1	Modeling of the Dec. 22 nd 2018 Anak Krakatau volcano lateral collapse and
2	tsunami based on recent field surveys: comparison with observed tsunami
3	impact
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10	Abstract
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19	The Dec. 22, 2018 lateral collapse of the Anak Krakatau (AK) volcano in the Sunda Straits of
20	Indonesia discharged volcaniclastic material into the 250 m deep caldera southwest of the volcano
21	and generated a large tsunami, causing runups of up to 85 m in the near-field, and 13.5 m in the
22	far-field, on the nearby coasts of Sumatra and Java. The tsunami caused at least 437 fatalities, the
23	greatest number from a volcanically-induced tsunami since the catastrophic explosive caldera-
24	forming eruption of Krakatau in 1883 and the sector collapse of Ritter Island in 1888. For the first
25	time in over 100 years, the 2018 AK event provides an opportunity to study a major volcanically-
26	generated tsunami that caused widespread loss of life and significant damage. Here, we present
27	numerical simulations of the collapse and tsunami generation, propagation, and coastal impact,
28	with state-of the-art numerical models, using both a new parametrization of the collapse and a
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29 near-field bathymetric dataset based on our 2019 field surveys and satellite images. These 30 subaerial and submarine data sets are used to constrain the geometry and magnitude of the 31 landslide mechanism, which show that the primary landslide scar bisected the AK edifice, cutting behind the central vent and removing 50% of its subaerial volume. The primary landslide volume 32 is estimated to range from 0.175 - 0.272 km³, based on uncertainties in the shape of the submerged 33 34 part of the failure plane. This is supported by an independent estimate of the primary deposit 35 volume of 0.214 ± 0.036 km³. Given uncertainties in the failure volume, and a secondary debris flow triggered by the collapse $(0.022 \pm 0.006 \text{ km}^3)$, we define a range of potential failure surfaces 36 37 that span these values in 4 collapse scenarios of volume ranging from 0.175 to 0.313 km³. These 38 AK collapses are modeled, assuming either a granular or viscous fluid rheology, together with 39 their corresponding tsunami generation and propagation. Observations of a single tsunami, with 40 no subsequent waves, are consistent with our interpretation of landslide failure in a rapid, single 41 phase of movement rather than a more piecemeal process, generating a tsunami which reached 42 nearby coastlines within ~30 minutes. Both modelled rheologies successfully reproduce near- and 43 far-field tsunami flow depth and runup observed in all post-event field survey results, tide gauge 44 records, and eyewitness reports to date, suggesting our estimated landslide volume range is 45 appropriate. This event highlights the significant hazard posed by relatively small-scale lateral 46 volcanic collapses, which can occur en-masse, without any precursory signals, and are an efficient 47 and unpredictable tsunami source. Our successful simulations demonstrate that current numerical 48 models can accurately forecast tsunami hazards from these events. In cases such as Anak 49 Krakatau's, the absence of precursory warning signals, together with the short travel time following tsunami initiation present a major challenge for mitigating tsunami coastal impact, 50 51 stressing the need to install early warning systems.

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53 1 Introduction

54 Over the past 20 years, catastrophic tsunamis in Papua New Guinea (1998), the Indian Ocean 55 (2004), and Japan (2011) have led to major advances in understanding and modeling tsunamis from submarine landslides, earthquakes, and dual mechanisms. These advances have been mainly based 56 57 on improved constraints on these recent events and their geographical distribution, together with 58 improved numerical tsunami modelling capability (e.g., Tappin et al., 2008; Grilli et al., 2007; 59 Ioualalen et al., 2007; Grilli et al., 2013; Kirby et al., 2013; Tappin et al., 2014; see Yavari-Ramshe 60 and Ataie-Ashtiani, 2016, for a recent review). Tsunamis from volcanic eruptions and collapses, 61 however, although having the potential for generating mega-tsunamis (Paris et al., 2020b), 62 resulting in significant loss of life and property (Day, 2015; Paris, 2015), and accounting for 20% 63 of all volcanic fatalities over the past 400 years (Auker et al., 2013), remain less well-studied 64 because, up until recently, there were few well-recorded and researched events.

65 Most of the earlier known volcanic collapse events are prehistoric, with no recorded direct 66 observations, and many were large-volume lateral volcanic collapses of ocean islands, such as in 67 the Canary Islands (e.g., Ward and Day, 2001; Day et al., 2005; Løvholt et al., 2008; Abadie et al., 68 2012; Giachetti et al., 2012) and Hawaii (e.g. McMurty et al. 2003), although some were smaller 69 scale events, such as Ritter Island 1988 (Ward and Day, 2003) and Stromboli 2002 (Tinti et al., 70 2000; Fornaciai et al., 2019). Of historical volcanic tsunamis the best studied are Krakatau, 71 Indonesia in 1883 (Verbeek, 1983, 1885; Siswowidjoyo, 1983) and Ritter Island, Papua New 72 Guinea in 1888 (Johnson, 1987; Ward and Day, 2003; Simkin and Fiske, 1983; Day et al., 2015). During the Krakatau eruption, there were 19 tsunamis, with the most destructive generated during 73 74 the final, cataclysmic, caldera collapse and the associated emplacement of pyroclastic flow

75	material into the sea, which destroyed the volcanic edifice and caused 33,000 fatalities. The Ritter
76	Island tsunami was generated by a ${\sim}5~\text{km}^3$ flank collapse, which is the largest recorded volume
77	lost from an island volcano in a single event historical times (Ward and Day, 2003; Watt et al.,
78	2019). Johnson (1987) estimated that 3,000 fatalities resulted from this event, but other (largely
79	upublished) evidence and anthropology studies (Dunbar, 1993), indicate a total number of death
80	of 700 to 1,000 is more likely. Because of the paucity of data on most volcanic events, the results
81	of their tsunami modelling have not been fully validated and both landslide source mechanisms
82	(e.g., Hunt et al., 2011; Ward and Day, 2003; Watt et al., 2019) and the generated tsunamis (e.g.,
83	Day et al., 2005; Løvholt et al., 2008; Abadie et al., 2012, 2020; Tehranirad et al., 2015), remain
84	poorly documented, so are a challenge to model. With remarkable prescience, Giachetti et al.
85	(2012) modeled a tsunami from a collapse of the SW flank of the Anak Krakatau (AK) volcano,
86	similar to that of Dec. 28th 2018, using a hypothetical 0.28 km ³ volume. The resulting wave heights
87	and arrival times along surrounding coastlines foreshadowed the 2018 event.

88 In the evening of December 22, 2018, at 20:55-57 local time, following a 6 month period of relatively heightened eruptive activity, a lateral collapse occurred on the southwest flank of the 89 90 AK volcano in the Sunda Strait, Indonesia (Figs. 1 and S1). The collapse generated a tsunami that 91 flooded the adjacent coastlines of Java and Sumatra within 30 minutes (Grilli et al., 2019), causing 92 up to 13.5 m runups and resulting in 437 fatalities, 13,000 people injured, 33,000 displaced and 93 thousands of buildings destroyed (AHA, 2018; Andersen, 2018; Muhari, 2018, 2019; Grilli et al., 2019; TDMRC, 2019). The AK event was the most damaging volcanically-generated tsunami 94 since the 1883 eruption of Krakatau and the 1888 lateral-collapse of Ritter Island. The numerous 95 observations made of AK's 2018 collapse and tsunami, including those previously unpublished by 96 97 the authors of this paper, provide a unique dataset for both understanding this event and testing 98 state-of-the-art tsunami modelling methodologies against direct observations, with the modelling 99 constrained by both volcanic tsunami source parameters and observations of the generated waves 100 and their coastal impact.

101 Here we report on and compare the validity of hypothetical volcanic lateral-collapse 102 scenarios with data from an actual event at AK. Our modelling is based on a comprehensive 103 subaerial and submarine data set, including from our 2019 marine survey, of the 2018 AK lateral 104 collapse (Hunt et al., 2020). The subaerial data on the collapse has been published (e.g., Williams 105 et al., 2019; Novellino et al., 2020; Perttu et al., 2020) and provided the basis for previous tsunami 106 modelling (e.g., Grilli et al., 2019). The new numerical modeling presented hereafter is also based 107 on this remote (mainly satellite) subaerial data but also, for the first time, on a hydroacoustic data 108 set of multibeam echosounder (MBES) bathymetry and seismic data acquired to the southwest of 109 the volcano after AK's eruption, in August 2019 (Privanto et al., 2020; Hunt et al., 2020).

110 An important aspect of our new modelling of the 2018 collapse and tsunami generation is 111 to use the latest version of the three-dimensional non-hydrostatic model NHWAVE (Zhang et al., 112 2021a,b), that features effects of vertical accelerations, not just in the water (as in earlier 113 implementations) but also within the slide material itself. Previous published modelling has 114 neglected vertical acceleration (i.e., non-hydrostatic) effects within the slide layer. We show that 115 including these effects is important for an accurate simulation of both wave generation from the 116 collapse and the near-field runups. These new simulations are also performed at a much higher 117 resolution, owing in part to the new high-resolution bathymetric and topographic data from our 118 2019 field survey and its subsequent analyses and reconciliations with the subaerial observations 119 (Hunt et al., 2020). Model results for both the near- and far-field tsunami generation, propagation 120 and coastal impact are validated against time series of sea surface elevation recorded at tide-gauges

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121	in the Sunda Straits together with all published field observations and eyewitness accounts to date	
122	of onland tsunami flow depth and runup, both on islands in close proximity to AK (including the	
123	August 2019 authors' drone survey), and in the far-field on the coasts of Java and Sumatra.	
124	The combined subaerial and marine data sets, and results presented here, constrain the style	
125	of the AK lateral collapse and also test current volcanic landslide-tsunami models, which can be	
126	used to predict the behavior of similar events at other volcanic islands. The results, therefore, are	
127	an important contribution towards improved assessment of tsunami hazard from analogous events	
128	in the future, and also provide an improved basis for developing mitigation strategies for volcanic	
129	tsunamis.	
130		
131	2 Background and earlier modeling work	
132	2.1. Geologic and volcanologic context	
133	AK (Fig. 1) is a composite volcanic cone that developed on the northeast margin of the 250 m deep	
134	flooded caldera formed by the 1883 eruption of Krakatau (Fig. 2d; Camus et al., 1987; Stehn,	
135	1929). It developed from and so is aligned with the feeder vents of the 1883 Krakatau eruption	
136	(Verbeek, 1885, 1983). During the past 90 years of frequent eruptive activity, AK has grown from	
137	a submarine volcano to a subaerial edifice, emerging in 1929. With a pre-2018 collapse height	
138	estimated at about 335 m (Grilli et al., 2019), it formed an island with a diameter of 1.5-2 km. On	
139	the SW flank of AK, coastline retreats of several hundred meters in 1934, 1935 and 1950 (Stehn,	
140	1929) imply long-lived instability of the edifice on this sector. The NW-SE orientation of the	
141	retreats align with both the underlying caldera-wall scarp and the 2018 collapse scar (Fig. 2d). The	
142	retreats are a result of: i) AK's location on the margin of the 1883 caldera; ii) the volcano is on the	
143	northeast margin of the 250 m deep intra-caldera basin; and iii) the asymmetrical pattern of island	

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145	first emergence of the island in 1929 was dominated by phreatomagmatic explosions (Umbgrove
146	1928; Stehn 1929). Similar explosions continued after first emergence and built-up a low-angle
147	tuff cone around a vent to which the sea continued to gain access until the 1960s. At that time the
148	vent dried out and further subaerial eruptive activity produced lava flows on the SW side of the
149	island, and Vulcanian and Strombolian scoria (Siswowidjoyo, 1983). This activity continued into
150	the 21st Century, with numerous small eruptions punctuated by more violent explosive episodes,
151	and built-up of a steep-sided central pyroclastic cone, with lava deltas extending the island on most
152	sides except the sheltered NE where the tuff cone was widest, but especially in the NW and SE
153	(Abdurrachman et al., 2018). During a subaerial eruption in 1981 (Camus et al., 1987), a ~2 m
154	high tsunami was recorded on Rakata, a remnant of the 1883 volcanic island and the southernmost
155	and largest island of the contemporary Krakatau archipelago (Fig. 1), which was inferred to
156	originate from a small flank landslide. The event highlighted the potential instability of the
157	southwest flank of the volcano (Camus et al., 1987) but, apart from this, no other tsunamis from
158	AK have been reported.
159	The recent period of AK volcanic activity started in June 2018 and continued into
160	December (https://volcano.si.edu/volcano.cfm?vn=262000), producing Strombolian explosions,
161	lava flows, and ash plumes reaching altitudes of up to 5 km (Anon, 2018; Fig. 2 in Paris et al.,
162	2020a; Figs. S1c,d;). On Dec. 22, 2018, a major lateral collapse occurred on AK's southwest flank
163	which discharged volcaniclastic material into the sea and triggered a destructive tsunami
164	(Andersen, 2018; PVMBG, 2018). Based on seismic records (Gurney, 2018), eyewitness reports
165	(e.g., Andersen, 2018; Perttu et al., 2020), and the agreement of modelled waves with tsunami

growth (see discussion in Hunt et al., 2020). The early submarine activity of AK before and during

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arrival times at tide gauges (Ina-COAP, 2019; see below; Fig. 1a, Table 2), Grilli et al. (2019)

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167	estimated that the collapse took place at 20:55'-57' (UTC + 7), a time range later confirmed and
168	used by other authors of numerical models (e.g., Borrero et al., 2020; Mulia et al., 2020; Paris et
169	al., 2020a; Zengaffinen et al., 2010), and confirmed in the interpretation of seismic signals by
170	Walter et al. (2019), who timed the collapse at 20:55'. Within 30 minutes of the collapse a tsunami
171	flooded the coasts of west Java and southeast Sumatra, causing up to 13.5 m on land runups. The
172	tsunami struck near high tide (+1.5 m over MSL in average at four tide gauge in Java and Sumatra;
173	Fig. 1a), which increased its impact (AHA, 2018; Muhari, 2018, 2019; Grilli et al., 2019; TDMRC,
174	2019).

175

176 2.2. Previous modeling of the 2018 AK event

177 In light of the modelling published by Giachetti et al. (2012), Grilli et al. (2019) performed the first comprehensive numerical simulations of the Dec. 22nd 2018 AK collapse, based on 178 179 satellite observations on the days following the event, drone and field surveys of near-field tsunami 180 impact conducted in early January 2019 (Reynolds, 2019; TDMRC, 2019; Fig. S1), and historical 181 data on the growth of AK (e.g., as in Fig. 2d). The modelling of AK's flank collapse and tsunami 182 generation was based on a range of failure surfaces with corresponding collapse volumes of 0.22-183 0.30 km³ and used the three-dimensional (3D) non-hydrostatic (NH) model NHWAVE (Ma et al., 184 2012, 2015; Kirby et al., 2016), in which the collapse was represented by a depth-integrated 185 (hydrostatic) layer of a granular material or dense viscous fluid. From the modelling it was 186 proposed that a 0.27 km³ collapse volume caused the tsunami that best reproduced the near- and 187 far-field tsunami propagation and impact, with the far-field modeling using the fully nonlinear and 188 dispersive Boussinesq model FUNWAVE (Shi et al., 2012).

189	Numerical simulations of the 2018 AK collapse and tsunami post-dating Grilli et al. (2019),
190	detailed in the following, were also based on hypothetical source parameters derived from a variety
191	of, mainly indirect, data. In these studies, the various assumed/hypothetical failure surfaces gave
192	collapse volumes in the range ≈ 0.14 -0.33 km ³ , which were both smaller and larger than the 0.27
193	km ³ of Grilli et al.'s (2019). In Ye et al.'s (2020) study, inversion of broadband seismic data was
194	used to infer a collapse volume of ≈ 0.20 km ³ . In some studies, an empirical analytical or
195	experimental (from laboratory tests) landslide source was specified directly on the free surface
196	without an actual modeling of the source (e.g., Heidarzadeh et al., 2020; Borrero et al., 2020). In
197	other modeling studies, new interpretations of subaerial observations were used (see Hunt et al.,
198	2020 for a discussion) and the flank collapse and tsunami generation modeled for a variety of
199	volumes and geometries, in a way similar to that of Grilli et al. (2019) (e.g., Mulia et al., 2020;
200	Ren et al., 2020; Omira and Ramalho, 2020; Paris et al., 2020a; Zengaffinen et al., 2020; Dogan
201	et al., 2021). In the latter models, tsunami generation was based on various rheologies (granular,
202	viscoplastic, Bingham) and simulated using a two-dimensional (2D) two-layer model. There were
203	also important differences in tsunami propagation models used in these various studies, with some
204	using a dispersive model as in Grilli et al. (2019) (e.g., Mulia et al., 2020; Paris et al., 2020a;
205	Borrero et al., 2020) and others using a non-dispersive tsunami propagation model (e.g.,
206	Heidarzadeh et al., 2020; Ren et al., 2020; Omira and Ramalho, 2020; Dogan et al., 2021). As
207	landslide tsunamis are typically made of shorter, more dispersive, wave trains, they often require
208	the use of a dispersive long wave model for their accurate modeling (e.g., Ma et al., 2012; Glimsdal
209	et al., 2013; Tappin et al., 2014; Grilli et al. 2015, 2017; Schambach et al., 2019). For the 2018
210	AK event, Paris et al. (2020a) concluded that dispersive effects were important during tsunami
211	generation and propagation, whereas Zengaffinen et al. (2020) found that they were not large in

212 the shallow water areas of the Sunda Straits (as could have been expected), to the north and south

213 of AK. More specifically:

214 In one of the more comprehensive recent studies, Zengaffinen et al. (2020) modeled the ٠ 215 tsunami using the rate of mass release, the landslide volume, the material yield strength, and 216 orientation of the landslide failure plane, together with the 2D two-layer depth-averaged 217 coupled model BingClaw, to identify different failure mechanisms, landslide evolution, and 218 tsunami generation. The depth-integrated landslide layer was based on a viscoplastic flow 219 rheology, coupled with depth-averaged long wave and shallow water type models to simulate 220 tsunami propagation. With a volume of 0.28 km³, identical to that of Giachetti et al. (2012), 221 the numerical simulations provided a reasonable match to the observed tsunami surface 222 elevation amplitudes and inundation heights in the far-field. Overall the results were consistent 223 with those of Grilli et al.'s (2019) preferred 0.27 km3 scenario, and discrepancies between the 224 simulated and observed arrival times at the offshore gauges were attributed to the (poor) 225 accuracy of the available bathymetry, rather than to their model. To match these, to the north 226 of Krakatau, Zengaffinen et al. (2020) arbitrarily increased the water depths in this area.

227 Paris et al. (2020a) used the 2D two-layer depth-averaged coupled model AVALANCHE, • 228 which features a granular rheology and a Coulomb friction for the slide description, with 229 dispersive effects for the water flow part. From pre- and post-collapse satellite and aerial 230 images, and a satisfactory comparison of the simulated water waves with far-field observations 231 (tide gauges and field surveys), they reconstructed a total (subaerial and submarine) landslide 232 volume of 0.15 km³, at the lower end of the volume range in the various studies described here. 233 Ren et al. (2020) applied a 2D two-layer depth-averaged coupled non-dispersive model 234 throughout, with the slide layer modeled as a dense fluid. Using two nested grids, the smaller

235		having a 30 m resolution and the larger a coarse 230 m resolution, and 0.2-0.3 km ³ collapse
236		scenarios, they showed a reasonable agreement with the first wave at the far-field tide gauges.
237	•	Mulia et al. (2020) integrated the landslide thickness over the estimated source area and,
238		assuming a failure surface similar to that of Giachetti et al. (2012), except for a slightly steeper
239		slope, obtained a collapse volume of 0.24 km^3 (slightly smaller than that of Giachetti et al.,
240		2012, and Grilli et al., 2019). Using the 2D two-layer depth-averaged coupled model VolcFlow
241		simulating avalanche dynamics (here assuming a constant retarding stress throughout), and
242		FUNWAVE for the far-field tsunami, their landslide generated higher than 40 m waves in the
243		vicinity of the volcano. As with other studies the tsunami attenuated rapidly as it propagated
244		away from the generation area, resulting in lower than 2 m wave heights at tide gauges around
245		the Sunda Strait.

Finally, in the latest study to date, Dogan et al. (2021) modeled a 0.25 km³ collapse (based on 246 ٠ 247 a maximum elevation for AK of only 260 m, smaller than used in other studies), and its tsunami 248 generation, using Imamura and Imteaz (1995)'s two-layer long wave model. Tsunami 249 propagation to the far-field was then simulated using the non-dispersive NSW model NAMI 250 DANCE, in an 80 m resolution grid. Little details are given of the parameterization of their 251 dense fluid rheology in the slide model or the rationale for defining the pre- and post-failure 252 volcano geometry, including the selected failure surface. However, they show a good 253 agreement with both arrival times and elevation time series measured at the 4 tide gages in 254 Java and Sumatra. Based on observed bathymetry changes in pre- and post-event surveys, they 255 model tsunami generation from additional submarine slope failures on the north and south 256 sides of the caldera, but conclude that these did not contribute to and hence were not 257 simultaneous with AK's 2018 event.

258 ADD TABLE SUMMARIZING EARLIER MODELING STUDIES AND THEIR FEATURES ?

259 All of these studies used different AK collapse scenarios and a wide spectrum of 260 approaches and tsunami modelling, but the differences in tsunami elevations predicted at the far-261 field tide gages were small; there were larger differences in predicted far-field runups, but some 262 of these could be explained by differences in grid resolution and model physics. While details of 263 a tsunami source become less important when the distance from the source increases, here, the 264 small differences in the predicted far-field tsunami impact between various modeling studies were 265 in great part because the landslide mechanisms were based on inverse methodologies and, hence, 266 were partly or wholly hypothetical. So, although the recorded far-field tsunami was reproduced, it 267 was not based on the actual collapse mechanism but, at best, on direct evidence such as from 268 satellite imagery, or indirect evidence such as from seismic observations of the subaerial collapse. 269 In all studies, submarine data to confirm the submarine components of the landslide source 270 mechanism was lacking. In the modelling studies using a semi-empirical landslide source (e.g., 271 Borrero et al., 2020; Heidarzadeh et al., 2020), the collapse volume and hence source strength were 272 adjusted based on field observations (e.g., near- and/or far-field runup and tide gauges). The 273 validation was then from the forward numerical modeling of the tsunami, which is rather circular. 274 Other modeling studies using an actual slide model also adjusted or confirmed their collapse 275 scenario and volume, among a range of those, based on achieving a good agreement of tsunami 276 simulations with far-field data.

While making some adjustments of the source to best match the far-field observations, most of the previous studies also demonstrated a moderate sensitivity of the predicted far-field tsunami impact to the landslide source characteristics. This shows that far-field tsunami observations alone cannot fully constrain the 2018 AK collapse parameters and, hence, stresses 281 the need for also using near-field tsunami data and, more importantly, marine surveys to do so, as

will be done in this work.

283

284 3 Methods

285 3.1 Study area, computational grids, and bathymetric/topographic data

Figure 1a shows the entire study area and the footprint of the two computational grids used in the simulations of: (G2) AK's collapse and tsunami generation/near-field impact with the 3D model NHWAVE; and (G1) tsunami propagation and far-field impact with the 2D model FUNWAVE, together with their bathymetric and topographic data.

290 The near-field Grid G2 is defined with a $\Delta x = \Delta y = 30$ m horizontal resolution (Table 1), 291 from the composite bathymetry developed by Hunt et al. (2020), based on the new multibeam 292 echosounding (MBES) bathymetry acquired during their August 2019 field surveys (Figs. 2a,b), 293 combined with: (i) unpublished Sparker seismic reflection profiles acquired in 2017; (ii) basin 294 bathymetry from Deplus et al. (1995) manually modified within the deep part of the caldera to add 295 up to 10 m of sediment infill between 1995 and 2018 (based on interpreted seismic profiles in Hunt 296 et al., 2020); (iii) an 8 m DEM for the islands of the Krakatau archipelago (from 297 http://tides.big.go.id/DEMNAS); and (iv) topography from Gouhier and Paris (2018) for AK itself, 298 based on the DEMNAS DEM, with modifications to account for island growth in 2018.

The far-field grid G1 (Fig. 1a; Table 1) is Cartesian with a 50 m resolution and its bathymetric and topographic data is interpolated from Giachetti et al. (2012)'s 100 m resolution dataset. Note that even though the bathymetric data is coarser than the model grid, using a finer model grid than, e.g., the 90 m resolution used by Grilli et al. (2019) allows for a more accurate resolution of the nearshore wave physics. Commented [SG4]: Simon suggests discussing uncertainties in processing bathymetric survey data. But this was done I think in the Nature Comm paper Regarding the reference mean water level (MWL), when the tsunami was generated, the average elevation at the four tide gauges (WG 6-9; Fig. 1a; Table 2) was approximately ± 1.5 m over mean sea level (MSL), to which the bathymetry is referenced. Hence, this value was added to the interpolated bathymetric data for both Grid G1 and G2, prior to performing tsunami simulations (i.e., MWL = MSL ± 1.5 m). When comparing to field data specified to be referenced to MSL, a 1.5 m correction was subtracted to the field data before comparing it to results of tsunami simulation.

311 3.2 Landslide source model

312 The landslide source model was defined on the pre-collapse bathymetry/topography grid G2 313 defined above, using constraints that drew on the post-collapse bathymetric survey of Hunt et al. 314 (2020), particularly to define the boundaries of the submarine failure surface, as well as an updated 315 interpretation of the subaerial failure plane. The latter was based on a sequence of Synthetic-316 Aperture Radar (SAR) satellite images collected in the days following the collapse, alongside 317 aerial imagery collected on Dec. 23rd 2018. These images proved particularly important in defining 318 the northern and southern bounds of the subaerial collapse scar, since their position could be 319 precisely defined based on the complex coastal shape of the lava deltas. The COSMO-SkyMed 320 SAR imagery from Dec. 23rd 2018 confirms the shape of the failure scar between these two coastal 321 points (cf. Hunt et al., 2020) and was used to pick both the upper line of the headwall and the point 322 where this intersected sea-level (i.e. the 0-m contour; e.g., Figs. 2d,e). These two boundaries were 323 used to define the subaerial dimensions of the modeled landslide failure plane, and we thus 324 consider this component of the failure surface to be fixed in the range of source models described 325 below.

326	To address the limitations of the published tsunami source models of the collapse	
327	mechanism and the landslide resulting from the 2018 AK flank collapse, MBES bathymetry and	
328	seismic data were acquired in the 250 m deep basin on the southwest flank of the volcano in August	
329	2019 (Hunt et al., 2020). From detailed analyses of this marine survey data (Fig. 2a-c) these authors	
330	mapped the submarine landslide resulting from the volcanic collapse and estimated the landslide	
331	outrun deposit volume at, 0.214 \pm 0.036 km³. Rather than being volcanoclastic material, the	
332	submarine deposit is mainly composed of large intact blocks (Figs. 2a-c), confirming that the event	
333	occurred as a single en masse slide with limited fragmentation, rather than in a more piecemeal,	
334	staged process. This mechanism is also confirmed by seismic data (Gurney, 2018). From these	
335	characteristics, while there were many large landslide blocks in the deposits (up to hundreds of	
336	meters across), a granular slide rheology was deemed more relevant in our subsequent modeling	
337	than a dense fluid rheology, which is more appropriate for debris flows (although both were	
338	simulated for completeness). An additional unit, to the southwest of this main deposit, with a	
339	volume of 0.022 ± 0.006 km ³ , was interpreted as a secondary sediment failure (i.e., debris flow),	
340	triggered by the primary landslide emplacement.	Commented [AMN5]: Perhaps say why?
341	Based on the marine survey, a range of volumes were identified for the 2018 AK collapse,	
342	in combination with new analyses of subaerial observations from high-resolution satellite imagery	
343	and aerial photography that estimated the subaerial collapse volume to 0.098 ± 0.019 km ³ (cf. Hunt	
244		
344	et al., 2020), and a new interpretation of the historical growth of AK (Fig. 2d). Thus, beneath sea	
345	et al., 2020), and a new interpretation of the historical growth of AK (Fig. 2d). Thus, beneath sea level, the lateral margins of the collapse scar were defined using bathymetric features on the	
345 346	et al., 2020), and a new interpretation of the historical growth of AK (Fig. 2d). Thus, beneath sea level, the lateral margins of the collapse scar were defined using bathymetric features on the submerged flank of AK evident on the post-collapse marine survey. A subtle step in the submerged	
344345346347	et al., 2020), and a new interpretation of the historical growth of AK (Fig. 2d). Thus, beneath sea level, the lateral margins of the collapse scar were defined using bathymetric features on the submerged flank of AK evident on the post-collapse marine survey. A subtle step in the submerged SW flank, at about -120 m, that may correspond to the base of the failure plane, was used to define	Commented [SG6]: Later on we use -80 m for the shallowe and -220 m for the deepest. We need to reconcile
 344 345 346 347 348 	et al., 2020), and a new interpretation of the historical growth of AK (Fig. 2d). Thus, beneath sea level, the lateral margins of the collapse scar were defined using bathymetric features on the submerged flank of AK evident on the post-collapse marine survey. A subtle step in the submerged SW flank, at about -120 m, that may correspond to the base of the failure plane, was used to define the minimum collapse volume scenario (Fig. 2e), which has a shallower failure surface than that	Commented [SG6]: Later on we use -80 m for the shallowe and -220 m for the deepest. We need to reconcile

volume of 0. 16 0.175 km³. Identification of this shallow failure surface was still uncertain,
however, because of burial by post-collapse deposits. Additional features on the NW and S flank
of AK, that align with the subaerial margins of the scar, alongside deeper features on the SW flank
(cf. Hunt et al., 2020), were used to define a larger, deeper-seated failure surface, whose volume
was estimated at 0.272 km³ (Fig. 2e). Both end point collapse volumes include a 0.098 km³
subaerial component.

of Grilli et al. (2019) for their minimum 0.22 km³ volume scenario, yielding a minimum collapse

349

Commented [sw7]: See my comments below – in Hunt et al. the value is 0.16, but we need to check the origin of this with James, as our final landslide surface (minimum) comes up with a volume of 0.175 which corresponds to your lowest modelled volume. It would be better to make the two papers consistent if James is happy to update his volume.

Commented [SG8]: To be reconciled with Hunt et al

356 Comparing the volumes estimated purely from the MBES survey to those estimated based 357 on the failure surface location and geometry, we find good consistency. The main part of the 358 landslide deposits forms a blocky mass, identified in the August 2019 MBES data (Figs. 2b,c) and 359 interpreted as representing material directly derived from the island flanks, with a volume of 0.214 360 ± 0.036 km³. A more distal part of the deposit is interpreted as a secondary debris flow, resulting 361 from sediment mobilized by landslide emplacement and seafloor incision, with a volume estimated 362 to, 0.022 ± 0.006 km³. The estimated primary deposit volume of 0.214 km³ lies between the two 363 end-point failure-surface-derived volumes described above (Fig. 2e). Given that the mass is likely 364 to have expanded upon fragmentation, and is potentially bulked via seafloor erosion, this suggests 365 that an increase in the volume of the landslide, compared to the maximum volume derived from 366 the shallower failure surface, can be accounted for by these phenomena. Additionally, some of the 367 failed mass could have remained within the scar region and been subsequently buried; hence, it 368 would not be included in the MBES estimate of deposit volume. Consequently, we cannot reject a 369 scenario with a deeper-seated failure plane and a larger volume of 0.272 km³ based on the MBES 370 surveys, although our interpretation is that the primary failure volume was likely closer to our 371 minimum estimate (0.175 km³).

Commented [SG9]: To be reconciled with Hunt et al.

372 Within the blocky landslide deposit (Fig. 2b,c), it can be assumed that transport of all 373 material derived from the volcano flanks was tsunamigenic. Because of the potential for expansion 374 and incorporation of seafloor material, we use the scar-derived volumes to define the range of 375 source-volumes for tsunami modelling. In addition to this, mobilization of seafloor sediment 376 triggered by primary landslide emplacement (forming the secondary debris flow deposit) may also 377 have contributed to tsunami generation. However, given that this must have followed the main 378 stage of landslide motion, was in relatively deeper water, and was an order of magnitude smaller 379 in volume, we assume that this material was not significant in contributing to the main tsunami 380 generation. The debris flow volume also falls well within the range of uncertainties of the estimated 381 landslide volume.

382 In the modeling, the above uncertainty in AK's collapse parameters is represented by 383 defining four landslide (and failure surface) geometry and corresponding volume scenarios, for 384 which we use the same subaerial pre-collapse geometry in every case (based on the SAR-derived 385 collapse-scar position), intersecting the NE flank at about 100 m elevation (Fig. 2e). For the 386 submarine surface, we use the minimum and maximum bounds of the failure surface described by 387 Hunt et al. (2020) and discussed above, along with two intermediate scenarios, defining maximum depths on the SW flank ranging from -80 to -220 m (Fig. 2e). All four failure surfaces cut the 388 389 active vent position at depths ranging from 25 to 40 m, which is consistent with the vigorous 390 Surtseyan eruptive activity that immediately followed the collapse. Based on the pre-collapse AK 391 topography (maximum 335 m), refined based on high-resolution satellite images (Novellino et al., 392 2020) the volumes associated with these 4 scenarios were computed to: (1) 0.313; (2) 0.272; (3) 393 0.224; and (4) 0.175 km³. The latter two compare closely with the deposit volume estimate, given 394 uncertainties and allowing for some degree of expansion, while the first two scenarios are larger,

Commented [sw10]: -80 based on my model, but again I'm just a bit concerned this is inconsistent with some of the text described by James, even though it uses the same definition of the failure limits. In his paper he describes the step at -100 to -120m, so I need to discuss with him so that we can ensure the two are as consistent as possible. He doesn't clearly show the maps of the failure surface in the main paper (perhaps in the supplement, but I don't have a copy of the final ones he submitted), but the model is based on the outlines agreed with him, so while I'm happy that they match we need to ensure that the descriptions match as far as possible between the two papers, as there is some inconsistency in the numbers at the moment.

Commented [sw11]: This matches the volumes I calculated, and they should also be the same as those described in Hunt et al

Commented [SG12]: Simon asks to discuss "degree of expansion". Dilation should be included in the model (I agree it would be good but it is not yet included in NHWAVE). We can add that to the discussion as a source of uncertainty to consider in future improvements.

395	but consistent with some bathymetric features and the possibility that some of the failure mass
396	remained within the collapse scar. The first scenario is close to the largest volume originally
397	simulated by Grilli et al. (2019), and the second is close to that of their likeliest scenario. Among
398	these scenarios, the third one, with a 0.224 $\rm km^3$ volume, is the estimated mean landslide deposit
399	volume plus 50% of the debris flow and hence is deemed our preferred (or likeliest) volume
400	scenario in terms of providing the best representation of the tsunamigenic mass movement
401	consistent with the MBES survey. The post-collapse bathymetry for this scenario is shown in Figs.
402	1c and 3b. Note, the latter figure shows that the specified failure surfaces are not planar but slightly
403	concave. This is a necessary shape given the relatively steep gradient (30-40 degrees) of the
404	subaerial failure plane (constrained from SAR imagery and consistent with the volcanic vent being
405	cut beneath sea-level) but the need for the foot of the failure to emerge within the submerged flank
406	of AK, and is also typical of the morphology of volcanic lateral collapses.

407

408 3.3 Tsunami generation and propagation simulations

409 3.3.1 Numerical tsunami models.

410 Two numerical models are used in simulations of AK's 2018 collapse and tsunami411 generation, propagation and coastal impact, which are briefly described below.

412 NHWAVE (Ma et al., 2012), a three-dimensional (3D) non-hydrostatic model, is used to 413 simulate both AK's volcanic collapse scenarios, and the corresponding tsunami generation and 414 near-field impact, on AK and surrounding islands, in Grid G2 with a 30 m horizontal Cartesian 415 grid with 1,155 by 9,55 cells, using 7 boundary fitted water layers in a vertical σ -coordinate system 416 (Figs. 1b,c; Table 1). With one layer, the model provides the same order of dispersion as a 417 Boussinesq model such as FUNWAVE, detailed hereafter, and higher-order dispersion effects Commented [AMN13]: This could be better worded.

418 when using more layers. NHWAVE has been used, and experimentally validated (e.g., Ma et al., 419 2012), to model tsunami generation from solid slides (landslides or slumps) (e.g., Grilli et al., 420 2015; Schambach et al., 2019) and from dual sources coseismic/solid submarine mass failures 421 (Tappin et al., 2014). NHWAVE was extended to simulate tsunami generation by deforming slides, 422 both submarine and subaerial, assumed to behave as either a granular medium or a dense 423 Newtonian fluid (Ma et al., 2015; Kirby et al., 2016). These NHWAVE models were applied to 424 case studies for deforming slide sources (e.g., Grilli et al., 2017b, 2019; Schambach et al., 2019), 425 and validated based on laboratory experiments for those studies (Grilli et al., 2017b), as well as 426 for dual sources involving a combination of coseismic and deforming underwater/subaerial slides 427 (e.g., Grilli et al., 2019; Schambach et al., 2020a,b).

428 Since the work of Grilli et al. (2019), a new version of NHWAVE has been developed 429 (Zhang et al., 2021a,b) that includes effects of vertical acceleration (i.e., non-hydrostatic pressures) 430 within the slide material layer, which was neglected in the earlier implementation (Ma et al., 2015). 431 Considering the steep slopes of both AK and the surrounding islands, it was anticipated that such 432 effects might be important. This was confirmed here by comparing, in Supplementary file #2, 433 simulations of the Grilli et al. (2019) preferred volume scenario (0.272 km³), with both a granular 434 and viscous rheologies, and with and without the non-hydrostatic effects included in the equations 435 for the slide layers. Results for both rheologies showed that slide motion and wave generation are 436 significantly affected, with larger waves generated and much larger runups occurring on the near-437 field islands, particularly Panjang and Sertung, when non-hydrostatic effects are neglected. When 438 comparing with near-field runup measured in field surveys, a much better agreement was obtained 439 with the newer version of the model that accounts for non-hydrostatic effects within the slide layer.

For these reasons, this newer version of NHWAVE by Zhang et al. (2021a,b) was used in thepresent study.

FUNWAVE-TVD (Shi et al., 2012; version 3.0 is used), a two-dimensional (2D) fully nonlinear Boussinesq wave model, is used to simulate far-field tsunami propagation and coastal impact in Cartesian Grid G1 with a 50 m resolution and 3,900 by 3,680 cells (Fig. 1a; Table 1), which is acceptable in view of the small geographic area considered here. To improve dispersive properties, the horizontal velocity used in this Boussinesq model is that at a depth z = -0.531 h. To prevent reflections from the open boundaries of grid G1 (Fig. 1a), 10.8 km (or 216 grid cells) wide sponge layers are specified along its 4 boundaries.

449 Both NHWAVE and FUNWAVE-TVD used a Courant number and Froude cap condition 450 to adaptively specify the time step in simulations to achieve optimal accuracy. In shallow water 451 and during runup this can lead to using prohibitively small time steps, which is prevented here by 452 specifying a minimum depth truncation of 1 m and 0.05 m in the NHWAVE and FUNWAVE 453 simulations, respectively. The 0.05 m minimum depth has little effect on FUNWAVE simulations 454 of the far-field tsunami impact. In the near-field, considering the very large waves and runups 455 modeled with NHWAVE, the 1 m minimum depth also does not significantly affect simulation 456 results. Both models are parallelized with MPI, allowing to efficiently run them on large computer 457 clusters. Here we typically used 20 processors to run each scenario. Finally, both models are open 458 source and available on github, together with their user manual and benchmarking examples.

459 3.3.2 Modeling methodology.

Simulations of AK's collapse and tsunami generation and near-field impact are first performed in grid G2 with NHWAVE, for the 4 volume scenarios, and for each of those, assuming either a granular or a dense fluid rheology. When the slide is fully at rest and waves approach the boundary 463 of grid G1 (Fig. 1a), NHWAVE results for surface elevation and horizontal velocity interpolated 464 at 0.531 times the local depth are used to initialize FUNWAVE simulations in Grid G1. These are 465 then run for another 2h of tsunami propagation time, to make sure all the diffraction, and multiple 466 reflection effects on the tsunami from the shores and many islands of the Sunda Straits are included 467 in the results.

468 In NHWAVE simulations, for each of the 4 specified collapse surfaces and volumes (Fig. 469 2e), we use the same parameterization of the slide rheology as in Grilli et al. (2019), i.e.: (i) a 470 Newtonian fluid of density $\rho_c = 1,550 \text{ kg/m}^3$ and the kinematic viscosity of a debris flow, $v_c = 0.5$ 471 m²/s; or (ii) a granular medium with $\rho_c = 1,900 \text{ kg/m}^3$ for the solid part and similar to Giachetti et 472 al. (2012), an internal friction angle $\phi_{ic} = 10^\circ$, a basal friction angle $\phi_{bc} = 2^\circ$, and a 40% porosity. 473 With this data, assuming a water density $\rho_w = 1,025 \text{ kg/m}^3$, the average density of the granular 474 medium is $\rho_{ac} = 1,550 \text{ kg/m}^3$. For each of these 8 scenarios, NHWAVE was run up to 410 s; 475 however, results showed that the time when the generated tsunami waves approach the boundary 476 of grid G1 is t = 380 s (e.g., Fig. 7h), which is used to prevent any perturbation of the solution.

477 Finally, in both model grids, in the absence of site-specific data we specify a constant 478 bottom friction coefficient $C_d = 0.0025$, which corresponds to coarse sand. While this value may 479 be too small to model friction on the rough walls of the caldera to the SW of AK, earlier work has 480 shown that bottom friction only significantly affects tsunami propagation (in particular reduces 481 tsunami elevations) over shallow areas where propagation distances are many dominant 482 wavelengths (Tehranirad et al., 2015). In this case, the bottom velocity caused by the long tsunami 483 waves (in terms of wavelength to depth ratio) is consistently large. Considering the fairly short 484 dominant period of the generated waves, here, bottom friction will only significantly affect tsunami 485 propagation towards Java and Sumatra, in the shallow eastern side of the Sunda Straits. In the

486 caldera to the SW of AK, however, both the water depth is large and the tsunami propagation 487 distances are short, and bottom friction effects are thus expected to be small; hence, the accuracy 488 of the selected C_d value is not important.

489 3.3.3 Computation of flow depth and runups based on model results

490 For all modeled scenarios, the pre- and post-processing of NHWAVE and FUNWAVE input data 491 and results was done using Matlab codes custom-developed for this purpose. Results produced 492 include most of the figures shown in this work, animations of model outputs provided in 493 supplementary material and, where available, a comparison of numerical simulations to field data. 494 In this respect, specific algorithms were developed to accurately extract the maximum tsunami 495 flow depth along the shore (0 m contour in MWL), and runup from model results. In both cases, 496 this was done for: (flow depth) by computing the location of the 0 m bathymetric contour in the 497 bathymetric/topographic data; and (runup) by computing the location of the 0 m elevation contour 498 in the maximum elevation minus bathymetric/topographic data set. Then, the envelope of 499 maximum computed surface elevation (e.g., Figs. 8 and 9d) was interpolated along both contours 500 to yield the flow depth and runup data along the selected coastline (e.g., Figs. 14-16). Note that to 501 clarify the visualization of these results, 4 classes of elevations were selected and color coded as: 502 (yellow) 0-1 m; (green) 1-2 m; (red) 2-4 m, and (pink) > 4 m; dots with this color code were plotted both 503 in plan view along the coast and in elevation figures to be compared with field data (e.g., Figs. 14, 504 16).

505

506 3.4 Tsunami field survey data

507 To validate our numerical model results, we used a comprehensive set of data, including marine 508 field surveys, satellite images, bathymetric data as discussed above, and onshore surveys of 509 tsunami impact (Figs. 12-16). The inshore survey data included: (i) the tree line drone survey 510 conducted on Rakata, Sertung and Panjang during our August 2019 field campaign (Figs. 11, 13), 511 and (ii) the runup and flow depth measurements made in the near- and far-field by TDMRC (2019), 512 Muhari et al. (2019), Putra et al. (2020), Borrero et al. (2020), and Heidarzadeh et al. (2020). In 513 addition, we used the extensive video made by Reynolds (2019), during his 01/11/2019 near-field 514 drone survey of AK and the three surrounding islands, of which salient images were extracted by 515 Grilli et al. (2019) (see their supplementary Fig. S8). One example is in Fig. S1f. These surveys 516 show that, in the near-field, the tsunami generated by AK's 2018 collapse caused up to 85 m runups 517 on the islands of Rakata and Sertung and, in the far-field, up to 13.5 m runups on the nearby coasts 518 of Java and Sumatra.

519 Additionally, as in Grilli et al. (2019) and all other modeling studies, time series of surface 520 elevations simulated for each scenario are compared with detided free surface elevations measured 521 at 4 tide gauges located at (Fig. 1a; Table 2): (5) Serang, Marina Jambu, (6) Ciwandan, (8) Kota-522 Angung, and (9) Panjang. Grilli et al.'s (2019) Supplementary file #3 explains how the raw data, 523 measured at a 1 min interval, was detided to obtain the tsunami signal (their Fig. S5) and shows 524 where each tide gauge was located (their Fig. S4), pointing out that each gauge is surrounded by 525 some reflective (or dissipative) coastal structures, not represented in the model grids, that can affect 526 tsunami signal in various ways (including seiching). Table 2 provides the location of each tide 527 gauge, its depth in grid G1 and the arrival time of both a 1 cm tsunami elevation and the first 528 significant wave crest. Fig. 10 shows the complete (detided) tsunami time series measured at each 529 gauge by two different independent instruments operating at each gauge (see details in Grilli et al., 530 2019); there are some differences (sometimes large) between the measurements of the two 531 instruments at each gauge, which allows quantifying experimental errors. The individual data 532 points in the time series illustrate the coarse 1 min temporal resolution of the measured signal.

533

534 4 Tsunami simulation results

535 4.1 Slide motion and deposits

Results of combined NHWAVE-FUNWAVE simulations in Grids G2 and G1 of the 4
volume scenarios ((1) 0.313; (2) 0.272; (3) 0.224; and (4) 0.175 km³), each with either a granular
or viscous rheology, are discussed hereafter.

539 Figures 4 and 5 show examples of slide motion and free surface elevations simulated with 540 NHWAVE. Fig. 4 first compares results in a vertical plane along a SW transect into AK, for the 541 preferred volume scenario (0.224 km³), using either a granular or viscous rheology. We see that 542 the change in rheology only moderately affects slide deformation for small times (t < 80 s), and 543 hence corresponding wave generation, but that differences in slide runout are much larger later in 544 time ($t \ge 120$ s), although this stage of motion is no longer tsunamigenic as the slide deposits are 545 too deep. At t = 200 s (Fig. 4h) the granular slide deposits have nearly stopped and have mostly 546 accumulated in the caldera, at the toe of AK's failure surface, whereas the viscous slide deposits 547 have moved further onto the caldera bottom and are still moving. While the granular slide deposits 548 appear to be located in the general area where the actual deposits were mapped during the August 549 2019 marine survey (Figs. 2b,c) (Hunt et al., 2020), the viscous slide deposits have moved beyond 550 this area; hence simulations based on the granular rheology appear to be more consistent with field 551 data than those with the viscous rheology. This is confirmed in Fig. 5, which shows greater details 552 of the 3D granular slide motion and deposits for the same volume scenario. Here in the last panel at t = 420 s (Fig. 5h), we see more clearly where the main slide deposits are located (i.e., their 553 554 runout) and how thick they are (up to 94 m) at the end of the motion, which appears to be consistent 555 with field observations (Fig. 2c). It should be noted that while the main collapse deposits are to

556	the SW, there is a layer of a few meters of granular material deposited on the opposite, NE side of
557	AK and to the NW and SE, which help cause small additional wave generation in those directions.
558	Given the low degrees of fragmentation evident from the very large blocks in the observed deposit
559	(Fig. 2), these features in model results may not be representative of an actual deposit distribution
560	and are more likely an artifact of a landslide model based on a continuous granular rheology. A
561	similar discrepancy between observed and modeled deposits was noted by Ward and Day (2006)
562	in their study of the 1980 Mount St Helens event, which caused a large debris avalanche.

Videos of computed slide motions with and without surface elevation, and for a granular
material or a viscous slide are given in supplementary material for the preferred volume scenario
(0.224 km³); see, AK_slide3D_gran.mp4, AK_slide3D_visc.mp4, AK_wave_slide3D_gran.mp4,
AK wave slide3D visc.mp4.

567

568 4.2 Near-field tsunami generation

Figures 4 and 7 show snapshots of free surface elevation at times t = 10, 20, 40, 80, 120,160, 200, and 380 s, computed for the preferred volume scenario and a granular or viscous rheology, and Fig. 6 compares time series of surface elevation computed at the 5 numerical wave gauges (Fig. 1b; Table 2) specified in grid G2, for the 8 modeled scenarios (4 volumes and 2 rheologies). Other snapshots of surface elevations for scenarios not shown here look qualitatively similar to those in Fig. 7. Videos of computed surface elevations are given in supplementary material.

Results in Figs. 4 and 7 show that, in the first 20 s of AK's collapse, a large-scale subaerial
slide motion occurs down the volcano, triggering a 50+ m horseshoe-shaped leading elevation
wave. From 20-80 s, as the slide moves mostly underwater (for all 4 volume scenarios), an up to

579 30+ m trough (negative elevation wave) forms near the volcano on the SW side, while the leading 580 elevation wave radiates as a cylindrical crest of decreasing height. Figure 6a shows that these 581 processes are well captured at WG 1, which is located directly SW of AK (Fig. 1b; Table 2); at 582 this site, depending on the scenario, a 25-33 m leading elevation wave arrives at $t \approx 60$ s, followed 583 by a 0-10 m trough. At WG 2 and 3, further NW and SW of the volcano, Figs. 6b,c show that, later 584 in time ($t \approx 175$ s), the large elevation wave and its trough (first depression wave) have essentially 585 propagated radially, with only a small decrease in the crest height. The propagation of the 586 horseshoe-shaped leading elevation and first depression waves, with their gradual directional 587 spreading and reduction in elevation, are clearly seen in Figs. 7c to 7h. As these waves propagate 588 away from AK, however, for t > 80 s (Fig. 7d), they start interacting with and running up both the 589 N shore of Rakata and S shore of Sertung, causing very large runups.

590 To the NE of AK, for t > 100 s, we see a significant tsunami impact occurring on Panjang's 591 southern tip (25+ m runup) and, for t > 150 s, a more moderate impact on its northern tip, that are 592 due to both the propagation and refraction around AK's bathymetry of the leading horseshoe-593 shaped wave (Figs. 7d-f) and later on its reflection off Rakata and Sertung. Finally, in Figs. 7g,h, 594 we see that large waves are propagating in the SW, E and N directions away from AK. For the 595 latter two directions, these waves are well captured at WG 4 and 5 (Figs. 6d,e), where we see 596 leading elevation waves of about 4 and 5 m, respectively. Fig. 7h also confirms that at 380 s, the 597 leading waves have not yet reached and interacted with the outer boundary of Grid G2.

598 Considering the 8 different scenarios, results at WG 1 to 5 in Fig. 6 show that while, overall, 599 all generated waves exhibit the same large-scale characteristics, both a change in collapse volume 600 and rheology affect wave elevation and phase to various extents. Between rheologies, the granular 601 rheology generates slightly smaller leading waves in all cases than the viscous rheology 602 (particularly to the SW), and the larger the collapse volume the larger the wave elevations. [Note 603 that the first conclusion is opposite to that of Grilli et al. (2019) who found that larger waves were 604 generated by a granular slide; this could result from the use here of a much higher grid resolution 605 and the new non-hydrostatic slide model.] At all wave gauges (WG 1-5), the larger leading wave 606 is followed by smaller waves of period as low as T = 30-50 s. Over the 250 m deep caldera, these 607 waves are fully or significantly dispersive. Waves in this period range would be dispersive for 608 depth $h < gT^2/400 = 3.5-9.8$ m, hence for most of their propagation to shore, which justifies using 609 a dispersive long wave model such as FUNWAVE to model AK's collapse far-field tsunami 610 propagation.

Fig. 8 shows the envelope of maximum surface elevation computed with NHWAVE in Grid G2 for the preferred volume scenario (0.224 km³) and a granular rheology; envelopes for the other scenarios look qualitatively similar and are not shown for the sake of brevity. The figure confirms the large wave generation SW of AK, and shows that large 50-100+ m runups occur on the exposed shores of Rakata and Sertung, and 25 m runup on the south shores of Panjang. These results will be detailed later and compared to field measurements.

617

618 4.3 Far-field wave propagation and coastal impact on Java and Sumatra

For each of the 8 scenarios, FUNWAVE simulations were initialized with results of NHWAVE in Grid G2 at 380 s (Fig. 7h), interpolated onto Grid G1, and tsunami propagation and coastal impact were simulated up to t = 7,580 s from the start of the event. Figures 9a-c show snapshots of surface elevation computed with FUNWAVE for the preferred volume scenario (0.224 km³) and granular rheology at t = 380, 1800 and 3600 s. Results for the other scenarios are qualitatively similar. After 30 min, Fig. 9b shows that leading tsunami waves have started impacting the SW coast of Java, 625 around the Kolijaah and Panaitan Island areas (Fig. 1a), are impacting the south facing coast of 626 Sebesi (Fig. 1a), and are about to impact the coastlines at Ujung Kulon and Serang, Marina Jambu 627 (tide gauge (WG) 5; Fig. 1a and Table 2). To the north of the grid, leading waves are also impacting 628 the SE tip of Sumatra; waves are also propagating in the direction of tide gauges (WG) 6-9 (Fig. 629 1a; Table 2). After 1h of tsunami propagation, Fig. 9c shows a complex pattern of waves in the 630 Sunda Straits, as a result of diffraction-refraction around islands and reflection off the coasts, 631 which justifies performing simulations for a long enough time to capture maximum runup at all 632 locations within Grid G1.

633 Fig. 9d shows the envelope of maximum surface elevation computed with FUNWAVE in 634 Grid G1, after 7,580 s of simulations, for the preferred (granular) scenario. AK's collapse 635 generated initial waves with a strong SW directionality and a secondary E and N directionality 636 (Fig. 7h), which translates upon far-field propagation into a maximum impact on the SW coast 637 Java and a relatively smaller impact eastward and northward on the coasts of Java and southern 638 Sumatra (see also Fig. 9b). Additionally, wave propagation is affected by a significant bathymetric 639 feature, the moderately steep S-N oriented (around Lon. E. 105.3) linear scarp that divides the 640 shallow eastern half of Sunda Straits from the much deeper Semangka trough to the west (Fig. 1a). 641 As can be seen in Fig. 9b (and in the animation of model results provided in supplementary 642 material), this bathymetric feature causes a wave guiding effect that reinforces waves to the south 643 onto Panaitan Island, where some of the largest flow depths and runups were measured, and also 644 guides some waves to propagate northward. Comparing bathymetric contours with the maximum 645 envelope in Fig. 9d, we see that little tsunami energy propagated west of Lon. E. 105.3, and that 646 bathymetric focusing also occurs towards Ujung Kulon (Fig. 1a), which is another area where very 647 large runups were measured (see later for details of runups).

648 Surface elevation time series were simulated for the 8 scenarios, combining the four 649 volumes and two rheologies, at the locations of the 4 tide gauges (6-9 in Fig. 1a; Table 2), which 650 are compared to the measured detided surface elevations in Fig. 10. Unlike in the near-field, only 651 small differences (including on arrival time) can be seen here between surface elevations simulated 652 for the 8 different scenarios, indicating that the predictions of the tsunami far-field and impact are 653 less sensitive to details of the collapse scenario assumed for AK (i.e., changes in volume 654 size/geometry and rheology). This was already pointed out by other authors in their discussion of 655 model results (e.g., Heidarzadeh et al., 2020; Borrero et al., 2020), and also explains why studies 656 that assumed an approximate empirical source for AK's collapse or only a 2D two-layer slide 657 model, with source parameters adjusted to match far-field data at the tide gauges and/or elsewhere, 658 performed reasonably well for predicting coastal impact. However, for future hypothetical 659 collapses, in the total absence of field data to calibrate these models, they might not have fared as 660 well in predicting tsunami impact, from a single forward model simulation.

661 Comparing numerical simulations to tide gauge data, Fig. 10 shows, overall, a good 662 agreement for any scenario, particularly earlier in the time series and more so for WG 6-8 (Figs. 663 10a-c). As summarized in Table 2, arrival times of the leading crest at each gauge are predicted to 664 within 15-78 s of observations. Considering the 1 min data sampling interval of the gauges, this 665 is an acceptable discrepancy. Later in each tide gauge time series, the phase difference between 666 simulations and observations increases, but the trough-to-crest height of the largest waves are well 667 predicted in the simulations. As indicated before, later in time, the signal at the tide gauges was 668 increasingly affected by any local effects and seiching not resolved and simulated in Grid G1, both 669 due to the limited 50 m resolution and the moderately coarse 100 m resolution of the available 670 nearshore bathymetry and topography. Finally, as reported by eyewitnesses, simulations predict that multiple large waves of fairly short period (2–10 min) impacted the coast, with the second orlater waves being the largest.

For each of the 8 scenarios, arrival time at the tide gauges is, to the first-order, governed by wave celerity, which strongly depends on bathymetry and to some extent on frequency for dispersive waves. An additional effect of amplitude dispersion may speed-up wave propagation for the largest waves in the near-field, but this effect will also be similar for all scenarios, as their near-field waves are quite similar (see Fig. 6). This explains the small range in arrival time difference, with the field data listed in Table 2 for the 8 scenarios.

679

680 4.4 Near-field runups

681 Grilli et al. (2019) pointed out the intense and continuous phreatomagmatic explosive activity that 682 immediately followed the collapse of AK, both obscuring the skies and discharging large volumes 683 of material that rapidly modified the post-collapse topography of AK and surrounding bathymetry. 684 Hunt et al. (2020) made a detailed analysis of these early stages of AK's post-collapse regrowth, 685 using both satellite images and submarine surveys, and quantified the large changes that took place 686 in AK's coastline and subaerial geometry (e.g., such as Fig. S1b and S1e for AK; see also Novellino et al., 2020). This post-collapse eruptive activity paused on Jan. 11th 2019, and Reynolds 687 688 (2019) was able to conduct a drone survey of AK and the islands of Rakata, Sertung and Panjang 689 (e.g., Fig. S1f and supplementary 4 in Grilli et al., 2019), that confirmed AK's coastline changes 690 inferred from SAR images. Arguably more important was their documentation of the large runups 691 the tsunami caused on the island of Rakata, Sertung and Panjang. Based on these images, Grilli et 692 al. (2019) estimated that 50+ m runups occurred on Rakata's N shore and Sertung's S shore. 693 Subsequent field surveys in 02/2019 by Borrero et al. (2020) and August 2019 by the authors

694	confirmed and quantified these early observations of near-field tsunami impact, and provided geo-
695	localized runup values reaching 85+ m on both islands (Fig. 11), with additional data on Panjang.
696	However, because Panjang was positioned downwind of AK, extensive ashfall-driven vegetation
697	damage, combined with the steep cliffs on the W coast made the runup line on Panjang difficult to
698	unambiguously identify. Finally, Borrero et al. (2020) also measured runup on Sebesi island, north
699	of Panjang, which we also consider to be part of the near-field tsunami impact (Fig. 1a).

700 For the preferred collapse volume scenario, with granular material, Figures 12a-c show 701 zoom-ins of the maximum envelope of surface elevation computed with NHWAVE (Fig. 8) onto 702 the NW shore of Rakata, SW shore of Sertung and S shore of Panjang, and Fig. 12d shows a zoom-703 in on Sebesi of the maximum envelope of surface elevation computed with FUNWAVE for the 704 same scenario (Fig. 9d). The location of our August 2019 drone tree line survey is marked on Figs. 705 12a,b, and the location of four runups/flow depth measurements made on Sebesi by Borrero et al. 706 (2019) are marked on Fig. 12d (7.5, 9, 2.8, 2.5 m from W to E, respectively); the latter values are 707 consistent with those we estimated during our August 2019 survey of Sebesi, in part based on 708 interviewing eyewitnesses. On both Rakata and Sertung (Figs. 12a,b), our predicted runup line 709 touches or goes over the 50 m contour and parallels the drone survey quite well, except at its 710 highest points; those however occur on steep, nearly vertical, cliff faces (Figs. 11a,c) that are not 711 well resolved with a 30 m horizontal grid. On Panjang, in Fig. 12c, our results show runups of 25-712 30 m on the island's SW tip, tapering to 8-10 m on the NW part of the western shore; the latter 713 values match those reported by Borrero et al. (2020), who could not make a precise survey due to 714 the difficulty in accessing the island, which is faced by steep cliffs on much of its western side 715 (Fig. 11e). In Fig. 12d, our model results show a close agreement with the 4 measured runups on 716 Sebesi's S and SE shore.

717 Figure 13 details the near-field runups computed on the 3 islands for the 8 modeled 718 scenarios (4 volumes and 2 rheologies), compared to available runup measurements and our drone 719 surveys. Overall, on Rakata and Sertung (Figs 13a,b), although all scenarios fare quite well, our 720 preferred volume scenario with a granular rheology appears to best match the quantitative field 721 data, as well as images from the 01/11/2019 and Borrerro's et al.'s (2020) 02/2019 field survey 722 (Figs. 11b,d) of these islands. On Panjang (Fig. 13c), all our model results are below our tree line 723 drone survey (Fig. 11e) but, again, this was done along a nearly vertical cliff face, a location where 724 it was difficult to estimate the runup line precisely and which is not well-discretized in our model 725 grid; hence, there is large uncertainty on both these runup measurements and their model 726 simulation. We note that all scenarios predict an 8 m runup on the NW side of the island as was 727 reported by Borrero et al. (2020).

728

729 4.5 Far-field runups

Far-field flow depth and runups were measured along the coasts most exposed to the tsunami in Java and Sumatra in several field surveys. The first one (TDMRC, 2019 took place in 01/2019, soon after the event) was the only such data available to Grilli et al. (2019) to validate their modeling. However, field surveys were also later performed by Muhari et al. (2019), Putra et al. (2020), Borrero et al. (2020), and Heidarzadeh et al. (2020). Figures 14 to 16 compare model results obtained for our preferred collapse scenario (granular rheology) with this data which, to our knowledge, is all such data available to date.

Figure 14 shows a zoom-in along the coast of Java (Fig. 1a) on the envelope of maximum surface elevation computed with FUNWAVE (Fig. 9d). As detailed in the methods section, both the maximum flow depth at the shore and the runup were extracted from these results and, for 740 clarity, color coded in 4 classes of surface elevation. Due to the complex geometry of the coast, 741 the same values of flow depth and runup were then plotted as a function of longitude and latitude 742 in 4 subfigures (Figs. 14a,b,d,e); on the plan view (Fig. 14c), the color coded flow depth values 743 were plotted along the shore. Fig. 14c shows that, as expected from the tsunami directionality, 744 wave guiding effects offshore, and wave refraction nearshore, leading to focusing/defocusing 745 effects, the alongshore variation of maximum tsunami impact is a highly irregular on SW Java; 746 this causes similarly large alongshore variations in flow depth and runup seen in Figs. 14a,b,d,e. 747 The field data for both flow depth and runup is plotted on top of the elevation figures showing 748 model results, in Figs. 14b,d and. 14a,e, respectively. Overall, there is a good agreement of model 749 results with the field measurements, and more so for flow depth at the coast, which is less sensitive 750 to irregularities of the terrain and the built-up elevation maps, that are not represented in our 50 m 751 resolution grid.

752 Figure 15 shows zoom-ins of results presented in Fig. 14 in three of the most impacted 753 areas along the coast of Java where field surveys were conducted, namely (Fig. 1a): (PI) Panaitan 754 Island; (UK) Ujung Kulon; and (K) Kolijaah. Model results for the preferred volume scenario 755 (granular rheology) are compared to the locations/values of measured maximum runups, wherever 756 available (Fig. 15e), or otherwise to field data measured by Borrero et al. (2020) marked onto 757 Google Earth images of each site (Figs. 15b,d). These measurements were provided as raw or 758 detided, so here we are plotting their raw values compared to our results with respects to MWL. 759 [Note, Borrero et al. only assumed a 2 cm tide throughout without justification, which will 760 introduce some uncertainty in the comparison; also, their measurements from UK (Fig. 15d) are 761 reported on Figs 14a,e as runup, since these values were measured inland.] At PI (Figs. 15a,b), 762 the model accurately predicts the 6-8.4 m range (referred to MWL) of maximum tsunami 763 elevations measured at the marked locations along an approximate N to S survey from the tip of 764 the island (Fig. 15b,; Borrero et al, 2020). At UK (Figs 15c,d) the model predicts slightly less (6.5 765 to 9 m) than the 6.9-11.5 m range (referred to MWL) of maximum tsunami heights measured at 766 the marked locations from N to S from the tip of the Peninsula (Fig. 15d; Borrero et al, 2020); 767 however, the largest flow depth was measured at an isolated tree (their Fig. 12) and our 50 m 768 resolution model grid cannot represent this level of detail. Finally, at K (Figs. 15e,f) the model 769 predicts most of the runups (both location and value) measured by Muhari et al. (2019), Putra et 770 al. (2020), and Heidarzadeh et al. (2020), reasonably well. Some of the reported measurement 771 locations show a mismatch, but the majority of measured runups in the 3 surveys align well with our predicted inundation limit. At the K location, the Google Earth image (Fig. 15f) is only 772 773 provided for reference.

Figure 16 shows results similar to those of Figure 14, for flow depth at the coast predicted along the SW shore of Sumatra for our preferred collapse volume scenario (granular rheology), compared to the available data from field surveys; the agreement between both is quite good here as well. The largest tsunami impact occurred in the area of Waymuli (W in Fig. 1a, around 105.6348 E), of which Fig. 16c shows a picture of the damage taken by Fritz et al. (2019) during their 02/2019 survey.

780

781 **5 Discussion and conclusions**.

New numerical simulations of AK's 2018 collapse and tsunami generation, propagation, and coastal impact were performed with state-of the-art numerical models, including a novel landslide tsunami model for granular and viscous slides that includes non-hydrostatic effects of vertical acceleration in the slide material. Results show that incorporating non-hydrostatic effects is important for accurately simulating tsunami generation and near-field impacts from the AK flank collapse. This is illustrated in the 8 scenarios we used, which combined 2 different rheologies (granular and viscous fluid material) and 4 different volumes obtained from a new parametrization of the collapse based on our August 2019 marine hydroacoustic survey (cf. Hunt et al., 2020), field observations and new interpretations of high-resolution satellite imagery.

791 Based on our improved knowledge/understanding of subaerial and submarine data, from 792 which we better constrained the geometry and magnitude of the landslide mechanism, we also 793 improved on previous interpretations of the primary landslide scar, which bisected the Anak 794 Krakatau edifice, cutting behind the central vent and removing 50% of its subaerial volume. From 795 our new combined subaerial and marine data sets we also provide a better validated estimate of 796 the landslide failure volume, which lies within a range of 0.175 to 0.272 km³, with a preferred 797 scenario of 0.224 km³. This volume is supported by our estimates of the actual landslide deposit 798 volume mapped in the basin to the SW of AK, which is comprised of a main volume of 0.214 \pm 799 0.036 km^3 , with a much smaller volume ($0.022 \pm 0.006 \text{ km}^3$) secondary debris flow. From our new 800 minimum and maximum bounds of the landslide failure surface and geometry, we defined 4 801 collapse scenario geometries, with volumes between 0.175 and 0.313 km³.

Observations of a single tsunami wave train, with no subsequently generated waves, are consistent with our interpretation of landslide failure, in a rapid, single phase *en masse* movement, rather than a more piecemeal process; this single event interpretation is also supported by seismic data. In the context of the uncertainty in field observations, all our scenarios successfully reproduced the near- and far-field tsunami flow depth and runup observed in all post-event field survey results published to date, as well as arrival times and time series of surface elevations at tide gauges, and eyewitness reports. This match between our model results and field observations Commented [TDR14]: Need to clarify that this is the evacuated volume

Commented [sw15]: As above, I think these need correcting in Hunt et al. to correspond to the value stated below shows that our estimated landslide volume range and material rheologies are appropriate to thecollapse event.

811 Despite an observed moderate sensitivity of tsunami impact to the range of modeled 812 landslide sources, particularly in the far-field, the granular rheology appears to yield slide deposits 813 in better agreement with the marine deposits mapped in the 2019 survey (both location and 814 thickness). Additionally, near-field runups are also better predicted using a granular rheology. 815 Regarding the collapse volume, the likeliest value inferred from the 2019 field survey, together 816 with a refined analysis of satellite images, is 0.224 km³ (referred to in this paper as our preferred 817 scenario), which appears to provide the overall best agreement with the near-field runup 818 measurements, as well as the far-field data. Hence, while the volume is harder to constrain using 819 far-field data, we conclude that tsunami modeling supports the likeliest scenario inferred from the 820 2019 marine geology survey, although the constraint is weaker than for the rheology.

821 The AK event highlights the significant hazard posed by relatively small-scale lateral 822 volcanic collapses, which occur en-masse, without any precursory signals, and are an efficient and 823 unpredictable tsunami source. Our successful simulations demonstrate that current numerical 824 models can accurately forecast tsunami hazards from these events, even assuming a large 825 uncertainty on the source parameters (e.g., collapse failure plane and volume); this is why the 826 precursor work of Giachetti et al. (2012) provided a reasonable forecast of the event that took place 827 at AK in 2018. In cases such as Anak Krakatau's, the absence of precursory warning signals 828 together with the short travel time following tsunami initiation present a major challenge for 829 mitigating tsunami coastal impact, stressing the need to install early warning systems. In their recent work on AK, Mulia et al. (2020) suggested that a high frequency (HF) radar could have 830 831 been useful in providing an early detection of the tsunami generated by AK's collapse. In fact,
Grilli et al. (2016, 2017a) proposed new algorithms for processing HF radar data to efficiently
detect tsunami signals; by performing modeling similar to that reported here, they demonstrated
that their algorithm could provide an early detection of landslide tsunamis. Guérin et al. (2018)
later applied the method to detect a large meteo-tsunami/surge in actual HF radar data, off of
Toffino, BC.

837	Finally, an important physical aspect not included in NHWAVE is slide dilation, which
838	results from water being sucked into the granular material during slide motion. While this effect
839	could affect tsunami generation, the good agreement observed in the near-field between the
840	measured and predicted runups would indicate that this was not significant during AK's event.
841	Additionally, the many large blocks seen in the debris deposits would indicate that the amount of
842	interstitial water may have been smaller than assumed in simulations and actual dilation effects
843	were small. Nevertheless dilation would be important to include in the model and study in future
844	work.

845

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855	Committee on Earth Observation Satellite's Earth Observation (CEOS)'s Volcano Demonstrator
856	(Order ID: 740794). FUNWAVE-TVD is open source software available at
857	http://github.com/fengyanshi/FUNWAVE-TVD/. NHWAVE is open source software available at
858	http://github.com/jimkirby/nhwave/. The Indonesian Ministry of Marine Affairs and Fisheries
859	(http://tides.big.go.id) is gratefully acknowledged for providing the authors with tide-gauge data,
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- 1107 validation. Ocean Modelling (minor revisions).

Tables

Grid	Mesh size	Resolution	SW Corner
	(N, M)	(m)	(Lat., Lon.)
G1	3680,3900	50	-7°, 104.4°
G2	955, 1155	30 (horiz.)	-6.2357°, 105.2916°
		7 σ (vert.)	

1109 Table 1: Parameters of grids used in simulations with NHWAVE (G2) and FUNWAVE (G1) (Fig. 1).

WG	Lon E.	Lat N.	Depth	t meas.	t meas.	t sim.	t sim.
	(Deg.)	(Deg.)	(m)	crest (s)	1 cm(s)	crest (s)	1 cm (s)
1.	105.4066°	-6.1234°	239.50	N/A	N/A	53-65	15.6-24.2
2.	105.3733°	-6.1524°	88.22	N/A	N/A	165.1-175.5	118.0-127.0
3.	105.4246°	-6.0691°	49.32	N/A	N/A	179-191	131.2-140.0
4.	105.4954°	-6.1279°	58.50	N/A	N/A	244.5-254.5	197.5-208.0
5.	105.3571°	-6.1361°	90.74	N/A	N/A	188.5-190.4	165.5-169.4
6.	105° 50' 15.0"	-6° 11' 21.5"	4.70	1980	1923	1995-2006	1967-1979
7.	105° 57' 10.8"	-6° 01' 02.5"	3.64	2700	2587	2712-2727	2617-2629
8.	104° 37' 08.5"	-5° 30' 01.2"	3.67	2520	2292	2550-2568	2358-2382
9.	105° 19' 06.1"	-5° 28' 08.7"	3.92	3600	3390	3660-3678	3564-3624

1110

1108

sampling interval. N/A: Not Applicable.

¹¹¹¹ Table 2: Parameters of numerical wave gauges (WG) 1-9 (Figs. 1a,b): Lat-Lon, depth (in grids G1

^{1112 (6-9)} and G2 (1-5), assuming a MWL=MSL+1.5 m, corresponding to the estimated average tide

¹¹¹³ elevation at the time of the event), and arrival time (1 cm elevation or first main crest),

¹¹¹⁴ measured/simulated range for 8 scenarios (Figs. 6,7). WG 1-5 have no measured time (Fig. 6), but

¹¹¹⁵ WG 6-9 are collocated with Tide Gauges (Fig. 10) at: (5) Serang, Marina Jambu, (6) Ciwandan, (8)

¹¹¹⁶ Kota-Angung, (9) Panjang. In simulations, AK collapse is assumed to take place at 20:57' local

¹¹¹⁷ time (UTC + 7). Simulated crest arrival time ranges at 9 WG for 8 scenarios are within 2-18 s.

¹¹¹⁸ Simulated differences in crest arrival time at tide gauges are 15-78 s, compared to the 1 min data

¹¹¹⁹





1138 Figure 2: DRAFT! (a,b,c) Pre- and post-collapse bathymetry/topography of AK and surrounding islands 1139 (Fig 1): (a) 1990 pre-collapse bathymetry with locations of 2017 seismic reflection profiles; (b) post-1140 collapse bathymetry from August 2019 showing landslide deposits and traces of collected seismic 1141 reflection profiles survey (Hunt et al., 2020); and c) rendering of (b) bathymetry showing large, blocky 1142 landslide deposits at the base of AK's SW flank. (d) AK historical profiles in SW transects, compared to 1143 August 2019 field survey (Hunt et al., 2020) with proposed failure surface with uncertainty region. (e) 1144 pre-collapse AK profile (SW direction, 225 deg. To N) used in simulations (no vertical exaggeration), 1145 with traces (dashed lines) of 4 failure surface scenarios modeled with NHWAVE, of total collapse 1146 volume: (red) 0.313; (blue) 0.272; (black) 0.224 (deemed the likeliest scenario; see Figs. 1c and 3b); and 1147 (green) 0.175 km³.



1155 outside of Krakatau islands and August 2019 field survey data (see Fig. 2) in the caldera and surrounding

islands (Hunt et al., 2020).











with NHWAVE in Grid G2 (Fig. 1), for 8 AK collapse scenarios with a granular (solid) or viscous (dashed) rheology, and volume (Fig. 2c): (red) 0.313; (blue) 0.272; (black) 0.224 (likeliest scenario; see Figs. 1c and 3b); (green) 0.175 km³. Time t = 0 is estimated collapse time, 20:57' local time (UTC + 7). Note, reference level in <u>simulations</u> is MWL = MSL + 1.5 m (tide elevation).



collapse scenario (granular, 0.224 km^3), at t = (a) 10, (b) 20, (c) 40, (d) 80, (e) 120, (f)160, (g) 200, and (h) 380 s (latter time is FUNWAVE initialization). Same case as Fig. 5. Reference level in simulations is MSL + 1.5 m.



simulations is MWL = MSL + 1.5 m.





1230Figure 10: Time series of surface elevations, in simulations with respect to MWL = MSL + 1.5 m, at1231numerical wave gauges 6-9 (a-d; Fig. 1a), computed with FUNWAVE for 8 AK collapse scenarios with1232line codes defined as in Fig. 6, compared to collocated detided observations (o) with 2 sensors, at 4 tide1233gauges (Table 2). Time t = 0 is estimated collapse time, 20:57' local time (UTC + 7).



Commented [MEMN18]: what if the wave-impact line is lower as shown with red dashed line (on top of the steepest slope; not on the tree line) so it might match the simulation?



1248Figure 11: Pictures from AK field surveys of near-field runups: (a,c,e) the authors August 2019 drone1249survey; (b,d) Borrero's et al. (2020) in 02/2019, with (a/b) Rakata's N/NW shores; (c/d) Sertung's SE/NE

Commented [sw19]: Need a scale on the drone survey images, let me know if you need me to redo any of these.



Figure 13: Maximum runup computed with NHWAVE along (Figs. 1c, 11, 12): (a) Rakata's N shore; (b)
Sertung's S shore; and (c) Panjang's W shore, for 8 AK collapse scenarios with line codes defined as in
Fig. 6, compared to (Fig. 11) the authors August 2019 drone field survey (pink line/circles) of tree line
and field measurements (yellow squares) of Borrero et al. (2020); note, the latter authors reported an 8 m
flow depth for north of Panjang. Black solid lines denote our preferred volume scenario with granular
rheology. Note, in simulations and the field data, zero elevation is MWL = MSL + 1.5 m (tide elevation).





1291 measurements of flow depth and runup, from: (**D**) TDMRC (2019), (**★**) Muhari et al. (2019), (**♦**) Putra et

1292 al. (2020), (♥) Heidarzadeh et al. (2020), and (●) Borrero et al. (2020) surveys.







1307 in (a,c,e) are bathymetry/topography in meter and color scale is maximum surface elevation in meter.









1316 Figure 16: Same results as in Fig. 14: (a) Envelope of maximum surface elevation (FUNWAVE, Grid 1317 G1; color scales in meter), for likeliest AK collapse scenario (granular, 0.224 km3; Fig. 9d), zoomed on 1318 Sumatra (Fig. 1); (b) comparison of computed flow depth at the shore with field surveys of: (■) TDMRC 1319 (2019), (*) Muhari et al. (2019); (c) View of Waymuli (W, Fig. 1a, 105.6348 E), looking east, from Fritz 1320 et al. (2019) 02/2019 field survey.

1321	Supplementary file for:
1322	Modeling of the Dec. 22 nd 2018 Anak Krakatau volcano lateral collapse and
1323	tsunami based on recent field surveys: comparison with observed tsunami
1324	impact
1325	
1326	Grilli ¹ S.T., Zhang ² C., Kirby ² J.T., Grilli ¹ A.R., Tappin ^{3,4} D., Watt ⁵ S.F.L., Hunt ⁶ , J.E.,
1327	Novellino ³ , A., Engwell ³ S., Nurshal M.E.M. ⁷ , Abdurrachman M. ⁷ , Cassidy, M. ⁸ , Madden-
1328	Nadeau A.L. ⁸ and S. Day ⁴
1329 1330	1. Department of Ocean Engineering, University of Rhode Island, USA
1331	2. Center for Applied Coastal Research, University of Delaware, USA
1332	3. British Geological Survey, Nottingham, UK
1333	4. University College, London, UK
1334	5. School of Geography, Earth and Environmental Sciences, University of Birmingham, UK
1335	6. National Oceanography Centre, Southampton, UK
1336	7. Bandung Institute of Technology, Indonesia
1337	8. Department of Earth Sciences, University of Oxford, UK
1338	



Commented [sw20]: We could perhaps show the key satellite (CSK SAR image) and the coastline images again here, as they are important for defining the landslide margine (the same ones shown in Hunt et al.).

Commented [AN21R20]: I agree, the COSMO-SkyMed data is critical considering what we are stating in Section 3.2

1355	S2: Non-hydrostatic versus hydrostatic NHWAVE results for near-field runups
1356	To illustrate the effect of including the slide non-hydrostatic pressure on the flank collapse
1357	motion and generated waves, two more simulations were carried out using both viscous
1358	and granular rheology with only the hydrostatic pressure effects included within the slide
1359	layer (as in the NHWAVE version used by Grilli et al., 2019), for the collapse scenario
1360	with a 0.272 km^3 volume (Fig. 2e) identical to Grilli et al.'s (2019) preferred scenario.
1361	Results were compared with those of the same simulation performed, as in the main text,
1362	with the non-hydrostatic pressure effects included within the slide layer.
1363	Figures S2 to S5 show instantaneous results of these simulations at different times,

in vertical profiles along a 213 deg. Transect from N (Figs. S2 and S3) and in plan view
(Figs. S4 and S5). In the first 20 s, for each rheology, the slide motion of the nonhydrostatic (NH) and hydrostatic (H) simulations are similar, except for a faster dynamics
of the failure of the volcano summit in the H simulations.






Figure S3: Same case as in Fig. S2, but for the viscous rheology.

1378Later in time, differences between the NH and H simulations increase as the slide1379tails still move at the same speed while the motion of the slide fronts are quite different,1380leading to a longer extent of the landslide runout and broader deposit spreading in the H1381simulation for both rheologies.





Figure S4: Results of same granular case as in Fig. S2, in plan view for (a,c,e) NH and (b,d,f) H simulations at t = (a,b) 20, (c,d) 40, and (e,f) 80 s. Solid blue line denotes slide deposit limit and color scale is slide thickness in meter.



Figure S4: Results of same viscous case as in Fig. S3, in plan view for (a,c,e) NH and (b,d,f) H simulations at t = (a,b) 20, (c,d) 40, and (e,f) 80 s. Solid blue line denotes slide deposit limit and color scale is slide thickness in meter.



Figure S6: Results of the same granular (blue)/viscous (red) cases as in Figs. S2,S4/S3,S5. Comparison of time series of surface elevations for NH (solid) and H (dashed) simulations at numerical wave gauges (Fig. 1b), 1, 2, 5, 4, 3, from top to bottom.

1409For the same granular/viscous cases as shown in Figs. S2,S4/S3,S5, Figure S61410compares the time series of surface elevations computed in the NH and H simulations at1411numerical Wave Gauges 1 to 5 (Fig. 1b, Table 2), showing the large influence on wave1412generation of simulating non-hydrostatic pressures in the slide layer. Overall, while as1413discussed in the main text effects of rheology are only moderate on wave generation, for1414both rheologies, significantly larger leading waves are generated in the H simulations than





in the more physically realistic NH simulations; and, in part due to amplitude dispersion
effects, the waves generated in the H simulations propagate faster and arrive ahead of the
waves generated in the NH simulations at all wave gauges. The same observations can be

1432made for the first and next troughs, which are much deeper in the waves generated in the1433H simulations than in the NH simulations. For waves propagating to the southeast (WG 4)1434and northeast (WG 5), in the H simulations, there is a large leading elevation wave that is1435not present in the NH simulations. This appears to result from the larger slide runout1436occurring on the backside of AK in the H simulations (Figs. S4 and S5).

1437 Figure S7 compares, for the same cases, the runups generated on the three 1438 surrounding islands of Rakata, Sertung and Panjang in the NH and H simulations, for both 1439 rheologies; as in Fig. 13, results are also compared to the August 2019 drone tree line 1440 survey and to runups measured by Borrero et al. (2020). The runups generated on Rakata 1441 island in the H simulations are nearly twice as large at most locations than those in the NH 1442 simulations, and show large discrepancies with the field data. The runups generated on 1443 Panjang island are also amplified in the H simulations, which is consistent with the higher 1444 waves computed at Wave Gauge 4 and 5; the H runups also significantly overestimate 1445 observations made on the north side of the island, where it was reported an 8-10 m flow 1446 depth/runup (Borrero et al., 2020). In contrast, the difference of the H and NH runups 1447 generated on Sertung island is less, except on the east side of the island where the H runups 1448 are much larger than the NH runups and again show a large discrepancy with the single 1449 runup measured in the field survey.

1450Based on the above, one can conclude that including non-hydrostatic effects in1451simulations of the slide layer motion in NHWAVE is important to accurately simulate the1452near-field waves and runups, as well as for simulating slide deposits and runout.