Ensemble flood simulation for a small dam catchment in Japan using 10 and 2 km resolution nonhydrostatic model rainfalls

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Abstract. This paper presents a study on short-term ensemble flood forecasting specifically for small dam catchments in Japan. Numerical ensemble simulations of rainfall from the Japan Meteorological Agency nonhydrostatic model (JMA-NHM) are used as the input data to a rainfall–runoff model for predicting river discharge into a dam. The ensemble weather simulations use a conventional 10 km and a high-resolution 2 km spatial resolutions. A distributed rainfall–runoff model is constructed for the Kasahori dam catchment (approx. 70 km²) and applied with the ensemble rainfalls. The results show that the hourly maximum and cumulative catchment-average rainfalls of the 2 km resolution JMA-NHM ensemble simulation are more appropriate than the 10 km resolution rainfalls. All the simulated inflows based on the 2 and 10 km rainfalls become larger than the flood discharge of 140 m³ s⁻¹, a threshold value for flood control. The inflows with the 10 km resolution ensemble rainfall are all considerably smaller than the observations, while at least one simulated discharge out of 11 ensemble members with the 2 km resolution rainfalls reproduces the first peak of the inflow at the Kasahori dam with similar amplitude to observations, although there are spatiotemporal lags between simulation and observation. To take positional lags into account of the ensemble discharge simulation, the rainfall distribution in each ensemble member is shifted so that the catchment-averaged cumulative rainfall of the Kasahori dam maximizes. The runoff simulation with the position-shifted rainfalls shows much better results than the original ensemble discharge simulations.

1 Introduction

Currently, short-term ensemble flood forecasting based on ensemble numerical weather predictions (NWPs) is gaining more attention in Japan, as evidenced by the commencement of a project for ensemble weather/flood forecasting using the new K computer in Kobe, Japan (Saito, 2013b). Here, short-term flood forecasting means flood forecasts with lead times of half to 1 day. Cloke and Pappenberger (2009) presented a comprehensive review of medium range (2–15 days ahead) ensemble flood forecasts; however, the review focused mainly on European weather/flood forecasting examples using global ensemble predictions.

Precipitation data from NWPs are usually not considered as primary data for flood forecasting because of their accuracy, especially in the disaster prevention purpose. In Japan, primary data are obtained using radar observations of precipitation calibrated by the Japan Meteorological Agency (JMA) AMeDAS (Automated Meteorological Data Acquisition System) surface rain gauges (Makihara, 2000) or by the rain gauges of the Ministry of Land, Infrastructure, Transport and Tourism (MLIT, 2012a). It should be noted that in Japan, NWP-based weather forecasting has shown success in predicting synoptic (spatial scale of O(1000 km)) weather systems and associated precipitation events. The difference between weather and flood forecasting arises because Japanese river basins are often too small for NWP models to provide accurate estimations. The largest catchment in Japan is the Tone river catchment, which is around 17 000 km², whereas many dam catchments are just several 100 km² or less. Thus,
the areas of concern for most river/dam administrators are too small for global NWP models.

In the aforementioned project (Saito, 2013b), the Meteorological Research Institute tested ensemble NWPs with 2 km resolution, finer than used previously for mesoscale ensemble forecasts (e.g., Saito et al., 2010, 2011). With such a resolution, complex topographies and mesoscale convective systems can be better represented. In addition, the atmospheric model does not apply cumulus convective parameterizations, which enables us to reproduce rainfall with more realistic intensities. Therefore, such high-resolution cloud-resolving ensemble weather simulations can produce probabilistic information of intense rainfall systems better than mesoscale models with lower resolutions (Duc et al., 2013). Using ensemble rainfall forecasts produced by the JMA non-hydrostatic model (JMA-NHM), the authors have performed a study on the ensemble flood forecasting for a real extreme flood event in Niigata, Japan, using a rainfall–runoff model, the results of which are presented in this paper.

Flood disasters occurred on 27–30 July 2011 in Niigata and Fukushima prefectures, Japan, following a severe rainstorm, characterized by two rainfall peaks. According to a report by the Niigata Prefecture (Niigata, 2011), the cumulative rainfall from the onset of the rainfall until 13:00 JST (04:00 UTC) on 30 July 2011 reached 985 mm at the Kasahori Dam Observatory. The cumulative rainfall at 68 rainfall observatories managed by MLIT, JMA, and Niigata Prefecture exceeded 250 mm. During this time, JMA announced “record-setting short-term heavy rainfall information” on 30 occasions. The hourly rainfall recorded from 20:00 to 21:00 JST on 29 July at the Tokamachi-Shinko Observatory reached 120 mm, which is an example of extreme record-setting rainfall within the region. Among the many local record-setting rainfall amounts, this paper focuses on the Kasahori dam catchment, which is a small sub-catchment of the Shinano river catchment.

The report by the Japan Weather Association (hereinafter JWA, 2011) indicates that the discharge forecasting system, operated at the Kasahori dam using short-term and very short-term rainfall prediction by a weather model, was effective for deciding the quantity of water release from the Kasahori dam. According to the report, at 03:00 JST on 29 July 2011 the discharge forecasting system predicted dam inflow of 846 m$^3$s$^{-1}$ at 13:00 JST on 29 July, in consideration of the observed inflow of 843 m$^3$s$^{-1}$. This information, together with a telephone consultation between the JWA and dam administrator, supported the decision for the preliminary dam release. Although this clearly demonstrates the usefulness of precipitation forecasts in dam control, it is not easy to produce an accurate deterministic forecast of precipitation for a small-scale dam catchment. Therefore, this paper studies the effectiveness of ensemble flood forecasting on the Kasahori dam catchment.

The structure of this paper is as follows. Section 2 describes additional details regarding the 2011 Niigata–Fukushima heavy rainfall. Section 3 briefly describes the Kasahori dam catchment and the Kasahori dam. Section 4 addresses the rainfall analysis using rain gauge and radar-derived rainfalls. Section 5 introduces the rainfall–runoff model. Section 6 presents the results of the rainfall–runoff simulations using both observed rainfall and ensemble predictions of rainfall. Section 7 presents the concluding remarks and aspects of future work.

2 The 2011 Niigata–Fukushima heavy rainfall

A local heavy rainfall event occurred in July 2011 over Niigata and Fukushima prefectures in northern central Japan. Record-breaking torrential rainfall of more than 600 mm was observed during 3 days from 27 to 30 July, which caused severe damages in the prefectures of Niigata and Fukushima. Six people were killed and more than 13,000 houses damaged by dike breaks, river flooding, and landslides.

Figure 1 (left) indicates a surface weather map at 09:00 JST (00:00 UTC), 29 July 2011. A distinct synoptic-scale stationary front runs from the northwest to the southeast over northern central Japan. The right panel of Fig. 1 shows the 3 h accumulated rainfall from 12:00 to 15:00 JST (03:00 to 06:00 UTC) (radar–rain-gauge precipitation analysis of the Japan Meteorological Agency). Torrential rain exceeding 100 mm per 3 h occurred over the small area along the stationary front. A detailed description of this rainfall event has been published by JMA as a special issue of the JMA Technical Report (JMA, 2013a).

3 Kasahori dam catchment

Figure 2 (left) shows the Shinano and Agano river catchments, Japan, and Fig. 2 (right) shows an enlarged view of the Kasahori and Otani dam catchments. These catchment data were obtained from the Digital National Land Information (hereinafter DNLI) of MLIT (MLIT, 2012b). The Kasahori dam catchment area is calculated as 72.7 km$^2$ from the DNLI; thus, the catchment is very small. The land use of the Kasahori dam catchment is shown in Fig. 3 (left), which reveals that most of the area is occupied by forest. Therefore, the model area is treated as entirely forested in the following modeling. The basic operation of the Kasahori dam is summarized as follows.

1. In the rainy season, the reservoir water level is decreased to the normal water level for the rainy season (elevation level (EL) 194.5 m).

2. If a flood risk due to extreme rainfall is expected by weather monitoring/prediction, the water level is further decreased to the preliminary release water level (EL 192.0 m).
3. When the inflow exceeds $140 \text{ m}^3 \text{s}^{-1}$, the gate opening is fixed so that the outflow amount is determined only by the water pressure. This is, in a broad sense, natural regulation. The gate opening is not adjusted until the water level reaches EL 206.6 m.

4. When the reservoir water level reaches EL 206.6 m, Tadashigaki (emergency) operation is taken: the outflow is set equal to the inflow.

Note that, after the flood event in July 2011, the dam has been under renovation to increase its flood control capacity.

4 Analysis of rainfall over the Kasahori dam catchment

The analysis of the rainfall over the Kasahori dam catchment is performed in this section. The rain gauge (RG) rainfall, JMA radar-composite (RC), and JMA radar–rain-gauge (RR) analyzed data are used for the investigation. The descriptions of the RC and RR data are as follows.

(a) The 1 km resolution RC data: the echo intensity, which can be converted to rainfall intensity, is observed by 20 meteorological radar stations of JMA and is available with 10 min temporal resolution.

(b) The 1 km resolution RR analyzed precipitation data: the rainfall intensity observed by the radar is corrected using rain gauge data (ground observation data) and they are available with 30 min temporal resolution.

See Nagata (2011) for the further details of the analysis data. Several previous studies have been published (e.g., Kamiguchi et al., 2010; Sasaki et al., 2008) using these precipitation analysis data.

(a) RG rainfall data: the time-series data of hourly rainfall of the Otani dam, Otani, Koumyozan, Kasahori dam, Kasahori, and Dounokubo rainfall observatories, shown in Fig. 3 (right), are used as the ground observation data. A Thiessen polygon is drawn based on the locations of the observatories, by which each observatory is assigned a representative area. Then, the hourly rainfall data are given to each representative area in the calculation. The cumulative and maximum hourly rainfalls for the period 01:00 JST 28 July to 24:00 JST 30 July were 955 mm and 83 mm $\text{h}^{-1}$ at the Kasahori dam, 722 mm and 71 mm $\text{h}^{-1}$ at Kasahori, 786 mm and 74 mm $\text{h}^{-1}$ at Koumyozan, and 723 mm and 78 mm $\text{h}^{-1}$ at Dounokubo, respectively.

Figures 5 shows the spatial distributions of the cumulative rainfall for the 2011 rainfall event around the Shinano and Agano river catchment by RC (upper left) and RR (upper right), while Fig. 5, lower left and right panels, shows those of the Kasahori dam catchment. It is apparent from Fig. 5 that the distributions by RC and RR show similar patterns in the mesoscale. However, it becomes slightly different when focusing on the small-scale Kasahori dam catchment. To verify whether the RC precipitation in this region is always smaller than RR, Fig. 6 show the rainfall patterns for another rainfall event in 2004, when flooding also occurred in the region. The damage by the flooding due to the 2004 event was even
Figure 2. Shinano and Agano river catchments (left). Kasahori dam and Otani dam catchments (right).

Figure 3. Land use of the Kasahori and Otani dam catchments (left). Rainfall observatories and Thiessen polygons of the Kasahori and Otani dam catchments (right).
Figure 4. Catchment-averaged rainfalls of the Kasahori dam catchment.

Figure 5. Spatial patterns of cumulative rainfalls around the Shinano and Agano catchments using radar composite (upper left) and radar–rain gauge (upper right) and around the Kasahori dam catchment using radar composite (lower left) and radar–rain gauge (lower right) for the 2011 rainfall event.
greater than that caused by the 2011 rainfall, although the total amount of rainfall in 2011 was larger. Figure 6 shows that the RC rainfall is larger than RR rainfall for the 2004 rainfall. The RR rainfall is obtained by correcting the RC using RG rainfall. Thus, the magnitude of the relation between the RC and RR rainfalls depends on the magnitude of the RG rainfall compared with the RC. The precipitation by RC is occasionally larger than the RR rainfall when the RG rainfall is smaller than RC and sometimes vice versa. As the RC can be obtained at 10 min interval with greater spatial coverage, it is considered more reasonable for use in future real-time purposes, though the authors do not carry out the operation. Thus, the calibration of the rainfall–runoff model is performed using RC rainfall.

5 Methods

5.1 Distributed rainfall–runoff (DRR) model

A DRR model was applied to the Kasahori dam catchment. The DRR model applied is that originally developed by Kojima and Takara (2003) called CDRMV3. The details of this DRR model can be seen in the work by Apip et al. (2011). In the DRR model, the surface and river flows are simulated using a 1-D kinematic wave model. The subsurface flow is simulated using a $q$–$h$ relationship developed by Tachikawa et al. (2004). The schematic of the $q$–$h$ relationship is shown in Fig. 7, where $q$ is the discharge per unit width and $h$ is the water depth, as shown in Fig. 7a. The mathematical expression is as follows:

$$q(h) = \begin{cases} v_m d_m \left( \frac{h}{d_m} \right)^\beta, & (0 < h \leq d_m) \\ v_m d_m + v_a (h - d_m), & (d_m < h \leq d_a) \\ v_m d_m + v_a (h - d_m) + \alpha (h - d_a)^m, & m = \frac{5}{3}, \quad (d_a < h). \end{cases}$$

(1)
where \( v_m = k_m i \), \( v_a = k_a i \), \( \alpha = \sqrt{i/N_{\text{slope}}} \), and \( D \) is the thickness of the layer, shown in Fig. 7a; \( d_a - d_m \) is the area of the saturated flow; \( d_m \) is the area of unsaturated flow; \( v_m \) is the unsaturated flow velocity; \( k_m \) is the hydraulic conductivity in \( d_m \); \( i \) represents the slope gradient; \( v_a \) is the saturated flow velocity; \( k_a \) is the hydraulic conductivity in \( d_a - d_m \); and \( N_{\text{slope}} \) represents the equivalent roughness coefficient of the slope. \( \beta k_m = k_a \) needs to be satisfied to establish the continuity of the \( q-h \) relationship. The initial discharge \( Q_i \) is set at the catchment outlet of the river. Normally \( Q_i \) is the observed discharge in the beginning of the simulation. Then, the \( Q_i \) is converted to the water depth \( h \) in each grid of the entire catchment depending on the ratio of the flow accumulation value for the particular grid and the flow accumulation value at the outlet. Thus, before the simulation, all the grids already have the initial water depth \( h_i \) depending on \( Q_i \). This could be considered as the base flow in the model concept. As mentioned in Sect. 4, the parameters of the DRR model are identified using the RC. The equivalent roughness coefficient of the forest, the Manning coefficient of the river, and identified soil-related parameters are described in Table 1.
Table 1. Equivalent roughness coefficient of the forest, Manning’s coefficient of the river, and soil-related parameters identified by the radar composite.

<table>
<thead>
<tr>
<th>Forest</th>
<th>River</th>
<th>$D$</th>
<th>$k_s$</th>
</tr>
</thead>
<tbody>
<tr>
<td>(m$^{-1/3}$)</td>
<td>(m$^{-1/3}$)</td>
<td>(m)</td>
<td>(m s$^{-1}$)</td>
</tr>
<tr>
<td>0.15093</td>
<td>0.004</td>
<td>0.320</td>
<td>0.0005</td>
</tr>
</tbody>
</table>

5.2 Mesoscale ensemble prediction system

Two 11-member ensemble forecasts with different horizontal resolutions (10 and 2 km) were conducted for the 2011 Niigata–Fukushima heavy rainfall event using JMA-NHM (Saito et al., 2006; Saito, 2012) as the forecast model. The 10 km ensemble prediction system (EPS) uses the JMA’s operational mesoscale 4D-Var analysis of 12:00 UTC (21:00 JST) on 28 July and the JMA’s global spectral model (GSM) forecast from the same time as the initial and boundary conditions of the control run, respectively. As for the initial and lateral boundary conditions, perturbations from the JMA’s 1-week global ensemble prediction from 12:00 UTC (21:00 JST) on 28 July were employed, whose detailed procedures are given in Saito et al. (2010, 2011). The 2 km EPS is a downscaling of the 10 km EPS with a 6 h time lag, using the forecasts of the 10 km EPS as the initial and boundary conditions (Fig. 8).

The bulk method that predicts the mixing ratios of six water species (water vapor, cloud water, rainwater, cloud ice, snow, and graupel) and the number density of cloud ice was adopted as the cloud microphysical process. The 10 km EPS applied the modified Kain–Fritsch convective parameterization scheme, while the 2 km EPS did not use convective parameterization. Other physical processes of the two systems were almost the same to those of the operational mesoscale model and the local forecast model of JMA (JMA, 2013b). The verification of the statistical performance of similar double-nested EPSs has been given by Duc et al. (2013).

Figure 9 (upper left) shows the 3 h accumulated rainfall from 12:00 to 15:00 JST (21:00 to 24:00 JST) by the control run of the 10 km EPS. Although the maximum value of the predicted rainfall (74 mm) is somewhat weaker than the observation (right panel of Fig. 1), the region of intense rainfall is simulated well. The upper right panel of Fig. 9 indicates the forecast by each member of the 10 km EPS. Seemingly, the result of each ensemble member resembles the others, and the basic characteristic features of the observed rainfall are simulated well. The maximum rainfall was obtained by member p02 (89 mm). A common feature seen in these figures is that weak fake rainfall appears over the coastal region facing the Sea of Japan, which is likely produced by the Kain–Fritsch convective parameterization.

Figure 9 lower panels shows the corresponding results by the 2 km EPS. The concentration of intense precipitation is produced more clearly, the maximum rainfall of which reaches 237 mm. The areas of weak rainfall over the western coastal region, appearing in Fig. 9 upper panels, no longer develop because of the removal of the convective parameterization. A detailed analysis of the two EPSs (ensemble spread and fraction skill scores) and the result of a sensitivity experiment to the orography have been presented by Saito et al. (2013a).

6 Results

First, the DRR model is verified by performing experiments with the observed rainfall data. Next, the ensemble rainfall forecast data are used to perform an ensemble flood forecast. Finally, additional experiments are performed to consider position errors of rainfall.

6.1 Rainfall–runoff simulations with radar and rain gauge rainfalls

The inflow to the Kasahori dam is simulated using the DRR model. The RG, RC, and RR data are used as the inputs to the runoff simulations. The three hydrographs with the parameters identified by the RC are shown in Fig. 10. The simulated hydrograph with the RC rainfall is in relatively good agreement with the observations, which is to be expected because the model parameters are calibrated against the RC rainfall. Using a straight line method for the base flow separation, the total discharge with RC in mm becomes 556.3 mm while the total rainfall is 568.5 mm.

The simulated hydrographs for the other two rainfalls are larger than the observations. We do not address the magnitude of the relationship in this paper because it is not possible to determine more accurate rainfall data. The RG, RC, and RR measurements all have strengths and weaknesses; however, we focus on the consideration of RC for use because of the frequency of the data, i.e., 10 min interval.

6.2 Ensemble rainfall–runoff simulations with raw output of JMA-NHM

Using the ensemble rainfalls from JMA-NHM, explained in Sect. 5.2, the ensemble flood simulation focusing on the Kasahori dam catchment was performed. A flowchart is shown in Fig. 11 to explain briefly again the overall procedure of the methodology for the ensemble simulations used in the paper.

The catchment-averaged ensemble rainfalls obtained from the 10 and 2 km resolution NHM are shown in Fig. 12. Figure 12 (upper) shows the control run and five negatively perturbed members, m01–m05 (m indicates minus), and five positively perturbed members, p01–p05 (p indicates positive), for the 10 km resolution.

It is apparent from the figure that the magnitude of the 10 km resolution ensemble rainfall is basically lower than the
Figure 9. Three-hour accumulated rainfall from 12:00 to 15:00 JST (03:00 to 06:00 UTC) on 29 July by the control run of the 10 km EPS (upper left). Same as in the left figure, but the forecast by each member of the 10 km EPS (upper right). The figures on the lower left and right are the same as in the upper figures but for the forecasts by the 2 km EPS.

Figure 10. Dam inflows for three rainfalls using the parameters identified with radar composite.
Figure 11. Flowchart of the overall procedure for the ensemble weather/flood simulation.

RC rainfall. Thus, the dam inflows, obtained from the RC parameters in Table 1 with the 10 km resolution ensemble rainfall, lead to lower magnitude discharge compared with the ground observations (shown later in the paper).

Figure 12 shows the control run, m01–m05, and p01–p05 for the 2 km resolution NHM. The figures reveal that the first peak in the 2 and 10 km resolution ensemble simulations appears 2–4 h earlier than that in the observation. The magnitudes of some 2 km resolution ensemble rainfalls are equivalent to that of the RC rainfall. Thus, dam inflows using the RC parameters in Table 1 with the 2 km resolution ensemble rainfall can indicate discharge with equivalent magnitude (shown later in the paper). Figure 13 shows the spatial patterns of the cumulative ensemble rainfalls from 03:00 JST on 29 July 2011 to 03:00 JST on 30 July 2011 by the 11 ensemble simulations (upper: 10 km resolution; lower: 2 km resolution). The figures indicate that the 2 km resolution NHM rainfalls are apparently larger than the 10 km resolution rainfalls. Tables 2 and 3 show the cumulative and maximum hourly rainfalls from the 10 and 2 km resolution NHMs, respectively, averaged over the Kasahori dam catchment, which show that the 10 km resolution rainfalls are smaller than the 2 km resolution rainfalls. The maximum cumulative rainfall of the 2 km resolution NHM is realized in p02: 175.5 mm. Table 2 also shows the average cumulative rainfalls of both the 10 and 2 km resolution NHMs. The average cumulative rainfall in the 2 km resolution NHM is greater than in the 10 km resolution NHM. With regard to the maximum hourly
rainfall in Table 3, p02 shows the highest values in both the 10 and 2 km resolution NHMs. The maximum hourly rainfall in the 2 km resolution NHM is also greater than that in the 10 km resolution NHM. This tendency is also true in the average maximum hourly rainfall shown in Table 3.

Figure 14 (upper) shows the simulated inflow to the Kasahori dam with the control run and positively/negatively perturbed rainfalls of the 10 km resolution NHM. Figure 14 (upper) shows that all the inflows to the Kasahori dam are lower than the observations; however, these inflows exceed the flood discharge of 140 m$^3$ s$^{-1}$, which is the threshold for the flood control operation (see Sect. 3 for the details of the operation).

Figure 14 (lower) shows the simulated discharge with the 2 km resolution ensemble rainfalls. Figure 14 (lower) shows that at least the first peak of the dam inflow in p02 shows a comparable value with that of the observed inflow; the peak discharge of the observation is 843 m$^3$ s$^{-1}$, whereas it is 779 m$^3$ s$^{-1}$ with the p02 of the 2 km resolution NHM. However, the occurrence of the first peak in the simulation is 4 h earlier than indicated by the observations. The fact that one of the ensemble flood discharges with the 2 km resolution NHM shows approximately equivalent magnitude of discharge with the observed first peak discharge, despite the forward shift in occurrence time, implies that the ensemble flood prediction with the 2 km resolution NHM could potentially be used as a reference in dam operations, although the discharge reproduction is still not fully satisfactory both in quality and quantity. The ensemble flood simulations with the 10 km resolution NHM could not reproduce the peak at all. Moreover, the first peak of the simulated inflow with the control run of the 2 km resolution NHM attains only 614 m$^3$ s$^{-1}$. A single value from a deterministic (i.e., control run only) NWP (i.e., prevailing prediction) might fail to capture a realistic discharge, whereas ensemble simulations produce additional prediction ranges that cover the higher observed discharge values.

In the actual operation of the Kasahori dam, the dam gate opening is fixed once the inflow exceeds 140 m$^3$ s$^{-1}$. The
Figure 13. Spatial distributions of cumulative ensemble rainfalls (upper: 10 km resolution; lower: 2 km resolution).

Table 2. Cumulative rainfall of 2 and 10 km resolution ensemble rainfall simulations.

<table>
<thead>
<tr>
<th></th>
<th>Cntl</th>
<th>p01</th>
<th>p02</th>
<th>p03</th>
<th>p04</th>
<th>p05</th>
<th>m01</th>
<th>m02</th>
<th>m03</th>
<th>m04</th>
<th>m05</th>
<th>avg.</th>
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<tbody>
<tr>
<td>10 km</td>
<td>108.8</td>
<td>130.2</td>
<td>140.6</td>
<td>113.5</td>
<td>140.2</td>
<td>97.9</td>
<td>111.6</td>
<td>93.5</td>
<td>102.2</td>
<td>101.2</td>
<td>100.5</td>
<td>112.7</td>
</tr>
<tr>
<td>2 km</td>
<td>156.7</td>
<td>124.6</td>
<td>175.5</td>
<td>128.5</td>
<td>165.1</td>
<td>93.9</td>
<td>111.3</td>
<td>98.2</td>
<td>169.1</td>
<td>86.9</td>
<td>148.8</td>
<td>132.6</td>
</tr>
</tbody>
</table>
Table 3. Maximum hourly rainfall of 2 and 10 km resolution ensemble rainfall simulations.

<table>
<thead>
<tr>
<th></th>
<th>Cntl</th>
<th>p01</th>
<th>p02</th>
<th>p03</th>
<th>p04</th>
<th>p05</th>
<th>m01</th>
<th>m02</th>
<th>m03</th>
<th>m04</th>
<th>m05</th>
<th>avg.</th>
</tr>
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<tbody>
<tr>
<td>10 km</td>
<td>26.8</td>
<td>17.6</td>
<td>41.7</td>
<td>27.9</td>
<td>18.7</td>
<td>18.2</td>
<td>27.5</td>
<td>16.0</td>
<td>21.9</td>
<td>28.9</td>
<td>29.4</td>
<td>23.5</td>
</tr>
<tr>
<td>2 km</td>
<td>41.4</td>
<td>32.4</td>
<td>49.5</td>
<td>31.8</td>
<td>37.0</td>
<td>27.6</td>
<td>28.8</td>
<td>29.8</td>
<td>42.5</td>
<td>28.2</td>
<td>30.8</td>
<td>34.5</td>
</tr>
</tbody>
</table>

Figure 14. Results of ensemble flood simulations with 10 km resolution rainfall (upper) and with 2 km resolution rainfall (lower).

Dam inflows from the control run and the other 10 ensemble rainfall predictions of both the 2 and 10 km resolution NHMs all predict that the dam inflow is above the flood discharge threshold (i.e., 140 m$^3$ s$^{-1}$). The single weather simulation produces solely a deterministic value, which does not reflect the uncertainty of the initial conditions, whereas ensemble simulations enhance confidence in the prediction by incorporating the uncertainty. The exceedance probability of 11/11 by the ensemble simulations is numerically the same as the probability of 1/1 by a single simulation. However, the physical implications of these two values are different in terms of confidence and significance.

All the dam inflow simulations, however, show that the second and third peaks of the inflow are much smaller than indicated by the observations. In the actual flood event, the so-called Tadashigaki operation was implemented at around the time of the second and third peaks. In the Tadashigaki operation, the dam outflow has to equal the inflow to avoid dam failure as the water level approaches overtopping of the dam body. The runoff simulations did not reproduce such
a critical situation this time because the second and third discharge peaks are not properly reproduced. This is a deficiency of the ensemble forecast method at this time. The accumulated inflow volume to the dam of both the observation and 2 km ensemble simulation from 03:00 JST, 29 July 2011, to 03:00 JST, 30 July 2011, is shown in Fig. 15. It can be seen that the inflow volumes are somehow comparable with the observations until the first peak is observed, though the discrepancy becomes larger afterwards. This will cause critical hardship for dam operation if the ensemble flood prediction were used in isolation, especially after the first peak.

Figure 15. Inflow volume into the reservoir based on observation and 2 km ensemble simulations.

Figure 16. Examples of the position shifts of the ensemble rainfalls.
6.3 Ensemble rainfall–runoff simulations with position-shifted rainfall

Numerical weather prediction have inevitable forecast errors. The current case has a large amount of accumulated rainfall within a limited area and is sensitive to the position error. Although ensemble simulation represents the uncertainty to some extent, the ensemble spread tends to be under-dispersive because of imperfect model/initial condition representations and limited ensemble sizes. Duc et al. (2013) verified the spatial–temporal fractions skill score of 10 km/2 km ensemble forecasts for heavy rainfall events occurring over central Japan from 3 July 2010 to 2 August 2010. They showed that a spatial scale of 60 km (positional lag of 30 km) should be considered to obtain a reasonable reliability from a high-resolution ensemble forecast. Thus, it is important to take into account the position error within a reasonable distance before input to the runoff model.

To improve the ensemble rainfalls in quantity and timing, the cumulative rainfalls of each ensemble member are calculated and the rain distribution is translated within 30 km from the original position so that the catchment-averaged cumulative rainfall for the Kasahori dam maximizes. The analysis is carried out using the 2 km resolution, 30 h rainfall after the simulation. This position change corresponds to consideration of a 30 km positional lag to detect a risk of the maximum rainfall amount. Figure 16 shows the examples of the position shifts for cntl, m02, p03, and p04. Although the ensemble forecasts produce high cumulative rainfall, the original peak lies to the south of the Kasahori dam in all four members shown in Fig. 16. Figure 17 shows the spatial distribution of the position-shifted cumulative ensemble rainfalls with the 2 km resolution. Comparing Figs. 13 and 17, it is apparent that the rainfall intensity becomes higher. The simulated discharges with these position-shifted rainfalls are shown in Fig. 18. Figure 18 indicates that the first peak discharge simulated becomes high enough compared with the observed discharge. Timing of the first peak is also improved, and, in particular, some members reproduce the exact timing. Figure 18 shows the ensemble mean of the discharge as well since the ensemble mean becomes more informative compared to that in the experiment without position shifting. Figure 19 shows the inflow volume into the reservoir based on the observation and position-shifted ensemble simulations; the simulated inflow volume becomes comparable to the observed inflow volume. These results indicate that the ensemble rainfall simulation with position shift brings better...
performance although testing with more cases is desirable to confirm that.

As indicated in Sect. 5.2, it is known that ensemble weather simulations can be useful in adding value to weather forecasts. In the current operational weather forecasting, it is not necessarily expected that the weather will be predicted accurately for any specific location. However, accurate prediction over dam catchments is the main concern of river dam administrators. In this regard, this paper shows clearly that although the original 2 km prediction forecast provides much better results than that with the 10 km resolution prediction, greater accuracy is still desirable. For example, in dam/reservoir operations, the reliable prediction of the peak timing, flood duration, and runoff volume is extremely important parameters necessary to avoid erroneous operation. The results with original ensemble rainfalls here do not match the current requirements; however, the position-shifted 2 km resolution ensemble rainfall could be a useful tool for supporting operational decisions after statistical validation with various rainfall events, which would not be possible based on previous simulations with coarser resolutions.

7 Concluding remarks and future aspects

This paper presents an example of short-term (lead times of less than a day) ensemble flood forecasting for a typical small-scale dam catchment in Japan. The Kasahori dam catchment (approx. 70 km$^2$) in Niigata, Japan, was selected as the study site. Japanese river catchments tend to be small and, thus, floods in such catchments are often in the category of flash flood of continental rivers. In other words, the rain-
fall over the small catchments and associated flood processes are too rapid to be captured well by coarse-resolution NWP models. Thus, JMA-NHM with the 2 km resolution was used to simulate the rainfall over the catchment. As the result, all 11 × 2 ensemble simulations (i.e., 10 and 2 km resolutions) predicted that the dam inflow would exceed the flood discharge of 140 m$^3$s$^{-1}$, which is the threshold quantity for flood control. However, only one out of 11 × 2 (2 and 10 km resolutions) ensemble predicted discharges, based on the ensemble rainfalls, reproduced in a broad sense the first peak of the observed discharge of the historically rare flood that occurred on 28–30 July 2011 with a 4 h lag in the occurrence time. Nevertheless, this is considered insufficient for the dam operations. In contrast, the position-shifted ensemble flood simulations (Sect. 6.3) show much better results and become comparable to the observation, indicating the importance of appropriate treatment of forecast uncertainties.

One of the strengths of the current study is the use of cloud-resolving ensemble NWPs. However, the cloud-resolving ensemble forecast is still too expensive for operational NWPs. Although this limits the number of experiments and their experimental periods in the current study, some previous studies also reported experimental use of similar NWP-based quantitative precipitation forecasts (QPFs) in the flood forecasting. For example, Yu et al. (2015) showed an improvement of rainfall and flood forecasting by blending NWP-based and radar-based QPFs. Their target was typhoon Talas of 2011 over the two catchments, Futatsuno (356.1 km$^2$) and Nanairo (182.1 km$^2$), of Shingu river basin (2360 km$^2$), Japan. In contrast, the target site (72.7 km$^2$) of the current study is much smaller and the target weather system is more confined in space, implying that the current case can be more challenging than the cases in Yu et al. (2015).

As far as we recognized, this study is the first trial of applying NWP-based ensemble QPF to such a small dam catchment of less than 100 km$^2$, showing potential benefits and difficulties at this spatiotemporal scale. Particularly, we demonstrated that the position error correction of ensemble QPF plays an important role for reliable flood forecasting. As a matter of course, bias correction and blending radar-based and NWP-based QPFs will also improve QPF up to lead times of several hours (e.g., Sun et al., 2014) and would bring further improvements in flood forecasting. For instance, Bowler et al. (2006) showed a good example of short-term blending ensemble rainfall prediction. Further research for the generalization of the proposed method and validation with more cases are needed. Likewise, the study of the optimization for the dam operational rule remains as the future work.

In any case, overall results are considered on some level helpful for decision-making related to flood control, especially as a supporting tool in addition to discharge observations and forecasting with radars. Likewise, improving the accuracy of original rainfall forecasted by high-resolution state-of-the-art numerical models, dense observation networks, and advanced data assimilation techniques is still essential.

8 Data availability

JMA-NHM is available under collaborative framework between MRI and related institute or university. Likewise, the
DRR model is available under collaborative framework between Kobe, Kyoto Universities and related institute or university. The JMA’s operational analyses and forecasts, radar rain gauge analyses, and radar composite analyses can be purchased at http://www.jmbsc.or.jp/. The rain gauge data were provided by MLIT, Niigata Prefecture and JMA.

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