

Author's response and marked up version

Editor report:

5 Thank you very much again for submitting your brief communication manuscript 'Hurricane Dorian: automated near-real-time mapping of the "unprecedented" flooding on the Bahamas using SAR'. Two referee reports are available with major comments raised regarding the scope of your work and the validation of the applied method, and accordingly major revisions are needed.

Reading your responses to these comments I am positive that a revised version of your manuscript will address these points appropriately.

10 As a next step, I kindly ask you to provide a revised marked up (track changes) version of your manuscript to make clear how you include the changes in response to the referee report. Please also take note of the specifications regarding pages, figures etc. for brief communication manuscripts (https://www.natural-hazards-and-earth-system-sciences.net/about/manuscript_types.html).

I look forward to receiving the revised version of your manuscript.

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Author's response:

We appreciate the Editor's positive view on our responses to the referee reports.

We are providing, as requested, a marked up (track changes) version of our manuscript, in which the implementation of the changes made to address the reviewers' comments is marked in red.

20 In order to meet the brief communication specifications, we reduced the abstract length to 97 words, we merged Figure 2 and Figure 3 into a unique Figure 2 (since we added Table 1), we included the sections Author Contribution and Competing Interests. Given that reviewers asked us to add more references, we are asking flexibility on the number of references allowed (we currently include 29 references).

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35 **Brief communication:**

Hurricane Dorian: automated near-real-time mapping of the “unprecedented” flooding on the Bahamas using SAR

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Abstract. ~~Lack of real time, in situ data on the extent of flooding in many parts of the world can hinder efficient disaster response. With the advent of satellite based synthetic aperture radar (SAR) sensors, we can deploy techniques to identify flooded areas worldwide while storms are occurring.~~ In this communication, we present ~~an~~ application of the automated near-real-time (NRT) system called Radar-Produced Inundation Diary (RAPID), ~~applying it~~ to European Space Agency Sentinel-1 SAR images to produce flooding maps for Hurricane Dorian in the northern Bahamas. ~~Images from RAPID maps,~~ released two days after the event, show that coastal flooding in the Bahamas reached areas located more than 10 km inland, covering more than 3,000 km² of continental area. RAPID flood estimates from subsequent SAR images show the recession of the flood across the islands, and present high agreement scores when compared to Copernicus Emergency Management Service (EMS) estimates.

1 Introduction

Hurricane Dorian was the strongest Atlantic hurricane at landfall in terms of maximum sustained winds (185 mph, 83 ms⁻¹), tied with the 1935 Labor Day Hurricane (Landsea et al., 2014). Dorian’s first record-breaking landfall occurred at 16:40 UTC and its second at 18:00 UTC on September 1, 2019, in the Abaco Islands in the northern Bahamas (NHC 2019). A third landfall occurred at 03:00 UTC on September 2 at the eastern end of Grand Bahama and was characterized by maximum sustained winds of 180 mph. Tropical storm conditions battered the northern Bahamas for 72 hours, and locations in northeastern Grand Bahama suffered hurricane conditions for more than 40 hours. Between 08:00 UTC on September 2 and 14:00 UTC on September 3, the National Hurricane Center issued 30 consecutive hourly public advisories (NHC 2019) indicating Hurricane Dorian was either moving at 1 mph or was stationary, resulting in prolonged extreme conditions over the same areas. The prolonged damaging and record-breaking winds were just one aspect of this storm - one that was measured directly and in near real time (NRT) by aircraft missions (HRD 2019). In addition, the combined effect of storm surge and heavy precipitation brought about the extensive flooding that was the major cause of “unprecedented and extensive devastation,” as described by Bahamian Prime Minister Hubert Minnis, and of deaths whose number has yet to be determined. Neither precipitation nor coastal surge could be directly measured because of the lack of a ground-based observational network.

In this and other remote areas around the world where ground-based measurements are not ~~available~~obtainable, precipitation can be assessed using near-real-time (NRT) satellite estimates available through the National Aeronautics and Space Administration (NASA) Global Precipitation Measurement (GPM) Integrated Multi-satellitE Retrievals for GPM (IMERG), version 06 (Huffman et al. 2019). Without these automated estimates, gaining prompt situational awareness becomes difficult, which can cause delays in rescue operations.

Several space missions acquire synthetic aperture radar (SAR) data derived from low Earth observation (LEO) satellites that can be used for NRT systems: Sentinel-1, from the European Space Agency (ESA; Torres et al. 2012), TerraSAR-X, from the German Aerospace Center (DLR; Werninghaus and Buckreuss 2009), CosmoSkyMed, from the Italian Space Agency (ASI; Covello et al. 2010), RADARSAT-2, from the Canadian Space Agency (CSA; Morena et al. 2004), and Alos-2, from the Japan Aerospace Exploration Agency (JAXA; Kankaku et al. 2013). Among these, only Sentinel-1 provides public access to SAR data that can be used to estimate flooding.

The nature of LEO satellites makes SAR data sparse, however, and, unlike with precipitation products, gaps cannot be filled using geosynchronous satellites. In fact, the spatial resolution needed for accurate flooding maps is three orders of magnitude higher than that required for a global precipitation product. Even when observations are present, no detailed processing methods exist for real-time retrieval of flooding data because underdetection or overdetection issues necessitate manual labor (Shen et al. 2019a).

Recently, we published the RADar-Produced Inundation Diary (RAPID) NRT fully automated system (Shen et al. 2019a), which involves the processing of high-resolution (10 meters) SAR images to allow the creation of rapid and efficient flood inundation maps by addressing both underdetection and overdetection. Differently than optical sensors, SAR images are not influenced by adverse weather conditions. As discussed in the methodology section below, the system is triggered (Yang et al. 2019) by IMERG precipitation estimates (Huffman et al. 2019), and it processes Sentinel-1 SAR data.

In this brief communication, we present the early results we delivered with the RAPID NRT automated system two days after Hurricane Dorian hit the Bahamas, just a few hours after SAR data for the event became publicly available. We provide a short description of the methodology, and we detail the extent of flood inundation by analyzing RAPID maps.

90 2 Methodology

Only a few handful SAR-based flood delineation methods (e.g. Horritt et al. 2003, Martinis et al. 2009, Matgen et al., 2011, Giustarini et al. 2012, Lu et al. 2014, Chini et al. 2017, Cian et al. 2018) have the potential to be fully automated (Shen et al. 2019b). The RAPID algorithm (Shen et al. 2019a) is an automated system capable of producing NRT inundation maps by processing SAR observations. Because of the considerable computation, storage, and data transfer time that would be needed to run the RAPID algorithm blindly for every SAR image worldwide, we implemented a zoom-in triggering mechanism (Yang et al. 2019) that allows selection of areas of the world where flooding is possible. Areas are defined by the availability of SAR images associated with land that has received at least 60 mm of accumulated precipitation during the previous day or potentially fluvial flood areas indicated by hydrological station observations. Within the continental United States (CONUS) area, we used the National Oceanic and Atmospheric Administration (NOAA) NEXt-Generation RADar (NEXRAD) precipitation product (NOAA 1991) and the U.S. Geological Survey (USGS) WaterWatch ~~system~~ (<https://waterwatch.usgs.gov/>), ~~and w~~We used NASA's IMERG, version 06 (Huffman et al. 2019), for the rest of the world (Yang et al. 2019).

After being triggered, the RAPID core algorithm (Shen et al. 2019a) handles both polarizations of SAR images in GRD mode through four steps: (1) identification of water and land pixels through a binary classification; (2) selection of water pixels connected to known water bodies and water areas not connected to known water bodies; (3) generation of a buffer region around the identified water bodies to reduce false negatives using less restrictive thresholds derived from the radar noise model; and (4) correction of the classification through a machine-learning algorithm that uses high-resolution topography (Farr et al. 2007), hydrography (Yamazaki et al. 2019), water occurrence (Pekel et al. 2016), and river bathymetry (Allen and Pavelsky 2018, Chen et al. 2019, Yamazaki et al. 2014) (Shen et al. 2019a). In step (2), the noise-free persistent water extent (know water body) is computed using at least 5 dry overpasses for each pixel.

The RAPID system has been quantitatively validated in past studies against manually derived flood maps using (overall, user, producer) agreement scores, representing (accuracy, true positive rate, precision) parameters of the confusion matrix. Specifically, for Hurricane Harvey, RAPID was validated against the DFO comprehensive flood map of August 30, 2017 (Shen et al., 2019) and against the USGS DSWE Northwestern flood map of June 25, 2019 (Yang et al. 2019). RAPID yielded consistently high agreement scores for Harvey (93%, 75%, 77%) and the Northwestern flood (96%, 84%, 76%). For

Hurricane Dorian, we are presenting a comparison between RAPID and the Copernicus Emergency Management Service (EMS) first estimate maps (available at <https://emergency.copernicus.eu/mapping/list-of-components/EMSR385/FEP/ALL>), both derived from the Sentinel-1 SAR observations. EMS flooding maps are not available for the entire SAR images, but only for the Abaco Islands on September 2, 2019, and for Grand Bahama on September 4, 2019.

120 3 Results

Because of the extremely high amounts of precipitation related to Hurricane Dorian (up to more than 1,400 mm over three days; see figure 1), the RAPID system was automatically triggered for the northern Bahamas. Sentinel-1 SAR data were available at 23:44 UTC on September 2 and at 11:09 UTC on September 4, 2019.

125 At the time of the first overpass on September 2, Dorian was located 20 km to the north of Grand Bahama (figure 1). Sentinel-1 data covered the northeastern sector of Grand Bahama and all of Great Abaco, both located in the southeastern sector of the hurricane and therefore affected by southwesterly winds. Lower-elevation areas on the west coast of Great Abaco were suffering onshore tropical storm force winds, and flooding in these territories was extensive, covering 518 km² of land, or 26% of the island (figure 2a).

130 That same day On September 2, offshore hurricane-force winds affected lower-elevation areas along the northern coast of Grand Bahama, which had been affected by onshore hurricane-force winds during the previous day. Despite the blowing of the winds away from the coast, proximity to the center of the hurricane did not allow the storm surge to retreat significantly. For this reason, these locations were also still experiencing extensive flooding: 138 km² of the 308 km² covered by the SAR images were flooded, amounting to 45% of eastern Grand Bahama (figure 2a). Since the first overpass occurred several hours after the passage of the hurricane, the flooded area shown in figure 2a represents a conservative estimate.

135 The second Sentinel-1 overpass entirely captured Great Abaco, Grand Bahama, Andros, New Providence, and other smaller islands of the archipelago on September 4, when Dorian was located 300 km to the north of Grand Bahama. Despite the absence of storm surge at the time of the overpass, 17% of Great Abaco was still flooded, while flooding on Grand Bahama had mostly receded (14% of the island was still flooded).

140 On the previous day September 3, high amounts of precipitation also fell on Andros Island, located approximately 200 km to the south of Dorian's path. The flooding map resulting from the automated trigger of the RAPID algorithm also showed extensive flooding on the low-lying terrain of this island, which received onshore winds during the entire duration of the event (figure 2b3). On September 4, the inundated area was 2,193 km² (37% of the island), and flooding reached more than 10 km inland. RAPID flooding estimates of area and inland extent on the Andros Island are in agreement with the coarser resolution product composited from VIIRS (375m) and ABI (1km) passive radiometers, displayed on the International Charter "Space and Major Disasters" website at https://disasterscharter.org/image/journal/article.jpg?img_id=3519568&t=1568272371731.

145 The agreement (overall, user, producer) scores between RAPID and EMS flooding maps for the Abaco Islands on September 2 and September 4, derived from the confusion matrix shown in Table 1, were (77%, 90%, 41%) and (89%, 61%, 86%), respectively. The high overall and user agreement scores for the September 2 flooding are also depicted in the flood maps of Figure 2b indicating a very good overlap of the two products over the coast of Great Abaco, while the relatively low producer agreement comes from the lack of flood detection by the EMS algorithm over the multiple near-sea-surface-elevation islands, located in the front of the western coast of Great Abaco. The relatively low user agreement score between the two products on September 4 is due to the fact that RAPID classifies some non-flooded areas within the EMS flooded boundary, which are expected to occur as a consequence of the flood recession.

155 According to the conservative flooding estimates shown in these maps, the total area covered by flooding in the Bahamas exceeded 3,000 km², spread over areas hundreds of kilometers away from each other. To assess the inundation extent over such vast and dispersed areas, recognition flights take days, and they cannot operate during such extreme weather conditions as were presented by the long-lasting hurricane-force winds in the Bahamas. In contrast, a system such as RAPID can provide flooding estimates for any area of the world within hours of the data's becoming available. RAPID has the potential to be a fundamental tool for a fast and efficient emergency response.

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4 Closing Remarks

Hurricane Dorian heavily damaged the northern Bahamas with extreme winds and precipitation and extensive flooding, causing deaths whose number is yet unknown. When extensive weather-related devastation occurs in areas of the world that do not have in situ observation networks, an assessment of the situation based on hydrometeorological parameters is often difficult.

In this brief communication, we analyzed the flooding related to Hurricane Dorian in the Bahamas at 10 meter resolution through RAPID, which is an automated system that can identify in near real time areas of the world at risk for flooding and create inundation maps for them producing near-real-time flood maps across the globe based on SAR images. Specifically, RAPID identifies possibly flooded areas using near-real-time, high-resolution precipitation products. The RAPID core algorithm and then processes SAR images to compute inundation maps that may be used promptly for effective emergency response.

For Hurricane Dorian, RAPID inundation maps showed that, several hours after the passage of the storm, 26% of Great Abaco, 45% of eastern Grand Bahama, and 37% of Andros were flooded, for a total area exceeding 3,000 km². We compared RAPID inundation maps with EMS maps finding high agreement scores and we discussed the differences between the two products for the case in exam. We believe ~~its~~ RAPID system's ability to map such a large area of inundation ~~so~~ quickly as soon as SAR observations were available makes RAPID-it the fastest fully automated method for assessing flood extension and providing situational awareness.

The main limitation of the system is the occasional unavailability of timely satellite overpasses in conjunction with heavy precipitation events. For Hurricane Dorian, Sentinel-1 images were not available at the peak of the event in the most affected area. This limitation can be overcome through international collaborations, such as the International Charter "Space and Major Disasters", Sentinel Asia, NASA-ISRO SAR Mission, and Copernicus Emergency Management Service – Mapping, that may would increase the availability of data from other satellite missions.

Future extensions of this work may will allow us to combine the rapidly derived inundated areas with high-resolution terrain elevation to identify flood levels and inversely estimate the surges that caused the flooding. Given Using this information, we would ~~have been be~~ able to extend the flood inundation estimates outside the SAR coverage, e.g. derive the for September 2 flooding over northwestern part of Grand Bahama, where Sentinel-1 observations ~~we~~ are not available. Furthermore, an estimate of the surge level ~~would have been~~ can be valuable information for comparing ~~ng~~ son against model the surge forecasts for this event.

Author contribution: DC: conceptualization, supervision, writing – original draft. QY and MK: software, formal analysis, data curation. XS and EA: conceptualization, project administration, writing – review and editing.

Competing interests: The authors declare that they have no conflict of interest.

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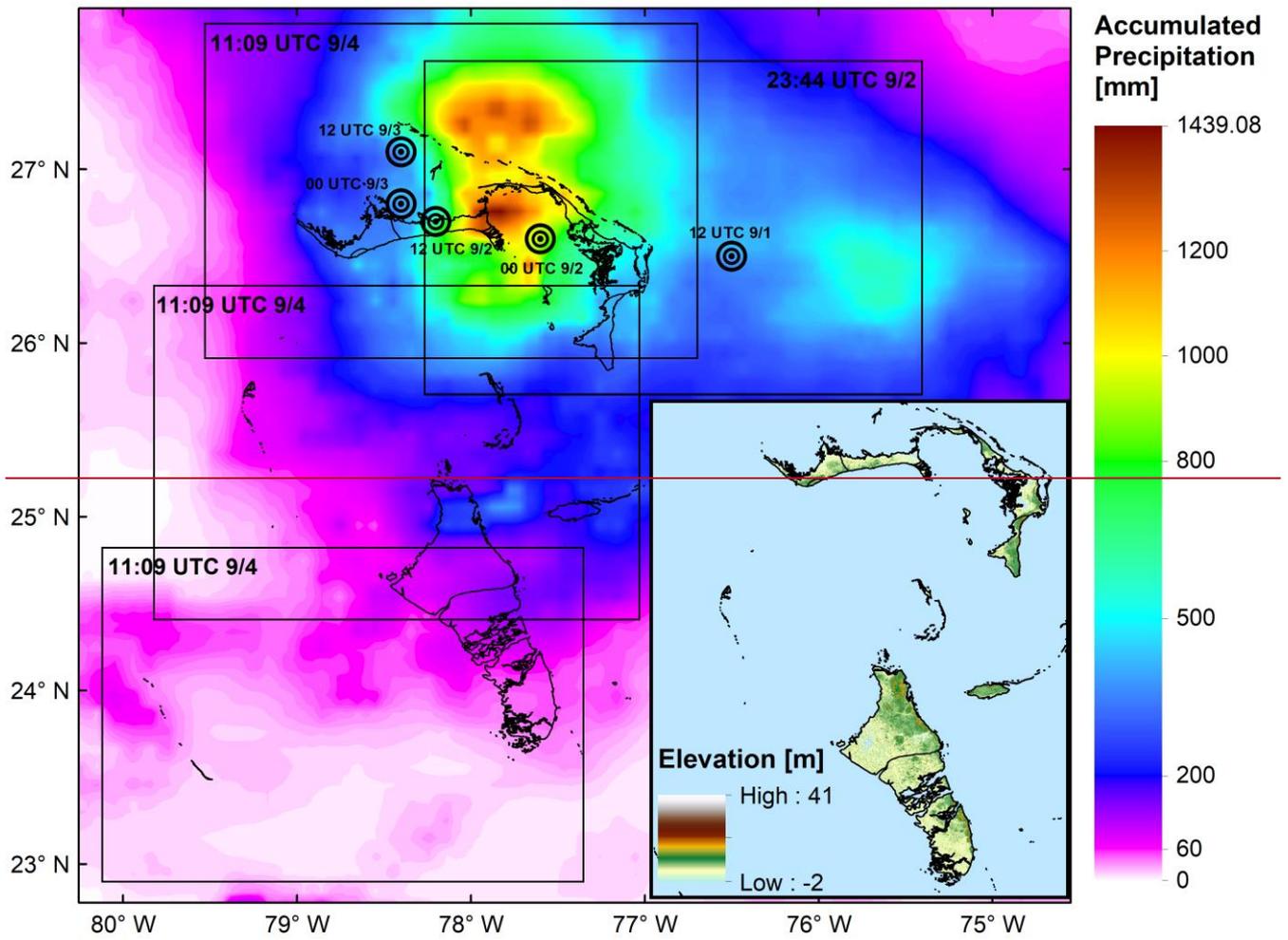
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Total Accumulated Precipitation between 00 UTC 01/09/2019 and 00 UTC 04/09/2019

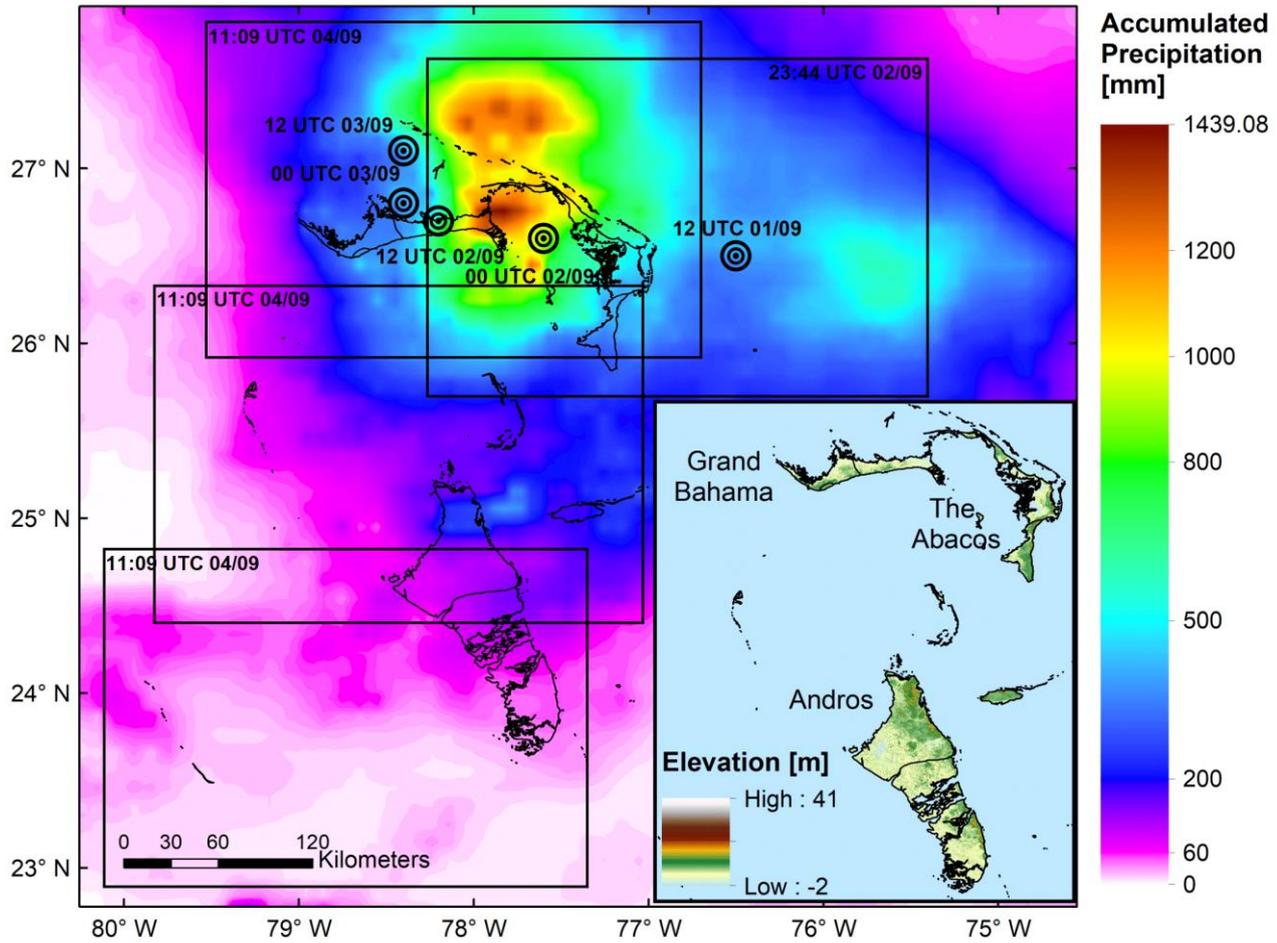
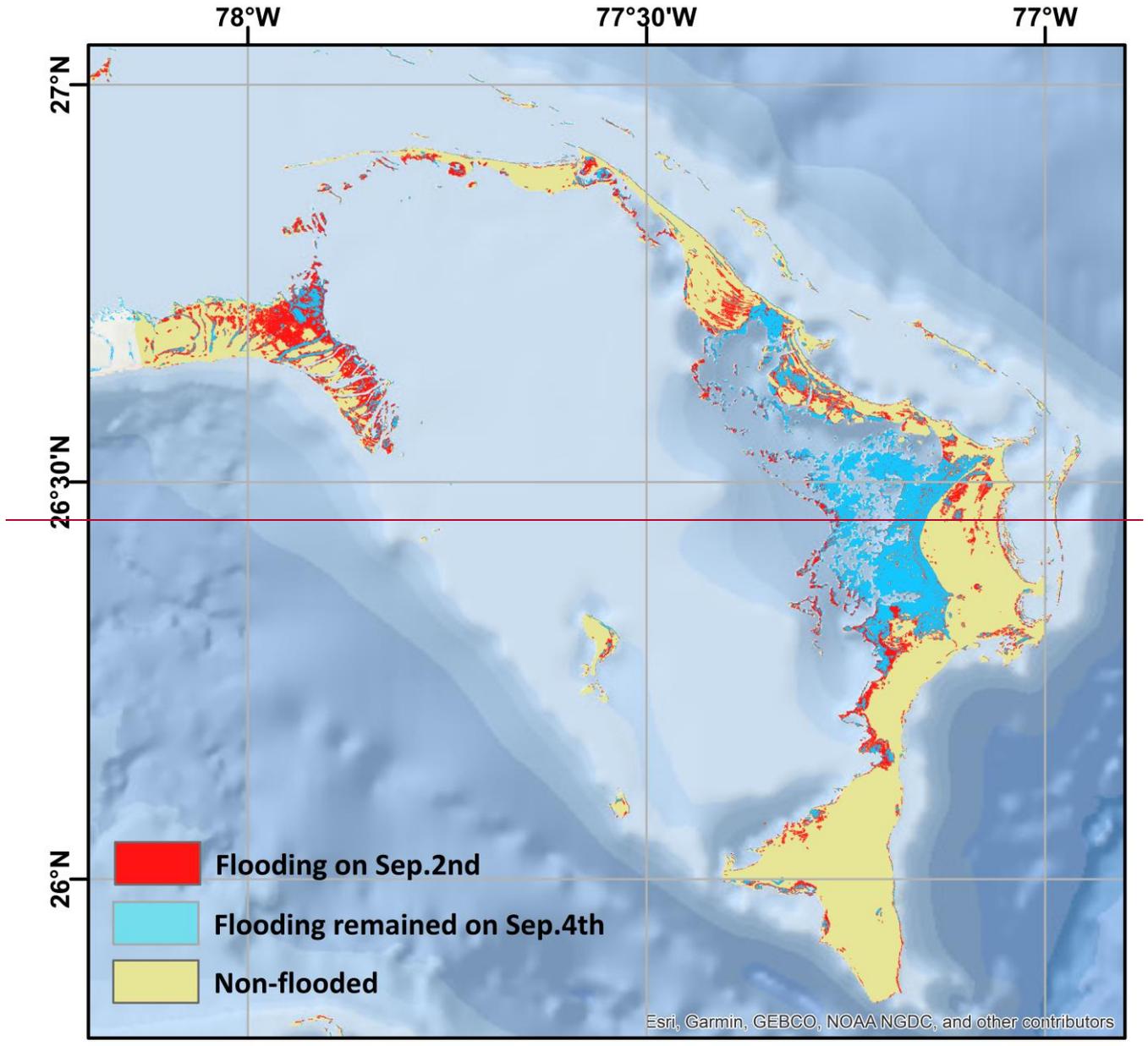


Figure 1: Background map: IMERG total accumulated precipitation between 00 UTC September 1 and 00 UTC September 4, 2019 (shaded), boundaries of Sentinel-1 SAR images (rectangles), and location (targets) of Hurricane Dorian from National Hurricane Center public advisories (NHC, 2019). Bottom right front map: elevation (Farr et al. 2007) of the study area. In both maps, Bahamas boundaries are delineated using the Global Administrative Areas dataset (Hijmans et al. 2015).



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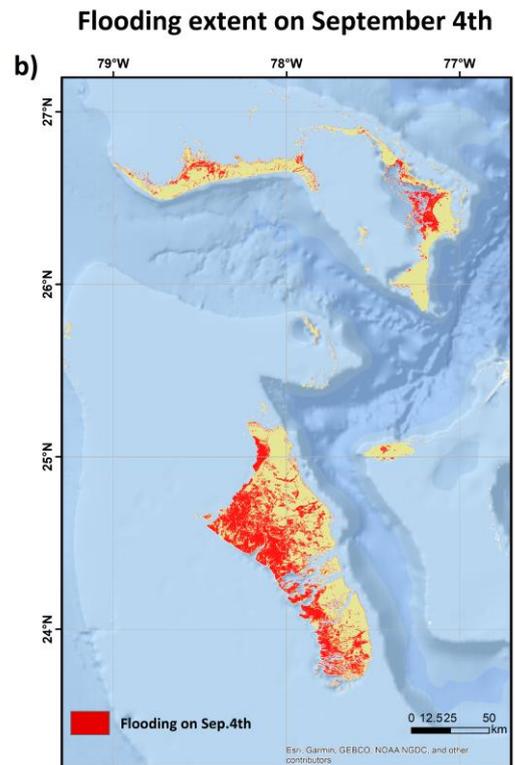
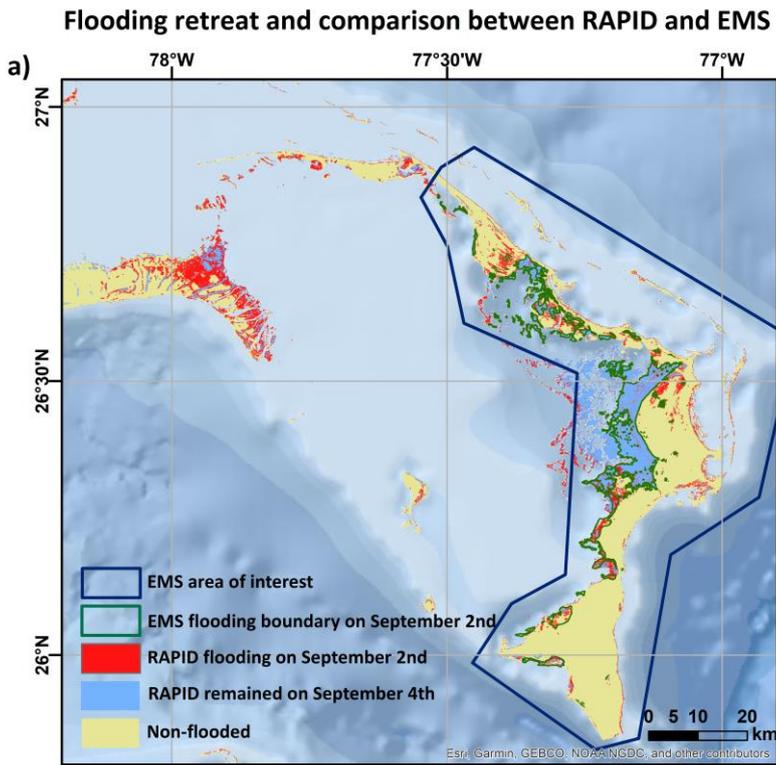


Figure 2: ~~Ocean background from World Ocean Base map (ESRI et al. 2014; list of contributors available at http://downloads.esri.com/esri_content_doc/da/WorldOcean_ContributorsDA64.pdf).~~ a) Flooded and non-flooded areas on September 2 and September 4, 2019 derived from the RAPID algorithm that processed SAR data from the Sentinel-1 overpasses, and flooded boundary on September 2 from EMS. b) RAPID flooding map for the entire Northern and Central Bahamas, for the September 4, 2019 Sentinel-1 overpass. For both images, ~~o~~ Ocean background from World Ocean Base map (ESRI et al. 2014; list of contributors available at: http://downloads.esri.com/esri_content_doc/da/WorldOcean_ContributorsDA64.pdf).

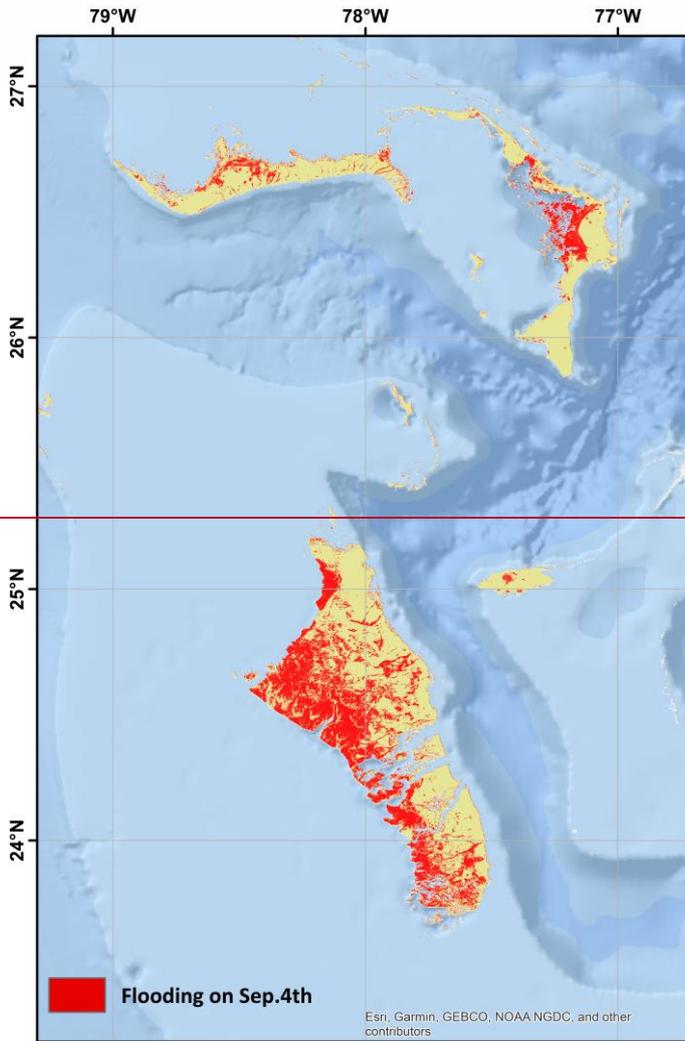


Figure 3: Same data sources as in Figure 2. Here, the SAR flooding map is shown for the entire Northern and Central Bahamas, for the September 4, 2019 Sentinel 1 overpass.

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<u>Confusion Matrix</u>		<u>September 2 – Great Abaco</u>		<u>September 4 – Grand Bahama</u>	
		<u>EMS</u>		<u>EMS</u>	
		<u>Flooded</u>	<u>Non-flooded</u>	<u>Flooded</u>	<u>Non-flooded</u>
<u>RAPID</u>	<u>Flooded</u>	<u>2,274,927 (14.5%)</u>	<u>3,318,143 (21.1%)</u>	<u>1,880,609 (13.2%)</u>	<u>32,989 (2.3%)</u>
	<u>Non-flooded</u>	<u>260,335 (1.7%)</u>	<u>9,847,017 (62.7%)</u>	<u>1,219,786 (8.6%)</u>	<u>10,710,519 (75.9%)</u>

Table 1: Confusion matrix between RAPID and EMS flooding products for September 2, 2019 overpass over Great Abaco (left) and for September 4, 2019 overpass over Grand Bahama (right). For each matrix, number and percentage of pixels is reported.