1 <u>Reply to Referee 1, Pedro Costa (NHESS)</u>

2 Major comments:

3 Interactive comment on "Reconstruction of flow conditions from 2004 Indian Ocean tsunami

- 4 deposits at the Phra Thong island using a deep neural network inverse model" by Rimali Mitra
- 5 et al.,

6 We thank the reviewer for the critical assessment of our manuscript and for the numerous comments

7 and suggestions. We have provided answers to your questions as listed below (in bold italics).

Q1: The Science is there but English must be revised extensively and above all there must be a clear 8 9 clarification on how and what exactly field data was used to validate the model. I believe you used Fujino et al. (2010) data but sometimes when reading the manuscript, one feels puzzled to confirm 10 that you did use it and which values have you used it. For instance, thickness, grain-size curve, etc. 11 Therefore, I cannot agree with the title proposed because it was not well-establish that the regressive 12 model used geological data. Sometimes the reader feels, the models fed and validated each other, 13 and no solid, extensive and accurate field data was used. I assumed that this might just be a language 14 and writing problem. Even if it is that, you need to address it. Sometimes the text is confusing and 15 one wonders what you trying to transmit. For example, when you state the model was validated by 16 "observed" flow depths in several locations along the studied profile...in fact, you are saying that the 17 model agrees well with previous modelling exercises for flow depth establishment. The meaning of 18

both sentences in totally different regarding field validation and this is crucial for this manuscript.

20

21 RE: The authors would like to thank Dr. Costa for his comments. We made a substantial effort

22 on the clarifications and the overall organization of the paper. We have done the English

language and grammar checking on our manuscript by an English proof-reading service agency
 for journals.

25 Yes, we have used the data set from Fujino et al. (2010) and the data set is given as 26 "Thai_gs5_revised_1.csv" in https://doi.org/10.5281/zenodo.4511317. However, we have revised

the representative diameters for grain-size classes in our revised manuscript and the detailed calculation is provided in the above-mentioned repository.

- We added an additional diagram as Figure 2 and 3 on page no. 5 and 6, on the distribution of the
- thickness of the samples with the distance along with the mean grain size and segregation of the
- 31 grain size classes from the distribution. Hence, we decided to keep the same title as it uses the real
- 32 field data set in the inverse model and the model uses mean squared error for the regression 33 algorithm.
- Here, "observed" flow depth implies the measured flow depth. We have unified the terminologies for all measured values to avoid further confusions.
- 36
- Q2: There are many other aspects I raised on the annotated version and I suggest you analyse them
 critically. I might have misunderstood some wording (which means that you need to make it clear)
 or I might have perceived things correctly (which means you need to change the structure and scope
 of the manuscript). One example, is sediment concentration. How can you validate flow sediment
- 40 of the manuscript). One example, is sediment concentration. How can you variate flow sediment 41 concentration from the deposit? Only if you look at grain-size curve, spatial distribution and packing
- 41 concentration from the deposit? Only if you look at grain-size curve, spatial distribution and packing 42 (inner architectural arrangement) of the deposit. You never mention this along the manuscript which
- 42 (Inner architectural arrangement) of the deposit. Four never mention this along the manuscript which 43 means that Low puzzled how you reconstruct addiment concentration on the inverse model. It is easy
- 43 means that I am puzzled how you reconstruct sediment concentration on the inverse model. It is easy

to understand how you do it with the forward model but departing from sediments (withoutmentioning the characteristics above) is baffling.

46

47 RE: Thank you for your feedback. Regarding sediment concentration, we did not intend to 48 validate the sediment concentration as it is almost impossible to evaluate the reconstructed values 49 of sediment concentration because there are no available observational data. We only 50 reconstructed the values of sediment concentration using DNN inverse model. However, Goto et 51 al. (2014) used the entire thickness and measured inundation depth to estimate the sediment 52 concentration which was around 2% in the inundation flow of the 2011 Tohoku-oki tsunami. 53 Hence, we added the reference of sediment concentration on Page 10., line 205.

54 We have provided the explanation of all comments in reply of annotated document.

55 56

Minor comments (annotations):

57

Interactive comment on "Reconstruction of flow conditions from 2004 Indian Ocean tsunami
deposits at the Phra Thong island using a deep neural network inverse model" by Rimali Mitra
et al.,

61

We thank the reviewer for the critical assessment of our manuscript and for the numerous commentsand suggestions. Please find our responses to each comment below (in bold italics).

64

65 1. The title does not clearly correspond to the content. Either the authors change the manuscript
accordingly and provide clear field validation information or they must remove "tsunami deposits"
from the title.

68

69 RE: Thank you for your comment. In accordance with your comment, we have added figure 3 to show the detailed information of the analyzed sediments. We have used the field data from Phra 70 Thong island, Thailand to reconstruct the flow conditions of 2004 Indian Ocean tsunami. Yes, we 71 have used the data set from Fujino et al. (2010) and the data set is given as 72 "Thai gs5 revised 1.csv" in https://doi.org/10.5281/zenodo.4511317. However, we have revised 73 the representative diameters for grain-size classes in our revised manuscript and the detailed 74 calculation is provided in the above-mentioned repository. The codes include the same data set to 75 predict the results. So, we decided to keep the title unchanged. 76

77

78 2. Are you referring to grain-size? When I look at Figure 9 that is what you are presenting. One thing
79 in sediment concentration and grain-size distribution on the incoming tsunami waves, another
80 totally different is grain-size and packing on the tsunami deposits. Throughout the text is not clear
81 what you are determining, because it was not possible how you estimate sediment concentration on
82 the wave from the deposit if your are solely relying on the inverse model. So, this needs to be
83 clarified and the fact that you did not incorporated any reference to "sediment concentration " on
84 the (minimal) conclusions provide further increases my confusion.

85

86 **RE:** Thank you for the comment.

87 P.5, 6, we added two new diagrams as Figure 2 and 3 to show thickness and grain size distribution

88 from the study area. Here, we refer to the sediment concentration but in Figure 9 we presented

- volume per unit area and spatial grain-size distribution. We reconstructed sediment concentration
 of the tsunami deposits.
- 91 Our inversion model estimates values of sediment concentration of tsunami that best explain
- 92 distribution of thickness and grain size of tsunami deposits. The inverse model is trained from the

93 results of the forward model calculation.

- 94 In response to your comment have added the reference Goto et al. (2014) reference and 95 clarification in our revised manuscript as follows,
- 96 P. 10, line 202, The range of parameters adopted in this study is applicable to most of the large-
- 97 scale tsunami-inundated areas as the ranges have been selected with several case studies of
- 98 tsunamis that includes mostly field measurements, survivor video and numerical analysis (Mori et
- 99 al., 2011; Wijetunge, 2006; Szczuci nski et al., 2012; Matsutomi and Okamoto, 2010; Abe et al.,
- 100 2012; Fritz et al., 2006; Nandasena et al., 2012; Goto et al., 2014).
- 101
- 3. Please rewrite. How did you determined the post-tsunami concentration? Was this data based onpost-tsunami survey? Please make the text more fluent.
- 104
- 105 Thank you for the suggestion. We determined the post-tsunami concentration using the DNN 106 inverse model, which automatically finds the distribution of sediment concentration that best 107 realizes the actual observed distribution of thickness and grain size of the tsunami deposit. We will
- 108 modify the text in our revised manuscript.
- 109 Revised text: P.1, line 5, The DNN inverse analysis reconstructed the values of flow conditions
- 110 such as maximum inundation distance, flow velocity and maximum flow depth, sediment
- 111 concentration of five grain-size classes using the thickness and grain-size distribution of the
- 112 tsunami deposit from the post-tsunami survey around Phra Thong island.
- 113
- 114 4. Abstract, Agree but the greater challenge is to study older deposits and reconstruct physicalparameters from them.
- 116
- 117 *RE: We agree with the reviewer.*
- 118
- 119 5. P-1, Line 15, This sentence is somewhat confusing. Please rewrite.
- 120
- 121 **RE:** Thank you for the suggestion. We modified the text in the revised manuscript as follows:
- 122 P.1, line 18, The total damage was estimated to amount to around USD 508 million, which equates
- 123 to 2.2% of GDP while the number of deaths was 4225, with the injured and missing cases.
- 124
- 125 6. P-2, Line 25, Please add citation.
- 126
- 127 RE: Thank you for identifying this. We have added the citation Suppasri et al., 2015.
- 128
- 129 7. Line 35, Please revise English. The wroding and figures are not well-structured.
- 130
- 131 *RE: Thank you for the suggestion. We have revised the text.*
- 132

133 8. Please see Costa et al., 2011, Earth Surface Processes and Landforms, Costa et al., 2012, The 134 Holocene and Moreira et al., 2017, Marine Geology

135

136 *RE: Thank you for the suggestion. We considered adding costa et al., 2011 and Moreira et al.,*137 2017.

138

P-3, Line 73, The big problem with this manuscript is that is not clear how the inverse model was
validated. Figure 9 shows something but much further detail needs to be provided much earlier in
the text for the reader to understand what the authors achieved or are aiming.

142

143 *RE: Thank you for the comment. We have provided a paragraph on the workflow of the inverse* 144 *model for detailed clarifications.*

145 P.3, Line 86, Here, we conduct an DNN inverse analysis of the tsunami deposits measured at Phra

146 Thong island and reconstruct theflow conditions such as the maximum inundation length, flow

147 velocity, maximum flow depth and sediment concentrations offive grain-size classes. The inverse

148 model was based on the forward model, which was proposed by Naruse and Abe (2017). The

149 forward model calculations were iterated at random initial flow conditions to produce artificial

150 training data sets that represent depositional characteristics such as the spatial distribution of 151 this has a set of the spatial distribution of

151 thickness and grain-size composition. Using the artificial903

152 training data sets, the DNN was then trained to establish a relation between the depositional 153 characteristics and the and the flowconditions. The post-trained DNN model was ready to predict

flow conditions from the tsunami deposits after the performance of the trained DNN was verified

155 using test data sets. The 1-D cubic interpolation was applied to the field data sets of PhraThong

156 island to fit the data set to model grids. Finally, this DNN inverse model was applied to the field

157 data sets from thePhra Thong island, Thailand to reconstruct the flow conditions of 2004 Indian

158 Ocean tsunami. Our inverse model was already95validated to be effective for 2011 Tohoku-oki

159 tsunami deposits distributed in Sendai Plain (Mitra et al., 2020). In case of PhraThong island, we

160 validated the results by the field measurements of the tsunami flow depth. Also, the estimated 161 thickness and grain size distribution of tsunami deposits were compared with the actual

161 measurements. Our inverse analysis results couldbe used for designing future tsunami hazard

163 assessments and disaster mitigation strategies in Thailand.

164

165 10. P-3, Line 83, If you mention this, then one would expect to see grain-size variations and deposit 166 thickness data.

167

168 *RE: Thank you for the suggestion. We have added Figure 2 and 3 which represents grain-size* 169 *distribution and thickness data.*

170

17111. Line 105, This is a big simplification...

172

173 *RE: Thank you for the comment. This simplification was done by Naruse and Abe, 2017 and this*

174 was further used by Mitra et al., 2020. For the details of the simplifications and step by step

175 procedures, please refer to Naruse and Abe, 2017.

176

178 you depart from field data and use it to reconstruct the physical parameters of the tsunami. This is not what you are explaining here. You use the forward model to produce tsunami hydraulic features 179 and then based on the neural network you check where the data produced is nore robust and assume 180 181 the values produced. I was expecting a regression from deposits to flow characteristics. It does not 182 seem to be the case here. 183 184 RE: Thank you for the comment. Yes, we are using the field data to reconstruct the physical parameters using the DNN inverse model. The forward model is used to generate artificial spatial 185 grain size distribution and volume per unit area data of tsunami deposits which was used to train 186 inverse model. We checked the robustness and precision of the inverse model using the artificial 187 188 test data set, and then used the produced inverse model to reconstruct the flow conditions of the tsunami from the actual deposit. 189 190 191 13. P-5, Line 119, I understand it but you need to clearly show where is the data coming from. can you please add a map and grain-size data. 192 193 194 RE: We assume that this comment is related to comment number 10. Please refer to the reply in 195 comment number 10. 196 197 14. P-5, Line 134, Why 0 to 2% concentration? I do not understand where this value comes from? Is it 198 random? From observations? Where? All the data I have access to suggest (much of it publiched by several authors) much higher sediment concentrations. 199 200 201 RE: Thank you for the comment. This range was considered from Goto et al., (2018) sediment concentration analysis which indicates that the total sediment concentration for tsunami deposits 202 is usually around 2%, and therefore the concentration for each grain size class seems unlikely to 203 exceed 2%. In response to your comment have modified our text by adding this reference. Please 204 refer to the reply of comment 2 where we have written the modified text. 205 206 207 15. P-5, Line 145, This needs to be explained earlier and more clearly. 208 209 **RE:** Thank you for the suggestion. We have revised the text in the revised manuscript. P.11, line 215, To apply the inverse model to the measured values of field data set from Phra Thong 210 island in 1-D vectors, the collected datapoints must be fit into that fixed coordinate system of the 211 model. 212 213 214 16. P-6, Line 157, I understand the seminal character of the work by Mitra et al (2020) and the need to many citations along the manuscript but is some cases I believe this could be avoided. 215 216 **RE**: Thank you for the suggestion. We have removed the sentence from the revised manuscript. 217 218 219 17. P-7, Line 161, What were their charateristics? 5

17712. P-5, Line 115, What one expects from an inverse model (like you are suggesting in the title) is that

220	
221	RE: The characteristics of artificial data sets were depositional characteristics such as volume per
222	unit area and grain-size distribution.
223	
224	In response to your comment we have revised the text in the revised manuscript.
225	P. 11, line 231, The DNN was trained using artificial data sets which were the depositional
226	characteristics such as volume per unit area and grain-size distribution.
227	
228	18. P-7, Line 165, And between 2000 and 3000 what is the gain?
229	
230	RE: The loss function was optimized already to its lowest value, the calculations continue up to
231	3000 epochs, the progress will remain the same. There will be no change as it already converged
232	before 2000 epochs.
233	
234	19. P-7, Line 169, this is obvious. No need to add it.
235	
236	RE: Thank you for the suggestion. We have removed the sentence from the revised manuscript.
237	
238	20. P-8, Line 188, Please add a aerial image with the sampling point clearly marked.
239	
240	RE: We agree with the reviewer. We have added a google earth image with the marked sampling
241	points (Figure 2).
242	
243	21. P-8, Line 191, add reference. Heights were reported or modeled?
244	
245	RE: Thank you for the suggestion. We have added the references Jankaew et al., 2008, Fujino et
246	al., 2010. Heights were reported from the above-mentioned literatures.
247	
248	22. P-8, Line 192, the nearshore area?
249	
250	RE: Yes. The area was nearshore.
251	
252	23. P-8, Line 193, thin and finer?
253	
254	RE: The deposits became thinner and finer in the landward direction. We have revised the text by
255	adding both phenomena in P. 4, line 122.
256	
257	24. P-8. Line 196, along the analyzed profile, right? Otherwise it contradicts what is stated above
258	regarding maximum thickness of the deposit.
259	
260	RE: Yes, all the thickness mentioned here is along the analyzed profile.
261	
262	25. P-8, P 198, This clearly needs to be provided here as well. This is pivotal for the validation and
263	must be displayed in this manuscript as well.
264	

RE: We assume that this comment is related to comment number 10. Please refer to the reply in comment number 10.

267

26826. P-10, Line 200, Not sure if I understood this last sentence correctly. Please provide frutherinformation. So, you stopped using field data?

270

271 *RE: Thank you for the comment. We have done subsampling test to check the effect of irregularly*

spaced field data sets on the accuracy of the inversion. The details on the subsampling procedure

273 is given in Mitra et al. (2020). In response to your comment, we have modified the text as follows,

P. 15, line 262, This test was done to check the effect of irregularly spaced field data sets on the
accuracy of the inversion. The details on the subsampling procedure is given in Mitra et al. (2020)

27727. P-14, Line 222, Please clarify how you obtained these values.

278

279 *RE: Thank you for the comment. These values are obtained from table 1. We have added the*280 *reference of table 1 in P. 15, line 275.*

281

282 28. P-14, Line 227, Discussion needs to be rewritten.

283

284 RE: Thank you for the comment. We have modified the overall discussion accordingly.

286 29. P-14, Line 232, So model was not validated against field data?

287

285

RE: The model reconstructed the tsunami flow characteristics from the actual tsunami deposit, 288 and the predicted results were validated with the field measurements of flow depths. This model 289 was trained with artificial data sets of tsunami deposits and validated with the field data. The 290 present model is estimating fair results using inexpensive artificial data for training of the neural 291 network and avoiding the difficulties to gather large amounts of data sets of tsunami deposits with 292 in-situ measurements of flow velocity and depth. For tsunami deposit data set in terms of grain 293 294 size distribution perpendicular to coastline is not easily obtainable and sometimes good data sets are difficult to obtain due to obstructions in field areas, and measurements of flow hydraulic 295 parameters such as velocity were quite rare. Therefore, training the model with the real 296 measurements may not be most viable option in terms of expense and performance. Indeed, all of 297 298 previous studies on inverse analysis were not developed from the relationship between measurements of deposits and flow parameters because it is practically impossible. Instead, their 299 models were depended on the simplified hydraulic modeling of tsunamis. Even if we can train 300 model with smaller number of training data obtained by measurements in the field, the model tends 301 to overfit which results into poor performance of the model on the observed value. 302 303 To this end, we employed very different approach, which uses the calculation of results in the

304 forward model as the training data sets.

305

306

30730. P-14, Line 239, when you mention true values are you referring to field data?

309 *RE:* We agree that this true value might create confusion. We referred the true values as the test
310 data set from artificial data that determines the performance of the model.

P. 20, line 295, Regarding the deviation of the predicted values from the true values which are
artificial test data set, the sample standard deviation values were relatively small for all parameters.

31431. P-16, Figure 9, For me this is the key findings of the manuscript and are somewhat lost in the
structure. This is a key figure that should be move further up. It will also important to acess
raw/original grain-size curves retrieved from the field by Fujino et al. (2010)

317

318 *RE: Thank you for the comment. We considered shifting this diagram in the upper part of the* 319 *section.*

320

321 32. P-17, Line 257, This sub-section is crucial and is poorly explained above and consequently also
here. The authors should make an effort to clarify the field validation. I read the manuscript 3 times
and struggle to fully understand what field data you used. And if it is an regressive exercise why
you do not use thickness and grain-size curve or D50. This needs to be clear for the reader from the
start of the manuscript.

326

RE: Thank you for the comment. We understand that there must be lack of clarification in the description of validation regarding the application of the field data. We have modified the text as follows:

P. 21, line 335, The maximum and measured flow heights from Phra Thong island were reported 331 7.1 m and 5.5 m respectively (\url{http://www.nda.ac.jp/~fujima/TMD/fujicom.html}). The

corresponding maximum and minimum values of elevation are 3.1 and 1.1 m respectively (Jankaew et al., 2008, 2011; Brill et al., 2012b). Hence, the approximate estimate of measured

(Jankaew et al., 2008, 2011; Brill et al., 2012b). Hence, the approximate estimate of measured
 maximum flow depth is ranged from 2.4 m to 6.0 m. Considering the bias correction of 0.43 m, the

reconstructed value of maximum flow depth (5.3 m) falls within the range of measured maximum

- 336 *flow depth values.*
- 337

338 33. P-17, Line 266, measured or modeled?

339 340

RE: The flow heights were measured using water mark in the field.

341

342 34. P-17, Line 269, observed or modeled? Did someone measure flow height at these specific points or
these values are the result of forward modeling exercises? I suspect it is the latter. So change the
text, please to be accurate.

345

346 RE: Yes, the mentioned groups measured the flow heights at specific locations. The details are

347 available in the link, http://www.nda.ac.jp/fujima/TMD/fujicom.html.

348 In response to your comment we have unified the terminologies of measured flow heights.

349

350 35. P-17, Line 274, But you used Fujino et al., 2010, right?

351

352 **RE:** We applied the data set of volume per unit area and grain size distribution in our model and

353 the model results were validated with several reported or measured values by different researchers.

Here, the distance 400 m indicates the locations of the measured value by Jankaew et al. (2011,
2008) from our study area.

356

357 36. P-18, Line 296, But there is a wide range of values presented even in this manuscript introductionand study area?!?!

359

RE: The DNN inverse model estimates only a single value. We have added the uncertainty or error estimations. We used a range of artificial values to generate artificial data set of depositional characteristics in the forward model. Using that artificial data sets the inverse model was trained and the final predicted result was a single value for one sampling window size. For example, we used 1700 sampling window size to reconstruct the single values for each parameter of flow conditions.

366

36737. P-18, Line, 307, Please explain this idea better and in greater detail.

368

369 **RE:** Thank you for your comment. We have modified the text as follows:

370

P. 22, line 370, In addition to the source model, this model also includes tsunami sediment transport calculation that consists of bed load layer and suspended load layer. However, the calculated value of the sediment thickness was overestimated as the assumption of movable bed for

a large area caused excessive erosion of the ground (Masaya et al., 2019).

375 Line 379, The model calculation of Masaya et al. (2019) relies on the estimation of a385single set

of fault parameters, which were not widely explored to obtain the optimal parameters. In future,

377 Model TUNAMI-N2can be potentially used as the forward model in DNN inverse model to consider

378 two-dimensional behavior of tsunamis. To do so, the model needs to be modified for considering

379 sediment transport of multiple grain size classes.

380

381 38. P-18, Line 310, Where have you provided this information? Where is the detailed variation oftsunami deposit thickness variation in the field and its comparison with the model results?

383

RE: Figure 9 shows the tsunami deposit volume per unit area and its comparison with the model results.

386

387 39. P-18, Line 315, Please rewrite the Conclusions.

388

389 *RE: Thank you for the suggestion. We have revised the conclusion as follows:*

390 *P. 23, line 386, The value of maximum flow depth including the additional bias correction was 5.3*

391 *m* that was within the range 2.4 *m* to 6.0 *m* which was the approximate estimate of measured

392 maximum flow depth at Phra Thong island. The value of flow velocity was also close to the reported

393 values using the video footage from the vicinity of the Phra Thong island. The uncertainty of the

394 results using jackknife method also indicated that simulated results did not contain a large range

- 395 of values. Phra thong island was one of the most well preserved and historically important area
- 396 for paleotsunami deposits. Hence, the application of the DNN inverse model was suitable to

397 reconstruct flow conditions of 2004 Indian Ocean tsunami from Phra thong island. The DNN 398 inverse model also represented the comparison of the calculated and measured spatial distribution 399 of volume per unit area along the transect at the island. This model can be applied to any areas of 400 modern and ancient tsunami deposits consisting of low land or flat areas to successfully 401 reconstruct the tsunami flow conditions and can serve as a tool for tsunami hazard assessment and 402 disaster resilience at coastal cities.

403

404 Reply to Anonymous Referee 2, (NHESS)

405

Interactive comment on "Reconstruction of flow conditions from 2004 Indian Ocean tsunami deposits at the Phra Thong island using a deep neural network inverse model" by Rimali Mitra et al.,

409

410 We thank the reviewer for the insightful assessment of our manuscript and for the numerous 411 comments and suggestions. We have provided answers to your questions as listed below (in bold 412 italics).

Q1: The figures presented as Figure 1 in this paper have already appeared in M2020 (their Figs. 1 and 2). This is at least acknowledged in the caption. Figure 2 also appears identical to Fig. 3 of M2020 (apart from some color changes), but this does not seem to be indicated in the Figure caption. Is this indeed essentially the identical figure? If so, this ought to be acknowledged. (It could just be very similar, and the differences not perceptable).

419 RE: Thank you for identifying this. We agree that Figure 1 in this manuscript looks similar to the appeared figure in Mitra et al., (2020). However, we have made slight changes in the Figure 1 by 420 changing the number of grain size classes from six o five, hence the number of output nodes 421 should be 8, here we did the typo by mentioning the number of nodes as 9. We have corrected the 422 typo and added the reference of Mitra et al., (2020). In the revised manuscript, although the 423 figure 4 looks similar and imperceptible, we changed the number of training datasets and the 424 performance of loss function is slightly different from the previous paper as the grain size 425 distribution is different for 2004 Indian ocean tsunami at Phra Thong island. Thus, we generated 426 separate artificial dataset and that Figure 4 shows that performance. 427

428

429 Q2: Several details of the numerical model used to generate test data sets in Section 2.1 are 430 seemingly missing. These include, the following issues: How are the friction velocity (u_*) , the 431 setting velocity (w_s,i) , the sediment entrainment coefficient (E_si), and other variables (r_0i and 432 F_i) determined?

433

RE: Thank you for your suggestion. We did not add the details of the parameters mentioned as
we thought it would be repetitive as we the details are already in Naruse and Abe, 2017 and Mitra
et al. 2020. In response to your comment we have added the reference in our revised manuscript.

438 *P.7, line 154, The details of the parameters and variables are provided in Naruse and Abe (2017).*439

Q3: Regarding the friction velocity (u_*) , in particular, it should be noted that great care ought to 440 441 be taken for this quantity if standard (based on steady flow) friction formulas are used, as several recent research papers have shown that tsunami induced bounday layers may span only a fraction of 442 443 the water depth, and hence these may well be invalid. See e.g. Lacy et al. (DOI: 10.1029/2012JC007954), Williams & Fuhrman (DOI: 10.1016/j.coastaleng.2015.12.002), Tinh & 444 Tanaka (DOI: 10.1080/21664250.2019.1672127) or Larsen & Fuhrman (DOI: 445 10.1016/j.coastaleng.2019.04.011). Please clarify this point, and if this is indeed being done, this 446 447 potential defficiency ought to at least be acknowledged.

448

RE: We understand the point the reviewer makes. We have used standard friction formula for our model. In response to your comment, we have added the suggested references and revised the text as follows:

452

P. 7, line 145, Here, we employed the flow resistance law to obtain friction velocity using the
friction coefficient, which is widely used in general. A few researchers recently reported that
tsunami induced boundary layers may span only a fraction of water length formula (Williams and
Fuhrman, 2016; Lacy et al., 2012; Larsen and Fuhrman, 2019). The importance of the resistance
law for the inverse analysis, considering such non-steady conditions, maybe a subject for future
study.

Q3: Can a definition sketch of the model domain (etc.) being used for the generation of the training
data sets please be provided? This will help readers immensely to get an idea as to the actual setup
being used. Plots just showing performance (like Figs. 2 and 3) fail to provide this.

464 *RE: Thank you for the comment. We have added a diagram (figure 5) on the explanation of model*465 *configuration in the revised manuscript.*

466

463

467 Q4: I do not find that the DNN architecture being used is presented with sufficient clarity. In Section 2.2 (top) it is stated that the DNN model accepts grain-size and thickness distribution at an 468 input layer, and that the outputs are the "tsunami characteristics through several hidden layers". This 469 is rather unclear. Further clarification is also provided in Fig. 1, though it is not clear if this is the 470 actual architecture or just intended as an example. Please (just in a sentence or two) summarize the 471 DNN architecture i.e. clarify precisely the no. of inputs, the number of hidden layers (and nodes in 472 each layer), and the number of outputs to remove any ambiguity. Such details are rather important 473 should one attempt to reproduce this work. 474

475

476 *RE: Thank you for the valuable suggestion. We agree with the reviewer that we should add the*477 *details of input and output layers. We have added the details of the DNN architecture in the revised*478 *manuscript. The details are as follows:*

479

480 P. 8, line 179, The DNN structure includes the input layer which consists of input nodes where 481 the input values are the volume per unit area of each grain-size class at the spatial grids. Thus, 482 expression of the input nodes numbers is presented as $M \times N$ where M and N are the total number 483 of spatial grids and grain-size classes, respectively. In this inverse model, the total numbers of 484 layers were five among which, the number of hidden layers were three with the 2500 nodes (Mitra et al., 2020). Finally, the output layer consists of the predicted parameters of flow conditions. The
details of hyperparameters selection is provided in Mitra et al. (2020).

487

489

488 **<u>Reply to Anonymous Referee 3, (NHESS)</u>**

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493

505

We thank the reviewer for the insightful assessment of our manuscript and for the numerous comments and suggestions. We have provided answers to your questions as listed below (in bold italics).

497 Major comments:

Q1: One of the major concerns is the technical problems on writing. Sentences in the body text are often complicated and difficult to understand the author's intent. A complete English proofreading by professional services or native speakers is needed. Section 3.2.1, which explains the sedimentary data from the Phra Thong island, must be placed before the description of the inversion results. In addition, earlier papers sometimes were inappropriately cited, and the order of some figures (and insets) are not consistent with the structure of the paper. Therefore, comprehensive reorganization and correction are required to improve the readability of the paper.

RE: Thank you for the suggestion. This is to inform the reviewer that we have done the English proof reading by professional editing service for journals. We have moved the previous section 3.2. 1 to section 2. We have rechecked and revised the references and overall organization of the manuscript in order to improve the readability of the paper.

510 511 Q2: Another concern is that whether the model assumption is valid for the study area. For example, both the transect of the tsunami deposit sites and the reference line (Figure 6) is oblique to the 512 coastline, meanwhile the model assumes that the coordinate x for the forward simulation is 513 perpendicular to the shoreline (equation 2). I'm curious that how likely the direction of tsunami 514 inundation was consistent with these lines. Satellite imageries show that the geometry and directions 515 of the sandy ridges are quite complex, implying the tsunami inundation might have been affected 516 the local topography. Fujino et al. (2010) mentioned that the measurement of the flow direction was 517 not many and in fact only single measurement was made near the coastline of the transect. There 518 may be an uncertainty in the tsunami inundation direction (and probably the sediment source). If this 519 is the case, additional computation of the forward and DLNN model using reference line with 520 different directions are needed. 521

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RE: Thank you for the comment. The transect was considered along the flow direction the flow direction and because of the complicated local topography the perfectly perpendicular locations were not possible to collect. In the previous paper Mitra et al., 2020 it was presented that the DNN inverse model works well perpendicular to the shoreline and Sendai plain was one of the most well-preserved tsunami deposits from 2011 Tohoku-oki tsunami. Moreover, practically in recent or ancient tsunami deposits, it is often difficult to obtain samples from perfectly shore 529 perpendicular transects due to anthropogenic disturbances or complicated topography. 530 Comparison between the model estimates and the observed flow depths represents that even if we 531 are unable to obtain completely shore-perpendicular transect, the DNN model still predicts 532 reasonable and realistic flow conditions.

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534 Q3 With regard to Section 4.3, I think the flow speed comparison is problematic, since the model assumes a constant flow speed over the inversion region and it is not clear whether it represents 535 either an average, maximum or something else. This also applies to the measured flow speeds. 536 Unless the attributes of the measured flow speeds (i.e. average, maximum or other) are specified, 537 the measured values cannot be compared with inversion results. The comparison to the inversion 538 results of the TsuSedMod also needs a careful discussion, since the TsuSedMod employs different 539 model assumptions and formulations. It is not clear how the comparison of the two different 540 inversion results are justified. 541

542 543 **RE:** The flow velocity in this model has been considered as constant, and this reconstructed velocity should be compared with the spatially averaged velocity of the field observation. In Mitra 544 et al. (2020), the result of velocity reconstruction was compared with the field measurements that 545 varied in space. As a result, the estimated value matched the spatial average of the measurements. 546 Our reconstructed values are the only estimates available for this region, and we do not consider 547 that these existing velocity values validate our model because of the different regions where the 548 549 measurements were taken. In addition, it is true that we do not know the timing of the velocity values that we referred in this paper. Nevertheless, we believe that referring to estimates of 550 velocities in surrounding areas may be useful in interpreting whether the velocity of the tsunami 551 inundation flow in this region was unique. 552 553 We have removed the comparison with the TsuSedMod as the model assumption and formulation of TsuSedMod is different from our model. 554 555

Q3: The idea of coupling DLNN with other tsunami hydrodynamic model, such as the well validated
TUNAMI-N2, is very interesting. Although it must be computationally expensive, the DLNN
inversion can include much more physically plausible hydrodynamic models to improve the model
performance. I suggest to expand on this aspect, such as outlining a road map and future challenges.

561 *RE: Thank you for the suggestion. We agree that our proposal might be computationally* 562 *expensive. In our revised manuscript we have added few details about this. The modified text is* 563 *as follows:*

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P. 22, line 370, In addition to the source model, this model also includes tsunami sediment transport calculation that consists of bed load layer and suspended load layer. However, the calculated value of the sediment thickness was overestimated as the assumption of movable bed for a large area caused excessive erosion of the ground (Masaya et al., 2019).

- 569 Line 379, The model calculation of Masaya et al. (2019) relies on the estimation of a single set of
- 570 fault parameters, which were not widely explored to obtain the optimal parameters. In future,
- 571 Model TUNAMI-N2can be potentially used as the forward model in DNN inverse model to

572 573 574 575 576 577	considering sediment transport of multiple grain size classes.			
578 579	1.	Abstract: I suggest to use 'distance' rather than 'length'.		
580 581 582		RE: Thank you for your suggestion. We decided to change the 'length' to 'distance' in the entire document.		
583 584	2.	P-1, Line 22, Out of context?		
585 586 587		RE: Thank you for the comment. We have removed the sentence at line 22 considering the sentence as out of context.		
588 589 590 591 592	3.	P-2, Line 25, I can understand what the authors intended to say, but the wording is unusual. Consider to revise, line 28, I understand what the authors intended to say, but the structure of the sentence is not easy to follow. Consider to revise, line 31, I understand what the authors intended to say, but the structure of the sentence is not easy to follow. Consider to revise.		
593 594		RE: Thank you for the suggestion. We have revised the text as follows,		
595 596 597 598		P.2, line 25, Indeed due to the lower tsunami risk and the higher return period of high magnitude tsunamis (600 years) (Suppasri et al., 2015), the degree of preparedness, for example, effective evacuation techniques, and appropriate awareness are still in the early stageof development in Thailand (Suppasri et al., 2012).		
599 600 601 602 603		Line 28, Suppasri et al. (2012) reported that, the nation has implemented post-tsunami precautionary measures such as, the construction of evacuation shelters at a safe height and distance from the coastline along with the evacuation routes with evacuation regulations, memorial parks, appropriate structural design and land use management which were aimed at dealing with tsunami waves.		
604 605 606 607		Line 33, To propose further regulations for evacuation plan and tsunami hazard mitigation, evaluating the extent of tsunamis with the, flow velocity and the maximum height that the tsunamis could reach is important (Pignatelli et al., 2009).		
608 609	4.	P-2, Line 25, lower?		
610 611 612		RE: In the sentence we tried to explain that the tsunami preparedness was not appropriate because of higher tsunami return period that resulted into extensive damages.		
613 614	5.	P-2, Line 36, The sentence is incomplete.		

615 **RE:** Thank you for identifying this. We considered revising the sentence as "Meanwhile, other 616 flow parameters, such as flow velocity and depth, remain largely unknown".

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6. P-2, Line 40, I understnad what the authors intended to say, but the structure of the sentence is not easy to follow. Consider to revise.

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621 *RE: Thank you. We have considered revising the sentence.*

P.2, line 43, It is important to obtain the flow conditions essential to tsunami hazard
mitigation in terms of devising future resilient structural measures by investigating tsunami
deposits, which provide crucial information on the flow discharge and the extent of the
tsunami inundation (Dawson and Shi, 2000; Udo et al., 2016; Sugawara and Goto,
2012;45Furusato and Tanaka, 2014; Sugawara et al., 2014; Koiwa et al., 2018; Masaya et al.,
2019).

- For example, this paper is
 a report of forward tsunami propagation and inundation modeling. Check all other citations
 carefully whether they are used appropriately.
- 633 **RE:** Thank for the comment. We will recheck the citations and we revise the sentence as 634 follows:

P. 3, line 58, To reconstruct quantitative values of tsunami characteristics from the deposits,
various numerical forward and inverse models which incorporate sediment dynamics, and
transport and depositional equations have been established.

- 8. P-3, Line 60, The structure is not easy to follow. Compare the estimated and observed values for
 the inundation distance and flow speed and depth.
- 642 *RE:* We agree with the reviewer. We added few more sentences to compare the estimated and 643 observed values. These sentences are added
- Line 65 "The DNN inverse model predicted the tsunami flow conditions. The reconstructed inundation length was 4,045m which is close to the original maximum inundation length of approximately 4,020 m, values of run-up flow velocity were 5.4 m/s which was close to the spatial average of the measurements which ranged from 1.9 to 6.9 m/s, and the estimations of
- 648 *the inundation depth was 4.11 m which was also within the range of the in-situ measured values*
- 649 from Sendai plain."
- 650
- 651 9. P-3, Line 65, The sentence is complicated. Consider to revise.

Line 74, "The Phra Thong island is one of the locations where the tsunami deposits were preserved without a great amount of topographic irregularities with almost no anthropogenic disturbances in the island."

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⁶⁵² *RE: Thank you for the comment. In response to your comment we have revised the text as* 653 *follows:*

^{658 10.} P-3, Line 67, This sentence must be divided into two sentences.

RE: Thank you for the comment. In response to your comment we have revised the text as follows: Line 66, The coastlines of Phra Thong island were severely eroded and retreated by the 2004 tsunami. However, the presence of widespread mangrove forests with other waterborne plant debris helped in the identifications of the extent and direction of the flow.

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- 665 11. P-3, Line 75, Difficult to understand how the flow conditions can be used to estimate sediment 666 characteristics.
- 668 *RE: Thank you for the comment. We added the following paragraph to modify the text,*

P.3, Line 86, Here, we conduct an DNN inverse analysis of the tsunami deposits measured at Phra 669 Thong island and reconstruct the flow conditions such as the maximum inundation length, flow 670 velocity, maximum flow depth and sediment concentrations of five grain-size classes. The inverse 671 model was based on the forward model, which was proposed by Naruse and Abe, (2017). The 672 forward model calculations were iterated at random initial flow conditions to produce artificial 673 training data sets that represent depositional characteristics such as the spatial distribution of 674 thickness and grain-size composition. Using the artificial training datasets, the DNN was then 675 trained to establish a relation between the depositional characteristics and the and the flow 676 conditions. The post-trained DNN model was ready to predict flow conditions from the tsunami 677 deposits after the performance of the trained DNN was verified using test data sets. The 1-D cubic 678 interpolation was applied to the field data sets of Phra Thong island to fit the dataset in to model 679 grids. Finally, this DNN inverse model was applied to the field data sets from the Phra Thong 680 island, Thailand to reconstruct the flow conditions of 2004 Indian Ocean tsunami. We also used 681 the reconstructed flow conditions to estimate the spatial distribution of the volume per unit area 682 and grain-size composition from Phra Thong island and compare the distribution with the 683 measured data. Our inverse model was already validated to be effective for 2011 Tohoku-oki 684 tsunami deposits distributed in Sendai Plain. In case of Phra Thong island, we validated the 685 results by the field measurements of the tsunami flow depth. Also, the estimated thickness and 686 grain size distribution of tsunami deposits were compared with the actual measurements. Our 687 inverse analysis results could be used for designing future tsunami hazard assessments and 688 disaster mitigation strategies in Thailand. 689 690

- 12. P-4, Line 102, Expression for the rate of total sedimentation is missing.
- 693 *RE: Thank you for identifying this. The typing error was corrected as equation 5 in the revised* 694 *manuscript.*
- 13. P-4, Line 104, Consider to explain how likely the assumptions are valid for this case.
- 698 *RE: Thank you for your comment. The explanation is as follows:*
- 699 *P. 8, line 161, The velocity of the run-up flow of the tsunami, U is assumed as uniform and* 700 *steady, but the inundation depth varies in time and space. Hence, this model simplification is* 701 *called the auasi-steady flow assumption (Naruse and Abe, 2017).*
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703 704 705	14. P-4, Line 113, Elaborate how the grain-size classes were determined. Measurement of tsunami deposit? Complete descriptions for the data (number of data points and their locations, thickness, grain-size distribution and other sedimentological features) are needed
706 707	RE: Thank for the suggestion. We have added a new figure (figure 3) on the thickness and grain size distribution and also moved the study area section before the methodology section.
708 709	15. P-5, Line 119, Figure 1a must be placed at first. Consider to exchange the position of Figure 1a and 1b, and swap the names. Or just revise the structure of this section
710 711 712	RE: Thank you for the suggestion. We have done the necessary changes with the position of the diagrams.
713 714 715	16. P-5, Line 120, measured (or observed)
716 717 718	RE : Thank you for the comment. These are all measured deposit. We have replaced the word "natural" with "measured".
718 719 720 721 722	17. P-5, Line 132, Complete descriptions are needed for the basis for selection of the range of the input values. The description may include appropriate reference to field observations or experimental data.
723 724	RE: Thank you for the comment. We have added the necessary references for the selection of the range of input values.
725 726 727 728 729 730	P. 10, line 202, The range of parameters adopted in this study is applicable to most of the large-scale tsunami-inundated areas as the ranges have been selected with several case studies of tsunamis that includes mostly field measurements ,survivor video and numerical analysis (Mori et al., 2011; Wijetunge, 2006; Szczuciński et al., 2012; Matsutomi and Okamoto,2010; Abe et al., 2012; Fritz et al., 2006; Nandasena et al., 2012; Goto et al., 2014).
731 732	18. P-5, Line 146, The sentence is complicated. Consider to revise.
733 734	RE: Thank you for the comment.
735 736 737 738	P. 11, line 215, To apply the inverse model to the measured values of field dataset from Phra Thong island in 1-D vectors, the collected datapoints must be fit into that fixed coordinate system of the model.
739 740	19. P-7, Line 165, Difficult to understand what 'epochs' indicates. Iterations?
741 742 743 744	RE: Thank you for your comment. The number of epochs indicates the number of times that a full data set has passed the optimization calculation. The specific number of epochs was determined based on the rates of the progress of the training. We can modify the text as follows:
745 746 747	<i>P. 11, line 236, The training process proceeded with a certain number of epochs that indicates the iterations of the optimization calculation by the full dataset.</i>

20. P-8, Line 173, Why the flow depth was biased? 748 749 **RE**: Thank you for the comment. We have added the clarification in the discussion section. 750 P. 20, line 291, The bias was caused by the internal algorithm and neural network structure, 751 but we hope the biasness will be sorted if we improve the neural network structure in future. 752 753 In future studies, the algorithm of the neural network structure can be improved to eliminate 754 or reduce the bias of the parameter. 755 756 21. P-8, Line 178, This section must be moved to somewhere in the section 2. Not in the result section. Readers may want to see whether model assumptions for the sediment source is valid in 757 this island. Were the tsunami deposit along the transect totally brought from the sea? 758 759 **RE**: Thank you for your suggestion. We decided to move the section before methodology. 760 761 762 22. P-8, Line 180, Swap Figures 5 and 6. Figure 5 must be appeared first in the text. Give name of 763 764 insets (a, b and c) for current Figure 6 for the convenience of readers. 765 766 **RE:** Thank you for the comment. We have revised the positions and labelling of the figures. 767 768 23. P-8, Line 191, However, the reference line seems not to be perpendicular to the coastline, according to Figure 6. Explanations are needed regarding the compatibility to the model 769 assumption (i.e. inversion transect must be perpendicular to the coastline). 770 771 772 **RE**: We are assuming that this comment is the same comment or related to Q2. Please refer 773 to the reply of Q2. 774 775 24. P-8, Line 192, The sentence is complicated. Consider to revise, Line 204, The sentence is 776 complicated. Consider to revise. **RE**: Thank you for your comment. We have modified the text as follows: 777 P. 4, line 119, The sediment from shallow seafloors were transported and deposited in large 778 volumes of sand sheet deposition widely along the coast, with the deposit is largely composed 779 of medium to fine sand. The deposit became thinner and finer in a landward direction, 780 becoming very fine at the landward limit of the inundation. 781 P. 14, line 256, The jackknife standard error was calculated for different sampling window 782 sizes of the datasets. Figure 9 represents that the error decreased as the sampling window 783 784 was increased. 785 786 25. P-10, Line 206, How do we interpret the increasing trend for the maximum flow depth? 787 788 RE: Thank you for your comment. Figure 9c shows that the jackknife standard error of 789 maximum flow depth has an increasing trend up to sampling window size 1500 m. After

790 791		sampling window size 1500 m, the error started to decrease and after 1600 m is was very low yet stable.
792		
793	26.	P-11, Line 212, The sentence is quite complicated and is not easy to follow in terms of English
794		writing. Consider to revise.
795		
796		RE: Thank you for your comment. We have revised the section of the paragraph as follows:
797		P. 15, line 264, The subsampling test demonstrated that the inversion model had a mean bias
798		of 10.82 m for maximum inundation distance (Figure 10) while the predicted result by DNN
799		was 1700 m. Likewise the predicted results for the flow velocity was 4.63 m/s and it was 4.82
800		m for the maximum flow depth, with the mean bias obtained from the subsampling results
801		being 0.14 m/s for flow velocity and -0.43 m for maximum flow depth, which were exactly in
802		line with the values obtained from the testing of the trained DNN model without the
803		subsampling test.
804		
805		
806	27.	P-11, Line 212, This figure must be appeared after current Figure 7.
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808		RE: Thank you for the suggestion. We have rearranged the figures.
809		
810	28.	P-12, Figure 6, This is not a appropriate caption for this figure.
811 812		RE : Thank you for the comment. We apologize for the typo in the caption. We have revised the caption and labelling as follows:
813 814 815		P. 5, Figure 1. (a) Location of study area in southwestern Thailand. (b) Phra Thong island and adjacent landmark areas where 2004 Indian ocean tsunami inundated. (c) Locations of study sites at Phra Thong island. The 2004 tsunami inundated about 2 km inland.
816 817	29.	P-14, Line 234, Are there any reason for the biased results for the flow depth?
818		RE: We assume that this comment is related to comment number 18. Please refer to the
819		comment 18.
819		comment 16.
820 821	30	P-17, Line 264, Is this value incudes the bias of -0.38m?
822	50.	1-17, Line 204, is this value includes the blas of -0.36in?
823		RE: Thank you for the comment. This value does not contain the additional bias -0.38 (now
		<i>RE. Thank you for the comment. This value does not contain the additional bias -0.38 (now revised -0.43 m). We have added the text in line 322.</i>
824 825		reviseu -v.45 m). We nuve uuueu ine lexi in une 522.
825 826	21	P 17 Line 267 Elaborate what is Tauji and KSCOE http://www.nde.ee.in/_fujime/TMD/
826 827	51.	P-17, Line 267, Elaborate what is Tsuji and KSCOE- http://www.nda.ac.jp/~fujima/TMD/.
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RE: Thank you for the comment. To address this point, in the revised version we have added the details and reference of the Tsuji and KSCOE group on P. 21, line 324-326.

831 32. This URL is no more available.

833 Thank you for identifying this. We have updated the URL link in the revised manuscript.

- 835 33. P-17, Line 276, Considering the local topographical variations, topographic elevation is needed
 836 for other sites to calculate and compare the flow depths.
- RE: Thank you for the suggestion. Previously we averaged the elevation of the area. To
 address this point, we will add the ranges of elevation and the respective flow depths. The
 revision will be as follows:
- P., 21, line 335, The maximum and measured flow heights from Phra Thong island were
 reported 7.1 m and 5.5 m respectively (http://www.nda.ac.jp/~fujima/TMD/). The
 corresponding maximum and minimum values of elevation are 3.1 and 1.1 m respectively
 (Jankaew et al., 2008, 2011; Brill et al., 2012b). Hence, the approximate estimate of measured
 maximum flow depth is ranged from 2.4 m to 6.0 m. Considering the bias correction of 0.43
 m, the reconstructed value of maximum flow depth (5.3 m) falls within the range of measured
 maximum flow depth values.

- 850 34. P-17, Line 283, I think flow speed comparison is problematic, since the model assumes that the
 851 flow speed is constant over the inversion region and it is not clear the inversion result represents
 852 either an average, maximum or something else. This also applies to the measured flow speeds.
 853 Unless the observational condition is specified, the measured values cannot be compared with
 854 inversion results. The question is that the attributes of the measured flow speeds (i.e. average,
 855 maximum or other).
- *RE: Thank you for the comment. We assume this comment is related to Q2. Please refer to the reply of Q2.*
- 859
 860 35. P-17, Line 286, Is this mean the tsunami flow speed was measured using video footages from aircrafts? Rossetto et al. (2007) did not mention about that.

- *RE: We apologize for mentioning aerial footage. Rossetto et al. (2007) mentioned about the video footage only. We have corrected the statement in line 350 of the revised version.*
- 866 36. P-18, Line 288, TsuSedMod uses different model assumptions and formulations. It is not clear
 867 how the comparison of the two different inversion results are justified.

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RE: We agree with the reviewer. We decided to remove the comparison with TsuSedMod.

871 37. P-18, Line 306, Before mentioning this, outlines of Masaya et al. (2019) must be introduced. It
872 seems that the paper includes not only the tsunami hydrodynamic simulation but also sediment
873 transport simulations.

RE: Thank you for the suggestion. In our revised manuscript, we have added the outlines of
the model (Masaya et al., 2019) including hydrodynamics and sediment transport.
P. 22. line 370. In addition to the source model, this model also includes tsunami sediment

P. 22, line 370, In addition to the source model, this model also includes tsunami sediment
transport calculation that consists of bed load layer and suspended load layer. However, the
calculated value of the sediment thickness was overestimated as the assumption of movable
bed for a large area caused excessive erosion of the ground (Masaya et al., 2019).

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882 38. P-18, Line 311, The idea is very interesting. Although it must be computationally expensive,
883 the DNN inversion can include much more physically plausible hydrodynamic models to
884 improve the performance. Better to expand on this, such as outlining a load map and future
885 challenges.

887 **RE:** We thank the reviewer for appreciating the idea. Currently, the model calculation of 888 Masaya et al., 2019 based on the single assumption of measured parameters which are not 889 optimized. Model TUNAMI-N2 could be integrated with the forward model of DNN inverse 890 model but the model needs to be modified for multiple grain size classes with the optimized 891 parameters. We have added these details in our revised manuscript as follows:

- P. 22, Line 379, The model calculation of Masaya et al. (2019) relies on the estimation of a single set of fault parameters, which were not widely explored to obtain the optimal parameters.
 In future, Model TUNAMI-N2can be potentially used as the forward model in DNN inverse model to consider two-dimensional behavior of tsunamis. To do so, the model needs to be
- 896 *modified for considering sediment transport of multiple grain size classes.*