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# A coastal vulnerability assessment for planning climate resilient infrastructure

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1 **A coastal vulnerability assessment for planning climate resilient infrastructure.**

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15  
16 **Highlights**

- 17 • Changing threats to coastal populations and infrastructure are found.  
18 • Features that enable coastal resilience are identified.  
19 • An approach to develop a stakeholder-focussed decision-support tool is presented.  
20 • Physical process understanding and real options analysis are combined.

21  
22 **Abstract**

23 There is a good understanding of past and present coastal processes as a result of coastal  
24 monitoring programmes within the UK. However, one of the key challenges for coastal managers  
25 in the face of climate change is future coastal change and vulnerability of infrastructure and  
26 communities to flooding. Drawing on a vulnerability-led and decision-centric framework (VL-DC)  
27 a Decision Support Tool (DST) is developed which, combines new observations and modelling to  
28 explore the future vulnerability to sea-level rise and storms for nuclear energy sites in Britain. The  
29 combination of these numerical projections within the DST and a Real Options Analysis (ROA)  
30 delivers essential support for: (i) improved response to extreme events and (ii) a strategy that  
31 builds climate change resilience.

32

33

34 **Key words:** Decision Support Tool (DST); Real Options Analysis (ROA); Flood hazard modelling;  
35 Storm impact monitoring; Human intervention.

36

### 37 1. Introduction

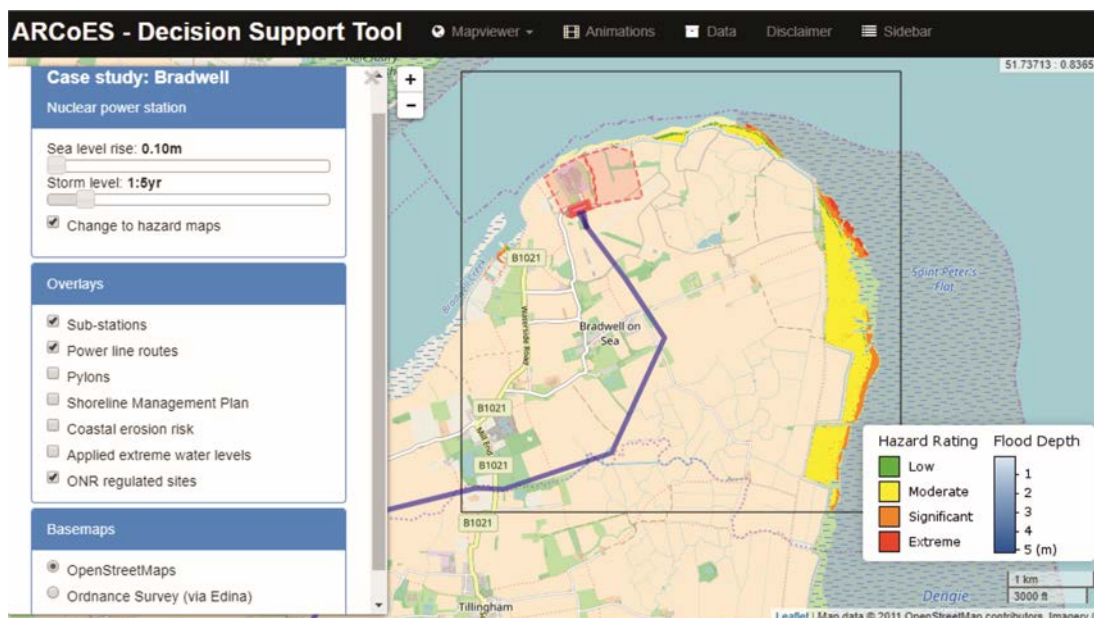
38 Energy security is a fundamental requirement for well-functioning modern societies (Morrissey et  
39 al., 2018). Due to its prevalent location in coastal areas, climate change, sea-level rise and extreme  
40 events represent significant challenges to the global energy infrastructure and supply chain  
41 (Reichl et al., 2013; Morrissey et al., 2018; Prime et al., 2018). The UK Energy Networks  
42 Association (ENA) identifies the biggest pressure to be from coastal flooding - if an electrical  
43 substation is flooded costs in clean up and repair can be high, and on-going costs from disruption  
44 and loss of supply have the potential to add to this significantly (Energy Network Association,  
45 2009). There is already a good understanding of past and present coastal processes, particularly  
46 at locations for present and planned nuclear power stations. However, to ensure that coastal  
47 populations and the necessary infrastructure required to sustain these populations are resilient  
48 in the future, tools that can inform adaptive management are required (Silva et al., 2017; Wadey  
49 et al., 2017; Lam et al., 2017). However, this is a complex problem as shoreline resilience to  
50 changes in the physical environment varies spatially and temporally in response to factors such as  
51 changing beach volume (Castelle et al., 2015), reduction in sediment supply (Guangwei, 2011),  
52 and the degradation of coastal wetlands (Lotzel et al., 2006), as well as to human interventions  
53 that are socio-economically, politically and culturally determined (Ratter et al., 2016). To be  
54 effective, management tools require the capacity to monitor and project a variety of interlinked  
55 physical and societal processes including sea-level rise, storm magnitude/frequency relationships,  
56 changing sediment budget (Brown et al., 2016) and population change and economic activity  
57 (Prime et al., 2018).

58

59 Developed for the UK energy sector as part of the Adaptation and Resilience of Coastal Energy  
60 Supply (ARCoES) project, this paper presents a web-based geospatial Decision-Support Tool (DST),  
61 the ARCoES DST (Fig. 1). Leaflet, an open source Javascript library, is used to construct the DST to  
62 enable the end user to interrogate the matrix of model results using slider bars and tick box  
63 options to toggle between hazard or inundation maps and overlay different infrastructure or map  
64 views (Knight et al., 2015). As described in this paper, the ARCoES DST is used in combination with

65 modelling and monitoring of different coastal environments to better understand future coastal  
66 vulnerability. Drawing on the interdisciplinary skills of the ARCoES researchers, the ARCoES DST is  
67 combined with an economic framework, Real Options Analysis (ROA), to provide an assessment  
68 of when it is most cost-effective to implement a new management approach. From a policy  
69 perspective, the data produced by the DST, when combined with a Real Options Framework can  
70 be used to initiate discussions with coastal practitioners to identify how future vulnerability to  
71 coastal flooding may be mitigated through appropriate and timely intervention and adaptation.  
72 Importantly, although the methodology is designed for the nuclear energy sector the DST could  
73 also be applied for other coastal management needs.

74



75

76 Fig. 1. The ARCoES DST, available at <http://arcoes-dst.liverpool.ac.uk/>.

77

78 Within this context, the aim of this paper is to demonstrate the usefulness of the ARCoES DST in  
79 understanding the physical and economic impact of sea-level rise and storms across 4 nuclear  
80 energy sites located along the coast of the UK. These sites include Seascale (representing Sellafield  
81 in the northwest), Lilstock (representing Hinkley Point in the southwest), Sizewell (in the east),  
82 and Bradwell (in the southeast). We also focus on Fleetwood (in the northwest) as an example of  
83 its application to a coastal community. The paper continues as follows: the methods used to  
84 deliver this holistic assessment are presented in Section 2. In Section 3 a selection of results to  
85 demonstrate the application and capabilities of the resulting DST at different sites is provided. The  
86 way in which this DST can be used to conceptualize shoreline management requirements to pose

87 questions at a high level for specialized studies to address is discussed in Section 4, before the  
88 conclusions about the future resilience of UK coastal energy are drawn in Section 5.

89

## 90 **2. Site Descriptions**

91 Although applied to a number of locations, here we focus on five study sites with different coastal  
92 geomorphology and hazard exposure. This national application demonstrates the development of  
93 a DST for the management needs of an industry with infrastructure in multiple locations rather  
94 than in response to site-specific coastal conditions. Each site requires a slightly different model  
95 configuration (see Section 3) but uses the same approach.

96

97 The coastline at Seascale/Sellafield faces the Irish Sea, the actual location is quite exposed  
98 (offshore  $H_s, 10\% = 2$  m; max  $H_s = 5.7$  m; data from British Oceanographic Data Centre (BODC) wave  
99 buoy MCMBE-OFF 1974–1976), with a maximum tide range and 1% storm surge height during  
100 winter of 7 m and 1 m, respectively. However, the beach morphology fronting the facility is  
101 characterised by a reflective high tide gravel/cobble beach with an extremely dissipative sandy  
102 intertidal zone. A storm monitored in January 2013 that more or less coincided with spring high  
103 tide had therefore insignificant impact on the beach (Almeida et al., 2014).

104

105 At Lilstock/Hinkley Point, located in the Bristol Channel, the site is not fully exposed to the Atlantic  
106 waves, but wave conditions can be relatively energetic (offshore  $H_s, 10\% = 1.8$  m; max  $H_s = 3.7$  m;  
107 data from BODC wave buoy SEVERNEST A 1979–1981). This is a mega-tidal environment with a  
108 maximum tide range of 10.7 m and a 1% storm surge height during winter of 0.8 m. However, in  
109 common with Sellafield, the wide and low gradient intertidal zone, here a rocky platform instead  
110 of a sandy beach, is extremely dissipative, limiting the wave energy levels impacting the high tide  
111 gravel/cobble beach. A storm monitored in December 2013 had therefore very limited  
112 morphological impact.

113

114 The gravel beach at Sizewell faces the North Sea. Wave conditions are relatively mild (offshore 10%  
115 exceedance  $H_s = 0.6$  m; max  $H_s = 2.2$  m; data from BODC wave buoy ALDEBURG 1975–1977) and  
116 the maximum tide range and 1% storm surge height during winter are 2.4 m and 1 m, respectively.  
117 During the 5-year duration of the ARCoES project, not a single extreme wave event occurred at  
118 Sizewell, but some measurements were made during a relatively modest storm event in March

119 2013. These revealed that the subtidal bar morphology at this site provides significant protection  
120 to the high tide gravel beach from large waves and that the main morphological changes occurred  
121 due to longshore sediment transport processes. The most significant wave events along the North  
122 Sea coast are from the northeast quadrant, but Sizewell is partly sheltered from such storms  
123 because the coastline aligns south-southwest to north-northeast, and potentially the most  
124 damaging waves for Sizewell are extremely rare storm waves from the southeast. Interestingly,  
125 the storm surge event in 2013, and which caused much erosion and flooding along the east coast  
126 of England (Wadey et al., 2015), was not an event of significance at Sizewell where  $H_s$  at the peak  
127 of the storm surge were  $< 1.5$  m.

128

129 The Bradwell site is characterised by a narrow gravel coastal plain fronted by the silty tidal flat and  
130 is located on the southern bank of the Blackwater estuary. The maximum tide range here is 4.8 m  
131 and the 1% winter storm surge is 0.9 m. The site is extremely sheltered and this is demonstrated  
132 by the results of a long-term deployment (Oct 2015 –Mar 2016) of pressure sensors at the base  
133 of the gravel beach and around low tide level. Mean wave conditions were characterised by  $H_s =$   
134 0.1 m and the most energetic event that occurred during this period had a  $H_s$  of 0.45 m.

135

136 Observing the physical processes at the sites above has found that they have a low vulnerability  
137 to storm impact. Seascale/Sellafield and Lilstock/Hinkley Point are relatively exposed sites, the key  
138 aspect limiting their vulnerability to extreme wave events is their highly dissipative intertidal zone  
139 (sand at Sellafield and rock at Hinkley Point). The very wide ( $> 200$  m) and low-gradient ( $< 0.015$ )  
140 surface fronting the high tide gravel/cobble beach and coastal structures at both sites greatly  
141 reduces the wave energy levels and wave runup around high tide, and therefore the risk of  
142 flooding and erosion, even under the largest offshore waves. Sizewell is sited such that it is not  
143 exposed to the most frequent North Sea storm wave conditions from the northeast quadrant. In  
144 addition, the low gradient and barred subtidal zone effectively dissipates storm wave energy, and  
145 the high and wide inter- and supratidal gravel beach also provides a significant buffer to extreme  
146 wave action. The site is perhaps most vulnerable to longer-term coastal dynamics, specifically  
147 alongshore redistribution of sand and gravel due to littoral drift. Bradwell is sited in an extremely  
148 sheltered location with very limited fetch and potential for wave generation. A low gradient  
149 subtidal zone and gravel ridges also fronts the facility, which adds additional protection.

150

151 In addition to sites of nuclear infrastructure the ARCoES DST was also developed to assess  
152 community vulnerability to coastal hazards. Our example site at Fleetwood, northwest England, is  
153 used here to demonstrate how flood hazard management of a community's electricity distribution  
154 has to consider the influence of shoreline management plans on the inland flood hazard to  
155 electricity substations to ensure the supply is resilient. The coastal conditions at this site include  
156 a mega-tidal regime (exceeding 10 m during spring tides), surge events that can reach 2 m and  
157 offshore wave conditions that can exceed 5.5 m (Brown et al., 2010). Our study region has a 'hold  
158 the line' shoreline management policy to protect the community from flood hazards. Within our  
159 study area this policy is implemented by a sea wall, thus understanding when a future 'tipping  
160 point' in wave overtopping hazard may occur for the existing scheme under rising sea levels is  
161 important.

162

### 163 **3. ARCoES DST**

164 There is often a good understanding of past and present coastal processes as a result of coastal  
165 monitoring programmes within the UK. However, one of the key challenges for managers in the  
166 face of climate change, is future coastal change and vulnerability of infrastructure and  
167 communities to flooding. A vulnerability-led and decision-centric framework (VL-DC) (Armstrong  
168 et al., 2015), the ARCoES approach combines new observations and modelling to explore the  
169 future vulnerability to sea-level rise and storms for nuclear energy sites in Britain. As will be  
170 outlined below, the resulting DST provides inundation mapping via LISFLOOD-FP, XBeach, XBeach-  
171 G and SWAB modelling. The data are then combined in a ROA framework to provide an  
172 assessment of when it is most cost-effective to implement a new management approach.

173

#### 174 *3.1 Inundation Mapping*

175 Inundation mapping is a key component of the ARCoES DST. While a general overview of the  
176 model application is provided here, more detailed studies focusing on individual sites (e.g., Prime  
177 et al., 2015a; 2015b; 2016) have considered sensitivity analysis of the model results to ensure the  
178 approach is robust for the purpose of the DST. A "soft" coupling approach is adopted where a  
179 storm impact model provides the input to an inundation model. Here we use models that are  
180 frequently used in flood and erosion risk studies (e.g., Lewis et al., 2013; Phillips et al., 2017; Poate  
181 et al., 2016).

182

183 LISFLOOD-FP (Bates et al., 2005) has been applied as a coastal inundation model to map depth,  
184 extent and velocity of floodwaters for extreme coastal and riverine events under rising sea levels.  
185 The horizontal model resolution varies from 20 m to 50 m depending on the size of the domain  
186 (which range from sites of critical infrastructure to the regional scale for supply network  
187 assessments) to allow efficient computation time and to capture the required level of detail for  
188 the management needs. Data on the time-varying storm tide alone, or combined storm tide and  
189 wave overwashing or overtopping volumes are used to generate the hazard imposed at the coastal  
190 boundary within LISFLOOD-FP, which propagates the floodwater landward across the floodplain.  
191 The positioning of the coastal boundary is domain dependent as is the boundary input data. At  
192 sites where wave hazard is considered negligible the low water contour is imposed as the coastal  
193 boundary and forced by storm tide water levels at 15 minute time intervals. At sites where wave  
194 hazard is considered important, through overtopping or overwash, the crest of a defence line  
195 (natural or engineered) is set as the coastal boundary and a wave resolving storm impact model  
196 is used to provide the (10 minute average) inflow discharge. In all cases the implemented models  
197 are run for a tidal cycle starting from low water. The inland model boundary is set some distance  
198 from the coast to ensure the flood pathways and area of inundation are generally contained within  
199 the domain. The boundary is set to allow through flow so under very extreme events the water is  
200 not restricted in a way that will cause it to inaccurately build-up. For the Fleetwood case high river  
201 flows have also been imposed as a discharge at the boundary points that cross the river Wyre (see  
202 Prime et al., 2015a). This allows the user to explore a range of flood hazard combinations (sea-  
203 level rise, coastal storms and high river flow).

204

205 At sites with wave hazard, overwashing or overtopping volumes have been calculated for various  
206 defences: hard engineered (SWAB, McCabe et al., 2013), sand dune (XBeach, Roelvink et al., 2010)  
207 or gravel barrier (XBeach-G, McCall et al., 2014, 2015). The use of the XBeach and XBeach-G  
208 models enables the role of storm-driven morphology and features within the cross-shore profile  
209 to be considered within the impact assessment. These models are applied as 1DH (horizontal)  
210 cross-shore profile models for present-day morphologies within the DST, while hypothetical future  
211 morphologies (such as changes in saltmarsh extent, barrier beach morphologies or subtidal bar  
212 geometries) are considered in more focused site-specific applications to determine potential  
213 changes in a system's response to storm impact (e.g., Prime et al., 2015b). The Shallow Water  
214 Boussinesq Model (SWAB) has also been used for a site with a sea wall (Prime et al., 2015a).

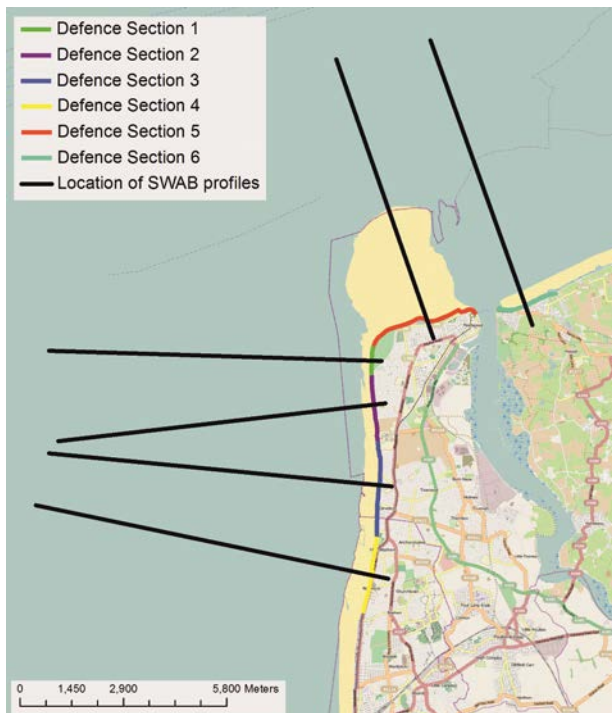


215 Although XBeach and XBeach-G can consider a fixed structure within the profile SWAB has been  
216 developed and validated with field observations to account for random wave breaking, impact  
217 and overtopping of sea walls (McCabe et al., 2013).

218

219 The initial profiles in the 1DH simulations are based on a combination of the latest available  
220 bathymetric data and beach profile surveys obtained for the site. The modelled cross-shore  
221 profiles have been selected to capture alongshore variability in the present-day coastal defence.  
222 At sites of energy infrastructure with a natural defence (gravel barrier or dunes) a 1 km spacing  
223 between the profiles with 50 m spacing closer to the nuclear power station is used to capture the  
224 alongshore variability in the beach-barrier system (Prime et al., 2016). For sites with sea walls a  
225 centrally positioned transect perpendicular to each defence section is chosen to simulate the  
226 flood hazard for each of the different defence designs (Prime et al., 2015a). An example set-up is  
227 shown in Fig. 2, where the sea wall provides protection to the local community behind. For sites  
228 where the 1DH models have been used to incorporate wave impact the wave direction is always  
229 assumed to be directly onshore to generate the worst case scenario.

230



231

232 Fig. 2. The LISFLOOD-FP model domain used to simulate flood hazard around the Fylde peninsula,  
233 northwest England. SWAB is applied in this example for each cross-section to simulate the wave-  
234 water inflow at the defence crest level (Prime et al., 2015a).

235

236 Within the ARCoES DST the flood maps were developed using data available to coastal managers.  
237 This includes the most recently available airborne laser altimetry (LiDAR) collected by the  
238 Environment Agency (EA) and observational data collected by national monitoring programs  
239 where available. These data include shoreline profile information collected by the EA or local  
240 authorities, the UK tide gauge network record (established in 1953), owned and operated by the  
241 EA, and the WaveNet record, a UK network of wave buoys (established in 2002) operated by the  
242 Centre for Environment Fisheries and Aquaculture Science (CEFAS). These real-time systems  
243 provide a long-term data archive to which a joint probability analysis can be applied to generate  
244 wave-water level combinations representative of a range of storm severities. Where observations  
245 are not available tidal predictions are obtained from the POLTIPS3 software, available from the  
246 national tide sea level facility, and wave data are obtained from long-term (40-year) hindcasts,  
247 such as the UK Climate Predictions 09 (UKCP09, Lowe et al., 2009) and the global wave hindcast  
248 produced in preparation of the European Centre for Medium Range Weather Forecasts (ECMWF,  
249 2016) next reanalysis (ERA5).

250

251 Where observations are limited to within the last decade (e.g., wave monitoring) or where only  
252 waves or water levels are monitored, archived data from climate modelling systems can be utilized  
253 to lengthen the datasets. The longer the data record the greater the confidence in the extreme  
254 value analysis. This research has used the European Centre for Medium-Range Weather Forecasts  
255 (ECMWF) 30-year wave ECWAM cycle 41R1 model data to lengthen the wave records. These  
256 numerical data are validated against existing wave observations prior to use in the analysis.

257

258 For the UK energy sector, events ranging from typical (1 in 1 year return period) to extreme (1 in  
259 10,000 year return period) conditions are considered. The joint probability analysis is performed  
260 using JOIN-SEA (Hawkes and Gouldby, 1998). This software uses the generalised Pareto  
261 distribution (GPD) model and simultaneous records of significant wave height ( $H_s$ ) and water level  
262 ( $WL$ ) at the time of the observed high water. In most cases the combined observational record  
263 covered a period of the order of a decade, the limitation often being related to the deployment  
264 of the wave buoy. For each return level a range of wave-water level conditions are generated.  
265 These cover conditions that transition from lower  $WL$  and higher  $H_s$  to higher  $WL$  and lower  $H_s$ .  
266 The conditions that pose greatest flood hazard along the probability curves are selected from an

267 ensemble of 1DH storm impact simulations that generate a range of inflow conditions to impose  
268 into LISFLOOD-FP (Prime et al., 2016). This generates the database of flood maps behind the DST.  
269 In this respect, the DST operates as a look-up table.

270

271 Once the required wave-water level combination has been ascertained a storm tide is created to  
272 force the offshore model boundary. The storm tide comprises a spring tide and a surge curve,  
273 available for all UK Class A tide gauge locations from the EA (McMillan et al., 2011). The surge  
274 curve is used to scale the tide such that the total high WL reaches the required extreme value.  
275 The time-varying water levels are combined with the required wave conditions within the 1DH  
276 storm impact model. Although the  $H_s$  is kept constant, a JONSWAP (Joint North Sea Wave  
277 Observation Project) spectrum is applied to create a time-varying wave field. This approach  
278 represents the worst-case scenario as the wave conditions maintain the desired extreme value for  
279 the duration of the simulation, a complete tidal cycle. An appropriate peak wave period ( $T_p$ ) is  
280 selected from the wave data for each  $H_s$ . At many sites around the UK there is a bimodal wave  
281 climate related to the wind sea and swell wave components. For each wave condition the longest  
282  $T_p$  associated with each  $H_s$  is used to simulate the highest wave runup levels.

283

284 Future sea-level projections are incorporated into the still water level of each event to take into  
285 consideration sea-level rise and explore future change in the inundation hazard. The projections  
286 are chosen to represent the high-end emission scenarios up to 2500AD (Jevrejeva et al., 2012).  
287 Incremental increases in mean sea level are considered at 10 cm intervals up to a rise of 2 m and  
288 then at 25 cm intervals to a rise of 5.5 m (Knight et al., 2015). The higher resolution is considered  
289 for levels representing plausible projections that could occur over the next 100 years, consistent  
290 with the long-term shoreline management planning framework. A lower resolution is then applied  
291 for the more bespoke longer term (c. 500 year) projections for the energy industry.

292

293

### 294 *3.2 Monitoring*

295 Alongside the numerical applications, storm surveys were performed at three nuclear sites across  
296 the UK, including Seascale (representing Sellafield in the northwest), Lilstock (representing  
297 Hinkley Point in the southwest) and Sizewell (in the east), as well as a long-term wave gauge  
298 deployment at Bradwell (in the southeast). This extreme event monitoring is used to assess the

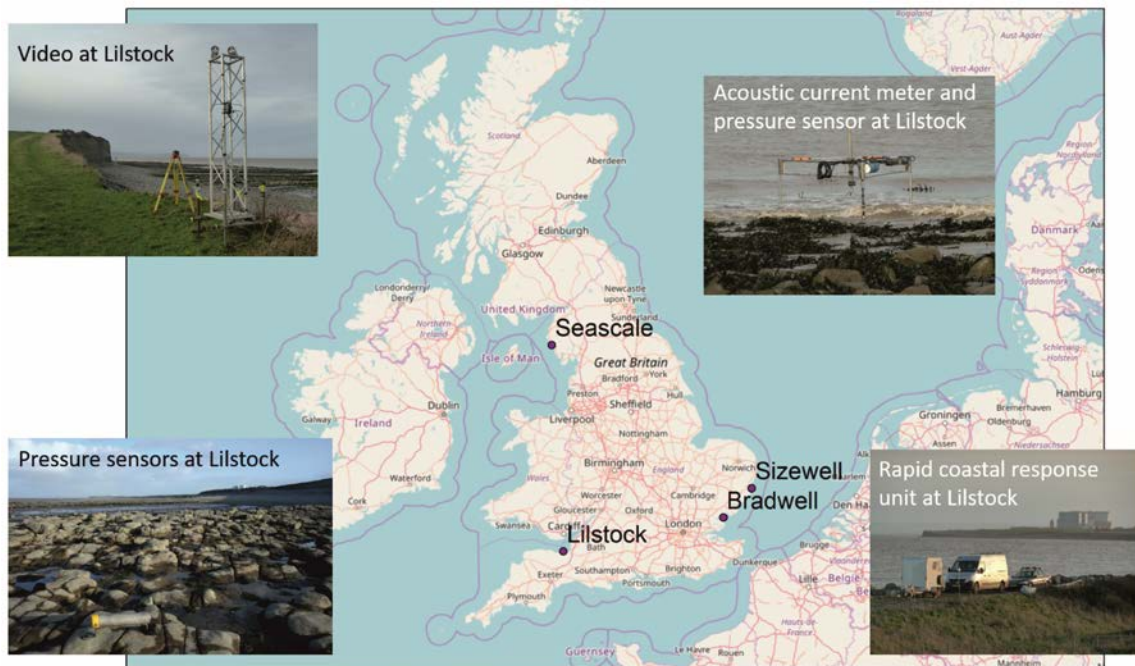
299 present-day vulnerability and disturbance-recovery behaviours of the sites. In order to  
300 compliment short-term survey campaigns that aim to characterise coastal response to storms, a  
301 cost-effective method of providing continuous observation of morphological change by  
302 automatically mapping large coastal areas has also been developed using a standard marine  
303 navigational radar (Bell et al., 2016; Bird et al., 2017a).

304

### 305 *3.2.1 Surveys*

306 Storm surveys over a tidal cycle were used to assess the response of different coastal systems and  
307 identify features that make them resilient or resistant to storm impact. During an event pre-,  
308 during and post-storm topographic data were collected (using a dGPS on a staff pole at low tide)  
309 alongside in-situ measurements and remote sensing observations. The in-situ instruments (e.g.,  
310 Fig. 3) were deployed pre-storm and retrieved after the storm. These included two low water  
311 scaffold rigs with pressure transducers and current meters together with five scaffold tubes with  
312 pressure transducers deployed alongshore at equal spacing (< 1 km) on the intertidal terrace.  
313 These instruments recorded the wave and tide elevations and the current velocities during the  
314 storm. Remote sensing techniques included a tower with two video cameras and a second tower  
315 with a laser-scanner. The video cameras were positioned to continuously record alongshore  
316 variability of wave runup during the storm (Poate et al., 2016). The laser-scanner tower was  
317 deployed on the beach face to measure morphological change and swash hydrodynamics along a  
318 cross-shore transect throughout the storm (Almeida et al., 2015; Almeida et al., 2017).

319



320

321 Fig. 3. Location map of the storm survey sites and examples of the instrumented rigs and towers  
 322 deployed.

323

### 324 3.2.2 Long-term monitoring

325 A new monitoring technique has been deployed, which uses a radar-imaged sea surface and an  
 326 accurate record of tidal elevations (such as a nearby tide gauge) as an altimeter to measure tidally-  
 327 driven water level elevations at each pixel in a radar scan. By knowing the position of the waterline  
 328 and the tidal elevation a bathymetric survey of the intertidal area can be produced. This  
 329 methodology was used to observe seasonal changes in morphology over a 3-year period and  
 330 assess storm impacts on beach volume and intertidal bedforms (Bird et al., 2017a). With the  
 331 ambition of applying this radar technique to multiple locations a semi-mobile radar survey system  
 332 has been developed during the ARCoES project by *Marlan Maritime Technologies Ltd*. This system  
 333 is powered by solar panels and a wind turbine and provides a stable radar tower, CCTV camera  
 334 and data recorder, enabling coastlines with limited power infrastructure to be monitored  
 335 effectively. This system continuously monitors beach topography within a few kilometres of the  
 336 radar for the entire duration of the deployment, which can then potentially update intertidal  
 337 bathymetry and waterline levels in near real-time. Study sites are shown in Fig. 4.

338



339

340 Fig. 4. Location map of the radar monitoring sites and the radar systems deployed.

341

342 A previous application to the Dee estuary, northwest England, has demonstrated the capability of  
 343 the radar to monitor complex geomorphological environments (Bird et al., 2017b). The tidal range  
 344 in this estuary is in excess of 10 m on high spring tides. The morphology is very complex and  
 345 includes large areas of intertidal sandflats, subtidal channels, mud banks, saltmarshes and rock  
 346 outcrops. Using a 2.5 m radar antenna intertidal topography was derived with a 3 m spatial  
 347 resolution over a 4 km range from the radar. Comparison with LiDAR showed radar-based system  
 348 was able to derive the major features of the topography including complex channels and bedforms  
 349 with a vertical accuracy of +/- 20 cm (although limitations with the LiDAR data should also be  
 350 acknowledged in any error analysis) (Bell et al., 2016). This surveying system therefore provides  
 351 advanced warning of adverse morphological change, volumetric information on sediment  
 352 movements (especially useful for monitoring beach nourishment schemes or identifying erosion  
 353 hotspots), bedform migration and broad-scale indications of a beach system health. Following the  
 354 development of this rapidly deployable remote-sensing survey platform (Rapidar), planned winter  
 355 deployments at sites of critical energy infrastructure (2017-18 for Minsmere, E coast UK, and  
 356 2018-19 for Dungeness, SE coast UK) will collect data to assess longer-term resilience of these  
 357 sites. These will also be complemented by additional storm surveys to assess the response of

358 these coastal systems to a winter season. This will help to identify and assess the role of shoreline  
359 response and morphological evolution within flood hazard assessments, enabling better  
360 understanding of some of the uncertainty surrounding modelled flood maps.

361

### 362 *3.2.3 Real Options Approach (ROA)*

363 The financial viability of investment projects or the selection of investment alternatives is typically  
364 assessed by cost–benefit analysis. The most widely used method is updating the future cash flows  
365 generated by the coastal scheme. This method is often referred as Discounted Cash Flow (DCF).  
366 However, it is widely acknowledged that the DCF leads to suboptimal decisions when irreversible  
367 investments are subjected to uncertainty (Pringles et al., 2015), such as large-scale infrastructure  
368 investment. Parallel to the modelling and monitoring of the physical processes, a Real Options  
369 Analysis (ROA) was developed to identify which energy infrastructure will benefit from flood  
370 management investment, and the optimal time to invest in this infrastructure (Prime et al., 2018).  
371 ROA is an adaptation of financial options analysis applied to valuing of physical or real assets  
372 (Pringles et al., 2015). ROA assesses the implied value of flexibility that is embedded in many  
373 investment projects. Flexibility acknowledges that investment plans are modified or deferred in  
374 response to the arrival of new (though never complete) information or until the uncertainty is  
375 fully resolved (Pringles et al., 2015). Using Monte Carlo simulation, the ROA values the options to  
376 defer or invest based on a set of pre-defined decision rules and option valuation (see for example  
377 Pringles et al., 2015). The analysis provided by the ROA is used to form a cost-benefit decision-  
378 support tree.

379

380 The next section presents a series of applications of the ARCoES DST to demonstrate the versatility  
381 of information that can be generated for planning coastal adaptation to climate change.

382

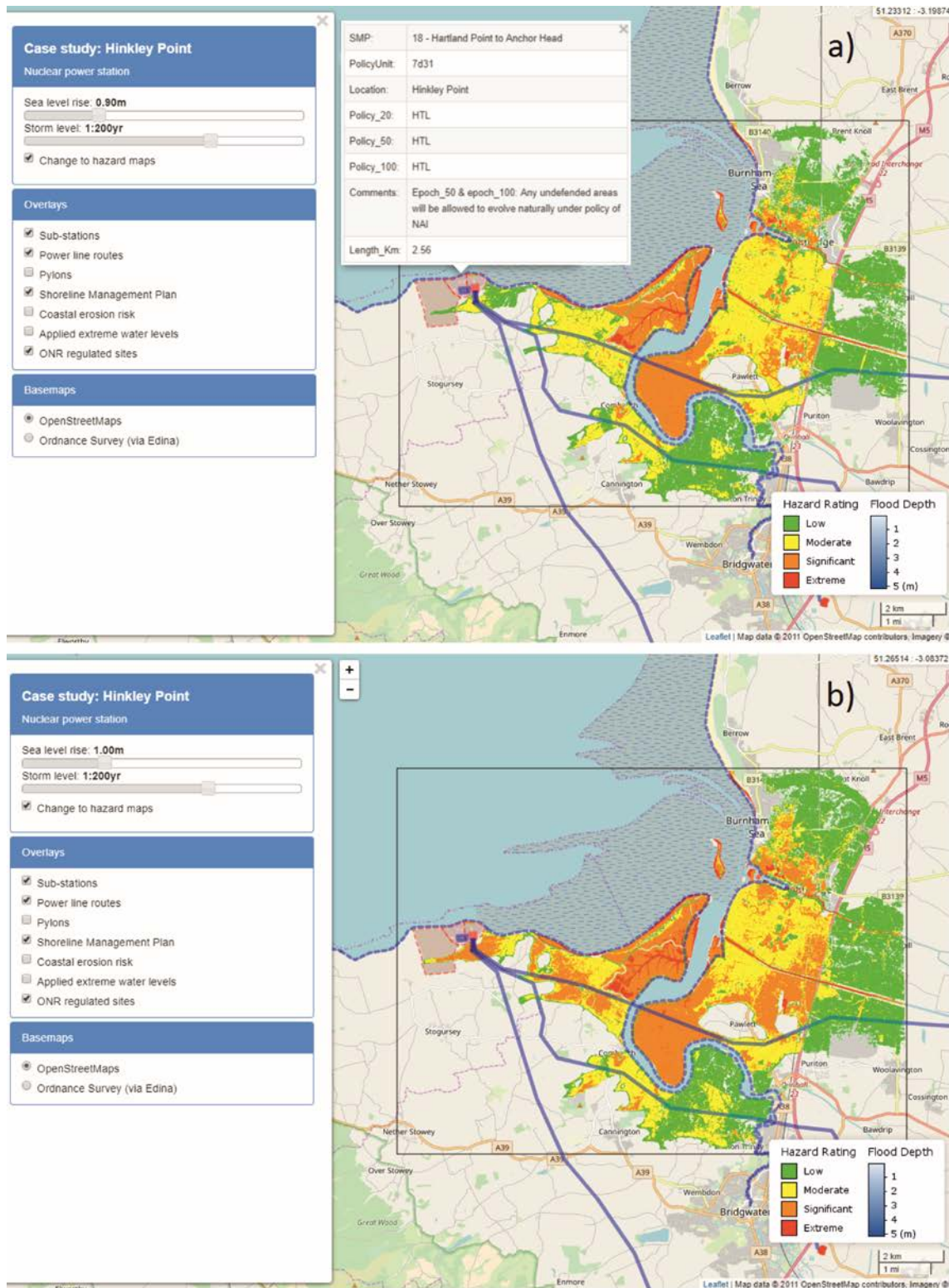
## 383 **3. Results**

### 384 *3.1 ARCoES DST*

385 The examples presented use LISFLOOD-FP (alone) in applications within the Bristol Channel and  
386 Severn Estuary, southwest England. At Hinkley Point (Fig. 5) the shoreline management policy is  
387 ‘hold the line’ (HTL Fig. 5a). By selecting a 1 in 200 year storm condition, typical of UK defence  
388 standards, we identify a tipping point in the storm hazard rating to people (from low/moderate,  
389 Fig. 5a, to significant, Fig. 5b, for road and power line route access) at around 1 m of sea–level

390 rise. At this site the flood hazard occurs due to inundation of lowlands towards the east of the site.  
391 This type of information highlights the need to reassess operational strategies in the future,  
392 particularly for first responders or workers using access routes or working on the electricity  
393 transmission lines.





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397

Fig. 5. Hinkley Point, showing a tipping point in the hazard to people from moderate to significant over access and electricity routes for a 1 in 200 year storm event and a change in mean sea level from a) 0.9 m to b) 1.0 m. Panel a also shows a pop-up window displaying the SMP metadata for

398 a defence section fronting the nuclear power station.

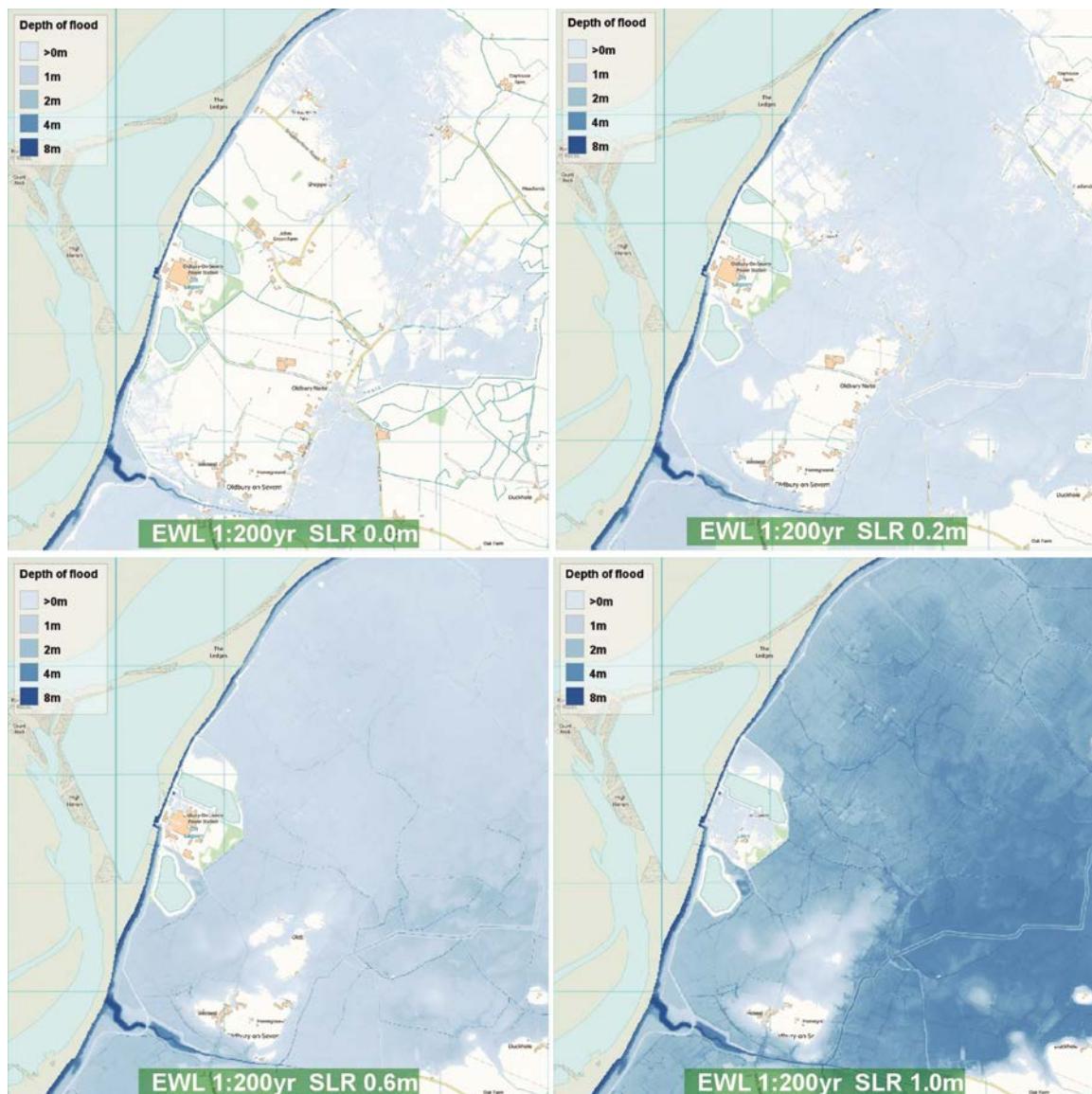
399

400 Animations are also available online for incremental sea-level rise and storm return period for  
401 certain nuclear power station sites. Fig. 6 shows screen shots of the online animations for the  
402 Magnox nuclear power station at Oldbury-on-Severn. The screen shots show increasing sea-level  
403 rise and a constant 1:200 year storm level. The base map used for these images in Ordnance  
404 Survey (OS, 2014). A 1:200 year storm level under present-day sea level (no increase) results in  
405 inundation of agricultural land of less than 1 m. A 1:200 year event, accompanied by 0.2 m sea-  
406 level rise results in more extensive inundation. However, the depth of inundation remains up to 1  
407 m. The Oldbury-on-Severn site remains unaffected, as do some residential properties in the towns  
408 of Oldbury-on-Severn and Oldbury Naite to the south. Around 0.6 m sea-level rise results in a  
409 greater extent of inundation up to 1 m, particularly agricultural land to the southeast of the model  
410 domain. Again, the nuclear site remains unaffected as well as some small areas around Oldbury-  
411 on-Severn. Widespread inundation results from 1.0 m sea-level rise and low lying inland areas  
412 become vulnerable as the flood water propagation is no longer restricted to limited pathways  
413 during tidal high water. All transport and access routes within the area are flooded, as well as local  
414 amenities, agricultural land and residential properties. These images show how the DST can be  
415 used to simulate increasing sea-level rise superimposed on a 1:200 year event and the resulting  
416 depth and extent of inundation, and thus identify where the vulnerability to flooding undergoes  
417 a step change. This information is simulated with no change to present-day flood defence. It can  
418 therefore identify where intervention may be required in the future, showing flood pathways to  
419 help inform the optimal locations to invest in defence infrastructure.

420

421

422



423

424 Fig. 6. Animation screen shot of a scenario with a 1:200 year extreme water level (EWL) and 0.0  
 425 m, 0.2 m, 0.6 m and 1.0 m sea-level rise (SLR) for the Oldbury model domain.

426

427 The DST is currently set-up to provide a simplified estimate of costs calculated from a depth-  
 428 damage curve for different land uses considering inundation by saltwater (Fig. 7a). The DST  
 429 displays the flooded area (km<sup>2</sup>) and cost (£M) for arable land, residential housing, roads, industry  
 430 and the total area of inundation for the selected storm event and sea-level value. Using this  
 431 information appropriate timeframes to implement new management strategies based on the  
 432 relative costs of flooding and the benefits of implementing resilience measures can be planned  
 433 (Prime et al., 2015a).

434

435 *3.2 Real Options Analysis (ROA)*

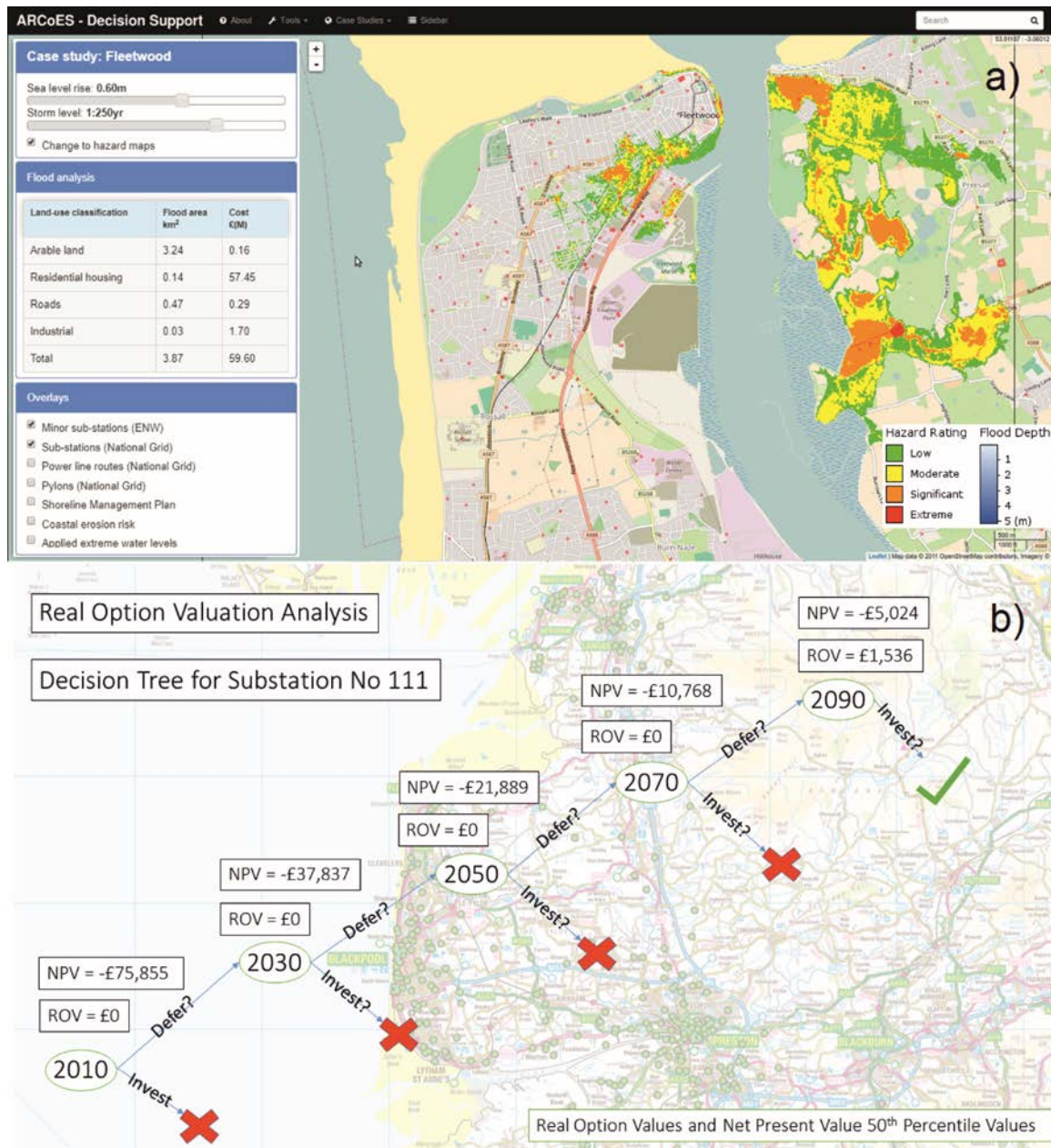
436 By identifying electricity distribution substations that are vulnerable to future flooding using the  
437 DST a ROA can be applied to assess when the implementation of any resilience measures would  
438 be cost-effective. The ROA combines the flood hazard exposure maps simulated for the sea-level  
439 projections with the economic data associated with the investment decision such as inflation,  
440 building costs, maintenance costs, clean-up costs and savings in relation to deferring a project  
441 (Prime et al., 2018). Fig. 7b illustrates a classic Net Present Value (NPV) calculation based on the  
442 most widely used investment decision tool, Discounted Cash Flow (DCF) analysis. According to  
443 DCF-based calculation any substation that has a positive value should go ahead with flood defence  
444 investment. However, NPV calculations based on DCF approaches do not value any flexibility in  
445 the management process. Using ROA a flexible NPV is also calculated. Based on the more flexible  
446 ROA methods, investment in flood defense for substation 111 should only go ahead in 2090.

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451

452 Fig. 7. Examples of a) the DST cost-benefit information for Fleetwood, northwest England, (the red  
453 symbols indicating where sub-stations are present) and b) the real options analysis decision tree  
454 for a substation in the northwest England.

455

### 456 3.3 Monitoring

457 While the DST explores future scenarios identifying when tipping points in flood hazard for the  
458 current management practice occur and the ROA enables assessment of when it is most cost-  
459 effective to implement a new management approach, observations inform us of the present-day  
460 disturbance-recovery behaviours of coastal environments (cf. Almeida et al., 2015). The ARCoes

461 project found that all four nuclear power station sites that were observed (see Section 2) currently  
462 experience limited vulnerability to extreme storm events due to the combination of their siting  
463 and geomorphology, as well as any site-specific interventions required as part of their pre-  
464 operational and operational safety cases as a requirement of their licencing approval.

465

466 From this understanding we can cast the coastal flooding and erosion risk to nuclear power station  
467 into a Source – Pathway – Receptor framework (Narayan et al., 2012; Sayers et al., 2002) and make  
468 two general statements. Firstly, all nuclear power station locations have limited potential for the  
469 occurrence of extreme wave conditions (i.e., Source) due to their siting. At the same time, the  
470 sites have a common morphology (i.e., Pathway), characterised by a reflective and permeable  
471 gravel/cobble high tide beach fronted by a wide and low gradient dissipative feature. This ensures  
472 that even if the site experiences extreme wave energy levels, potential damage to the nuclear  
473 power station site (i.e., Receptor) due to flooding and erosion would be limited. With uncertainty  
474 surrounding the consequence of climate change and sea-level rise (the Source) at the coast,  
475 monitoring of the morphology (Pathway) is recommended, using techniques such as Rapidar, to  
476 provide early warning to trigger a review of the current management strategy to maintain the  
477 required standard of protection (to the Receptor). Through understanding of the present-day  
478 processes, critical evolution within the system can be identified for consideration in sensitivity  
479 modelling using the models that make up the DST. One example would be the update and  
480 exploration of time-evolving beach profiles within the numerical approach that generates the  
481 hazard maps. Such studies continued study will highlight areas for continued development within  
482 the DST.

483

#### 484 **4. Conclusions**

485 The ARCoES DST and parallel ROA presented in this paper provide a resource that can be used to  
486 initiate discussions with coastal practitioners to identify how future vulnerability to coastal  
487 flooding may be mitigated through appropriate and timely intervention and adaptation. Such a  
488 forum for dialogue is required to improve the transfer of knowledge between costal researchers  
489 and decision-makers, to enable science based evidence to underpin choices made when setting  
490 new coastal management strategies. The DST enables maps of potential flooding, and associated  
491 costs, from increments of sea-level rise and storm magnitude to be explored by a wide range of  
492 users to identify key locations and ‘tipping points’ where and when the increased vulnerability to

493 flooding challenges current operations, emergency plans and long-term management strategy.  
494 When combined with understanding gained from present day observations informed monitoring  
495 programmes to support management decisions can be put in place and site inspections can be  
496 focused on assessing geomorphic change that has the potential to change a sites vulnerability to  
497 storm impact. The detailed understanding of the local processes also allows the limitations of the  
498 'static' morphology within the DST to be put in context though the identification of how  
499 uncertainty within the mapped results could occur. A key area for expansion of the ARCoES  
500 framework would be to incorporate shoreline evolution within the projections of future coastal  
501 flood hazard. By using freely accessible models and mapping systems within the DST continued  
502 development can be facilitated, enabling incorporation of such information in the future.

503

504 Within a policy context, project outputs have already provided practice and policy  
505 recommendations for national and regional decision-makers on building coastal resilience to sea-  
506 level rise and storms (please see the Living With Environmental Change (LWEC) partnership policy  
507 and practice notes, Plater and Brown, 2016). In this respect, the DST and associated resources  
508 provide a framework for engagement and dialogue across research and stakeholder communities  
509 for the co-production of future plans (e.g., Armstrong et al., 2015). Over the longer term, the DST  
510 provides energy infrastructure stakeholders with a roadmap for planned investments that address  
511 resilience to future change in sea level and extreme events. This would include measures such as  
512 the relocation of substations, raising transformers and other hardware above ground, and  
513 replacing ageing assets (e.g. circuit breakers) that may be more sensitive to water. The DST  
514 therefore delivers essential support for: (i) improved response to extreme events and (ii) a strategy  
515 that builds climate change resilience. Both offer the consumer greater confidence in the constancy  
516 of energy supply and an awareness that their money is being spent effectively in combating  
517 present and future risks from flooding.

518

519 Finally, the ARCoES DST platform is an effective example of inter-disciplinary collaboration across  
520 physical, natural, and social sciences on one axis, and across research, energy and infrastructure  
521 sectors, coastal management authorities, environmental regulators, and coastal communities on  
522 another. Interactive dissemination of the DST has revealed its value in discussions that centre on:  
523 (i) future changes in coastal geomorphology and how this may be managed to promote 'natural'  
524 coastal resilience, (ii) engagement of stakeholders with projections of flooding due to sea-level

525 rise and other forcing factors, and uncertainties therein; and (iii) interventions that mitigate  
526 impacts in an appropriate (according to location and scale of challenge), timely and cost-effective  
527 way. The DST is therefore presented as a resource for framing dialogue and exploring solutions,  
528 rather than providing simplistic answers out of context. Rather than this being viewed in negative  
529 terms by decision makers, the DST has been received positively as providing a focus for the sharing  
530 of knowledge, perspectives and priorities.

531

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548

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