



Flood occurrence mapping of the middle Mahakam lowland area using satellite radar

H. Hidayat^{1,2}, D. H. Hoekman³, M. A. M. Vissers⁴, and A. J. F. Hoitink^{1,5}

¹Hydrology and Quantitative Water Management Group, Wageningen University, Wageningen, The Netherlands

²Research Centre for Limnology, Indonesian Institute of Sciences, Cibinong, Indonesia

³Earth System Science and Climate Change Group, Wageningen University, The Netherlands

⁴SarVision BV, Wageningen, The Netherlands

⁵Institute for Marine and Atmospheric Research Utrecht, Department of Physical Geography, Utrecht University, Utrecht, The Netherlands

Correspondence to: H. Hidayat (hidayat.hidayat@wur.nl)

Received: 2 December 2011 – Published in Hydrol. Earth Syst. Sci. Discuss.: 22 December 2011

Revised: 16 May 2012 – Accepted: 6 June 2012 – Published: 3 July 2012

Abstract. Floodplain lakes and peatlands in the middle Mahakam lowland area are considered as ecologically important wetland in East Kalimantan, Indonesia. However, due to a lack of data, the hydrological functioning of the region is still poorly understood. Among remote sensing techniques that can increase data availability, radar is well-suitable for the identification, mapping, and measurement of tropical wetlands, for its cloud unimpeded sensing and night and day operation. Here we aim to extract flood extent and flood occurrence information from a series of radar images of the middle Mahakam lowland area. We explore the use of Phased Array L-band Synthetic Aperture Radar (PALSAR) imagery for observing flood inundation dynamics by incorporating field water level measurements. Water level measurements were carried out along the river, in lakes and in peatlands, using pressure transducers. For validation of the open water flood occurrence map, bathymetry measurements were carried out in the main lakes. A series of PALSAR images covering the middle and lower Mahakam area in the years 2007 through 2010 were collected. A fully inundated region can be easily recognized on radar images from a dark signature. Open water flood occurrence was mapped using a threshold value taken from radar backscatter of the permanently inundated river and lakes areas. Radar backscatter intensity analysis of the vegetated floodplain area revealed consistently high backscatter values, indicating flood inundation under forest canopy. We used those values as the threshold for flood occurrence mapping in the vegetated area.

1 Introduction

Flood maps that include the occurrence of flooding are an important source of input for integrated assessment of flood dynamics, ecological processes, and vulnerability required in planning, designing, and operating flood works, nature reserves, and land management policies (Qi et al., 2009). Regular acquisition of remotely sensed inundation extents allows for mapping flood occurrences over a large area. In areas with less cloud cover, data from optical sensors such as Landsat imagery can be used for this purpose (e.g. Qi et al., 2009; Ran and Lu, 2012). Ran and Lu (2012) noted that images with cloud cover of less than 5 % are preferred. However, for the humid tropics area, the preferred limit of cloud cover is hard or even impossible to be satisfied.

Radar imagery is useful for the identification, mapping, and measurement of streams, lakes, and wetlands in humid tropical areas as it is unconstrained by cloud cover (Romshoo, 2006; Henderson and Lewis, 2008; Hoekman, 2009). Most surface water features are detectable on radar imagery because of the contrast in returns between the smooth water surface and the rough land surface (Lewis, 1998). The advantages of using space-borne radar in environmental monitoring in these areas are that radar measurements can remotely acquire data in poorly accessible areas with 24 h per day functioning, and to a certain extent, that radar can penetrate vegetation cover. The latter functionality allows to observe flooding under a closed forest canopy. Detection of flooded forest is enabled by the bright appearance

of the inundated forest on radar images as a result of double bounce reflections between water surfaces and tree trunks or branches (Hess et al., 1990).

The Phased Array L-band Synthetic Aperture Radar (PALSAR) is one of the remote sensing instruments onboard of the Advanced Land Observing Satellite (ALOS). PALSAR is a polarimetric instrument operating at a wavelength of 23.6 cm with a 46-day satellite cycle period for global environmental monitoring (Rosenqvist et al., 2007). The modes of observation include Fine Beam Single Polarization (FBS), Fine Beam Dual Polarization (FBD), Polarimetric (POL), and ScanSAR. FBS and FBD are designed for land cover changes and forest monitoring. POL is dedicated to research related to polarimetry and polarimetric interferometry. ScanSAR is intended for seasonal phenomena studies, such as inundation extent monitoring and rice-field mapping. All land areas on the globe are covered at least once every year by the FBS, FBD, and ScanSAR modes. The typical repetition frequency for most areas is three to five times per year (Rosenqvist et al., 2007).

The application of remote sensing and GIS plays an important role in filling the gaps in wetland inventory and could reduce uncertainties due to data availability constraints (Rebello et al., 2009). Radar data sets are gradually obtaining a more prominent role in wetland mapping, not only in scientific projects but also in operational practices that require information on wetlands' presence, extent, and conditions (Henderson and Lewis, 2008). SAR technology has been used for remote monitoring of inundation patterns, duration of hydroperiods and computation of surface water level changes in the Everglades wetlands in Florida (Bourgeau-Chavez et al., 2005; Wdowinski et al., 2008). Rosenqvist and Birkett (2002) investigate the use of JERS-1 SAR mosaics in the Congo river basin for hydrological application. They proposed that SAR mapping missions need at least three repetitive observations within one year to describe the hydrology in such a complex region. Recently, Salvia et al. (2011) used Envisat ASAR and AMSR-E data to estimate flooded area extent and mean water level in the wetlands of the Parana River Delta floodplain. They found that both active and passive microwaves data can be used to estimate water level in flooded vegetation. The specular reflection of microwaves from open smooth water bodies resulting in dark tones on the SAR image can be used to evaluate 1-D or 2-D flood inundation models (Schumann et al., 2008).

The suitability of radar techniques for spatial characterization of floods (e.g. Hess et al., 1990; Oberstadler et al., 1997) draws researchers' interest to explore and apply methods for extracting flood information from satellite radar images. Based on backscattering coefficients of the L-band JERS-1 SAR images, Hess et al. (2003) used a pixel-based classifier to map wetland vegetation and flooding state for the central Amazon basin. They delineated wetlands and classified the masked area into "cover states" that consisted of vegetation cover classes and inundation states for high- and low-water

conditions. They obtained a cover-state map for large regions with an accuracy of 78–91 % for open water, nonflooded forest, and flooded forest cover-state classes. Lower accuracies were reported for aquatic macrophytes and for flooded woodland. Hostache et al. (2009) proposed a SAR image analysis method for spatiotemporal characterization of flood events that includes extracting flood extent limit, estimating water level, and constraining water level estimates using hydraulic coherence concepts. The method was applied to an ENVISAT SAR image during the January 2003 flood of the Alzette river. They showed that SAR imagery offers the possibilities to obtain distributed remote-sensing-derived water levels over a large area with sufficient accuracy for calibration of a hydraulic model. Matgen et al. (2011) proposed a hybrid approach to automatically extract flood extent from SAR images by estimating the statistical distribution of open water backscatter values from SAR images of floods, radiometric thresholding to extract the core of the water bodies, and region growing to extract all water bodies. They proposed a change detection procedure using pre- or post-flood SAR reference images to remove over-detection of inundated areas. The methods were evaluated through the 2007 flood of the Severn river (using ENVISAT SAR images) and the 1997 flood of the Red river (using RADARSAT-1 images). Their study showed that the automated method that includes a change detection procedure yields the same performance as optimized manual approaches.

This study explores the use of PALSAR imagery for floodplain dynamics mapping of the Mahakam River Basin (MRB). The MRB is located in Kalimantan, the Indonesian part of Borneo, between 2° N to 1° S and 113° E to 118° E. It represents a poorly gauged meso-scale river basin with a complex land cover mosaic. Part of the runoff that feeds the Mahakam is derived from peat domes, which are difficult to monitor. In studies at the plot scale, piezometers are commonly used to analyze water levels in peat, which are labour intensive to maintain (Devito et al., 1996; Baird et al., 2004; Fraser et al., 2001). SRTM data, time series of radar images and field measurement can be combined to study the temporal dynamics of lake water mixtures, and fluxes between the river and the floodplain adjacent to the river (Bonnet et al., 2008). Using the difference in mean brightness levels from SIR-B L-band radar images of coastal lowlands in East Kalimantan, Ford and Casey (1988) characterized three classes of forest canopy (swamp, coastal lowland forest and tidal forest) and two classes of open land cover (wetland and clear cuts). Ford and Casey (1988) also noted the enhanced backscatter returns from the inundated area covered by mangrove and nipah swamps. Time series of L-band radar data such as the JERS-1 SAR, the predecessor of PALSAR, have been applied to acquire information on the hydrology in Central Kalimantan peat swamps (Hoekman, 2007).

Hoekman et al. (2010) used multi-temporal (dry and wet season) PALSAR images of 2007, which covered the whole of Borneo, for land use/land cover mapping based on the

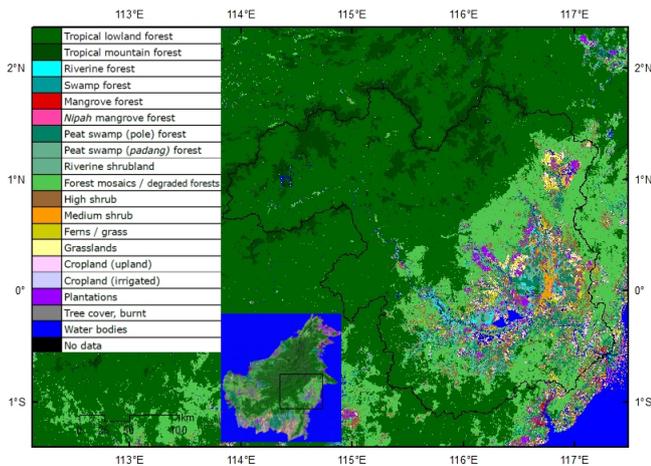


Fig. 1. Land use/land cover map of the Mahakam river basin based on PALSAR images of 2007 (Hoekman et al., 2010). The black line indicates the catchment boundary.

classification of FBS and FBD polarization (path) image pairs. The classification was performed with 18 land cover classes including seven forest types, two woodland types, two shrubland types, two grassland types and three anthropogenic vegetation types. They obtained good results for a sub-continental high resolution map, except for grassland, cropland and shrubland. Based on the confusion matrix analysis, the obtained land use/land cover (LULC) map was for 85.5 % fully in agreement with the independent reference dataset, for 7.8 % in partial agreement, while 6.7 % was in disagreement. Herein, partial agreement is defined as confusion with an adjacent class along a continuum with a fairly similar biophysical characterization, which may be related to the dynamic behaviour of low biomass areas (Hoekman et al., 2010). For a wide coverage area of about 750 000 km² and a relatively detailed classification, the overall result is promising. Application of longer or denser time series were expected to improve results significantly. Figure 1 shows the LULC map of Borneo, zoomed in on the MRB area. The middle Mahakam region has a complex land cover mosaic that mainly consists of degraded forest, riverine forest, shrub, agriculture area, swamp and peat forest, which is strongly related to the inundation pattern and anthropogenic disturbances.

In this study we aim to extract flood extent and flood occurrence information from a series of radar images of the middle Mahakam lowland area, including both open water and areas under vegetation. We mapped the flood occurrence as counts of flood marks obtained from the evaluation of radar backscatter from PALSAR data series against flooding thresholds. The remainder of this paper is structured as follows. Section 2 describes the study area and a brief hydrological background of the middle Mahakam area. Descriptions of data collection and data processing are presented in

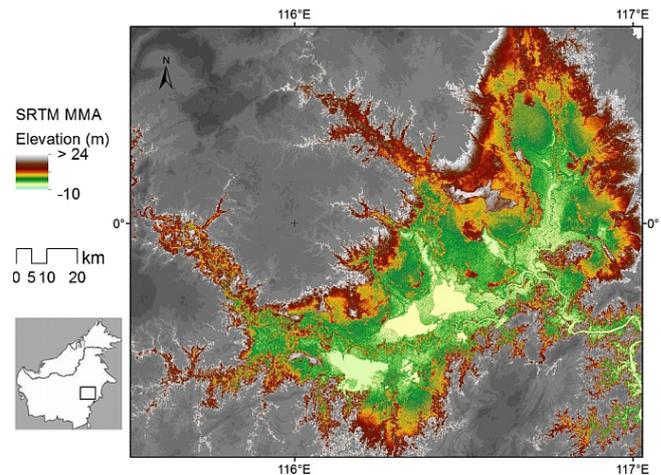


Fig. 2. Study area in East Kalimantan with a colour coded SRTM digital elevation model for the middle Mahakam lowland area.

Sect. 3. Section 4 presents the results discussion and finally Sect. 5 presents the conclusions.

2 Study area

The Middle Mahakam Area (MMA) is characterized by low relief (Fig. 2) with around 40 shallow lakes on both sides of the river. Lakes Semayang, Melintang, and Jempang are the three largest lakes in this region. Lakes in the MMA regulate the discharge in the lower Mahakam area. They have a function as a buffer by storing water during the high flow conditions, and releasing it during low flows. Lake filling and emptying mechanisms play a role to shave water level peaks downstream of the Mahakam lakes area (Hidayat et al., 2011b). The MMA is also surrounded by peatland, as part of the Kutai lowland spreading over an area of 35 km NW–SE by 130 km SW–NE, with elevations of around sea level in the Mahakam River to ca. 24 m a.s.l. (Hope et al., 2005). The peat in Kalimantan is of the ombrogenous type (Jaenicke et al., 2010; Page et al., 2004). This type of land potentially stores more water than any other type of land because of the sponge nature of the pores.

The climate in Kalimantan is strongly influenced by the Indo-Australian Monsoon driven by the Inter-tropical Convergence Zone, and El Nino-Southern Oscillation (Meehl and Arblaster, 1998). A record from a meteorological station in Kotabangun in the central part of the MMA showed that in general, the mean daily temperature varies between 24 and 29 °C, relative humidity between 73–99 % and the mean annual precipitation is approximately 2300 mm. During El Nino-Southern Oscillation (ENSO) years such as in 1997, precipitation can be as low as nearly half the mean annual value. Due to the global air circulation and the regional climate, the Mahakam catchment has a bimodal rainfall pattern with two peaks of rainfall, which generally occur

in December and in May. The normal dry season lasts from June to September. However, it may begin as early as March and last until November, as was the case during the ENSO event in 1997.

The flooding regime in the MMA is generally characterised by long duration floods during the peak of the rainy season in December through January and in May, with several short duration high water events in between. Water levels at the upstream stations are fluctuating primarily in response to rainfall in the catchment, while more downstream they are also influenced by the tidal motion. Bank overtopping occurs during a flood situation in Penyinggahan (Fig. 3). During this period, the Penyinggahan floodplain is flooded and water flows through the floodplain to Lake Melintang, then to Lake Semayang, and finally meets the Mahakam again through a tie-channel. Downstream of the middle Mahakam area, water level fluctuation is relatively low. Beside the lakes, the vast area of the Kutai wetland is believed to also control the River Mahakam water level and discharges downstream. The information on flood duration and inundation extent obtained in the context of the present study will be used in a future stage to model the hydrological functioning of the area, which has not been quantified to date.

3 Methodology

A series of PALSAR images with HH polarization and a pixel spacing of 75 m was provided by the Japan Aerospace Exploration Agency (JAXA), covering the middle and lower Mahakam areas for the years 2007 through 2010 (Table 1). The PALSAR images were radiometrically calibrated, orthorectified using 3'' SRTM data and corrected for slope illumination effects. Next, these geocoded data were chronologically stacked into a layered multi-temporal radar image suitable for time-series analysis. Figure 4 shows the spatial and temporal dynamics of radar backscatter from PALSAR images of the study area during wet and dry seasons. These dynamics are related to flood conditions and soil moisture.

The flood occurrence was determined by evaluating pixels in the images used as input against the lower and the upper threshold values. The pixel was flagged as flooded if its backscatter value falls within the range between the lower and upper thresholds. An image with pixel values of counts of the flooded flag was obtained, which was then colour mapped. As radar returns for flooding in open areas and for flooding under vegetation result in contrasting behavior (dark for the former and bright for the latter), the two types of flooding were mapped separately. The crucial part in the thresholding step is to determine the upper threshold value for open water and the lower threshold value for flooding under vegetation cover. From the input images we acquired radar backscatter statistics in regions covering the main river and lakes that are known to be permanently inundated, to determine threshold values for open water flood occurrence



Fig. 3. (top) River bank overtopping in Penyinggahan during the high water event in March 2009. The red line indicates water level marks on trees, from a previous flood event. (bottom) A view to the floodplain upstream of the MMA during low water event in August 2009. Flood marks on trees (red rectangle) indicate the height of the previous inundation in the vegetated floodplain area.

mapping. The minimum was taken as the lower threshold, and mean plus one standard deviation was taken as the upper threshold. From the statistics of the region of interest, we obtained the threshold value of -25.1 to -11.2 dB for open water flooding detection. During the field campaign, we carried out an extensive bathymetry survey in the Mahakam River, yielding cross-river depth profiles with an interspacing of about 200 m. We used these bathymetry data for the assessment of flood occurrence threshold values. We also carried out a stream reconnaissance, which has provided a qualitative confirmation of the results presented. Since the river width is about 300 m and the pixel size is 75 m, continuous river bank lines should occur in the flood occurrence map, along the boundaries of river areas with maximum flood counts. This is indeed the case. Figure 5 shows flood occurrence maps obtained using different threshold values, illustrating that the threshold we chose results in a good correspondence

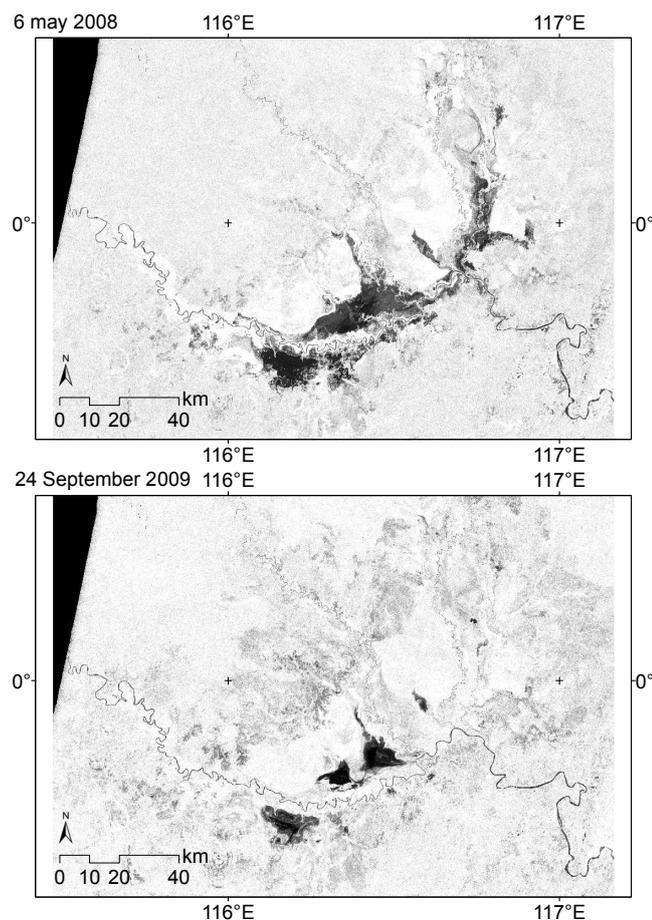


Fig. 4. PALSAR images of the middle Mahakam area during a wet season (top) and a dry season (bottom). Dark areas indicate open water extent, bright areas indicate inundation under vegetation cover.

between the river banks lines from bathymetry mapping and those apparent in the flood occurrence map.

It is generally accepted in SAR literature that flooded vegetation results in enhanced backscatter returns (e.g. Ford and Casey, 1988; Hess et al., 1990; Townsend, 2002). Our field observations confirmed that floodplain areas in Penyinggahan showing bright returns are indeed flooded during high water conditions (Fig. 3 top panel). During our stream reconnaissance survey further upstream of the MMA, we also noted several flood marks on trees that provide an indication of the maximum height of the previous inundation in the vegetated floodplain area (Fig. 3 bottom panel). The latter inundation event resulted in a consistently enhanced radar backscatter, as further discussed in Sect. 4.2. For flooding under vegetation cover, we determined the threshold from backscatter statistics obtained from about 7900 pixels sampled from the floodplain region upstream of the MMA that was mainly covered by riverine forest, riverine shrubland, and cropland classes. Taking the mean as the lower threshold and the maximum as the upper threshold, we obtained

Table 1. PALSAR images used in this study.

Image No.	PALSAR data dates
1	19 March 2007
2	4 May 2007
3	4 August 2007
4	19 September 2007
5	4 November 2007
6	20 December 2007
7	4 February 2008
8	21 March 2008
9	6 May 2008
10	21 June 2008
11	6 August 2008
12	21 September 2008
13	22 December 2008
14	9 May 2009
15	24 September 2009
16	9 February 2010
17	27 March 2010
18	12 May 2010
19	27 June 2010
20	27 September 2010

the threshold values -7.52 and -2.26 dB for detection of flooding under vegetation cover. Based on Hope et al. (2005), the upstream boundary of the flood occurrence maps are at 24 m a.s.l. (see Fig. 2), which is considered to be the highest elevation in the MMA. The downstream end of the mapped area is the region of Senoni village, where the relief shifts to a steeper terrain that marks the transition to a different geological setting in the downstream region.

Speckle filtering was applied prior to flood occurrence mapping. In a preliminary flood occurrence map derived from unfiltered images, contours of the river and lake extent, which coincide with the circumferences of the areas with maximum flood occurrence, were rather noisy. The enhanced Lee filter was used to reduce speckle in the radar images while preserving texture information. We applied the filter using a 3 by 3 pixel window. The enhanced Lee filter uses coefficients of variation within an individual filter window. All pixels are divided into three classes: homogeneous, heterogeneous, or point target. Each class type is treated differently. The pixel value is replaced by the average of the filter window, replaced by the weighted average, or is not changed for homogeneous, heterogeneous, and point target classes, respectively (Lopes et al., 1990).

Pressure transducers to measure water levels were installed in the peat forest near Lake Melintang, at the shrub-covered peat swamp floodplain in Penyinggahan, along the river in Melak (upstream of MMA) and in Muara Kaman (downstream of MMA), and in the main lakes (Jempang, Semayang, Melintang). Local flood levels at the water level gauge locations were determined by evaluating the relative

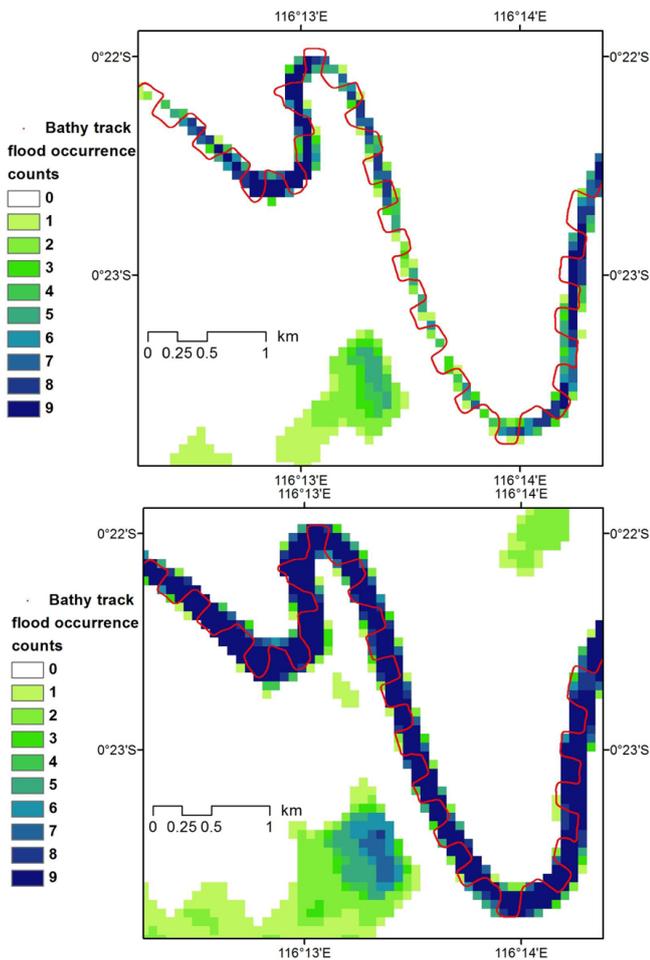


Fig. 5. Bathymetry measurement tracks in the Mahakam River plotted on open water flood occurrence maps using the upper threshold values of -14.9 dB (top) and -11.2 dB (bottom).

position of the pressure transducers and water levels to the ground surface. Bathymetry data of lakes were collected using a single beam echosounder.

4 Results and discussion

4.1 Water level and radar backscatter relationship

An analysis of the relationship between radar backscatter and water levels is reported in Hidayat et al. (2011a). For lakes and shrub covered floodplain peatland, where the range of water level variations was large, high water level-backscatter correlations of 0.87–0.98 for the lakes and 0.72–0.99 for the floodplain peatland were obtained. In forest covered peatland subject to a small range of water level variation, water level-backscatter correlations were poor ($r = 0.18$ –0.48). This poor correlation between water level and radar backscatter for this area could be related to the limitation of PALSAR over certain thresholds of biomass on the forest canopy. The

backscatter changes induced by water level variation at locations of water level gauges are likely due to vegetation succession related to inundation dynamics in the study area. This is reflected by high correlations at gauges located in the lakes and floodplain areas with seasonal cycle shrub-type vegetation, and low correlations at gauges located in the peat forest.

Radar backscatter intensity depends on land cover type. The land use/land cover map by Hoekman et al. (2010) identified 19 classes of land cover types including water bodies. Figure 6 shows time series of radar backscatter intensity for several key land cover types in the middle Mahakam area, revealing a markedly diverging response to flooding. Regularly inundated medium shrub and high shrub were very sensitive to flooding, yielding low backscatter returns when fully inundated. Other land cover classes representing vegetated areas generally feature a moderate increase of radar backscatter intensity with flood inundation.

4.2 Temporal analysis of PALSAR data

A temporal analysis of radar backscatter was carried out based on the statistics from 20 images. A standard deviation image captures the backscatter intensity standard deviation for all pixels in the image. From the maximum and minimum value of radar backscatter we obtained the range image. The average absolute difference of consecutive images resulted in the mean change image. The color composite of the latter three images (Fig. 7) reveals a zoning that provides details in an area that was poorly classified in the work of Hoekman et al. (2010), for its complexity. The map shows land cover types that are related to the dynamics of inundation, i.e. the regularly inundated shrubland, reeds/sedges growing during dry season and floating vegetation, such as water hyacinth in lake areas, and rice growing in lake areas in years when flood levels are not too high.

From the mean of three images representing the wettest conditions, we obtained a map with the maximum inundation extent of the MMA, covering inundated areas in both open water and in vegetated areas. Figure 8 shows that a vast area of the MMA is inundated during the flood conditions, including villages along the Mahakam river and floodplain area, which appear as open water. The open water inundation extent was indicated by dark radar returns. Flooding under vegetation showed bright radar returns that were colour mapped in Fig. 8 to distinguish flooding under vegetation from open water flooding.

4.3 Flood occurrence mapping of open water extent

Nine images from 2008 through 2009 were used to make a flood occurrence map, and served as the basis for accuracy assessment using field measurements taken in the corresponding period. Figure 9 shows the open water flood occurrence map of the MMA from nine filtered PALSAR images using backscatter statistics from the permanently inundated

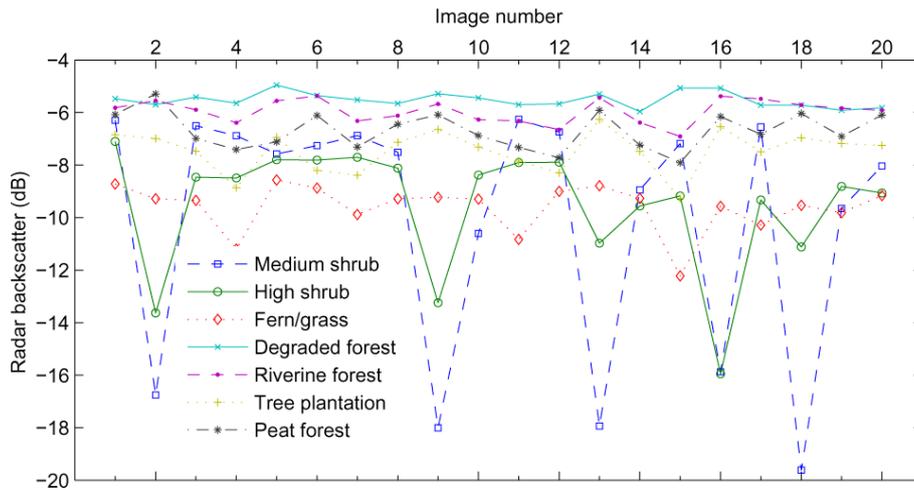


Fig. 6. Radar backscatter mean values of land cover classes in the Mahakam lakes area. Image numbers correspond with image acquisition dates in Table 1.

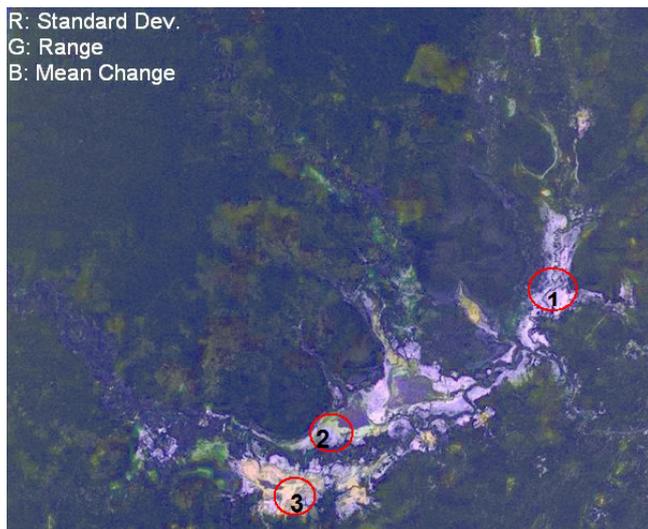


Fig. 7. Results of a temporal analysis of 20 PALSAR images collected in the period between 19 March 2007 and 27 September 2010. Light purple indicates the regularly inundated shrubland (1), green indicates reeds/sedges (2), yellow at the southern lake indicates rice growing areas in years when flooding is not too high (3).

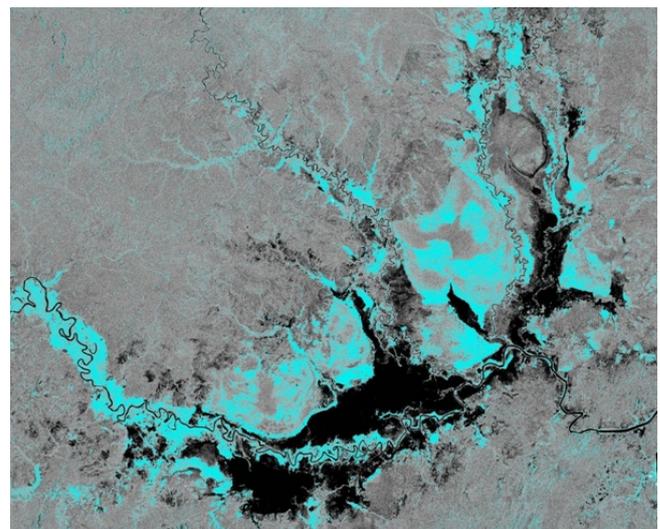


Fig. 8. Inundation extent obtained by averaging three images in wet periods (4 May 2007, 22 December 2008, and 9 February 2010). Dark areas indicate open water inundation extent, cyan indicates flooding under vegetation.

main river and lakes, by taking the minimum as the lower threshold and mean plus one standard deviation as the upper threshold. Permanently inundated lake areas and main river sections were well-mapped with the maximum probability of flood occurrence. During the peak flooding events in the rainy seasons in May 2008 and December 2008, a vast area of the MMA was inundated, including villages and cities along the river. These radar-inferred flood events were qualitatively confirmed by flood marks on houses and trees in the study area, besides water level records from pressure sensors. Quantitatively, the open water flood occurrence was

validated using data from bathymetry measurements of lakes (Fig. 10) taken during the high water period in March 2009. Sums of flood occurrence were sampled from the flood occurrence map along the bathymetric track, by extracting values on the map at the respective depth sampling point. Overall, flooding occurrence was well-correlated with lake depth (Fig. 11) resulting in r values of 0.83–0.85, deeper areas were more frequently flooded than shallower areas at the shores and near islands. At some places, emergent and floating plants could occupy large and patchy parts of the lakes, especially in the shallower area, creating abrupt changes in flooding occurrence between neighboring cells.

Table 2. Correlation between lake depth and flood occurrence mapped using different periods of PALSAR images for lake sections depicted in Fig. 10.

Period of images	# images	r_{A-B}	r_{C-D}	r_{E-F}	r_{G-H}
2008–2009	9	0.85	0.84	0.83	0.83
2007–2010	20	0.39	0.41	0.54	0.82

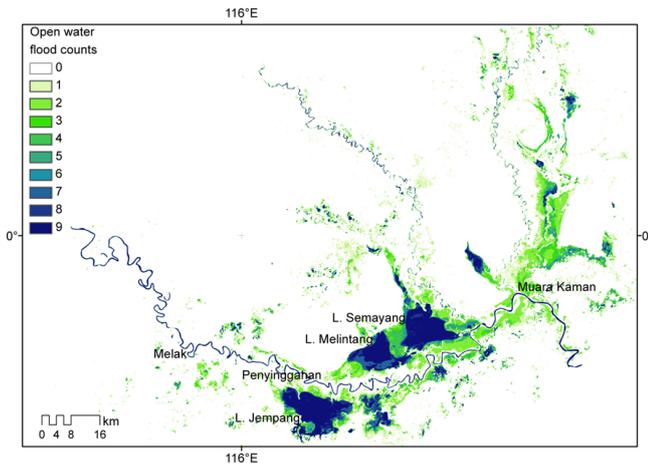


Fig. 9. Flood occurrence map of the middle Mahakam area obtained from the filtered (enhanced Lee) PALSAR images collected in 2008 and 2009.

To investigate the occurrence of floods in the MMA over a longer time period, open water flood occurrence maps were produced further using 20 images from the period between 2007 and 2010, adopting the threshold values described in the previous sections. Flooding extents are generally similar between the two flood occurrence maps. However, significant differences can be observed when zooming into the lakes area (Figs. 12 and 10). Table 2 shows the correlations between lake depth and flooding occurrence for the two maps. Depth-occurrence correlations remained relatively high for Lake Melintang ($r=0.82$) but dropped for all bathymetry sections in Lake Jempang ($r=0.39$ – 0.54). This may be attributed to the stronger influence of human activity in Lake Jempang relative to Lake Melintang. During the dry season, large parts of Lake Jempang are used by local farmers to grow rice. This agricultural activity is not suitable in Lake Melintang due to the low pH values of water derived from peat forest. Compared to 2007, 2008 was a wet year and continued to remain wet in the first half of 2009, which resulted in relatively high lake water levels so that rice growth was not possible. In the second half of 2009, however, a relatively long dry period occurred. In addition, floating and emergent aquatic vegetation has been growing fast during the dry season and, to some extent, was maintained by local fishermen in Lake Jempang as nesting grounds for fish. During the lake emptying period following the end of the flood

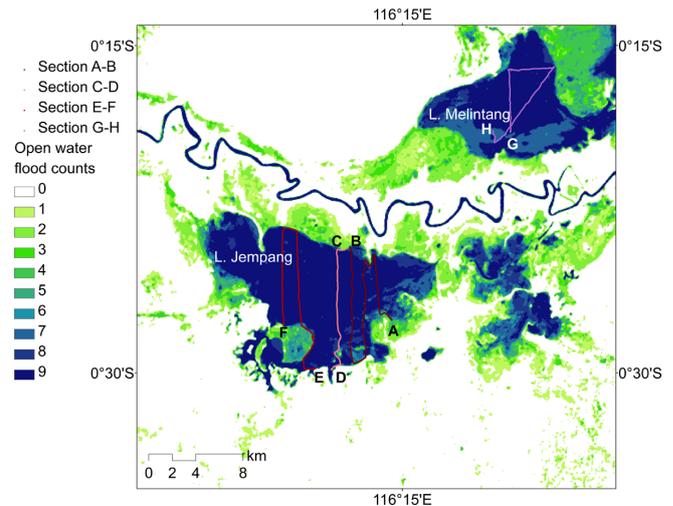


Fig. 10. Bathymetry measurement track in Lake Jempang and Lake Melintang plotted on the flood occurrence map from nine PALSAR images in 2008 and 2009.

event, large parts of the floating plants were flushed downstream as a drifting bulk of biomass (Fig. 13). These vegetation dynamics and anthropogenic factors hindered validation of flood occurrence maps using lake bathymetry data.

4.4 Flood occurrence mapping of area under vegetation

It requires a large effort to map flooding under forest canopy, considering the poor correlation between radar backscatter and water level under such circumstances. An image obtained from combining the minimum, mean, and maximum backscatter values of 20 PALSAR images from 2007 to 2010 shows a clear signature of flooding under vegetation in the Mahakam floodplain, upstream of the MMA. The flooding occurrence under vegetation was obtained by evaluating pixels in the filtered images against the lower and the upper threshold values taken from the mean and maximum of backscatter values of the regularly inundated floodplain upstream of the MMA. Figure 14 shows (ground) water level records at the peat forest near Lake Semayang, plotted along with radar backscatter values, which shows enhanced radar backscatter in this flooded forest. Backscatter values are relatively high during the wet period and relatively low during the dry period, except for the image acquired on 21 September 2008, which coincided with the dry period. For the latter image, backscatter values at the water gauging points were high for low water level conditions. This may relate to the relatively high soil moisture of the peat forest compared to the rest of the area, which could lead to a false detection of flood under vegetation. Flood detection in the shrub covered peat swamp is more problematic due to the dynamics in water levels and vegetation cover. On the one hand, shrub vegetation that are mainly reed and sedges grow and cover this area during the dry season. On the other hand, this area is fully

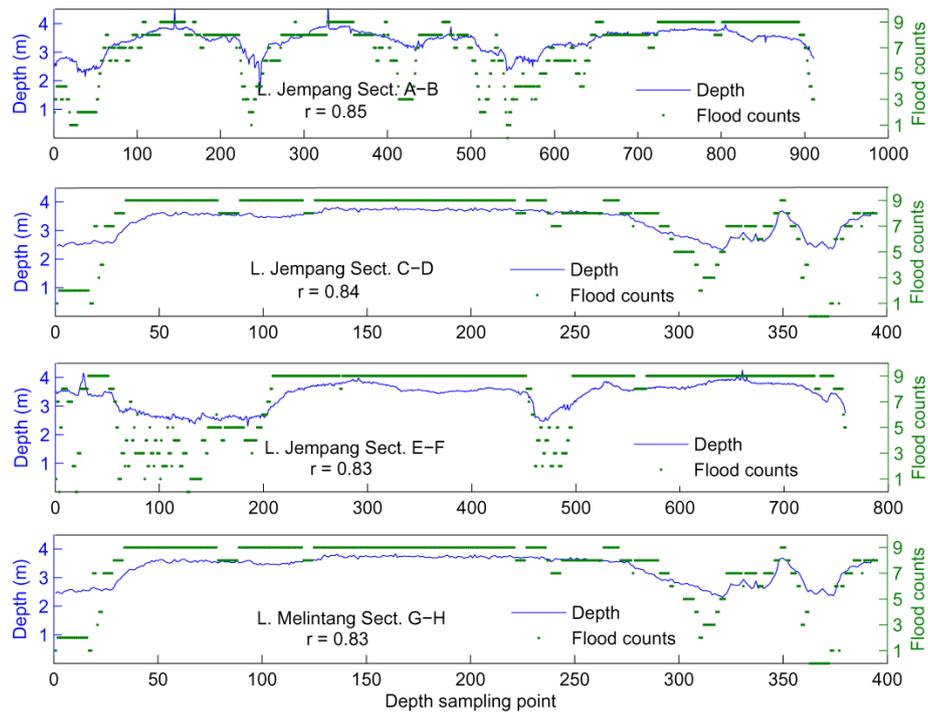


Fig. 11. Lake depth vs. flood occurrence from nine PALSAR images collected in 2008 and 2009, r values indicate correlation coefficients.

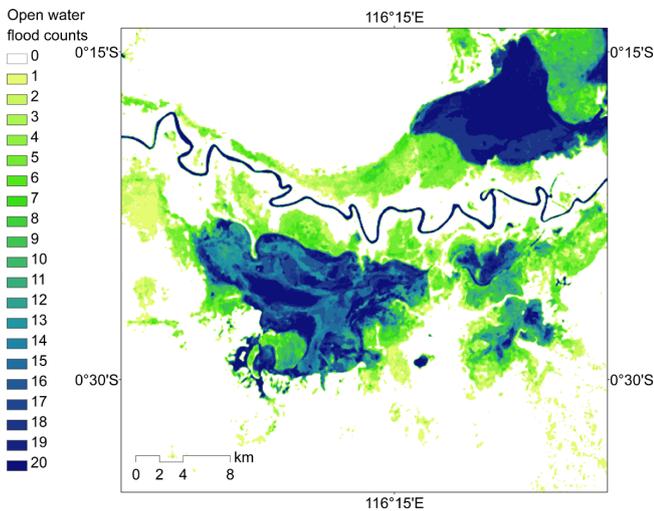


Fig. 12. Flood occurrence map from 20 PALSAR images collected in 2007 through 2010, zoomed in to the lakes area.

inundated during the high water period, such that it becomes an extension of the lakes. When the flood recedes, shrub vegetation falls down and the vegetation succession starts again. Consequently, some of the underestimation or overestimation of the extent of the inundated areas cannot be avoided, as noted previously by Romshoo (2006). Figure 15 shows the composite flood occurrence map of the MMA, overlaying results for open water and for inundation under vegeta-



Fig. 13. Bulks of floating vegetation flushed off Lake Jempang, drifting downstream during a lake emptying period.

tion. The flood occurrence of the peat swamps in Fig. 15 was mapped using the open water results, which neglects flooding of the peat swamp areas under vegetation. This problem was solved by using a simple procedure to count the total number of flood cases per pixel, which represents open water and flooding under vegetation canopy (Fig. 16). From Fig. 16, it can be inferred that at least 91 % of the mapped area was inundated during the wettest period, and only about 12 % of the mapped area was inundated during the driest period.

The flooding occurrence under vegetation estimates were validated using water level measurement taken in the peat forest near Lake Melintang, by checking if the radar-based assessments could be confirmed from the level gauges. The

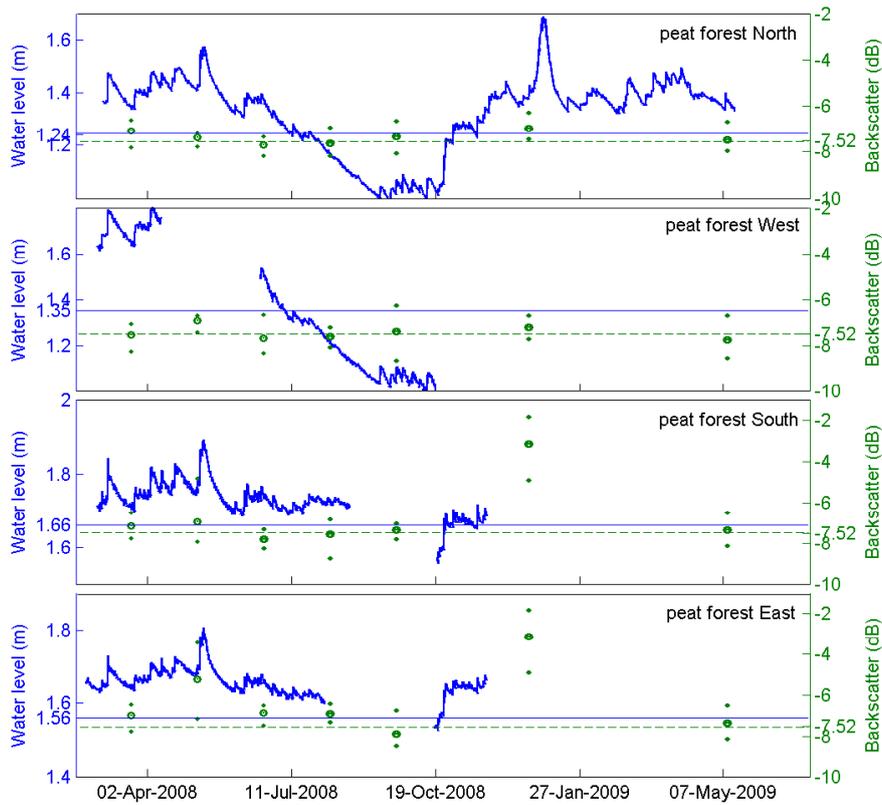


Fig. 14. Water level and radar backscatter return values obtained from a block of nine pixels at the water level gauge locations in the peat forest near Lake Melintang. Blue lines indicate ground levels at the gauge locations. Water levels are with respect to the subterrain position of each pressure sensor. Green circles, upper and lower dots indicate mean, maximum and minimum backscatter values, respectively. The dashed green line indicates the ground level for the radar backscatter return.

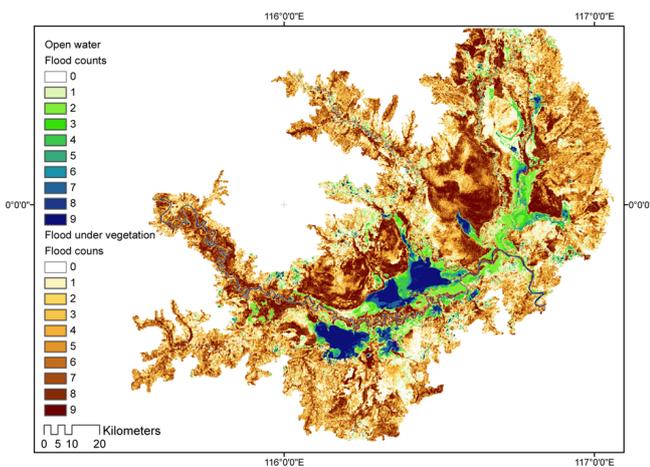


Fig. 15. Composite flood occurrence map of the Kutai wetlands from nine PALSAR images collected in 2008 and 2009 as an overlay of results for open water and for inundation under vegetation cover.

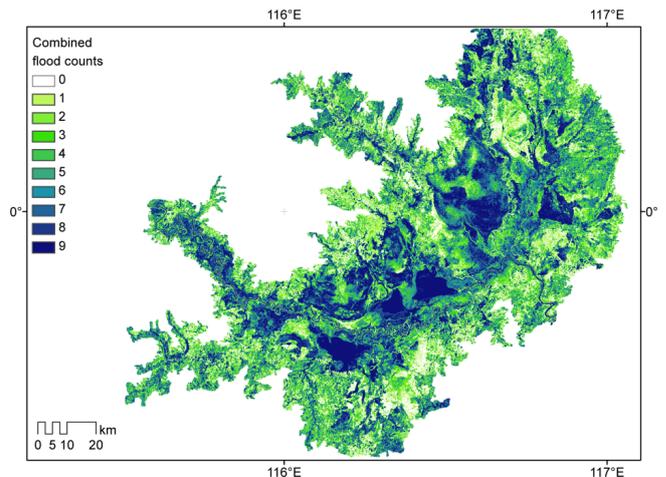


Fig. 16. Combined flood occurrence map of the Kutai wetlands from nine PALSAR images collected in 2008 and 2009, counting the total number of flood cases per pixel, which represents both open water and flooding under vegetation.

local gauge location is assumed to represent the complete pixel area, which features local topographic irregularities. Consequently, the gauge location may better represent one of the eight pixels surrounding the pixel in which the gauge is located. Figure 14 therefore shows the mean, minimum and maximum backscatter return values within a block of nine pixels at the gauge location, relative to the flooding under vegetation threshold line, and compares those values with the water level relative to the local ground level. In 78 % of the cases, radar-based assessments from one or more of the nine pixels correctly indicated the water level to be above or below the ground level. This gives merely a rough indication of the quality of the map produced, at least showing radar backscatter to offer a promising means of monitoring flooding under vegetation. A much larger suite of level gauges and detailed information about the micro topography would be required to achieve a more quantitative accuracy assessment.

5 Conclusions

Flood extent and flood occurrence information were inferred from a series of radar images of the middle Mahakam lowland area. Relative radar backscatter levels sampled in regions of interest and a land use/land cover map showed that different land cover types yield markedly different backscatter returns in response to flooding. Regularly inundated medium shrub and high shrub were sensitive to flooding, with decreasing radar backscatter during flood periods, while peat forest, riverine forest and tree plantation signatures featured a slight increase of radar backscatter with flood inundation.

Image statistics of the river and main lakes areas were used to determine thresholds to map open water flood occurrence. The minimum was taken as the lower threshold, and the mean plus a standard deviation was taken as the upper threshold. The bathymetry of lakes obtained during a high water period served as validation data for the open water flood occurrence map. High correlations between lake depth and flooding occurrence were obtained for the map derived for the wet year of 2008–2009, which is the period when field surveys were carried out. For the map derived from input images of 2007–2010, covering a longer dry period, the correlation between lake depth and flooding occurrence dropped for one particular lake, which was subject to high vegetation dynamics and rice growing activity. In a separate procedure, the average radar backscatter value of the floodplain area was used as a threshold to map flooding under forest canopy. Validation with water level data gave an indication that the flood occurrence map under vegetation cover is fairly accurate.

The flood occurrence map obtained from a newly developed algorithm offers a detailed insight into the hydroperiod and flood extent of the Mahakam lowland area, illustrating the added value of radar remote sensing to wetland hydrological studies. As a side effect, we found that simple statis-

tics of the radar backscatter maps such as mean, total range and mean change can be used to improve land classification maps. This may help future efforts to classify flat and complex wetland areas such as the one under study, where existing classification methods fall short.

Acknowledgements. This research has been supported by the Netherlands organization for scientific research (NWO grant number WT76-268). This work has been undertaken in part within the framework of the ALOS Kyoto & Carbon Initiative. ALOS PALSAR data have been provided by JAXA EORC. The help from Maximiliano Sassi, Bart Vermeulen, Fajar Setiawan, Unggul Handoko, Ulfah Karmilasari and Supriadi in data collection is gratefully acknowledged. Wawan Kustiawan and Y. Budi Sulistioadi (Mulawarman University) are thanked for facilitating the field campaign in many ways. We thank Remko Uijlenhoet for the discussion and comments on the early draft of this paper. Renaud Hostache, anonymous reviewers, and the Editor are thanked for their comments and suggestions that helped improve the manuscript.

Edited by: L. Pfister

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