M., Fujimoto, H., Noguchi, S., Shinbo, T., et al. (2016). Outline of Seafloor Observation Network for Earthquakes and Tsunamis along the Japan Trench (S-Net). EGU General Assembly 2016, Vienna, Austria, EGU2016–13832.
- Wessel, P., and Smith, W. H. F. (1998). New, Improved Version of Generic Mapping Tools Released. *Eos Trans. AGU* 79 (47), 579. doi:10.1029/98EO00426
- Wu, S., Yamada, M., Tamaribuchi, K., and Beck, J. L. (2015). Multi-events Earthquake Early Warning Algorithm Using a Bayesian Approach. *Geophys. J. Int.* 200 (2), 791–808. doi:10.1093/gji/ggu437
- Yamada, M., Tamaribuchi, K., and Wu, S. (2014). Faster and More Accurate Earthquake Early Warning System. J. JAEE 14 (4), 21–24. doi:10.5610/ jaee.14.4_21
- Yamada, M., Tamaribuchi, K., and Wu, S. (2021). The Extended Integrated Particle Filter Method (IPFx) as a High-Performance Earthquake Early Warning System. Bull. Seismological Soc. America 111 (3), 1263–1272. doi:10.1785/ 0120210008
- Zhang, H., Jin, X., Wei, Y., Li, J., Kang, L., Wang, S., et al. (2016). An Earthquake Early Warning System in Fujian, China. *Bull. Seismological Soc. America* 106 (2), 755–765. doi:10.1785/0120150143

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