

# 1 Estimation of insurance related losses resulting from 2 coastal flooding in France

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## 9 10 Abstract

11 A model has been developed in order to estimate insurance-related losses caused by coastal  
12 flooding in France. The deterministic part of the model aims at identifying the potentially  
13 flood-impacted sectors and the subsequent insured losses a few days after the occurrence of a  
14 storm surge event on any part of the French coast. This deterministic component is a  
15 combination of three models: a hazard model, a vulnerability model and a damage model. The  
16 first model uses the PREVIMER system to estimate the water level resulting from the  
17 simultaneous occurrence of a high tide and a surge caused by a meteorological event along the  
18 coast. A storage-cell flood model propagates these water levels over the land and thus  
19 determines the probable inundated areas. The vulnerability model, for its part, is derived from  
20 the insurance schedules and claims database; combining information such as risk type, class  
21 of business and insured values. The outcome of the vulnerability and hazard models are then  
22 combined with the damage model to estimate the event damage and potential insured losses.  
23 This system shows satisfactory results in the estimation of the magnitude of the known losses  
24 related to the flood caused by the Xynthia storm. However, it also appears very sensitive to  
25 the water height estimated during the flood period, conditioned by the junction between sea  
26 water levels and coastal topography, the accuracy for which is still limited by the amount of  
27 information in the system,

# 1    **1    Introduction**

2    The Xynthia storm, which occurred in France in February 2010, renewed the awareness of  
3    coastal flooding risk on the French coast (Lumbroso and Vinet, 2011). This peril, resulting  
4    from the combination of high spring tide and severe meteorological conditions (maximum  
5    instant wind comprised between 100 and 140 km/h in the most affected areas), leads to very  
6    high levels of loss; both in term of death and injury and in resultant economic losses. The  
7    financial cost of Xynthia has been estimated to have been approximately 2.5 billion euros, of  
8    which 1.5 billion euros was covered by insurance and reinsurance companies (FFSA and  
9    GEMA, 2011). In terms of insurance payments, Xynthia is one of the most costly storm  
10    events to have occurred in France in the last thirty years

11    In France, the financial compensation for the victims of such natural disasters is governed by  
12    the national Natural Catastrophes system (thereafter the Nat Cat-system). This has provided  
13    approximately 713 million euros for the victims of Xynthia. This system engages CCR  
14    (Caisse Centrale de Réassurance) a reinsurance company, which is owned by the French  
15    State, to administer the state guaranty for Nat Cat events.

16    In order to meet the needs of the French State and its clients, and for its own income forecasts,  
17    CCR has developed models aiming to estimate the insured losses arising out of Nat Cat events  
18    (flood, coastal flooding, windstorm, drought, earthquake and volcanic eruption). The strategy  
19    of CCR is to develop its own modelling tools for natural disasters in order to have control  
20    over the different components of the model and to validate them with the available data. The  
21    development of such models is also made possible by the significant database of insurance  
22    schedules and claims that have been collated by CCR with the support of its cedants (cedant -  
23    this term designates the client companies of CCR).

24    Coastal flooding modelling faces several issues: scales (multi frequency and multi spatial  
25    scales processes); process interaction related to the forcing conditions (waves, surge, tide,  
26    river discharge etc.); coastal flooding process complexity (overtopping, interaction with  
27    structures, etc.); a detailed enough knowledge of the topography and associated structures.  
28    Several approaches have been developed to estimate coastal flooding, ranging from the  
29    simple approach of estimating flooding from topographic contours to the more accurate  
30    2D/3D, time-varying, full-process models. Estimating flooding from topographic contours

1 leads to an overestimation of the flood event, without providing any information on the  
2 velocity or the temporal dynamics (e.g, Breilh et al., 2013).

3 Recent modelling developments now permits realistic 2D simulations of urban areas  
4 conditions (Le Roy et al., 2014) but still with computation time exceeding several days to  
5 simulate a few event hours. Some regional models such as the ones developed over the last 5  
6 years in the United State (Bunya et al. 2010) have demonstrate that it was possible to simulate  
7 coastal flooding at a regional scale. These models could be an interesting alternative for  
8 operational purpose but require massive parallel techniques.

9 To tackle the issue of estimating coastal flooding on a national scale, a compromise has to be  
10 found between the quality of the results, feasibility (data available) and computational time.  
11 This leads to the necessity of developing alternative approaches, between those of basic static  
12 projection of water levels and the most advanced approaches used for projecting events in  
13 urban areas.

14 The requirements of CCR correspond to two objectives. The first is to determine the likely  
15 affected areas and the resultant cost of catastrophic events a few days after their occurrence.  
16 This estimation allows CCR to advise the French State and its clients of the potential financial  
17 magnitude of the event. The second objective is to assess the potential losses due to extreme  
18 events according to their return period. Successful achievement of this objective will allow  
19 effective evaluation of the financial exposure of CCR, the French State and of cedants to  
20 coastal flooding risk.

21 In order to fulfil the first objective for storm surge projection, a deterministic model is  
22 currently being developed by CCR. Numerous studies have been published on the estimation  
23 of flood damages (Jongman et al. 2012; Meyer et al., 2013; Meyer and Messner, 2005).  
24 Generally, these models combine a hazard model which characterizes the intensity of the  
25 event, and a susceptibility model, describing the exposure of the subject area and a damage  
26 model to estimate the financial cost. In the same way, some specific models for storm surge  
27 have been developed, especially for tropical cyclones in America (Genovese et al., 2011;  
28 Wang et al., 2014) or in the Netherlands-dealing with the problematic nature of polder areas  
29 (de Moel et al., 2012; Bouwer et al., 2009). These models are most often concerned with the  
30 global economic cost, but some more specific works have concentrated on the insurance cost.  
31 For example, Pistrika and Jonkman (2009) have studied the insurance cost due to hurricane

1 Katrina. Sousounis and Kafali (2010) in a recent study have used the AIR model to estimate  
2 the potential cost of the 1959 typhoon in Japan. Some models use a probabilistic approach,  
3 like Gaslikova et al. (2011) for consideration of future storm surge impact in the North Sea  
4 region. The approach of Czajkowski et al. (2013) presents a methodology to determine  
5 premium on the basis of property exposure to coastal flood risk.

6 For CCR, the loss consideration will not concern itself with damage to public infrastructure or  
7 assets or agricultural losses as these are not included in the Nat Cat compensation system.  
8 Thus, a specific calibration is needed such as it has already been established for inundation by  
9 CCR (Moncoulon et al., 2014).

10 In order to estimate the losses related to coastal flooding events, the deterministic model has  
11 been developed on the same structure followed by most of the insurance related catastrophe  
12 models: a hazard model, a vulnerability model and a damage (financial loss) model. This  
13 development should face two main challenges. The first one is the operational modelling of  
14 the inundation resulting from coastal flooding hazard at a large scale (the French coasts) and  
15 within a short time-frame and the second consideration integrates the hydrodynamics output  
16 and the limitations of the insurance related data.

17 The present paper examines the methodology developed for a timely estimation of the costs  
18 generated by coastal flooding: a hazard model is used to estimate flooded areas, and a  
19 vulnerability model then allows estimation of the exposure of insurance policies to coastal  
20 flooding, before finally calculating the potential cost of the event through a damage model.  
21 This methodology has been applied to the data relating to the major coastal flooding that  
22 occurred during Xynthia storm in France in 2010. A sensitivity analysis has permitted a full  
23 understanding of the effectiveness of the hazard model in interpreting/predicting the results,  
24 and permitted the application of the methodology to four other minor events shows the  
25 consistency of the results.

26

## 27 **2 Hazard model**

### 28 **2.1 Methodology**

29 The purpose of the hazard model is to estimate the water levels over the land areas where  
30 there is likely to be insurance policy coverage. During a storm event, water levels along the

1 coastline are influenced by three main processes. The first one is that of tide variation, which  
2 plays an important role along the Atlantic coast of France, the Channel and the North Sea.  
3 The amplitude can vary from 3 to 11 meters for spring tides (SHOM, 2012). The second  
4 process is atmospheric storm surge, resulting essentially from atmospheric pressure and wind  
5 effect, and its coincidence with high tides. For Xynthia, this surge was estimated to have been  
6 1.5 meters at La Rochelle (Bertin et al. 2012).

7 In this study, we chose to use the Previmer system ([www.previmer.org](http://www.previmer.org)) to determine potential  
8 water levels resulting from these two processes. This system provides both forecasts and  
9 observations of water levels, but also models waves, currents, temperatures and plankton  
10 content.

11 The third component is the wave setup which corresponds to the surge that is generated by the  
12 wave breaking over the beach. This surge component, not provided within the Previmer,  
13 model, could represent a significant part of the total surge effect (Holman and Sallenger,  
14 1985) and is estimated on the basis of the wave characteristics.

15 Once the total water levels have been estimated along the coastline, an inundation model  
16 propagates them over the landmass behind. For this purpose, an inundation model, similar to  
17 the Lisflood-FP model (Bates et al., 2010) is utilized.

18 Finally, this modelling chain allows estimation of the inundation process by assessing the  
19 volume of the overflowing seawater, but this is projected without the additional consideration  
20 of the consequences of wave overtopping or the destruction of sea defenses.

### 21 2.1.1 Tide, atmospheric storm surge and regional wave setup

22 The Previmer system, created in 2006 (Muller et al., 2014) uses the MARS-2D hydrodynamic  
23 model (Lazure and Dumas, 2008) and usually shows satisfactory results (Idier et al., 2012).  
24 The model is based on the resolution of shallow-water equations with the finite difference  
25 method. The model runs on three different levels of resolution: 2 km for the North East part of  
26 the Atlantic Ocean, 700 meters for the Channel and the Bay of Biscay and 250 meters for five  
27 smaller areas. The bathymetry used by Previmer is based on the NOOS 1° (North-West Shelf  
28 Operational Oceanographic System), EMODNET (European Marine Observation and Data  
29 Network) and the SHOM-Ifremer DTM at 100 meters and 500 meters of resolution. The  
30 MARS-2D model uses the FES2004 tide model (Lyard et al., 2006) for the two lowest

1 resolutions and the model cstFrance (Simon et al., 2011) with 115 harmonic constants at 250  
2 meters. The meteorological data is provided by Meteo France and is based on Arpege  
3 (Courtier et al., 1994) and Arome models (Seity et al., 2011) whose resolutions are  
4 respectively of  $0.5^\circ$  and  $0.025^\circ$

5 The validation of the model, presented in Muller et al. 2014, has shown that root mean square  
6 errors (RMSE), calculated on the basis of hourly data, are usually comprised between 5 and  
7 14 centimeters, with a mean of 9 cm, which is deemed to be satisfactory. However, these  
8 errors can be higher for exceptional events such as Xynthia. The difficulties of generating an  
9 effective estimation of the water levels for Xynthia could be explained by the fact that this  
10 event had characteristics which were quite particular. Indeed, Bertin et al., (2012) have shown  
11 that the associated surge was exceptionally high for this event in comparison to the intensity  
12 of the storm. This behavior could be explained by the trajectory of the depression that induced  
13 an Ekman transport directed toward the coast, enhanced by the presence of young and steep  
14 waves (Bertin et al., 2014).

15 The version of Previmer used in this study tends to underestimate water levels observed in the  
16 central part of the Bay of Biscay during the Xynthia event. According to the tide gauge  
17 measurements, this underestimation could reach 30 to 40 centimeters in some sectors. In order  
18 to use the Previmer outputs in the best conditions for our system, the water levels simulated  
19 have been adjusted to match the highest levels observed by tide stations during the simulated  
20 events. To realize this correction, an adjustment value is computed for each pixel located  
21 along the coast on the basis of a linear regression between the errors that are measured for the  
22 two nearest stations. It should be noted that tide gauge measurements include tide and  
23 atmospheric storm surge, but also regional wave setups. Indeed, for instance at La Rochelle,  
24 as shown by Bertin et al. (2015), the total water level result from these processes, with a  
25 regional wave setup, was between 5 and 10 cm at La Rochelle. Thus, our calibration method  
26 indirectly permits the taking into account not only tide and atmospheric surge, but also the  
27 regional wave setup. An illustration of two maregraphs obtained for Xynthia is presented for  
28 in Fig. 1 and shows that RMSE were generally between 10 and 30 cm for this event.

### 29 2.1.2 Wave setup

30 Waves could also play a key role in the inundation processes during storm surges with the  
31 phenomena of wave setup and wave runup (Kim et al., 2008; Ferrarin et al., 2013). However,

1 it is relatively difficult to estimate these parameters accurately, especially on a large scale and  
2 with a poor bathymetry. In addition, as discussed in the water level calibration step, two kinds  
3 of wave setup can be identified. Regional wave setup, occurring over large areas such as bays  
4 (see e.g. Bertin et al., 2015) - which can be reproduced utilizing a rather coarse grid (e.g. 50 m  
5 grid size), and local wave setup, resulting from localized waves breaking near the coast,  
6 requiring usually finer grids (5-10 m grid size). As discussed in the above, the regional setup,  
7 even if not actually calculated here, has been indirectly integrated into the adjustment of the  
8 simulated water levels against the actual water levels observed by the tide stations, which are  
9 included in this regional setup model. Thus the values of the local wave setup, given as  $\eta$  are  
10 estimated within the empirical formula of Stockdon et al. (2006):

$$11 \quad \langle \eta \rangle = 0.35 \beta_f (H_0 L_0)^{1/2} \quad (1)$$

12 The parameters of this formula (deep water wavelength,  $L_0$ ; the significant wave height  $H_0$ ),  
13 are computed using PREVIMER wave data (Ardhuin et al., 2010). The third parameter,  
14 representing the beach steepness,  $\beta_f$ , was estimated on the basis of the BD CARTO (IGN  
15 database). This database contains the limits of the low and high spring tide. The distance and  
16 the height between them are used to roughly estimate the beach slope. In order to measure and  
17 compare offshore parameters and those for beaches, vectors for the coastline were established  
18 every 250 meters and a cross-section was established to determine the coordinates of the  
19 parameters. A selection of these cross-sections was then scrutinized: the sections for which  
20 the bathymetry did not reach a 20 m depth within a 5 km range offshore were deleted from the  
21 wave setup computation. In the same ways, in harbours, the cross sections have been deleted  
22 and instead the statistical setup of the surrounding beaches is applied.

23 Fig. 2 shows the maximal wave setup height computed for the Xynthia event along the coast.  
24 The mean height often reaches 40 centimeters but extends to 80 centimeters for the most  
25 exposed areas. This empirical estimation of the local wave setup provides results of the same  
26 order as those obtained by Bertin et al. (2015) by coupling a spectral wave model (WWM)  
27 and the SELFE hydrodynamic model (local wave setup can locally exceed 0.4 m, even if the  
28 calculation resolution remains too coarse to capture the maximum setup along the coastline).  
29 These values justified the necessity of taking this phenomenon into account. However, the  
30 methodology presents two main limitations. In effect, the WWIII model used by Previmer  
31 simulates the refraction phenomenon but the spatial resolution does not always permit an

1 accurate representation of local refraction at the coast. For the moment, the actual model does  
 2 not take into account for the reverse shoaling effects. It could be interesting, in the future, to  
 3 use the correction of H0 proposed in Lecacheux et al. 2012 in order to be in the conditions of  
 4 the Stockdon et al. (2006) formulae (deep water and waves perpendicular to the coast). The  
 5 second limitation is that local wave setup, which could impact sheltered bays, is not  
 6 considered. However, as explained above, the regional wave setup is indirectly taken into  
 7 account, such that the wave setup along sheltered bays is not nullified. The potential impact of  
 8 these limitations will be discussed in Sect. 4.3.

### 9 2.1.3 Coastal flooding model

10 Water levels estimated along the coastline are then propagated onto the land using an  
 11 inundation model. Given the large scale and the operational aspects of the study, an  
 12 application of full-process based model (for example based on non-linear shallow water  
 13 equations or on Boussinesq equations) would have been too cumbersome to implement given  
 14 the need for extensive spatial information and significant computational times required. As a  
 15 result, a storage-cell flooding model has instead been used, similar to the Lisflood-FP model  
 16 (Horritt and Bates, 2001; Hunter et al., 2005; Bates et al., 2010) that has been tested for the  
 17 computation of coastal flooding by Bates et al. (2005) with good results. The model  
 18 developed in this study is based on the continuity equation relating to flow and volume  
 19 changes:

$$20 \quad \frac{\partial h^{i,j}}{\partial t} = \frac{Q_x^{i-1,j} - Q_x^{i,j} + Q_y^{i,j-1} - Q_y^{i,j}}{\Delta x \Delta y} \quad (2)$$

21 Where,  $Q^{ij}$  represents the flows (m<sup>3</sup>/s) between cells at the node (i, j), x stands for horizontal  
 22 and y for vertical flows,  $\Delta x$  and  $\Delta y$  are the cell dimensions (meters), h is the free water  
 23 surface and t is the time (seconds). The flows between cells are estimated according to  
 24 Manning Striker law with, for example in the x direction:

$$25 \quad Q_x^{i,j} = \frac{1}{n} (h_{f_{\max}} - z_{\max})^{5/3} \left( \frac{h^{i-1,j} - h^{i,j}}{\Delta x} \right)^{1/2} \Delta y \quad (3)$$

26 Where n is the Manning's friction coefficient,  $h_{f_{\max}}$  the highest water free surface in the two  
 27 cells, and  $z_{\max}$  the higher bed elevation.



1 The Digital Terrain Model (DTM) used for this study is a commercial DTM that presents a  
 2 spatial resolution of 25 meters. This product could be considered has a temporal solution  
 3 since a Lidar-based DTM called Litto-3D is developed over the French territory. This DTM  
 4 that would present a 1 meter spatial resolution is currently freely available in different sectors  
 5 such as Brittany or Mediterranean areas. However, according to the application scale of this  
 6 study, it seems preferable to use a homogeneous DTM, available all over the coastline. In  
 7 order to apply the model within operational perspectives, the French coast has been divided  
 8 into 39 sectors, exempted from interdependency, where the model is applied independently.

9 In order to take into account the roughness of the ground, the Manning's coefficient ( $n$ ) is  
 10 fixed according to the land use described in the Corine Land Cover database (2006), using the  
 11 coefficients proposed by Lopes et al. (2013) for the Ria de Aveio at Portugal.

12 Finally, the modelling process permits a chronological estimation of water levels for each cell  
 13 of the DTM. Accordingly it is possible to extract, for each cell, the highest free surface water  
 14 height presented by the damage model. This storage-cell model presents the dual advantages  
 15 of being easy to set up and is relatively fast in computational time.

## 16 2.2 Results

17 After the Xynthia storm, inventories of inundated areas have been achieved. These inventories  
 18 have been used in our study to compare the simulated flood areas and the observed inundated  
 19 areas. The results of this comparison, presented for the Charente Maritime department in Fig.  
 20 3, show that the hazard model allows globally a good representation of the inundated areas  
 21 with significant differences at the local scale: the CCR model makes projections of an  
 22 inundated area of 670 km<sup>2</sup> against 595 km<sup>2</sup> actually affected.

23 The fraction  $F$  of the domain classified correctly by the model was calculated on the basis of  
 24 the formulae proposed by Aronica et al.2002:

$$25 \quad F = \frac{\sum_{i=1}^n P_i^{D_0M_0} + \sum_{i=1}^n P_i^{D_1M_1}}{n} \quad (4)$$

26 Where  $n$  is the number of pixels in the domain,  $P_i^{D_0M_0}$  the pixels not inundated in the reality  
 27 and not detected by the model and  $P_i^{D_1M_1}$  the pixels inundated and detected by the model. The  
 28 fraction obtained is of 85% for the considered domain which represent 4363 km<sup>2</sup>. This result

1 confirm a satisfying behavior of the model at a global scale with however notable differences  
2 at a local scale.

3 The cartography presented in figure 3 highlights areas where the extent of the inundation is  
4 widely underestimated. This is the case, for example, of the municipalities of Charron or Saint  
5 Pierre d'Oléron. These deviations could be due to underestimation of sea water levels in  
6 sheltered bays, but perhaps mainly by the quality of the DTM used. Indeed, at 25 meters of  
7 resolution, dikes and protection walls have been smoothed, despite some corrections, and this  
8 could skew the real water level estimation. The quality of the DTM will be discussed in Sect.  
9 3.2. In such flatlands, a few centimetre differences in water levels induce very different  
10 estimations of the inundated areas (see Sect. 4.3).

11 There are contrary cases where the model seems to overestimate flooded areas as, for  
12 example, in the Isle de Ré or Isle de Noirmoutier. The consequences are opposite, but the  
13 origins of the problem could be the same: the approximate estimation of water levels in the  
14 sea and the quality of the DTM used.

15 The one-way nesting approach that has been developed in this study could also be limited by  
16 the problematic of water levels limitation during flooding. Indeed, several studies showed that  
17 the consideration of the hydrodynamic model boundary as walls could drive to water level  
18 over-estimations given the fact that water could not flood the lands (Towned and Pethick  
19 2002, Bertin et al., 2014). The analysis of this phenomenon realized by Waeles et al. (2014)  
20 for Xynthia show that this phenomenon could be very significant in the case of enclosed bays  
21 and estuaries and may exceed 0.5 meters.

22 However, in our case, it is difficult to evaluate the divergence given the fact that the spatial  
23 resolution of Previmer is lower than the one used by Waeles et al. (2014) and may  
24 underestimate locally water levels. The adjustment of simulations realized on the basis of tide  
25 measurements could also partially compensate the overestimation induced by water levels  
26 limitation.

27 Lastly, it appears also that the quality of the meteorological data used by Previmer in 2010  
28 was limited - there is a resolution of  $0.5^\circ$  and a time step of 6 hours (Arpege model). Since  
29 2012, the Previmer system uses Arome meteorological data with hourly time steps and a  
30 spatial resolution of  $0.1^\circ$ . This difference of the qualitative information explains why  
31 Previmer outputs underestimate water levels for the Xynthia event, in particular for bays such

1 as the Aiguillon Bay where a simulation realised with Arome data indicates that water levels  
2 could be underestimated by about 40 cm.

3 It is possible that the water level limitation effects are compensated by these two aspects and  
4 that, finally, water levels are underestimated given the low inundation spread observed in the  
5 Charron sector.

6 The results illustrate also an incorrect reproduction of the behaviour of inundation processes  
7 along the coastal inlets and estuaries. This is noticeable for the municipality of Rochefort for  
8 example where the inundation has been induced by the interaction between the river flow and  
9 the sea. This problem is not yet fully remedied in the actual version of the inundation model,  
10 but when so, this should probably improve substantially the results. Apart from in these areas,  
11 the model, even if basic, generally provides flooded areas being in a reasonable agreement  
12 with the observed ones.

13

### 14 **3 Vulnerability model and hazard validation**

#### 15 **3.1 Vulnerability database**

16 The vulnerability model developed by CCR is derived from schedule information and loss  
17 records termed henceforth “insurance policies”. A database was created under the framework  
18 of bilateral contracts with insurance companies under conditions of confidentiality. The fact  
19 that CCR provides cover for a very large proportion of individual French insurance policies  
20 offers a particularly broad vision of French exposure to coastal flooding risks.

21 This database describes the nature of the risk (house, building or apartment), its usage  
22 (residential or commercial) and its occupation (owner, tenant, co-ownership). For a large part  
23 of them, these risks have been geo-referenced. The latitude and the longitude are estimated on  
24 the basis of the location of the address in the street and not the real location of the building. Its  
25 real position could vary from a few meters to several thousand in some cases. In some cases,  
26 data sources do not permit a specific designation of the property locus, and the nearest co-  
27 ordinate- street or commune centre has therefore been adopted. Some complementary  
28 elements have been extracted from a DTM such as the distance to the coastline or the location  
29 elevation.

1 Additionally, each year, the insurance companies have undertaken, at renewal, to transmit  
2 historic claims details. It is thus possible to associate each claim with an historic flood loss  
3 event and so to estimate the global insured cost of each past events that are managed under  
4 the Nat Cat system. For the coastal flooding risk evaluation, the historic claims have been  
5 consolidated to three events for the studied area: Lothar and Martin (1999), Johanna (2008)  
6 and Xynthia (2010).

### 7 **3.2 Hazard validation**

8 The claims localisation could be considered as a spatial indicator of the extent of the flood.  
9 Initial geographic comparisons between observed and simulated inundated areas and claims  
10 were evaluated. These comparisons are shown in Fig. 4 locally for the commune of  
11 Châtelailon-Plage. This example shows first that the claims data has not always been  
12 effectively localized as examination of the detail shows the claims clustered in the town  
13 centre. Indeed, the delineation of inundated areas, often produced a few days after an event,  
14 on the basis of satellite data, could shows that claims data input, where not sited in its actual  
15 location, could miss some important areas, especially say in the marsh areas. Furthermore,  
16 these data inventories could miss sectors where flood water has just runoff and not remained  
17 as standing water to be recorded after the initial flood event.

18 This figure shows also some isolated claims not explained by overflowing simulation alone.  
19 This is the case for claims localised into the Vieux-Châtelailon sector where the elevation of  
20 the DTM is relatively important. Often, these problems are explained by the localization of  
21 the insurance policy details, but they could also result from local conditions such as streams  
22 or urban water network overflowing (as opposed to actual coastal inundation) and cellar  
23 inundations.

24 Two indicators (Wilks, 2011) have been computed to evaluate hazard model performances on  
25 the basis of its ability to detect claims:

- 26 - the probability of detection (POD): which is the ratio between the claims detected by  
27 the model and all of the actual claims. The POD evaluates the capacity of the model to  
28 detect claims, the best case being a POD of 100%;

- 1 - the probability of false detection (POFD): which is the ratio between false alarms  
2 (policies detected wrongly by the model as being flooded) and the number of policies  
3 that have not been the object of a claim. A POFD of 0% would be ideal.

4 Both indicators have been calculated for Xynthia at two scales: the national and the  
5 communal scale. At the national scale, the results give a POD of 58% for a POFD of 4%,  
6 which means that the system missed a significant number of claims (42%) and selected a non-  
7 negligible number of policies that have not been affected. However, the comparison of these  
8 policies with areas identified as flood affected allows us to put in perspective our scores as  
9 actual inundated areas enabled us to detect only 72% of claims and also generated false claim  
10 reports (POFD=1.2%).

11 The results obtained at the communal scale, presented in Fig. 5, also give additional  
12 information. Firstly, they appear strongly correlated with the results of the hazard model  
13 presented previously. Thus, municipalities with high POD and low POFD are generally well  
14 simulated in terms of hazard. In the same way, when both POD and POFD are low, the extent  
15 of the inundation is generally underestimated. This is the case for the municipalities of La  
16 Flotte and Charron that should normally appear as some of the the most affected areas. It also  
17 appears that hazard is sometimes overestimated when both POD and POFD are high, notably  
18 for in the western part of the Isle de Ré or in the sector of Noirmoutier.

19 The major limit of the hazard model is the quality of topographic data. This is particularly  
20 visible in the Fig. 6 that shows a comparison between the DTM used in this study and the  
21 terrestrial part of the Litto-3D DTM (RGE ALTI® IGN) for the same cross section located in  
22 the commune of Yves (south part of the Fig. 3). The RGE ALTI DTM has been interpolated  
23 with a 25 meters resolution with respect to coastal protections levels. The differences between  
24 the two DTMs seem very significant with an average value of one metre and up to two metres  
25 at the location of the dike whose width, narrower than the 25 m resolution, does not allow it to  
26 be well represented in the DTM used. The analysis of this DTM shows two shortcomings.  
27 The first is that it generally underestimates the elevation of coastal protections, especially  
28 when they are narrower than the DTM resolution, and the second is that it seems generally  
29 "smoothed" in urban area where it overestimates the elevation of the ground. However, it has  
30 the advantage of being more accurate than the ASTER GDEM (Rexer and Hirt, 2014) and to  
31 be available all over the French territory, while the Litto-3D DTM is not yet implemented all

1 over the French territory. Furthermore, the implementation of hydraulic connections such as  
2 conduits or bridges also needs to be addressed with a good quality DTM.

3 The vulnerability model could also explain the scores obtained. The first limit is that the data  
4 inventory is not always exhaustive. Indeed, when losses are lower than the amount of the  
5 insurance franchise (deductible), claims are generally not declared. This situation could  
6 explain the high number of false alarm computed. The second limit comes from the  
7 localization of the policies. As indicated in Sect. 3.1, this loss localization is deduced from the  
8 address transmitted by the cedant. A comparison with the real locus of the policy (and  
9 therefore of the loss) has shown that about 30% of policies are in reality located at a distance  
10 greater than 50 meters of the estimated coordinates. The third limit of the vulnerability model  
11 is that it often does not distinguish secondary residences from main residences. This could  
12 generate a significant number of both false alarms and non-detections. Despite these  
13 problems, claims give interesting and original information by localizing areas that have been  
14 inundated but perhaps not taken into account in the areas otherwise inventoried.

15

## 16 **4 Damage model**

### 17 **4.1 Description**

18 The damage model aims to explain the cost that could be generated according to the  
19 magnitude of the phenomena and the vulnerability (policy exposure).

$$20 \quad C_{e,i} = P_{c,i} \cdot T_{d,i} \cdot V_{a,i} \quad (4)$$

21 The aggregation of claim amounts enables the computation of the whole event loss, but also  
22 the losses per region, per municipality or per insurance company. According to Eq.4 the claim  
23 loss estimation of one insurance policy alone is not consistent: the sum of claim probability  
24 will be a good estimation of the real number of claims at the event scale, but will not enable  
25 identification of the individual claim loss value.

26 The destruction rate is a relationship between the loss amount and the hazard intensity.  
27 Indeed, it is often assumed that the higher the water levels are, the greater the loss will be.  
28 Numerous studies have tried to adjust such models on the basis of damage curves describing  
29 the loss (cost or destruction rate) as a function of physical parameters such as water levels,

1 velocity, discharge, etc. (Scawthorn et al., 2006; Messner et al., 2007; Prettenthaler et al.,  
 2 2010; Huttenlau et al., 2010, Boettle et al., 2011). In the case of coastal flooding, losses could  
 3 result from the contact of saltwater with buildings -generating salt and corrosion problems or  
 4 from the mechanical action of waves or currents in the most exposed areas (André et al.,  
 5 2013).

6 These works often show a very large spread of the losses, especially for high water levels and  
 7 it appears that costs depend also of the nature of the risk type. In our study, eight classes of  
 8 risk have been selected to calibrate the destruction rates: the individual owner (house or  
 9 apartment), the individual tenant (house or apartment), the houses non occupant owner, the  
 10 building non occupant owner, the agricultural companies, and the other companies. For each  
 11 of these classes, the destruction rate has been calibrated according a square root relationship,  
 12 giving the best results in our case:

$$13 \quad T_{D,i} = a \cdot \sqrt{WD_i} \quad (6)$$

14 For instance, only water levels are taken into account into the loss probability and the damage  
 15 function. However, some other variables such as flow duration, flow velocity or the hour at  
 16 which the phenomenon occur could be tested in the future but this requires a larger sample  
 17 field of events.

18 The claim probability gives the probability of the insured location to be damaged according to  
 19 its water level during the event. Indeed, the data shows that a large proportion of goods  
 20 located in an inundated area are not declared as damaged. This absence of claim could come  
 21 from the eventual protection of the risk (e.g. issues such as local elevation being more  
 22 important, or the presence of natural or artificial protections, barriers, etc.) or to the low  
 23 proportion of material damages by comparison to the insurance deductible. The result of the  
 24 calibration adopting a logistic behaviour, the following equation is used to calculate the claim  
 25 probability:

$$26 \quad P_{c,i} = \frac{e^{a+b \cdot WD_i + c \cdot S}}{1 + e^{a+b \cdot WD_i + c \cdot S}} d \quad (5)$$

27 where WD corresponds to the sea water depth and a, b, c, d denote four dimensionless  
 28 parameters. The variable S, qualifying the surge, corresponds to the difference between the  
 29 maximum water level in sea and the water level corresponding to a two years return period.

1 This variable is calculated for each of the 20 available tide stations and interpolated along the  
2 coastline. This interpolation is not designed to predict accurately the storm surge along the  
3 coastline but to give an idea of the event magnitude in a given sector and thus to compensate  
4 the lack of precision of the hazard model. This formula provides an estimation of the number  
5 of claims by summing the probabilities.

6 The localization of the policy locus has not always required specific accuracy to calculate the  
7 hazard, especially when the localization is at communal or street level precision only. Finally,  
8 this model does not enable accurate individual claim statements, but the aggregation of costs  
9 at a larger scale produces a representative result.

## 10 **4.2 Calibration and validation of the damage model**

11 The calibration of the equations (5) and (6) was realized on a small sample of risks for the  
12 Xynthia event. The sample chosen was composed of two cedants presenting good quality data  
13 and exhaustive claims details for the chosen event. Among these policies, only those geo-  
14 localized at the roof top were taken into account. This sample represents 71,778 policies with  
15 736 claims for the coastal municipalities of the French departments affected by Xynthia.

16 The validation was realized by comparing the total cost for the 12 major cedants affected by  
17 the event- representing about 38% of the market. These total costs are not the sum of claims,  
18 but information directly transmitted by the cedants after the event, to overcome the  
19 problematic of non-exhaustive claims databases. Like the other studies in the public domain, a  
20 high dispersion is observed in the relationship between destruction rate and water height  
21 (Prettenthaler et al., 2010; André et al., 2013)

22 The results of the calibration exercise, presented in the Table 1, show that the validation  
23 allows a good estimation of the total cost for the calibration sample, with a little bias of  
24 0.02%. In the same way, the total losses simulated for the validation sample is near the  
25 observed total, with a bias of 0.36%. This bias, caused by risks that are not localized to the  
26 address as well as hydraulic modelling approximations, appears acceptable according to the  
27 different uncertainties of the sources.

28 Finally, the global cost of Xynthia is estimated at 695.6 million euros which is close of the  
29 observed cost than has reach 713 million euros.



### 1 **4.3 Sensitivity analysis of the damage model**

2 The evaluation of the hazard model discussed previously has shown its limitations, in  
3 particular the problems of the DTM quality as well as the limitations of the loss vulnerability  
4 database. These limitations are, from our point of view, inherent in the majority of operational  
5 systems at such spatial scales. The question is finally to know how the loss estimates that  
6 constitute the final modelling product are sensitive to these limitations.

7 According to de Moel et al. (2012), uncertainties in damage estimation are as important as  
8 uncertainties in hydraulic boundary conditions. A test was undertaken to evaluate the  
9 sensitivity of the damage model to water levels along the coastline. Thus, the input water  
10 levels in of the inundation model have been modified throughout the flood duration (i.e. not  
11 only on the maximum water level) on the basis of four scenarios: plus or minus 10 cm and  
12 plus or minus 20 cm.

13 The results of these simulations, presented in the Table 2, show that the uncertainties in  
14 coastal water level data are prone to generate very different results. Thus, a difference of 10  
15 cm in water level could vary the total loss by -10.6 to 17.8%. In the same way, a difference of  
16 20 centimeters in level induces a variation in the cost of about 27.3 to 30.3%. These important  
17 deviations are explained by the increase of the number of risk properties brought into the  
18 inundated area. In effect, as illustrated in Fig. 7, a modification of water levels of 10 to 20  
19 centimeters could significantly change the extent of the inundation and thus, the number of  
20 risk properties affected. These differences specifically concern the flattest areas but could  
21 have a great influence on the final insurance cost, especially when urban areas are concerned,  
22 because of insured property densities.

23 Upon examining these results, it can be assumed that the influence of the quality of the DTM  
24 is in the same order of magnitude as the sea water levels. If the DTM elevation along the  
25 coastline presents an error of a few centimeters, all the adjacent area can be affected in terms  
26 of water height and loss estimates. The sensitivity to the sea water levels also illustrates the  
27 need to take into account the wave setup so as not to underestimate the losses. The fact that  
28 the wave setup that could affect the sheltered bays are not taken into account by the model  
29 could partially influence the final result. This problem could be overcome by coupling a wave  
30 model with the surge model however it could be time expensive and thus, difficult to put in  
31 place in operational conditions.

1 The system developed has been used to estimate five more event losses: Lothar and Martin  
2 (1999), the storms Johanna that affected Brittany in 2008, the storm Xaver that occurred at the  
3 end of 2013 in the North of France and two events that occurred in the Bay of Biscay at the  
4 end of January and at the beginning of March 2014 (Christine). For these last three events, the  
5 incurred costs are still not available, but according to the observation, they should be low  
6 even if these three events have severely impacted the coastline with damage affecting  
7 beaches, roads and dikes. The results of the costs estimated for each event and for the market  
8 are presented in Table. 3.

9 Globally, the order of magnitude between the observations and the simulation is respected  
10 even if there are some differences. For instance, damages are overestimated by 23 % for  
11 Lothar and Martin and by 58.8 % for Johanna. These differences could be explained by the  
12 fact that the model was calibrated only for an extreme case and not for less severe events. For  
13 Xynthia, the bias between observation and simulated cost is lower of 2.5 %, representing a  
14 satisfactory performance. The low costs of the three most recent events appear to be  
15 consistent with observations which indicate low to moderate damages. Finally, even if the  
16 model could still be improved, the estimations already constitute a useful indicator of the  
17 magnitude and of the extension of an event.

18

## 19 **5 Conclusions**

20 The system presented in this article permits the estimation of insurance related losses due to  
21 coastal flooding events in France, especially on the Atlantic coast. The modelling chain, even  
22 if limited and still in an early stage of development, is, when applied on a large scale  
23 effective in satisfactorily estimating the magnitude of the resultant financial losses two to  
24 three days after the event occurrence. The estimations are consistent with the few historical  
25 events available (especially for Xynthia storm in 2010), and for calibration and validation of  
26 the approach.

27 The coastal flooding modelling at a large scale and within an operational perspective  
28 constitutes a difficult exercise that needs to match the scale of the output results and the  
29 accuracy and the complexity of the represented processes. The results emphasise the  
30 importance of the quality of topographic information, which constitutes one of the main  
31 limiting factors of the system. In order to improve this aspect, the use of other topographic

1 data such as LiDAR (Litto-3D) will be investigated in the future. The results obtained by the  
2 sensitivity tests underline also the need for a good estimation of sea water levels and of the  
3 wave setup process, although a future addition of a wave model should improve the system.  
4 The validation of the model is also limited by the lack of data due to the limited number of  
5 events in France, since the origin of the national Nat Cat scheme.

6 Finally, this study illustrates the usefulness of hydrodynamic models for an operational  
7 modelling of storm surge. In this situation, the insurance claims data constitutes a good  
8 indicator of the extent of the inundated areas which could help in the validation of sea surge  
9 models. The deterministic model, presented in this article, constitutes the basis of the  
10 probabilistic approach under development at CCR. This probabilistic model will be used to  
11 evaluate the exposure of the French coast to coastal flooding risk. It is based on the  
12 combination of fictive events and hazard scenarios generated by the inundation model. This  
13 probabilistic model should allow estimation of the financial exposure of CCR, as reinsurer,  
14 the French state and of insurance companies to storm surge events and losses in the future, but  
15 also the potential consequences of sea level rises within climate change scenarios.

16

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22

## 1 **References**

- 2 André, C., Monfort, D., Bouzit, M., Vinchon, C.: Contribution of insurance data to cost  
3 assessment of coastal flood damage to residential buildings: insights gained from Johanna  
4 (2008) and Xynthia (2010) storm events. *Natural Hazards and Earth System Science* 13,  
5 2003–2012, 2013.
- 6 Ardhuin, F., Rogers, A., Babanin, A., Filipot, J. F., Magne, R., Roland, R., Westhuysen, A. V.  
7 D., Queffleulou, P., Lefevre, L., Aouf, L., and Collard, F.: Semi-empirical dissipation source  
8 functions for ocean waves: 5 Part I, definition, calibration and validation, *J. Phys. Oceanogr.*,  
9 40, 1917–1941, 2010.
- 10 Aronica, G., Bates, P. D., Horritt, M. S.: Assessing the Uncertainty in Distributed Model  
11 Predictions Using Observed Binary Pattern Information within GLUE. *Hydrological*  
12 *Processes* 16, 2001-2016, 2002.
- 13 Bates, P.D., Dawson, R.J., Hall, J.W., Horritt, M.S., Nicholls, R.J., Wicks, J., Mohamed  
14 Ahmed Ali Mohamed Hassan Simplified two-dimensional numerical modelling of coastal  
15 flooding and example applications. *Coastal Engineering* 52, 793–810, 2005.
- 16 Bates, P.D., Horritt, M.S., Fewtrell, T.J.: A simple inertial formulation of the shallow water  
17 equations for efficient two-dimensional flood inundation modelling. *Journal of Hydrology*  
18 387, 33–45, 2010.
- 19 Bertin, X., Bruneau, N., Breilh, J.F., Fortunato, A.B., Karpytchev, M.: Importance of wave  
20 age and resonance in storm surges: the case Xynthia, Bay of Biscay. *Ocean Modelling* 42, 16-  
21 30, 2012.
- 22 Bertin, X., Li, K., Roland, A., Bidlot, J.R.: The contribution of short-waves in storm surges:  
23 Two case studies in the Bay of Biscay. *Continental Shelf Research* 96, 1-15, 2015.
- 24 Bertin, X., Li, K., Roland, A., Zhang, Y.L.J., Breilh, J.F., Chaumillon, E.: A modeling-based  
25 analysis of the flooding associated with Xynthia, central Bay of Biscay. *Coastal Engineering*  
26 94, 80–89., 2014.
- 27 Boettle, M., Kropp, J.P., Reiber, L., Roithmeier, O., Rybski, D., Walther, C., 2011. About the  
28 influence of elevation model quality and small-scale damage functions on flood damage

1 estimation. *Natural Hazards and Earth System Science* 11, 3327–3334. doi:10.5194/nhess-11-  
2 3327-2011

3 Bouwer, L.M., Bubeck, P., Wagtendonk, A.J., Aerts, J.C.J.H.: Inundation scenarios for flood  
4 damage evaluation in polder areas. *Natural Hazards and Earth System Science* 9, 1995–2007,  
5 2009.

6 Breilh, J.F., Chaumillon, E., Bertin, X., Gravelle, M.; Assessment of static flood modeling  
7 techniques: application to contrasting marshes flooded during Xynthia (western France).  
8 *Natural Hazards and Earth System Science* 13, 1595–1612. 2013.

9 Courtier P., Thepaut J., Hollingsworth A. A strategy for operational implementation of 4D-  
10 VAR, using an incremental approach. *Q J R Meteorol Soc* 120:1367–1387, 1994.

11 Czajkowski, J., Kunreuther, H., Michel-Kerjan, E. Quantifying Riverine and Storm-Surge  
12 Flood Risk by Single-Family Residence: Application to Texas: Quantifying Catastrophic  
13 Flood Risk in Texas. *Risk Analysis* 33, 2092–2110. doi:10.1111/risa.12068, 2013.

14 De Moel, H., Asselman, N.E.M., Aerts, J.C.J.H.: Uncertainty and sensitivity analysis of  
15 coastal flood damage estimates in the west of the Netherlands. *Natural Hazards and Earth  
16 System Science* 12, 1045–1058, 2012.

17 Ferrarin, C., Roland, A., Bajo, M., Umgiesser, G., Cucco, A., Davolio, S., Buzzi, A.,  
18 Malguzzi, P., and Drofa, O.: Tide-surge-wave modelling and forecasting in the Mediterranean  
19 Sea with focus on the Italian coast, *Ocean Modelling*, 61, 38 – 48, 2013.

20 FFSA (Fédération Française des Sociétés d'Assurance) and GEMA (Groupement des  
21 Entreprises Mutuelles d'Assurance): La tempête Xynthia du 28 février 2010 – Bilan chiffre au  
22 31 décembre 2010, 19 pp., available at:  
23 <http://www.ffsa.fr/sites/upload/docs/application/pdf/2011-06/bilanxynthia28022011.pdf>, 2011

24 Gaslikova, L., Schwerzmann, A., Raible, C.C., Stocker, T.F. Future storm surge impacts on  
25 insurable losses for the North Sea region. *Natural Hazards and Earth System Science* 11,  
26 1205–1216. doi:10.5194/nhess-11-1205-2011, 2011.

27 Genovese, E., Hallegatte, S., Dumas, P.: Damage Assessment from Storm Surge to Coastal  
28 Cities: Lessons from the Miami Area, in: Geertman, S., Reinhardt, W., Toppen, F. (Eds.),  
29 *Advancing Geoinformation Science for a Changing World*. Springer Berlin Heidelberg,  
30 Berlin, Heidelberg, pp. 21–43, 2011.

1 Holman, R. A., Sallenger, A. H.: «Setup and Swash on a Natural Beach. *Journal of*  
2 *Geophysical Research*, 90, 945-953. 1985.

3 Horritt, M.S., Bates, P.D.: Predicting floodplain inundation: raster-based modeling versus the  
4 finite-element approach. *Hydrological Processes* 15, 825–842, 2001.

5 Hunter, N.M., Horritt, M.S., Bates, P.D., Wilson, M.D., Werner, M.G.F.: An adaptive time  
6 step solution for raster-based storage cell modelling of floodplain inundation. *Advances in*  
7 *Water Resources* 28, 975–991, 2005.

8 Huttenlau, M., Stötter, J., Stiefelmeyer, H. Risk-based damage potential and loss estimation of  
9 extreme flooding scenarios in the Austrian Federal Province of Tyrol. *Natural Hazards and*  
10 *Earth System Science* 10, 2451–2473. doi:10.5194/nhess-10-2451-2010, 2010.

11 Idier, D., Dumas, F., Muller, H. Tide-surge interaction in the English Channel. *Natural*  
12 *Hazards and Earth System Science* 12, 3709–3718. doi:10.5194/nhess-12-3709-2012, 2012.

13 Jongman, B., Kreibich, H., Apel, H., Barredo, J.I., Bates, P.D., Feyen, L., Gericke, A., Neal,  
14 J., Aerts, J.C.J.H., Ward, P.J.: Comparative flood damage model assessment: towards a  
15 European approach. *Natural Hazards and Earth System Science* 12, 3733–3752, 2012.

16 Kim, S. Y., Yasuda, T., and Mase, H.: Numerical analysis of effects of tidal variations on  
17 storm surges and waves, *Applied Ocean Research*, 30, 311 – 322,  
18 doi:http://dx.doi.org/10.1016/j.apor.2009.02.003, 2008.

19 Lazure, P., Dumas, F. : An external–internal mode coupling for a 3D hydrodynamical model  
20 for applications at regional scale (MARS). *Advances in Water Resources* 31, 233–250, 2008.

21 Lecacheux, S., Pedreros, R., Le Cozannet, G., Thiébot, J., De La Torre, Y., Bulteau, T.: A  
22 method to characterize the different extreme waves for islands exposed to various wave  
23 regimes: a case study devoted to Reunion Island. *Natural Hazards and Earth System Science*  
24 12, 2425–2437. doi:10.5194/nhess-12-2425-2012, 2012

25 Le Roy, S., Pedreros, R., André, C., Paris, F., Lecacheux, S., Marche, F., and Vinchon, C.:  
26 Coastal flooding of urban areas by overtopping: dynamic modelling application to the  
27 Johanna storm (2008) in Gâvres (France), *Natural Hazards and Earth System Sciences*  
28 *Discussions*, 2, 4947–4985, doi:10.5194/nhessd-2-4947-2014, 2014.

1 Lopes, C., Azevedo, A., and Dias, J.: Flooding assessment under sea level rise scenarios: Ria  
2 de Aveiro case study, *Journal of Coastal Research*, Special Issue No. 65, pp. 766–771, 2013.

3 Lumbroso, D.M., Vinet, F. : A comparison of the causes, effects and aftermaths of the coastal  
4 flooding of England in 1953 and France in 2010. *Natural Hazards and Earth System Science*  
5 11, 2321–2333, 2011.

6 Lyard, F., Lefevre, F., Letellier, T., and Francis, O.: Modelling the global ocean tides: modern  
7 insights from FES2004, *Ocean Dynamics*, 56, 394–415, doi:10.1007/s10236-006-0086-x,  
8 2006.

9 Meyer, V., Becker, N., Markantonis, V., Schwarze, R., van den Bergh, J.C.J.M., Boucher,  
10 L.M., Bubeck, P., Ciavola, P., Genovese, E., Green, C., Hallegatte, S., Kreibich, H., Lequeux,  
11 Q., Logar, I., Papyrakis, E., Pfuerscheller, C., Poussin, J., Przyluski, V., Thielen, A.H.,  
12 Viavattene, C.: Review article: Assessing the costs of natural hazards – state of the art and  
13 knowledge gaps. *Natural Hazards and Earth System Science* 13, 1351–1373, 2013.

14 Meyer, V., Messner, F.: National flood damage evaluation methods: A review of applied  
15 methods in England, the Netherlands, the Czech Republic and Germany (UFZ Discussion  
16 Papers No. 21/2005), 2005.

17 Moncoulon, D., Labat, D., Ardon, J., Leblois, E., Onfroy, T., Poulard, C., Aji, S., Rémy, A.,  
18 Quantin, A. : Analysis of the French insurance market exposure to floods: a stochastic model  
19 combining river overflow and surface runoff. *Natural Hazards and Earth System Science* 14,  
20 2469–2485. 2014.

21 Muller, H., Pineau-Guillou, L., Idier, D., Ardhuin, F.: Atmospheric storm surge modeling  
22 methodology along the French (Atlantic and English Channel) coast. *Ocean Dynamics* 64,  
23 1671–1692. 2014.

24 Pistrika, A.K., Jonkman, S.N.: Damage to residential buildings due to flooding of New  
25 Orleans after hurricane Katrina. *Natural Hazards* 54, 413–434, 2009.

26 Pretenthaler, F., Amrusch, P., Habsburg-Lothringen, C.: Estimation of an absolute flood  
27 damage curve based on an Austrian case study under a dam breach scenario. *Natural Hazards*  
28 and *Earth System Science* 10, 881–894, 2010.

29 Rexer, M., Hirt, C.: Comparison of Free High Resolution Digital Elevation Data Sets  
30 (ASTER GDEM2, SRTM v2.1/v4.1) and Validation against Accurate Heights from the

1 Australian National Gravity Database. *Australian Journal of Earth Sciences*, 61, 213-226,  
2 2014.

3 Scawthorn, C., Flores, P., Blais, N., Seligson, H., Tate, E., Chang, S., Mifflin, E., Thomas,  
4 W., Murphy, J., Jones, C., Lawrence, M.: HAZUS-MH Flood Loss Estimation Methodology.  
5 II. Damage and Loss Assessment. *Natural Hazards Review* 7, 72–81, 2006.

6 Seity Y., Brousseau P., Malardel S., Hello G., Bénard P., Bouttier F., Lac C., Masson V. The  
7 AROME-France convective scale operational model. *Mon Wea Rev* 139:976–991. 2011.

8 SHOM: *Références Altimétriques Maritimes*, Tech. rep., 2012.

9 Sousounis, P., Kafali, C., Insured loss estimation from wind and storm surge for a re-  
10 occurrence of typhoon Vera. Presented at the 29th Conference on Hurricanes and Tropical  
11 Meteorology, Tucson, p. 8. , 2010

12 Simon, G., Dumas, F., Duhaut, T. : Spectral analysis of mean flow and turbulence forced by  
13 waves in a horizontally homogeneous zone of the Iroise sea. *Ocean Dynamics*, 61, 1887–  
14 1903. 2011.

15 Stockdon, H.F., Holman, R.A., Howd, P.A., Sallenger, A.H. Empirical parameterization of  
16 setup, swash, and runup. *Coastal Engineering* 53, 573–588.  
17 doi:10.1016/j.coastaleng.2005.12.005, 2006.

18 Tolman, H., Chalikov, D.: Source terms in a third-generation wind wave model. *Journal of*  
19 *Physical oceanography* 26 2497–2518. 1996.

20 Townend, I. and Pethick, J.: Estuarine flooding and managed retreat. *Philos. Trans. R. Soc. A*  
21 *Math. Phys. Eng. Sci.* 360 (1796), 1477–1495, 2002.

22 Waeles, B., Bertin, X., Breilh, J.-F., Li, K., Le Mauff Dorn, B.: Limitation of high water  
23 levels in bays and estuaries during storm flood events, in: *SimHydro 2014*. Presented at the  
24 *Modelling of rapid transitory flows*, Sophia Antipolis, 9 p., 2014

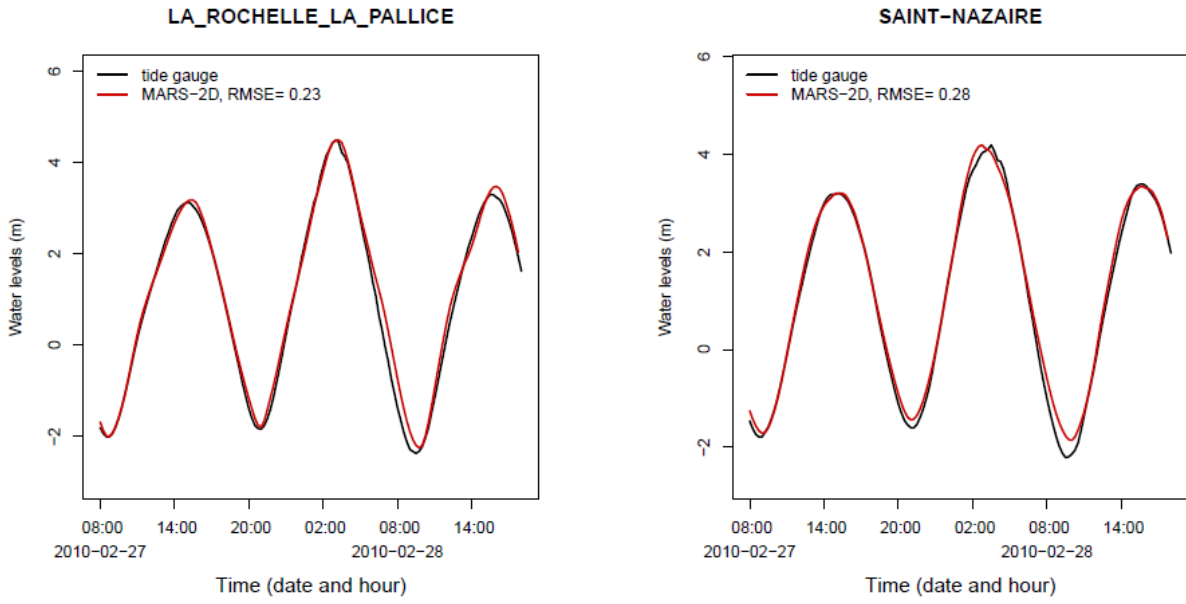
25 Wang, H., Loftis, J., Liu, Z., Forrest, D., Zhang, J.: The Storm Surge and Sub-Grid Inundation  
26 Modeling in New York City during Hurricane Sandy. *Journal of Marine Science and*  
27 *Engineering* 2, 226–246, 2014.

28 Wilks, D.S.: *Statistical methods in the atmospheric sciences*. Academic Press, Oxford;  
29 Waltham, MA, 2011.

30

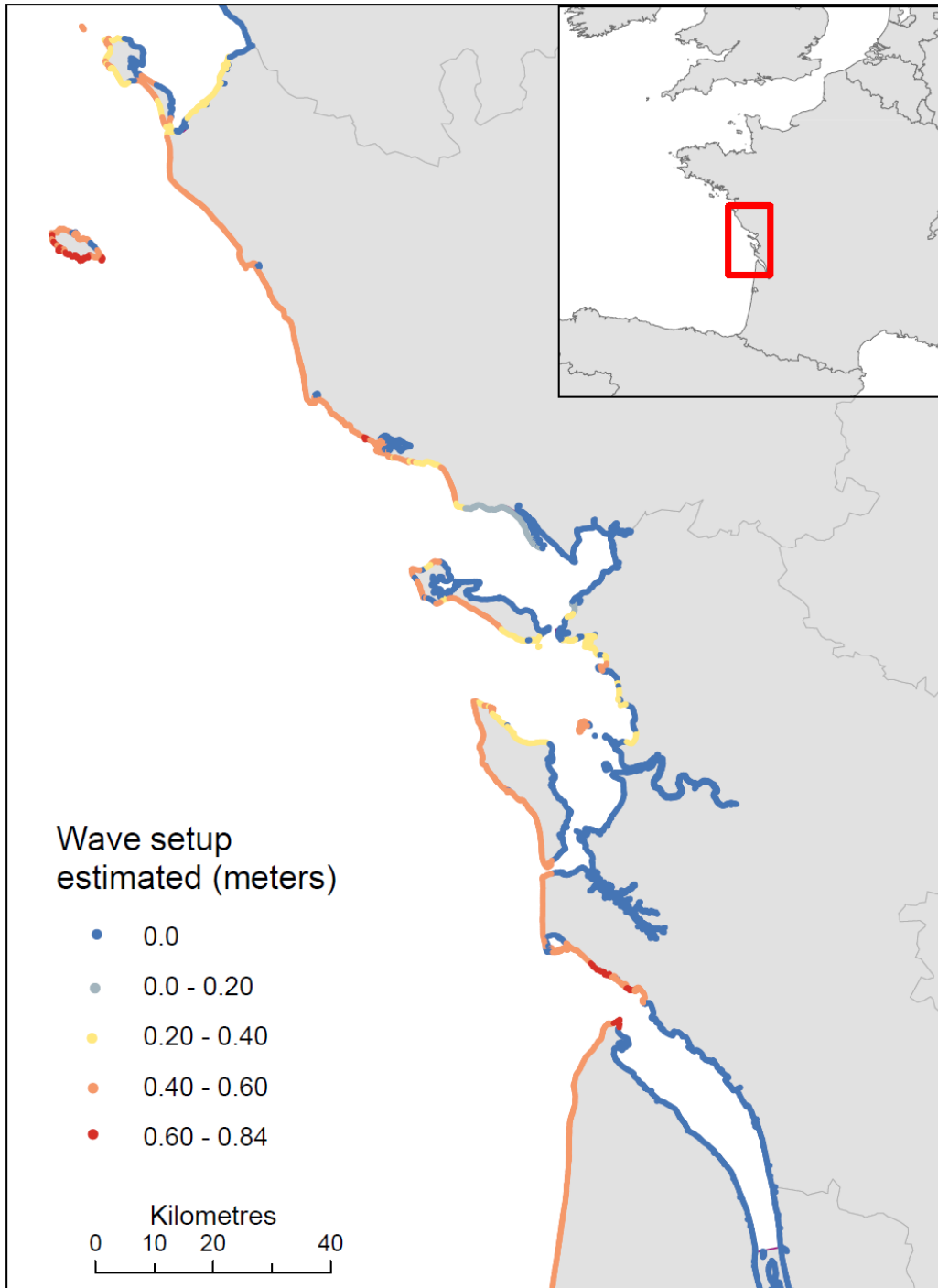


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2

3 Figure 1. Comparison between the observed tides and the water levels estimated by the  
4 PREVIMER for the Xynthia event at two gauge stations after maximal water levels  
5 adjustment.

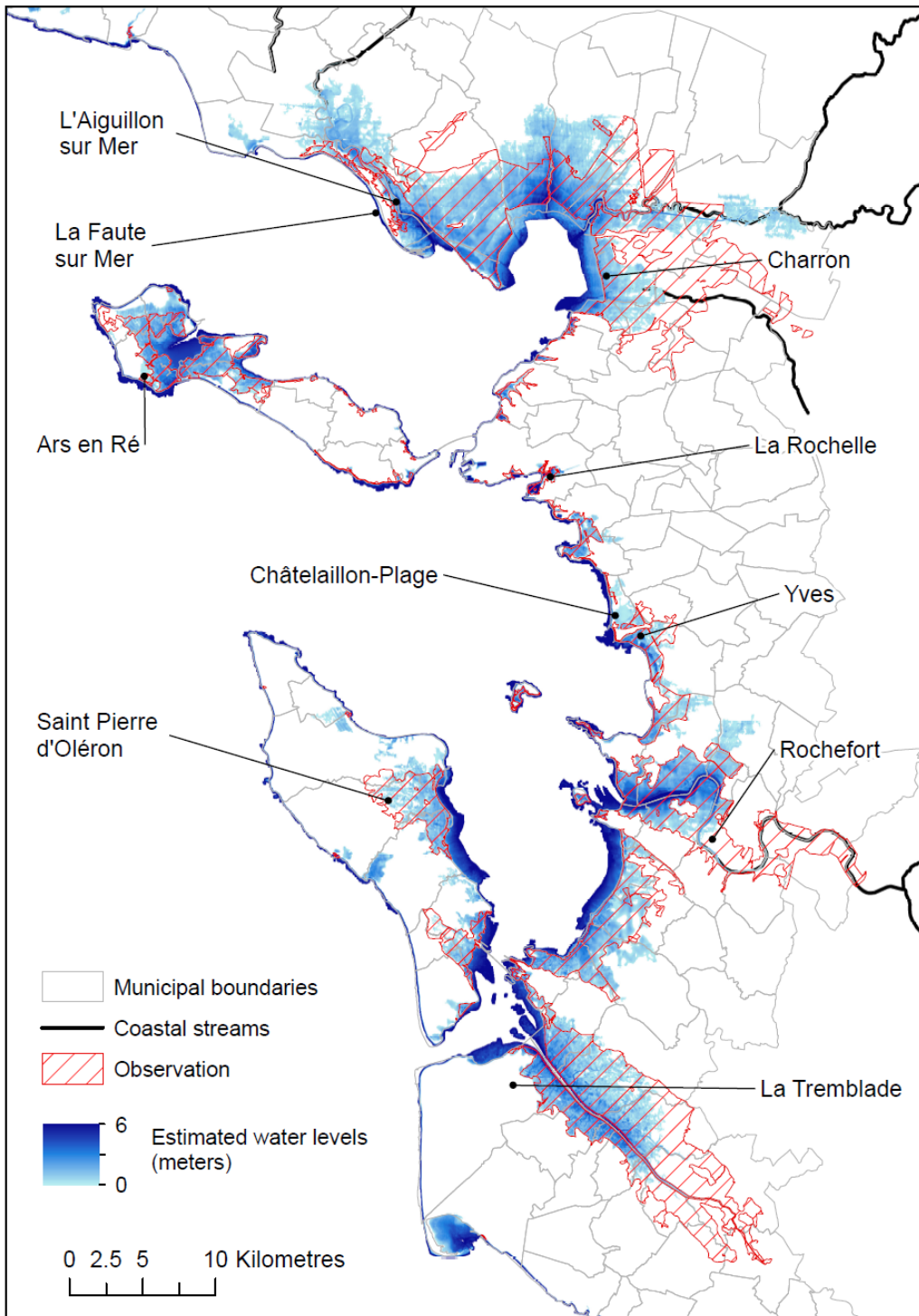


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2 Figure 2. Maximal estimated local wave setup, computed according to Stockdon et al. (2006)  
 3 along the coastline for the Xynthia event.

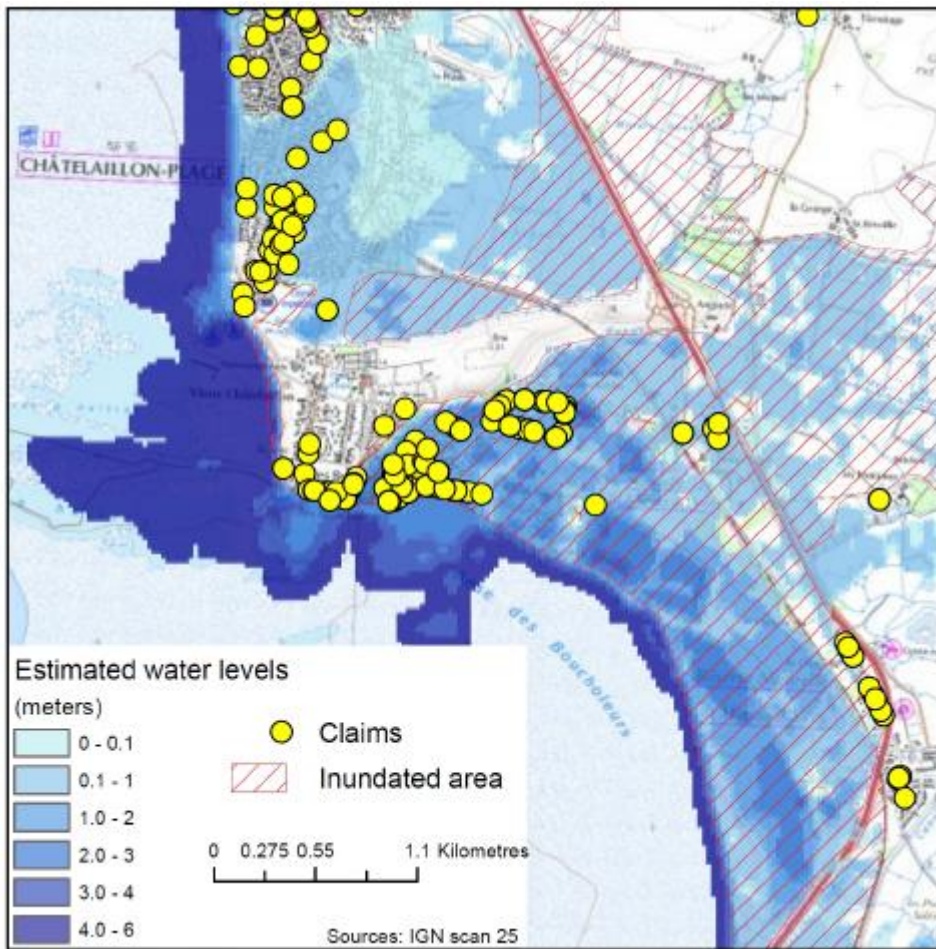
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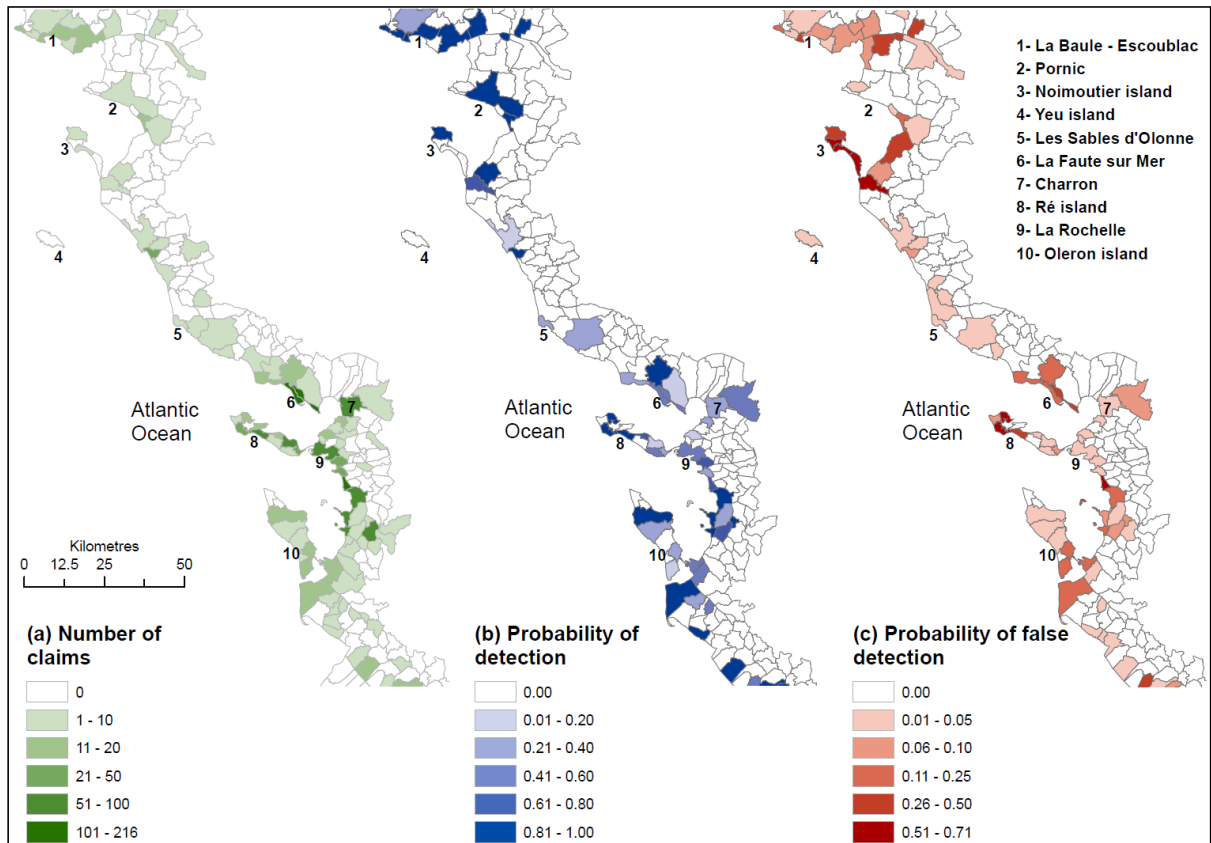
3 Figure 3. Simulated and observed inundated areas for the Charente Maritime County for the  
4 Xynthia event.



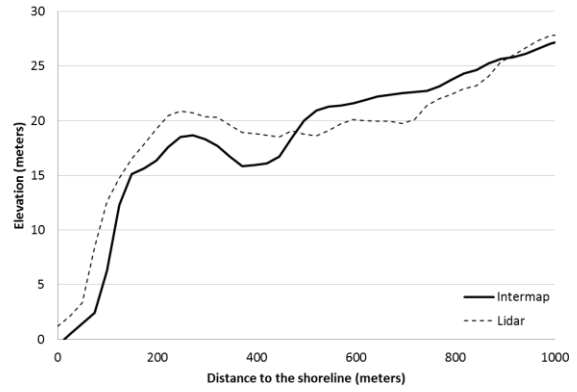
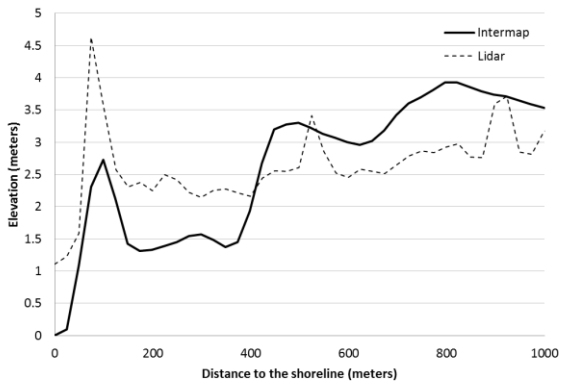
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2 Figure 4. Claims and inundated areas for the Chatelaillon-Plage municipality during Xynthia.

3



1  
 2 Figure 5. Performances of the system, obtained by comparing claims and insurance policies to  
 3 the water levels simulated.  
 4

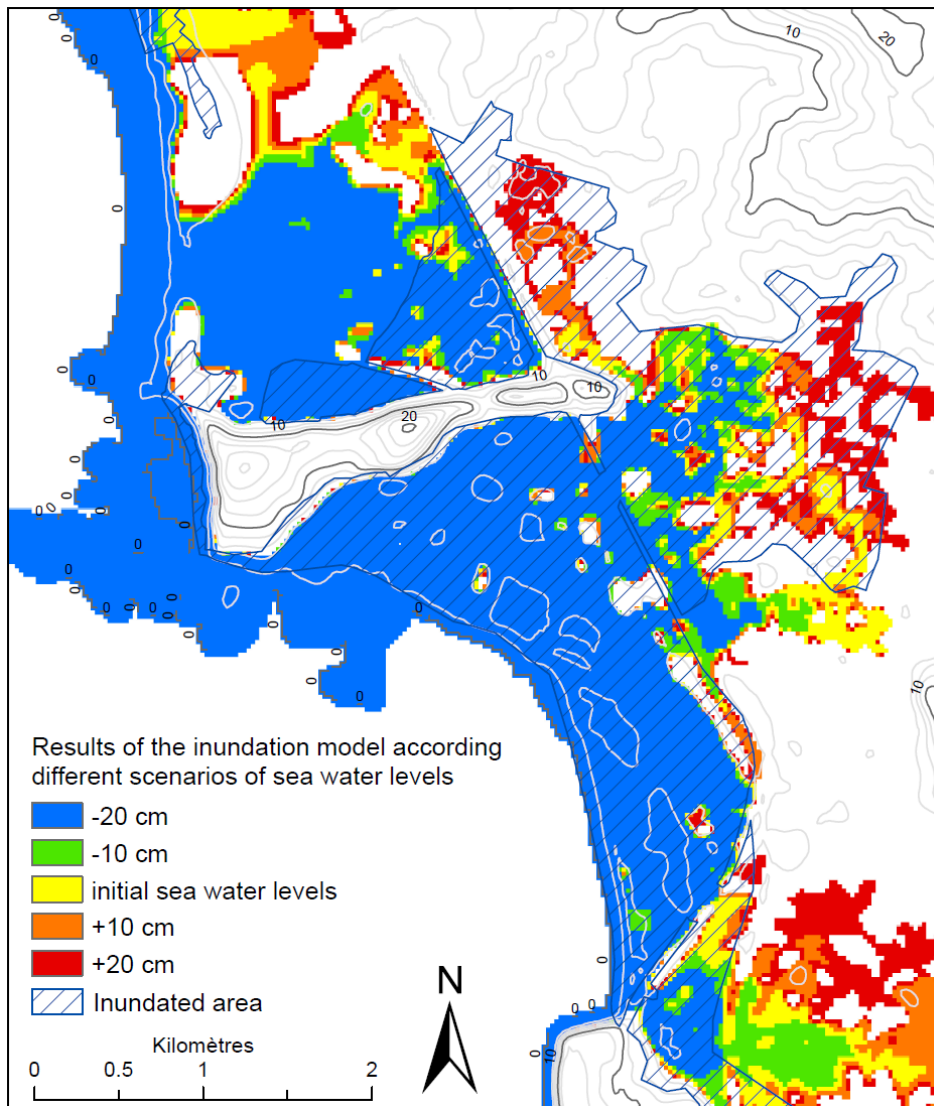


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2 Figure 6. Comparison between the DTM used in this study and a DTM elaborated from Lidar  
 3 observations, interpolated at a 25 meters resolution, for a cross section located in the  
 4 Boucholeur area (left) and near the Concarneau city (right).

5

6



1  
2 Figure 7. Results of the sensitivity test of the inundation model to the water levels estimated  
3 for the sea.

4  
5

1 Table 1. Observed and simulated losses at country scale.

	Simulated (Million euros)	Observed (Million euros)	Bias
Calibration sample	62.60	62.58	0.02%
Validation sample	218.74	217.96	0.36%
Total	281.34	280.54	0.28%

2

3

4 Table 2. Losses computed at country scale on the basis of several inundation scenarios for the  
5 calibration sample.

	Difference between the scenario and the initial water levels (centimeters)				
	-20	-10	0	+10	+20
Losses (Million euros)	484	571	696	769	885
Bias	-30.3	-17.8	0	10.6	27.3

6

7 Table 3. Comparison between the losses observed and simulated at country scale on five  
8 recent events for the market. The losses of the Xaver event and the two 2014 events are not  
9 yet available.

Losses (Million euros)	Lothar-Martin 1999	Johanna 2008	Xynthia 2010	Xaver 2013	Storm 30/01/2014	Christine 2014
Observed	45.3	16.2	713	-	-	-
Simulation	59.5	10.2	695.6	10.9	3.8	7.7

10