



12 **Abstract**

13 We describe the development of a block-structured, equal CPU-load, multigrid nesting in-
14 terface for the Boussinesq wave model FUNWAVE-TVD. The new model framework does
15 not interfere with the core solver, and thus the core program, FUNWAVE-TVD, is still a
16 stand-alone model used for a single grid. The nesting interface manages the time sequencing
17 and two-way nesting processes between the parent grid and child grid with grid refinement
18 in a hierarchical manner. Workload balance in the MPI-based parallelization is handled by
19 an equal-load scheme. A strategy of shared array allocation is applied for data management,
20 that allows a large number of nested grids without creating additional memory allocations.
21 Four model tests are conducted to verify the nesting algorithm, model accuracy, wetting-
22 drying treatment, and the robustness in the application to modeling transoceanic tsunamis
23 and coastal effects.

24 **Plain Language Summary**

25 The multiple-grid nesting technique is an important methodology used for modeling
26 transoceanic tsunamis and coastal effects. The traditional grid nesting approach is one-way
27 nesting, which is done manually grid by grid. In this study, we developed a two-way nesting
28 interface in a multigrid nesting system for the Boussinesq wave model, FUNWAVE-TVD.
29 FUNWAVE-TVD is a widely accepted open-source wave model for simulating surface wave
30 propagation and wave-driven processes in the nearshore region, as well as tsunami wave
31 propagation and evolution from the oceanic scale to nearshore scales. The new development
32 of the interface does not alter the core solver, and thus the core program is still a stand-
33 alone model used for single grid applications. Some strategies in workload balance, data
34 management, and parent-child communications in the MPI-based parallelization system are
35 utilized to guarantee the model efficiency and accuracy. Four model tests are carried out in
36 the paper.

37 **1 Introduction**

38 To improve the resilience of the world's highly populated coastal areas to tsunami haz-
39 ard when tsunamigenic events (typically earthquakes or landslides) occur, there has been
40 an increasing need for issuing early warnings and near- and far-field forecast of tsunami
41 coastal impact. This has led to a growing demand for accurate and efficient models of
42 transoceanic tsunami propagation, in multiple-nested grid systems that allow refining the



43 discretization towards shore, as depth decreases. Models predicting tsunami wave evolution
44 from generation at the source, to propagation at the oceanic basin-scale, transformation
45 over the shelf, and coastal inundation in the nearshore-scale have typically been based on
46 the non-dispersive nonlinear shallow water wave equations (NSWE; e.g., GeoClaw, George
47 and LeVeque, 2008) or on dispersive Boussinesq-type such as FUNWAVE (e.g., Shi et al.,
48 2012; Kirby et al., 2013) or non-hydrostatic wave equations such as NHWAVE (e.g., Ma et
49 al., 2012; Tappin et al., 2014; Grilli et al., 2019). Modeling studies of tsunami propagation
50 in the ocean with and without dispersion have indicated that, even for co-seismic tsunamis,
51 frequency dispersion effects can accumulate to a sufficient degree to change waveforms, al-
52 tering the spatial distribution of wave elevations and coastal inundation (Ioualalen et al.,
53 2007; Horrillo et al., 2012; Kirby et al., 2013; Glimsdal et al. 2013; Zhou et al. 2012; Kirby,
54 2016). Due to wave dispersion and nonlinearity, tsunami wave crests often evolve into un-
55 dular bores (a.k.a., dispersive shock waves) as they approach the shoreline, an effect which
56 may significantly increase tsunami impact (i.e., currents and forces) on coastal structures
57 (Madsen et al., 2008; Schambach et al., 2019). For landslide-generated tsunamis, wave-
58 lengths are relatively shorter, and thus wave dispersion effects cannot be neglected (e.g.,
59 Ma et al., 2012; Grilli et al. 2015, 2017, Schambach et al., 2019). As shown in the above
60 studies, the magnitude of dispersive effects at given locations is a priori unknown; hence, it
61 can only be estimated by performing simulations with a dispersive models for each specific
62 event, whether hypothetical, historical or in real time. With this realization, in the last
63 decade, modelers have gradually acknowledged the need for using a dispersive wave model
64 to accurately assess tsunami hazard, especially nearshore effects.

65 Although some models use irregular grids or adaptive mesh refinement, the traditional
66 way for carrying out multi-scale tsunami modeling has been to use nested grids, either with a
67 one-way nesting or two-way nesting method. The grid nesting method is usually performed
68 by nesting a fine grid within a coarse grid in a two- or multi-grid system with the hierarchical
69 structure from coarser (lower-level) to finer grids (upper-level). In a one-way nesting, the
70 model at an upper-level is forced by the boundary conditions obtained from the output of
71 the lower-level model. There is no feedback from the upper-level grid to the lower-level grid.
72 The nesting process can be done offline manually by running the model from the lower-level
73 grid to the upper-level grid without an additional interface developed in the model. Kirby
74 et al. (2013), Tappin et al. (2014), Schambach et al. (2019, 2020), and Nemati et al. (2019),
75 for instance, are recent examples of using many levels of one-way nested spherical and/or



76 Cartesian grids, with FUNWAVE and/or NHWAVE, varying from a few meters or tens of
77 meters nearshore, to 1 or 2 arc-min in the deep ocean. In a two-way nesting, the procedure
78 to force the upper-level grid model is the same as the one-way nesting, but the feedback from
79 the coarse grid to the fine grid is taken into account by updating the coarse grid solution
80 with the fine grid solution. To achieve this, the calculations at all grid levels have to perform
81 simultaneously. To this effect, an interface to handle the interactions between the nested
82 grids has to be developed, which involves a significant programming effort.

83 Multi-scale tsunami modeling may also be carried out using adaptive mesh refinement
84 (AMR). In an AMR model, the calculations at all grid levels have to perform simultane-
85 ously, in which the grid resolution is adaptively refined as a function of chosen features of
86 the flow field, such as a high spatial gradient in the solution. AMR can be implemented us-
87 ing either an unstructured (e.g., Sleigh et al., 1998, Skoula et al., 2006) or block-structured
88 scheme (Berger and Olinger, 1984; Berger and LeVeque, 1998; Liang, 2012). The latter is
89 very similar to the two-way grid nesting mentioned above, except that the grid refinement
90 is processed dynamically rather than prescribed using static sub-domains in the traditional
91 two-way nesting framework. Over the last decade, the AMR technique has found increasing
92 use in publicly available codes (see the review paper by Dubey et al., 2014). In tsunami
93 applications, the AMR technique has been used in the NSWE-based models such as Geo-
94 Claw (George and LeVeque, 2008, Watanabe et al., 2012, Arcos and LeVeque, 2014). For
95 Boussinesq-type wave models, however, the higher-order numerical schemes and tridiagonal
96 matrices, which are derived on a structured grid system, make it challenging to implement
97 a quadtree-structured AMR; although a block-structured AMR is relatively easier to imple-
98 ment, its efficiency may be penalized by the large data management and computational costs
99 when solving the complex nonlinear dispersive equations at multi-grid levels. Therefore, the
100 AMR technique has rarely been applied to solving Boussinesq-type wave equations.

101 In practical applications of multi-scale tsunami modeling using a dispersive wave model,
102 the traditional multi-grid one-way nesting approach has proved efficient and accurate when
103 focusing on nearshore effects, provided the nearshore grid refinement ratio (i.e., ratio of
104 discretization size from one nested grid to the next) is 4 or better. As the coastal area
105 of interest is usually predetermined when setting up the model, the grid refinement can
106 be generated at the beginning and remain unchanged throughout the entire simulation.
107 Besides other applications mentioned above, a typical recent example is Tehranirad et al.'s
108 (2020) FUNWAVE simulations of the far-field effects of the Tohoku-Oki 2011 tsunami in



109 Crescent City Harbor, California, using a nested grid system including the ocean basin,
110 regional, and nearshore harbor domains, as shown in Fig. 1. The basin-scale grid has a
111 2 arc-min resolution, covering the entire Pacific Ocean; the nested grids are then specified
112 in five levels along the U.S. west coast, with a hierarchical structure from a resolution
113 of 16 arc-sec to 1/6 arc-sec, downsizing towards the Crescent City Harbor domain. The
114 fully nonlinear Boussinesq model, FUNWAVE-TVD (Shi et al., 2012; Kirby et al., 2013)
115 is used in each individual grid with a one-way nesting scheme performed by applying the
116 boundary conditions obtained from a lower-level grid model. While this nesting process is
117 straightforward, it involves considerable post-processing effort to manipulate and interpolate
118 results from one level of nested grid to prepare data for simulating the next level grid. In
119 addition, the one-way nesting scheme may cause inconsistencies between different grids due
120 to wave reflection at each model boundaries.

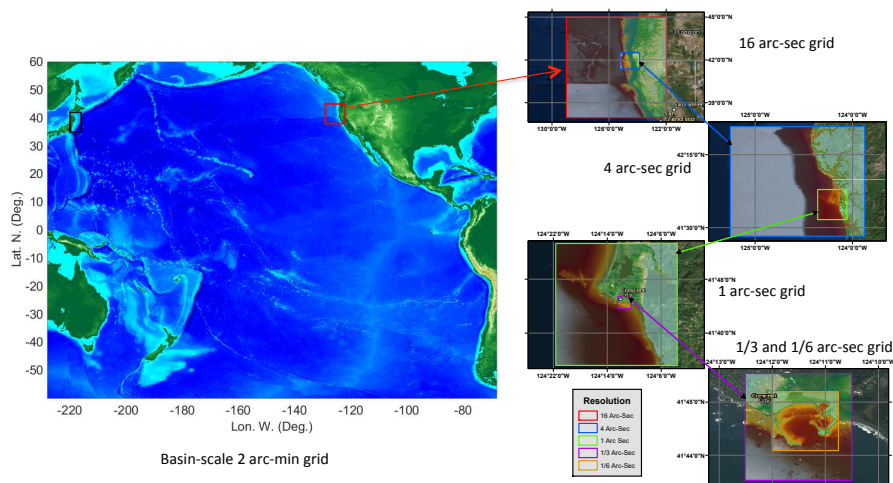


Fig. 1. The nested grids in the simulation of Tohoku-Oki 2011 tsunami impact on Crescent City Harbor, Oregon (Tehrani-rad et al., 2020). The nesting process scales down the grid resolution from 2 arc-min in the ocean basin domain to 1/6 arc-sec in the harbor domain. The same simulation was conducted using the present model nesting framework in section 4.4.



121 Yamazaki et al. (2010) implemented two-way nesting method in their dispersive depth-
122 integrated, non-hydrostatic wave model for tsunami applications. The nesting model frame-
123 work was based on a block-structured scheme with multiple prescribed nested grids. They
124 used this model to simulate the 2009 Samoa tsunami and the coastal inundation caused
125 in Pago Pago harbor, and reported good efficiency and accuracy of the two-way nesting
126 model framework. However, it is not clear whether this two-way nesting scheme was paral-
127 lelized and how the mega-data structure was handled in the nesting framework. Recently,
128 Chakrabarti et al. (2017) implemented the fully nonlinear and dispersive Boussinesq model
129 FUNWAVE-TVD in the block structured AMR framework CACTUS, which has been widely
130 used in the field of astrophysics (Löffler et al., 2014). They showed that shallow water waves
131 could be simulated at higher resolution, with a reasonable computational cost, which also al-
132 lowed using an improved higher-order representation of the vegetation drag force. However,
133 in this application, the nested grids were statically prescribed, to reduce the computational
134 cost from using dynamically adapted grids with a Boussinesq-type model. In addition, the
135 CACTUS-based version of FUNWAVE-TVD relies on a specific library package and config-
136 uration (Oler et al., 2016), limiting its general applications in the large user community.

137 There are significant challenges implementing an AMR and two-way multi-grid nesting
138 framework in a parallel computing environment. Load balance, communication between
139 parent and child grids, and mega-data management are major issues in Message Passing
140 Interface (MPI)-based programs. Load balance is important for CPU scaling, in terms of
141 synchronization of solutions across refinement levels. Dubey et al. (2014) reviewed load
142 balancing methods in several public-domain AMR packages and pointed out the difficulties
143 in achieving workload balance in the AMR framework. In a parallel multi-level grid system,
144 the parent-child grid communication is also critical to modeling efficiency. Strategies to build
145 direct communication between multi-level grids cross-ranks can be found in many AMR
146 packages (Dubey et al., 2014). A multi-level grid system makes the meta-data management
147 more complex, especially for tree-structured data. Finally, it is important to optimize the
148 amount of meta-data replication according to both the communication cost and memory
149 cost (Dubey et al., 2014).

150 The scope of the present work is to develop a multi-grid nesting framework for the
151 Boussinesq-type wave model FUNWAVE-TVD, a widely-used public domain model in the
152 nearshore and tsunami research community. FUNWAVE was initially developed by Kirby
153 et al. (1998) based on the fully nonlinear Boussinesq equations derived by Wei et al. (1995).



154 The development of the Total Variation Diminishing (TVD) version of the model was mo-
155 tivated by a growing demand for phase-resolving modeling of nearshore waves and coastal
156 inundation during storm or tsunami events. The model comprises a Cartesian mode (Shi
157 et al., 2012) and a spherical mode (Kirby et al., 2013), an appropriate mode can be se-
158 lected according to applications. The code was parallelized using the domain decomposition
159 method based on MPI, for CPU-based High-Performance-Computing (HPC) clusters, and
160 the GPU-accelerated program for single- and multi-GPU systems (Yuan et al., 2020).

161 The limitation of the prior CACTUS implementation of an AMR version of FUNWAVE-
162 TVD to a specific HPC platform (Chakrabarti et al., 2017), motivates the development of
163 a more platform-independent implementation of a two-way nesting scheme. The primary
164 objective for the present development is to provide a generic interface, which can be used
165 with any HPC platforms. The interface is developed separately from the core program
166 and does not interfere with the main solver of FUNWAVE-TVD. Hence, the package of the
167 combined interface and core program can be updated concurrently.

168 In the following, a brief description of the FUNWAVE-TVD model is given in section 2.
169 Section 3 describes the two-way nesting interface, including the general algorithm, workload
170 balance and flowchart of a MASTER program. Applications are presented in section 4.
171 Section 5 provides a summary of the study.

172 **2 FUNWAVE-TVD**

173 FUNWAVE-TVD is a Boussinesq-type wave model discretized by a hybrid method com-
174 bining finite-volume and finite-difference TVD-type schemes. The model was developed in
175 both the Cartesian coordinates (Cartesian mode, Shi et al., 2012) and spherical coordinates
176 (Spherical mode, Kirby et al., 2013). The Cartesian mode solves the fully nonlinear Boussi-
177 nesq equations, initially derived by Wei et al. (1995), with the second-order correction of
178 vertical vorticity by Chen (2006) and the moving reference level of Kennedy et al. (2001).
179 The Spherical mode solves the weakly nonlinear, weakly dispersive Boussinesq equations in
180 spherical coordinates (Kirby, et al., 2013). In tsunami applications, where nearshore waves
181 are expected to be strongly nonlinear, a combination of deep water spherical and nearshore
182 Cartesian grids has often been used in the one-way coupling nested grid framework (e.g.,
183 Grilli et al., 2013, 2015, 2017; Schambach et al., 2019, 2020; Tappin et al., 2014). Here, we



184 provide a brief summary of the governing equations, numerical schemes, and parallelization
 185 method.

186 **2.1 Conservative form of Boussinesq equations in the Cartesian and spher-**
 187 **ical coordinate systems**

Although the sets of equations in Cartesian and spherical coordinate systems are differ-
 ent, the two FUNWAVE modes were developed within the same numerical framework and
 using the same TVD-type solver. The combined form of the Boussinesq equations in the
 two coordinate systems can be written as:

$$\frac{\partial \Psi}{\partial t} + \nabla \cdot \Theta(\Psi) = \mathbf{S}, \quad (1)$$

188 where Ψ and $\Theta(\Psi)$ are the vector of conserved variables and the flux vector function,
 189 respectively, given by:

$$\Psi = \begin{pmatrix} H \\ U \\ V \end{pmatrix}, \quad \Theta = \begin{pmatrix} S_p P \mathbf{i} + Q \mathbf{j} \\ \left[\frac{S_p P^2}{H} + \frac{1}{2} S_p g (\eta^2 + 2\eta h) \right] \mathbf{i} + \frac{PQ}{H} \mathbf{j} \\ \frac{S_p PQ}{H} \mathbf{i} + \left[\frac{Q^2}{H} + \frac{1}{2} g (\eta^2 + 2\eta h) \right] \mathbf{j} \end{pmatrix}, \quad (2)$$

where (P, Q) are the horizontal volume fluxes:

$$(P, Q) = H(\mathbf{u}_\alpha + \bar{\mathbf{u}}_2), \quad (3)$$

where $H = h + \eta$ with h the water depth and η the surface elevation, \mathbf{u}_α is the horizontal
 velocity vector at a reference depth z_α , $\bar{\mathbf{u}}_2$ is the depth-averaged second-order horizontal
 velocity of $O(\mu^2)$, in which μ is the dimensionless parameter quantifying the magnitude
 of wave dispersion. The velocity components (U, V) combine \mathbf{u}_α and the time derivative
 dispersive terms \mathbf{V}_1 :

$$(U, V) = H(\mathbf{u}_\alpha + \mathbf{V}_1). \quad (4)$$

190 The velocity \mathbf{u}_α is obtained by solving a system of equations with a tridiagonal matrix
 191 formed by (4).

In (2), S_p is the spherical coordinate correction factor defined for the spherical mode
 as:

$$S_p = \frac{\cos \theta_0}{\cos \theta}, \quad (5)$$

192 in which θ and θ_0 are the latitude and the reference latitude, respectively (see, Kirby et al.,
 193 2013). For the Cartesian mode, $S_p = 1$. The last term \mathbf{S} in (1) contains the Boussinesq



194 source terms, which are detailed in Shi et al. (2012) for the Cartesian mode and Kirby et
195 al. (2013) for the spherical mode.

196 **2.2 Numerical schemes**

197 In FUNWAVE-TVD, the HLL Riemann solver with the fourth-order accurate MUSCL-
198 TVD scheme (Erduran et al., 2005) was implemented to discretize the leading order spatial
199 derivative terms of the equations, while the dispersive terms were discretized by a second-
200 order centered finite difference scheme. Choi et al. (2018) compared the performance of
201 the MUSCL-TVD, WENO and MLP schemes in FUNWAVE-TVD, and showed that the
202 MUSCL-TVD scheme with a van-Leer limiter provides an accurate and stable solution in
203 long-term simulations.

204 For time stepping the equations, FUNWAVE uses the third-order Strong Stability-
205 Preserving (SSP) Runge-Kutta scheme (Gottlieb et al., 2001), with an adaptive time step-
206 ping based on the Courant-Friedrichs-Lewy (CFL) condition prescribed as:

$$\Delta t = C_{\text{cfl}} \min \left(\min \frac{\Delta x}{|u_{\alpha} + \sqrt{gH}|}, \min \frac{\Delta y}{|v_{\alpha} + \sqrt{gH}|} \right), \quad (6)$$

207 where C_{cfl} is the Courant number, and Δx and Δy are grid sizes in the x and y directions,
208 respectively.

209 Although the conservative equations (1) are solved explicitly using the HLL Riemann
210 solver, a system of tridiagonal matrix equations derived from (3) needs to be solved to get
211 the velocity at the reference level, which is done with Thomas' algorithm (Naik et al., 1993).

212 Various boundary conditions were implemented in the model, including a wall boundary
213 condition, wave periodic boundary condition, wavemaker boundary condition, and absorb-
214 ing or partially absorbing boundary conditions. The wall boundary condition is the main
215 boundary condition, dealing with either full wave reflection or a moving shoreline. Ghost
216 cells are used in the grid to implement a mirror boundary condition.

217 **2.3 Parallelization**

218 The CPU code uses a domain decomposition technique to subdivide the problem into
219 multiple regions and assign each subdomain to a separate processor core. Each subdomain
220 region contains an overlapping area of ghost cells, which is three rows deep, as required by



221 the fourth-order MUSCL-TVD scheme. MPI with non-blocking communication is used to
222 exchange data in the overlapping region between neighboring processors. The tridiagonal
223 matrices are solved using the parallel pipelining tridiagonal solver described in Naik et al.
224 (1993).

225 Data exchanges between neighboring subdomains are conducted through the ghost cells
226 at every Runge-Kutta time step. To increase the model efficiency, the values of dispersive
227 terms, in addition to the major variables (η, P, Q) , are also exchanged at the ghost cell
228 boundaries.

229 3 Two-way nesting interface

230 As mentioned in Section 1, the goal of the development here is to build an interface,
231 which can be used as a MASTER program to couple the sub-models with different grid
232 resolutions in a nested, two-ways, interactive manner. This way, FUNWAVE-TVD can be
233 used either stand-alone on a single grid or in a multi-grid nested system.

234 3.1 Algorithm

For simplicity, we consider first a two-nested grid system containing the parent grid Ω_0 and the child grid Ω_1 as shown in Fig. 2. The parent grid has a larger grid size, Δx_0 , while the child grid has a smaller grid size, Δx_1 . The grid refinement ratio is thus defined as, $s = \Delta x_0 / \Delta x_1$. The boundary of the child grid is denoted by Γ , which has ghost cells. Following the general procedure for two-way nesting, such as detailed in Debreu and Blayo (2008), the partial differential equation (1) can be rewritten as:

$$\frac{\partial \Psi}{\partial t} = L(\Psi), \quad (7)$$

where $L(\Psi) = \mathbf{S} - \nabla \cdot \Theta(\Psi)$, represents a general operator. The equation is discretized in Ω_0 and Ω_1 grids, by:

$$\frac{\partial \Psi_0}{\partial t} = L_0(\Psi_0), \quad \frac{\partial \Psi_1}{\partial t} = L_1(\Psi_1), \quad (8)$$

235 respectively, where L_0 and L_1 denote the discretized form of the same operator L at a
236 different resolution. In the two-way nesting framework, the child grid solution is driven by
237 the lateral boundary conditions along Γ , while the parent grid is updated using the child
238 grid solution. Noting that both of the procedures need interpolation/mapping processes,
239 we define the interpolators, I_s and I_t , and the restriction operator, R . I_s and I_t perform
240 interpolations in space and time, respectively, at Γ , and R performs the mapping from the



241 child grid solution to the parent grid. Assuming the grid refinement factor, s , equals the
 242 time refinement factor based on the CFL condition (6), the two-way nesting can be described
 243 by the following pseudo code:

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 $\Psi_0^{n+1} = L_0(\Psi_0^n)$ 
loop  $i = 1$  to  $s$ 
 $\Psi_1^{n+\frac{i}{s}} = L_1(\Psi_1^{n+\frac{i-1}{s}})$ 
with  $\Psi_1^{n+\frac{i}{s}}|_{\Gamma} = I_t[I_s(\Psi_0^n), I_s(\Psi_0^{n+1})]$ 
end loop
 $\Psi_0^{n+1} \in \Omega_1 = R(\Psi_1^{n+1}).$ 
    
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244 Three ghost cells are used along Γ , which are required by the higher-order numerical schemes
 245 used in the model.

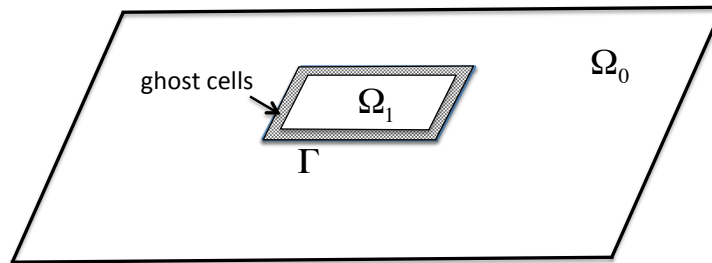


Fig. 2. Schematic drawing for two-way nesting method. The parent grid Ω_0 has a coarse resolution grid while the child grid Ω_1 is high resolution. Ghost cells (3 rows) are specified along the inter-grid boundary Γ .

246 3.2 Interpolator and restriction operator

247 In some two-way nesting methods used in 3D ocean models, the interpolators, I_s and
 248 I_t , and the restriction operator, R , are complex due to issues raised by mass/momentum im-
 249 balance, barotropic/baroclinic mode splitting, and the staggered grid configuration (Debreu
 250 et al., 2012). To ensure mass and momentum conservation during the two-way nesting, a
 251 correction may be needed at the nesting boundaries according to specific numerical schemes.



252 This especially occurs in a nesting scheme using discretizations of the unconservative forms
 253 of mass and momentum equations, based on finite differences, because the flux of mass or
 254 momentum is expressed by a nonlinear term. Hence, typically, the mass flux, $(h + \eta)u$, is
 255 no longer conserved when performing a linear interpolation individually for η and u at a
 256 nesting boundary.

257 FUNWAVE-TVD is based on the conservative forms of mass and momentum equations,
 258 in which advection is performed using the finite volume method (Shi et al., 2012). The latter
 259 makes it possible to use a linear or doubly-linear interpolator in the nesting method, without
 260 changing the conservative property of the equations. In the AMR application of a NSWE
 261 model, Liang (2012) demonstrated the conservative property of the linear operator used in
 262 the finite volume Godunov-type scheme and, later, pointed out that the operator preserves
 263 both mass conservation and the C-property (i.e. conservation property) as the wetting-
 264 drying process involves in the grid nesting. FUNWAVE-TVD uses a finite volume scheme
 265 similar to Liang’s (2012) and Liang et al.’s (2105) and, therefore, its conservative property
 266 should be maintained when applying a linear operator. Unlike Liang (2012), who used a
 267 second-order scheme, FUNWAVE-TVD applies a higher-order Godunov-type scheme, hence
 268 ghost cells must be used along nesting boundaries.

269 Consequently, a doubly-linear interpolation is applied to ghost points in the child do-
 270 main, using values from the parent grid. Thus, at a ghost point (X, Y) in the child do-
 271 main, which is surrounded by four points, (x_{ij}, y_{ij}) , $(x_{i+1,j}, y_{i+1,j})$, $(x_{i+1,j+1}, y_{i+1,j+1})$,
 272 $(x_{i,j+1}, y_{i,j+1})$, in the parent domain, a given variable φ_1 is interpolated as:

$$\varphi_1 = [t\varphi_{ij} + (1-t)\varphi_{i+1,j}]s + [t\varphi_{i,j+1} + (1-t)\varphi_{i+1,j+1}](1-s), \quad (9)$$

273 where,

$$t = \frac{x_{i+1,j} - X}{x_{i+1,j} - x_{ij}}, \quad s = \frac{y_{i,j+1} - Y}{y_{i,j+1} - y_{ij}}, \quad (10)$$

274 where φ_{ij} , $\varphi_{i+1,j}$, $\varphi_{i,j+1}$, $\varphi_{i+1,j+1}$ are values of the variable in the parent domain.

275 The restriction operator uses linear averaging, which guarantees the conservation of
 276 mass and momentum.



277 **3.3 Workload balance and data management**

278 The MPI parallelization of FUNWAVE-TVD uses a 2D Cartesian topology for the
279 domain-decomposition, which subdivides the computational domain into a 2D grid, each
280 cell of which is assigned to a processor. The size of the global arrays is not necessarily
281 divisible by the number of processors, but an evenly divisible configuration results in a
282 perfectly equal workload. To ensure workload balance in computations involving multi-
283 grid levels, we used the same domain-decomposition algorithm on all grid levels with the
284 same number of processors. This algorithm is especially efficient for block-structured or
285 patch-structured nesting schemes, as described in Debreu et al. (2012).

286 Fig. 3 gives an example of the domain-decomposition and communication at two-grid
287 levels in a system of 9 processors. Both the parent computational domain and the child
288 domain are decomposed evenly into a 3×3 grid, according to the standard 2D Cartesian
289 virtual topology used in the MPI library, with ranks named ID=1, 2, ..., 9. The commu-
290 nications between the parent and child grids are straightforward without a data-gathering
291 process using an additional processor or a global array. In this example, along the west
292 boundary of the child grid, the processors with ID=1, 2 and 3 communicate directly with
293 processors with ID = 4 and 5 in the parent grid. The parent-child proximity is pre-calculated
294 at the beginning of the model run, and hence will not require additional computational cost.
295 The same strategy is used by the restriction process.

296 As mentioned in the introduction, our goal in development is to make a generic grid
297 nesting interface without altering the main FUNWAVE-TVD code. To achieve this, we
298 treated the main program of the original model as a kernel, which performs computations
299 at all grid levels. The kernel is called from the MASTER program, which manages the time
300 sequencing and nesting processes. A strategy of shared array allocation is used, whereby
301 the arrays are allocated with the maximum dimension of all grids at the initialization stage
302 and grids at all levels share the same memory allocations. New arrays are created only
303 for the storage of boundary conditions at all levels. There is no additional data structure
304 implemented in the meta-data management.

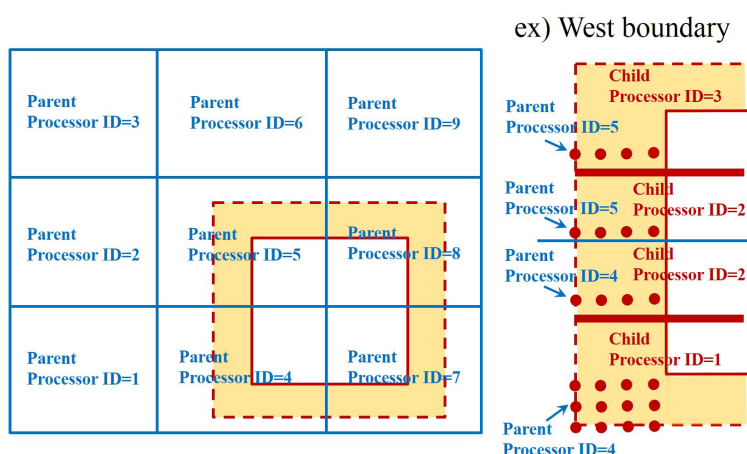


Fig. 3. Decomposed domain by MPI and the relation between parent processor ID and child processor ID.

3.4 Flowchart

Fig. 4 summarizes the flowchart of the MASTER program. After the MPI initialization, the program reads input data, including model parameters needed in the original model and nested grid information. As mentioned earlier, array allocation and initialization are performed based on the maximum dimension of all grid levels. Additional arrays for the storage of boundary conditions are also allocated at this stage. Then the program starts the main time loop based on time stepping of the background (first level parent) grid. The calculations at each grid level are conducted hierarchically inside the main time loop, with a time steps based on the grid refinement ratio s . At each grid level, the model is assigned by the initial condition (solution at last time level) and boundary conditions obtained from the I_s and I_t interpolation processes. Then the core FUNWAVE-TVD program is run at the grid level and stores boundary values for the child grid. All parent grids are updated based on the child grid results through the R process, after all subgrid levels computations are completed.

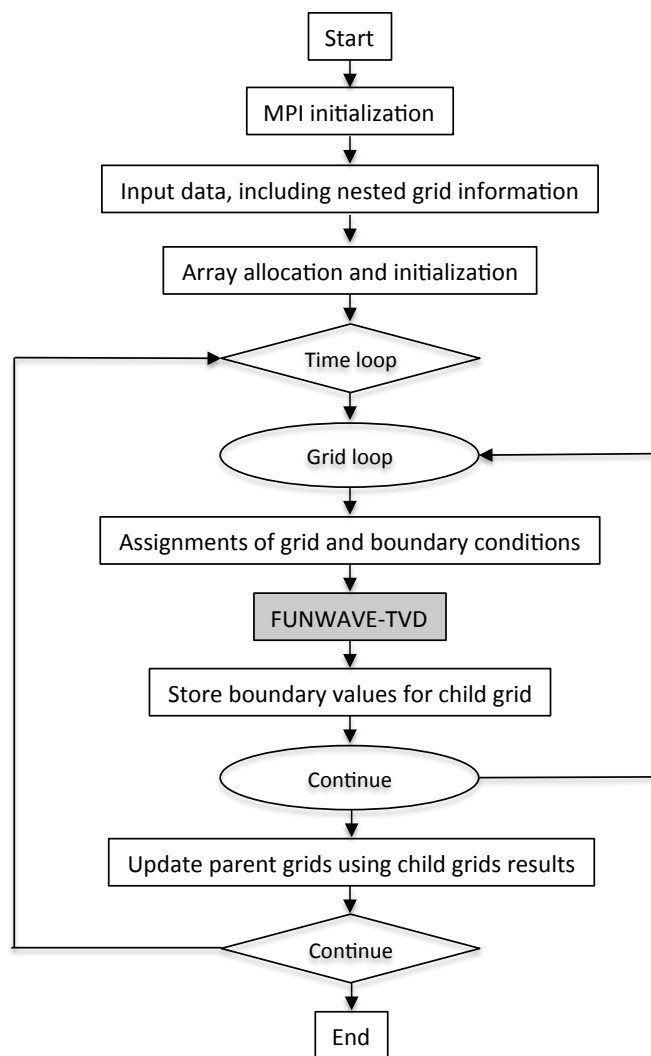


Fig. 4. Flowchart of the two-way nesting interface.

319 4 Applications

320 Hereafter we test our new two-way coupled nested grid solution with FUNWAVE-TVD
321 on a series of standard idealized or benchmarking applications, and then on the Tohoku
322 2011 tsunami case study discussed earlier.



323 **4.1 Evolution of an initial rectangular-shaped hump**

324 The evolution of waves generated from an initial arbitrary rectangular-shaped hump on
 325 the free surface is used to test the consistency of the multi-grid nesting system and effects
 326 of the higher-resolution resulting from the grid refinement. As shown in Fig. 5, a 100 m ×
 327 100 m hump with an elevation of 1 m is specified, with no initial velocity, at the center of
 328 a 500 m × 500 m rectangular domain with a 5 m water depth. Wall boundary conditions
 329 (fully reflective) are specified at the four boundaries of the domain. The initial still water
 330 level can thus be defined as:

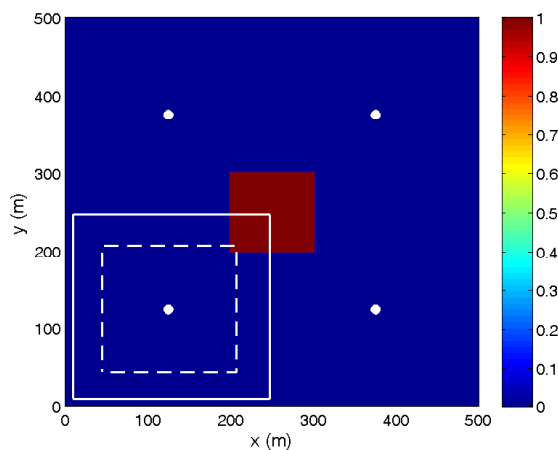


Fig. 5. Wave generation from an initial 1 m elevation still water hump. The parent Grid 1 covers the entire domain and solid/dashed lines mark the boundary of Grid 2/3. Bullets mark locations of numerical wave gauges for comparing free surface elevations. Color represents the initial surface elevation in meter.

$$\eta(x, 0) = \begin{cases} 1.0, & 200 \leq x \leq 300, 200 \leq y \leq 300, \\ 0.0, & \text{elsewhere.} \end{cases}$$

331 The consistency and accuracy of the two-way nested grid algorithm is first assessed by
 332 defining a three-level nested grid system with identical grid resolution $\Delta x = \Delta y = 2.5$ m
 333 in Cartesian coordinates, hence a grid refinement ratio $s = 1$. Grid 1 is the background



334 parent grid and Grids 2 and 3 are nested grids located in $10.0 \leq x, y \leq 247.5$ and 30.0
335 $\leq x, y \leq 207.5$, respectively. Because the refinement ratio is 1, the same numerical solution
336 is expected whether nesting is used or not. To verify this, surface elevation time series were
337 computed at 4 numerical wave gauges located at, $(x, y) = (125, 125)$, $(375, 125)$, $(125, 375)$,
338 and $(375, 375)$ m (Fig. 5). The bottom left gauge is located within the two nested grids
339 and its time series computed in Grid 3 is compared to those at the three gauges located in
340 Grid 1. Because of the symmetry of the initial solution, results at the four gauges should
341 be identical, which is verified in Fig. 6, hence assessing the consistency of the nested grid
342 model.

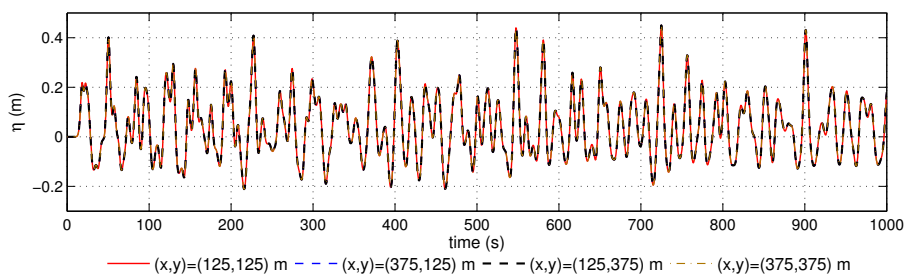


Fig. 6. Wave generation from an initial 1 m elevation still water hump. Comparison of surface elevation recorded at the four numerical wave gauges marked in Fig. 5. Nested grids have the same grid resolution of 2.5 m as the background parent grid.

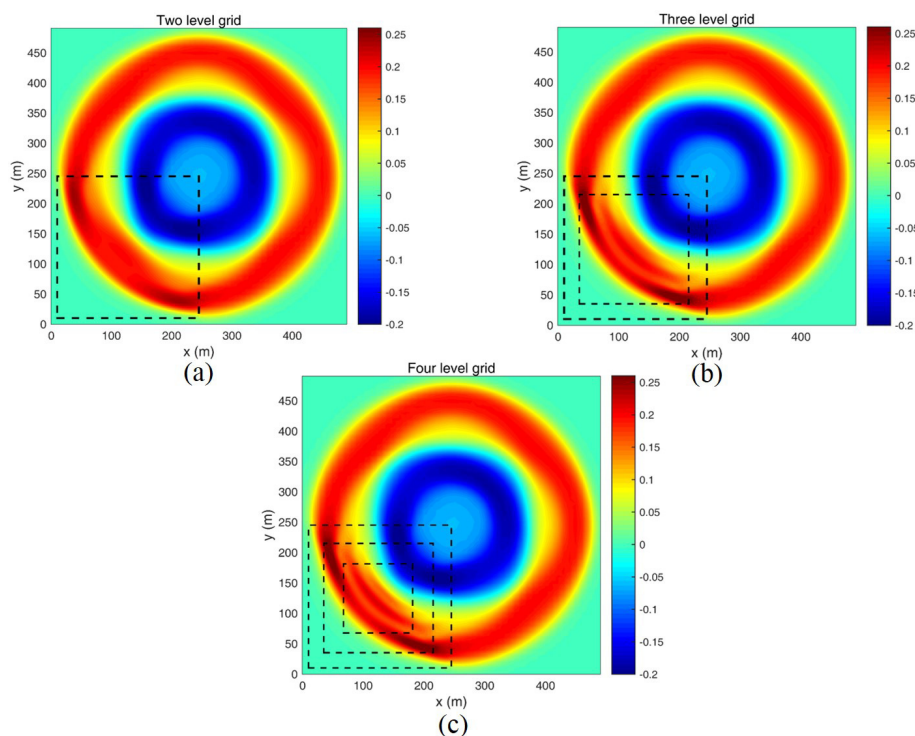


Fig. 7. Wave generation from an initial 1 m elevation still water hump. Snapshots of surface elevations computed at $t = 23$ s in: (a) two-level grids, (b) three-level grids, and (c) four-level grids, with discretization of 10, 5, 2.5, and 1.25 m, respectively ($s = 2$). Dashed lines denote the boundary of child grids.

343 Next, we examine the effects of grid refinement in a hierarchical nested grid system, for
 344 the same application. Here, the Grid 1 resolution is $\Delta x = \Delta y = 10$ m and Grid 2 and Grid 3
 345 are nested within Grid 1, as before, but with grid resolutions of 5 m and 2.5 m, respectively.
 346 Grid 2 is located in $10.0 \leq x, y \leq 245$ and Grid 3 in $35.0 \leq x, y \leq 215$. An additional grid,
 347 Grid 4 was added within Grid 3, located in $67.5.0 \leq x, y \leq 181.25$ with a grid resolution of
 348 1.25 m. Three computations were run using two- to four-level of nested grids, with surface
 349 elevations computed at $t = 23$ s shown in Fig. 7. Because wave dispersive effects are related
 350 to grid resolution, the solution in a finer grid is not exactly the same as in a coarser grid,
 351 resulting in asymmetric distributions of surface elevations in the figure. With a two-level
 352 nested grid system (Grids 1 and 2), Fig. 7a shows the appearance of sharper crests (dark



353 red) in Grid 2, as compared to the solution in Grid 1. As more levels of nested grids are
 354 used, Figs. 7b and c show that shorter waves increasingly appear in the finer grids.

355 4.2 Wave refraction-diffraction over a shoal on a sloping bottom topography

356 Although the main targeted applications of our new two-way grid nesting model system
 357 are tsunami simulations in multi-scale cases, the method can also be applied to the modeling
 358 of ocean wave transformations in coastal areas. This is demonstrated here by simulating the
 359 laboratory experiments of Berkhoff et al. (1982), for wave refraction-diffraction over a shoal
 360 on a 1/50 sloping bottom topography, both rotated by 20° off the y -axis (Fig. 8). This
 361 experimental dataset has served as a standard benchmark for assessing the accuracy and
 362 performances of numerical wave models for simulating wave shoaling, refraction, diffraction,
 363 and nonlinear dispersion. Shi et al. (2011) showed that the original version of FUNWAVE-
 364 TVD accurately reproduces measured wave heights in this experiment.

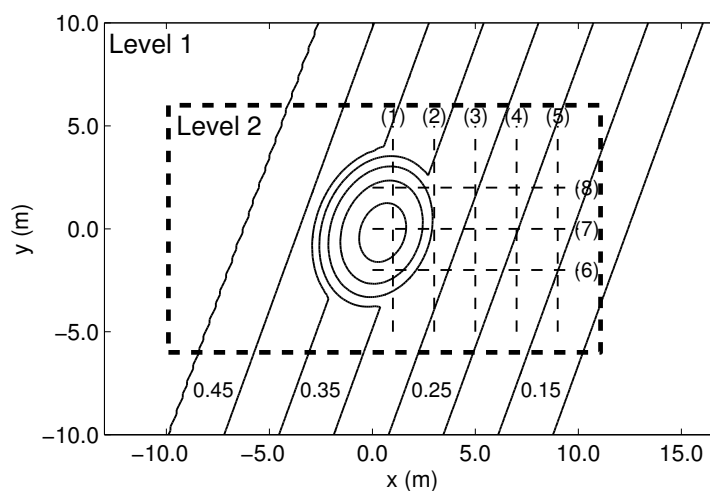


Fig. 8. Computational domain and bottom topography of Berkhoff et al.'s (1982) experiments of wave transformations over a tilted elliptical shoal on a sloping bottom. The bold dashed line box marks the area of nested Grid 2 (Level 2). Dashed lines (1) - (8) are transects for model/data comparisons.



365 Two numerical simulations were carried out for: (i) the original single grid model; and
366 (ii) a two-level nested grid model (Fig. 8). Both models are set up in a rectangular domain
367 with Cartesian coordinates, $-13 \text{ m} \leq x \leq 16.9 \text{ m}$ and $-10 \text{ m} \leq y \leq 10 \text{ m}$. The single
368 grid model has grid resolutions of $\Delta x = 0.025 \text{ m}$ and $\Delta y = 0.05 \text{ m}$. In the two-level grid
369 model, Grid 1 is coarser with $\Delta x = 0.1 \text{ m}$ and $\Delta y = 0.2 \text{ m}$ resolutions, and the finer Grid
370 2 is nested in the region of $-9.9 \text{ m} \leq x \leq 11.075 \text{ m}$ and $-6 \text{ m} \leq y \leq 6 \text{ m}$, with resolution
371 $\Delta x = 0.025 \text{ m}$ and $\Delta y = 0.05 \text{ m}$ identical to those of the single grid mode, corresponding to
372 a grid refinement ratio is thus $s = 4$. Total numbers of cells in Grid 1 and 2 are 30,300 and
373 202,440, respectively, which is much smaller than the 480,000 cells of the single grid model.

374 In both model setups, regular waves with a period $T = 1 \text{ s}$ and an amplitude $A = 4.64$
375 cm are generated, as in experiments, by a wavemaker located at $x = -10 \text{ m}$. Sponge layers
376 with a width of 2 m were specified on the left and right boundaries of both the single grid
377 domain and Grid 1 in the nested grid model.

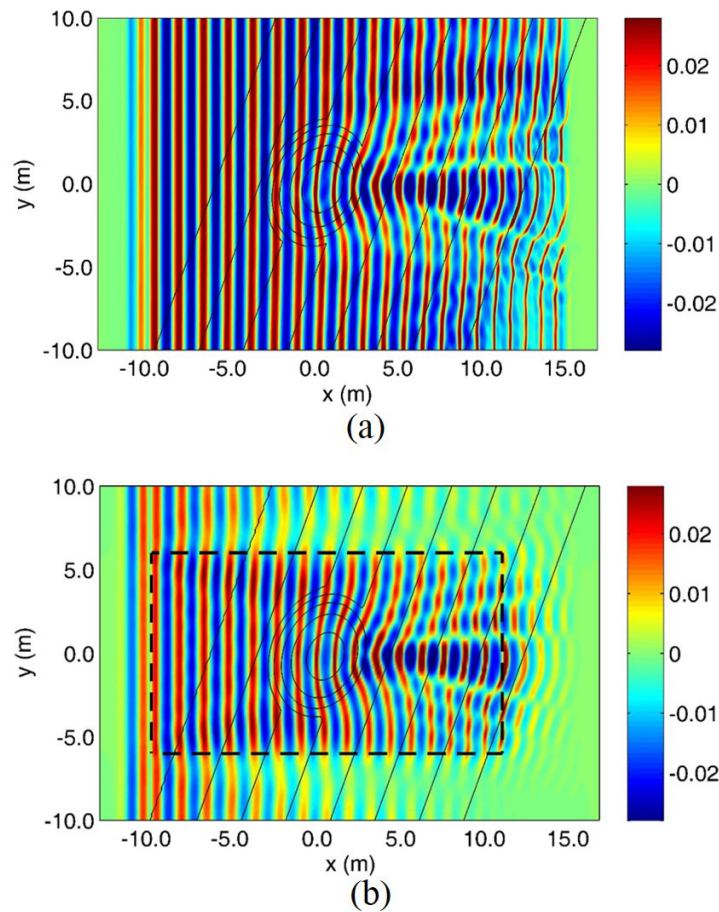


Fig. 9. Experimental benchmark of Fig. 8. Snapshots of free surface elevation (color scale in meter) at $t = 40$ s, computed in: (a) the single grid; and (b) the nested grid models.

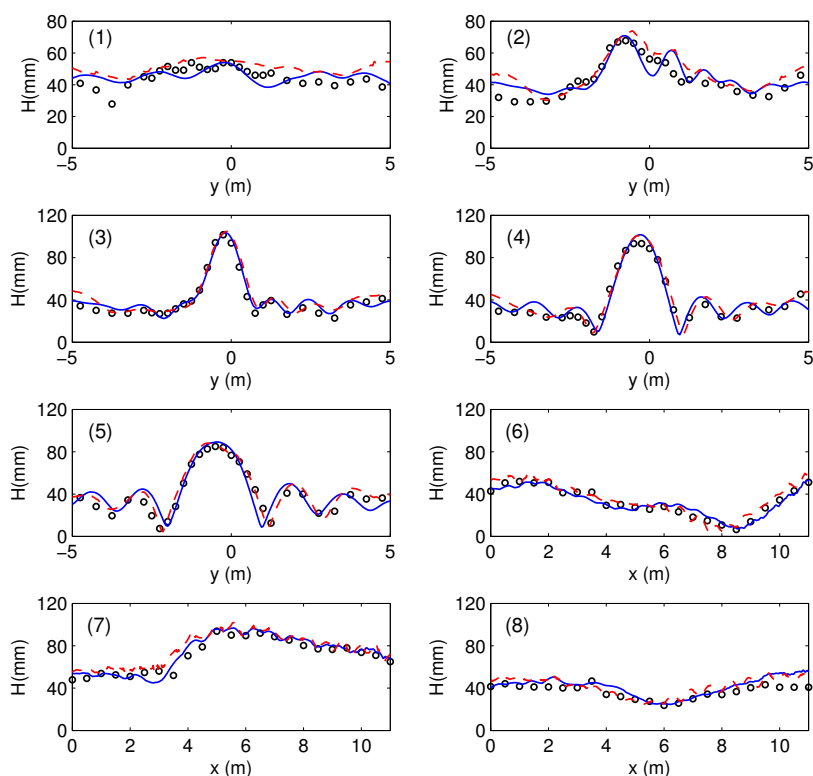


Fig. 10. Experimental benchmark of Fig. 8. Comparison of wave height distribution along transects (1)-(8) in: (o) experimental data and results of (dash) single grid, and (solid) nested grid model.

378 Fig. 9 shows snapshots of surface elevation computed at $t = 40$ s in both model
 379 setups. Compared to the single grid model, which uses the finest grid resolution over the
 380 entire domain, the nested grid model, which only uses it in Grid 2, shows that waves are
 381 numerically damped due to the coarse grid resolution used outside of Grid 2. Over the
 382 shoal and slope behind it, results in both the nested Grid 2 and the single grid show similar
 383 intense wave shoaling, refraction, and diffraction patterns. However, in the nested grid
 384 model, additional spurious wave diffraction effects can be seen around the lateral nesting
 385 boundaries due to the wave damping on the coarse grid side.

386 Fig. 10 shows comparisons of both model results with experimental data for the wave
 387 height variation along the transects marked in Fig. 8. For all transects, both the single



388 grid and nested grid model results agree well with the data. As expected from the spurious
 389 diffraction effects, compared to the single grid model, the nested grid model predicts slightly
 390 smaller wave heights at the ends of transects (1) – (5).

391 Regarding computational efficiency, in this application, the cost of the nested grid model
 392 is about 46.5 % that of the single grid model. It should be mentioned that this test is only
 393 for verification of the nested grid algorithm and is not a typical case for demonstrating the
 394 efficiency of the nested grid method.

395 4.3 Solitary wave runup on a shelf with an island

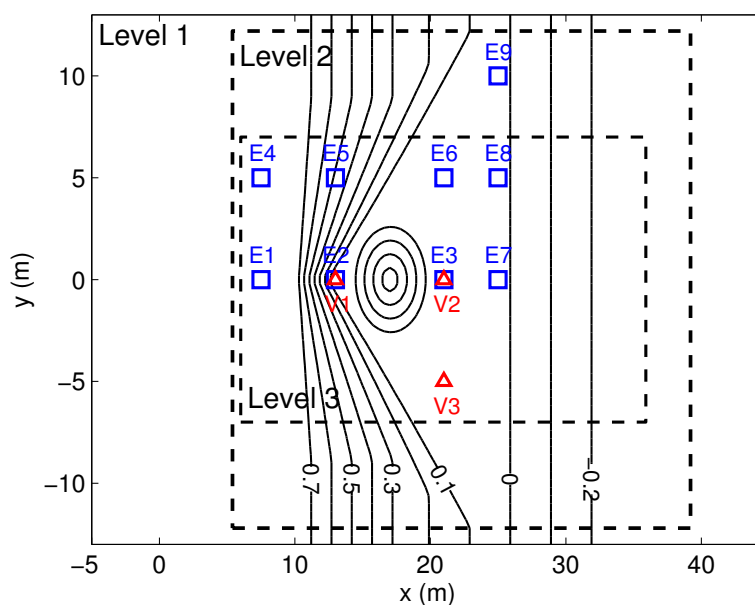


Fig. 11. Solitary wave runup on a shelf with an island (Lynett et al., 2010). Bathymetry contours (solid lines, in meters) and measurement locations in the computational domain. The parent Grid 1 (Level 1) covers the entire domain; blocks with dashed lines mark the boundaries of nested Grids 2 and 3 (Levels 2 and 3). Symbols mark locations of: (□) physical/numerical wave gauges E1–E9, and (△) velocimeters (ADVs), V1–V3.

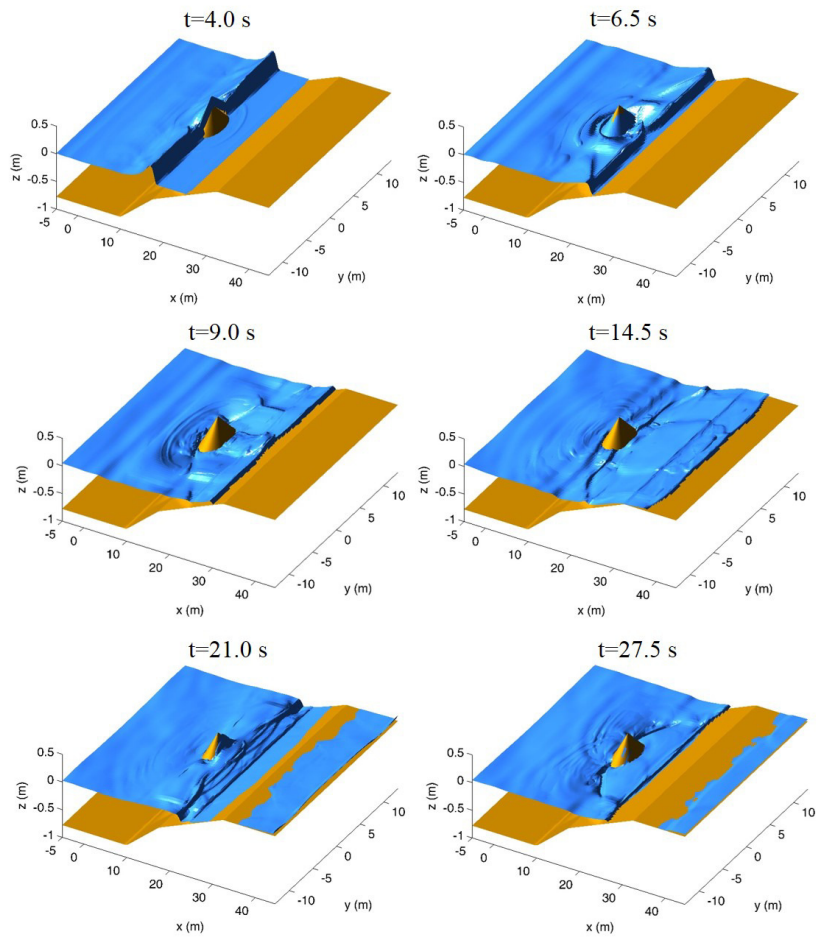


Fig. 12. Case of Fig. 11. Surface elevations simulated at $t = 4.0, 6.5, 9.0, 14.5, 21.0$ and 27.5 s in nested grid system.

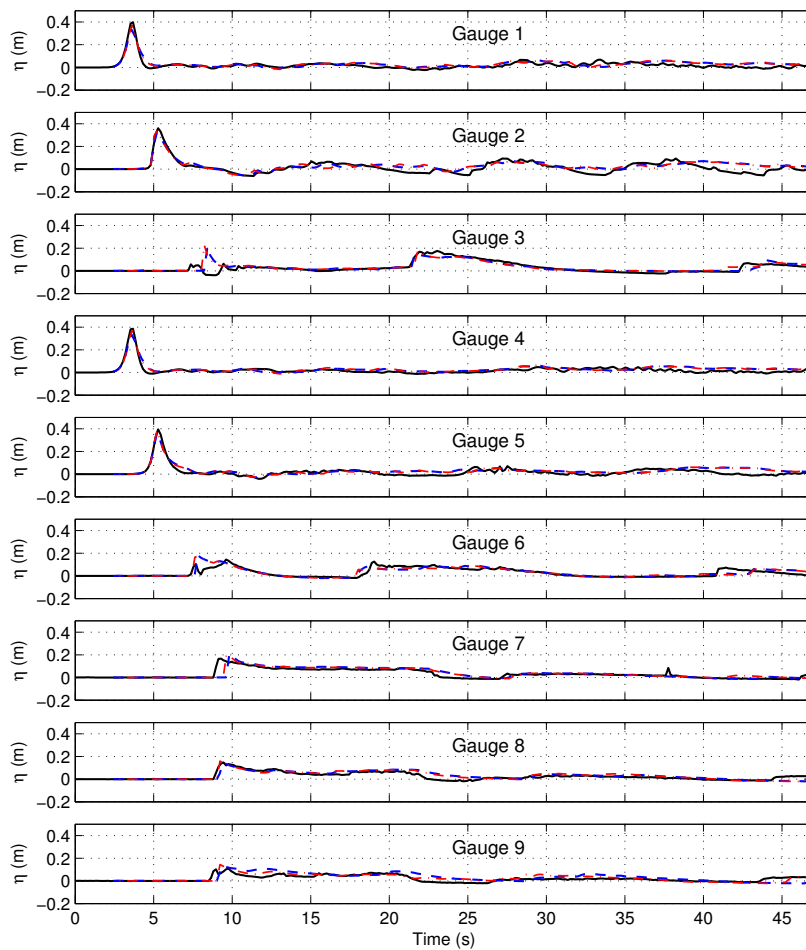


Fig. 13. Case of Fig. 11. Comparison of surface elevations at wave gauges in: (solid) experiments; (blue/red dash) present nested model/original FUNWAVE-TVD results.

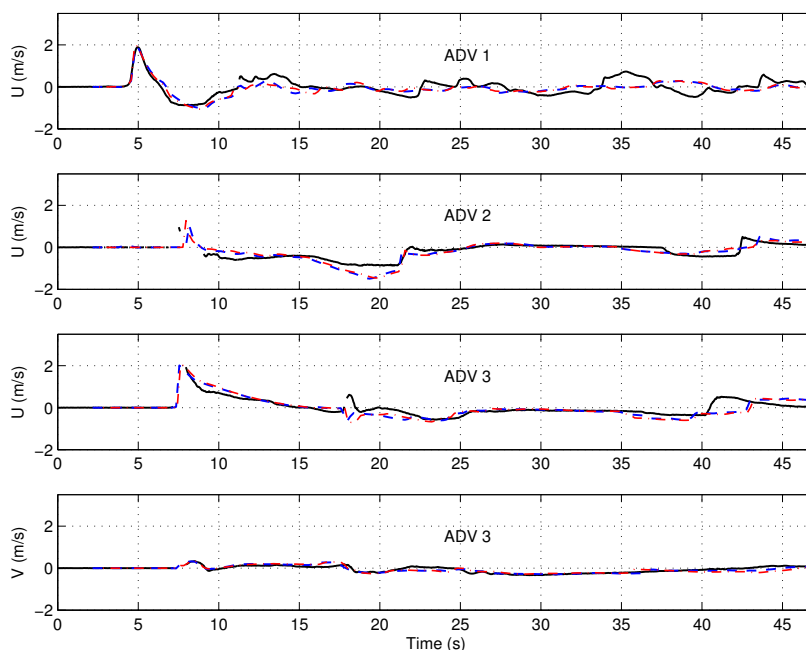


Fig. 14. Case of Fig. 11. Comparison of velocities at ADV locations in: (solid) experiments; (blue/red dash) present nested model/original FUNWAVE-TVD results.

396 A second experimental benchmark, for the runup of a solitary wave over a complex
 397 nearshore bathymetry with an island (Fig. 11), is simulated to assess the accuracy of the
 398 wetting and drying algorithm along the model shoreline, in a nested grid system. These
 399 experiments were performed in the large wave basin of Oregon State University's O.H. Hins-
 400 dale Wave Research Laboratory (Lynett et al., 2010). The 3D bathymetry was constructed
 401 in the 48.8 m long and 26.5 m wide basin, with a 2.1 m depth. It consists of a 1/30 plane
 402 slope connected to a triangular shelf with a conical island over the shelf (Figs. 11 and 12).
 403 Surface elevations were measured at nine locations using wave gauges (E1 – E9 in Fig. 11),
 404 and velocities were measured at three locations by Acoustic Doppler Velocimeters (ADVs)
 405 (V1 – V3 in Fig. 11). Details of the experiment can be found in Lynett et al. (2010). Shi
 406 et al. (2012) applied the original version of FUNWAVE-TVD to this case.

407 In the model simulations, a three-level nested grid system is set-up with Grid 1, Grid
 408 2, and Grid 3 shown in Fig. 12. The model setup for Grid 1 is similar to Shi et al.'s (2012),
 409 except that grid resolution is coarser, with $\Delta x = \Delta y = 0.4$ m, versus 0.1 m in the original



Table 1: Grid information for the solitary wave experiment.

Domain	x range (m)	y range (m)	$m_x \times m_y$	$\Delta x, \Delta y$ (m)
Level 1	-5.0 ~ 44.6	-13.0 ~ 13.0	66 × 125	0.4
Level 2	5.4 ~ 39.2	-12.2 ~ 12.2	123 × 170	0.2
Level 3	6.0 ~ 35.9	-7.0 ~ 7.0	141 × 300	0.1

410 model. The nested grids, Grid 2 and 3, have 0.2 m and 0.1 m resolution, respectively
 411 (hence, $s = 2$), and are centered in the middle of the domain where wetting and drying
 412 frequently occur due to the moving shoreline during runup. As measured in experiments,
 413 an incident solitary wave of height $H_o = 0.39$ m is specified in Grid 1, in the constant depth
 414 $h_o = 0.78$ m region on the left side of the model, from $-5 \text{ m} < x < 5 \text{ m}$, with its crest
 415 initially located at $x = 0$. The initial solitary wave condition is based on Nwogu’s extended
 416 Boussinesq equations (Wei, 1997). With $H_o/h_o = 0.5$ this represents a strongly nonlinear
 417 incident wave. A summary of the nested grid configuration is given in Table 1.

418 Fig. 12 shows snapshots of surface elevations simulated in the nested grid model,
 419 constructed using results from all grids, wherever the highest resolution results are available.
 420 Results show successively that wave breaking occurs at $t = 4.0$ s, edge wave collide behind
 421 the island at $t = 6.5$ s, a breaking bore forms at $t = 9.0$ s, with its front running-up and
 422 -down the upper slope and beach terrace, from $t = 9.0$ to 27.5 s. These are all quite complex
 423 processes that appear well-resolved in the nested grids.

424 Fig. 13 shows the comparison of model results with experimental data for surface
 425 elevations measured at the nine wave gauges (E1 – E9 in Fig. 11). Results from the single
 426 grid model (with grid resolution 0.1 m) are also plotted in the figure for comparison. Surface
 427 elevations simulated in the new nested grid model are quite close to those in the original
 428 single grid model, and both agree well with the experimental data. Slight differences between
 429 the nested grid and single grid models can be seen at Gauge 9, likely because this gauge is
 430 located in Grid 2, for which the resolution is lower than that in the single grid model; all
 431 the other gauges are located in Grid 3 which has the same resolution as the original single
 432 grid model.



Table 2: Grid parameters for the 2011 Tohoku-Oki tsunami simulation.

Domain	Range of longitude ($^{\circ}$ E)	Range of latitude ($^{\circ}$ N)	$m_x \times n_y$	Resolution (arc-min)
Level 1	132.0000 \sim 292.0000	-60.0000 \sim 60.0000	2400 \times 1800	4
Level 2	205.0833 \sim 238.4000	29.7500 \sim 49.7333	2000 \times 1200	1
Level 3	221.7000 \sim 236.6917	37.2000 \sim 45.5250	1800 \times 1000	1/2
Level 4	232.9250 \sim 236.2542	40.6750 \sim 43.1708	800 \times 600	1/4
Level 5	234.8292 \sim 235.9521	41.4750 \sim 42.4313	540 \times 460	1/8
Level 6	235.6563 \sim 235.9052	41.5833 \sim 41.8531	240 \times 260	1/16
Level 7	235.7677 \sim 235.8401	41.6844 \sim 41.7359	140 \times 100	1/32

433 Fig. 14 similarly compares time series of simulated and measured mean horizontal
434 velocity at 3 ADVs (V1 – V3 in Fig. 11). Results from the nested grid model are all close
435 to those of the original single grid model and both agree well with the data. We note that
436 all ADVs are located in Grid 3, which has the same grid resolution as the original single
437 grid model, hence results of both models are expected to be consistent.

438 4.4 Tohoku-Oki 2011 tsunami impact on Crescent City harbor, CA

439 As mentioned in the introduction, the multi-scale modeling of transoceanic tsunamis
440 is a typical application of the nesting grid technique. Tehranirad et al. (2020) used the
441 one-way nesting technique with six-level nested grids to simulate the impact of the Tohoku-
442 Oki 2011 tsunami, particularly morphological changes, in Crescent City Harbor, CA. This
443 harbor is known for its vulnerability to tsunamis due to wave-guiding effects caused by a
444 ridge feature in the bottom topography of the Pacific Ocean (Grilli et al., 2013). During the
445 2011 tsunami, Crescent City Harbor experienced extensive damage caused by a significant
446 inundation, but most of all strong currents induced within the harbor by successive long
447 waves in the incoming tsunami wave train. Tsunami-induced oscillations of the harbor, and
448 currents, were reported to have lasted for several days in the harbor (Wilson et al., 2012),
449 due to nearshore edge waves associated with the tsunami event. When using that many
450 levels of grids, the multi-scale modeling using the one-way nesting technique is particularly
451 cumbersome, in terms of the manual post processing it involves. Hereafter, we repeat this



452 simulation using the new two-way nesting framework. Unlike the three earlier tests, which
 453 used the Cartesian mode, this test uses spherical coordinates.

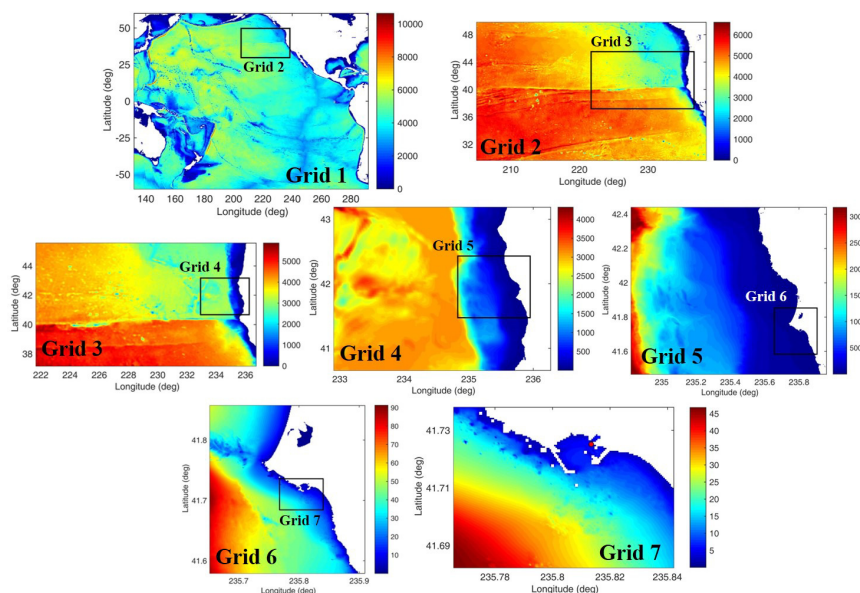


Fig. 15. Nested grid simulation of the 2011 Tohoku-Oki tsunami. Bottom topography (color scale in meter) and computational domains for Grids 1 to 7. Red circles in Grid 7 denote numerical wave gauge locations.

454 The nested grid system uses seven levels, with grid resolutions varying from 4 arc-min
 455 at the ocean basin scale to 1/32 arc-min around the harbor, and a nesting ratio $s = 2$.
 456 As shown in Fig. 15, with a high resolution of 1/32 arc-min in Grid 7 (or about 53 m),
 457 the model is able to resolve the harbor structures quite well. Following Tehranirad et al.
 458 (2020), the bathymetry used to define the model grids was constructed by combining 1 arc-
 459 min ETOPO-1 data (Amante and Eakins, 2009), 3 arc-second Coastal Relief Model (CRM)
 460 data (NGDC, 2003), and the local 10 m resolution tsunami DEM of Crescent City Harbor
 461 (Grothe et al., 2011). The tsunami was generated using the same source configuration as in
 462 Grilli et al. (2013) and Kirby et al. (2013). Model parameters were specified according to
 463 Tehranirad et al. (2020). Table 2 summarizes the locations, dimensions, and grid sizes of
 464 the nested grids.



465 Fig. 16 show snapshots of tsunami surface elevations computed in the basin-scale Grid
466 1 and the nested grids, Grid 5 and Grid 7, at $t = 11.2$ and 11.4 hr, when the water surface
467 elevation within the harbor reaches its maximum and minimum levels, respectively (Fig.
468 17). The model shows the generation of edge waves propagating along the coast (Grid 5),
469 which were not simulated in Tehranirad et al.'s (2020) one-way nesting computations. The
470 two-way nesting is a more relevant technique to model waves propagating across nesting
471 boundaries, without significant wave reflection from the boundaries.

472 Fig. 17 compares the modeled surface elevation with the data measured at the gauge
473 location within the harbor (red circle in Grid 7 in Fig. 15). Following Tehranirad et al.
474 (2020), the model result were shifted by 8 minutes backward to compensate for the time
475 delay identified in earlier studies, which was possibly caused by compressibility and earth
476 elasticity effects (Allgeyer and Cummins, 2014, Wang, 2015, Abdolali and Kirby, 2017,
477 Abdolali et al., 2019). Overall, the model shows a good agreement with the data, although
478 the largest wave crests are slightly over predicted.

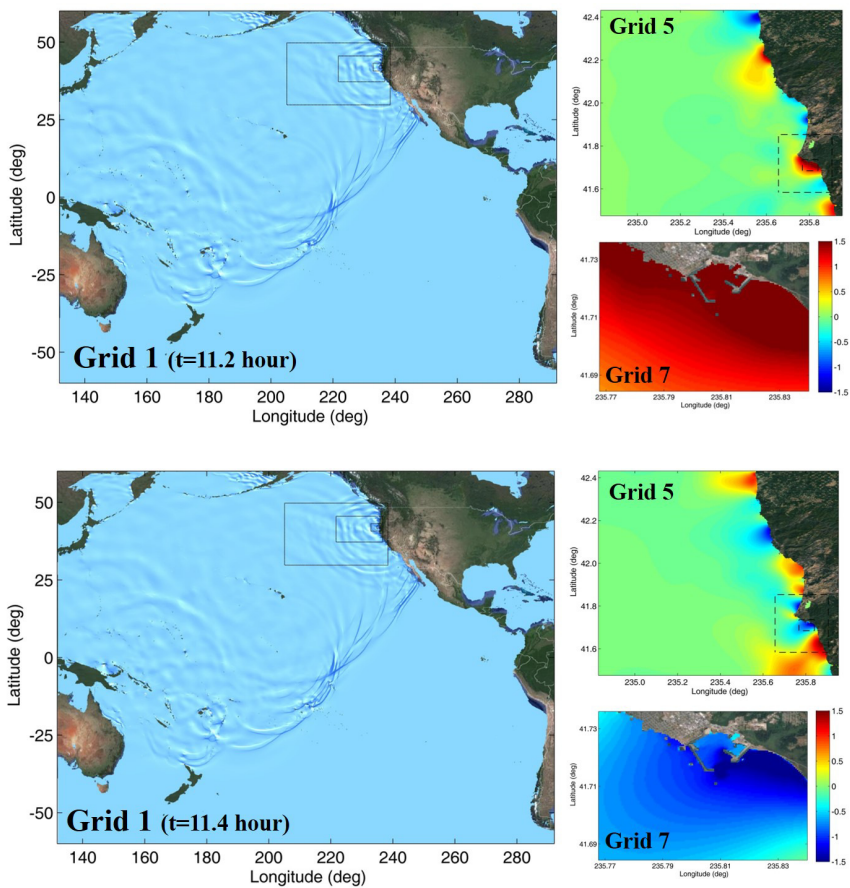


Fig. 16. Same case as Fig. 15. Snapshots of tsunami surface elevations simulated in Grid 1, Grid 5, and Grid 7 at $t = 11.2$ and 11.4 hr.

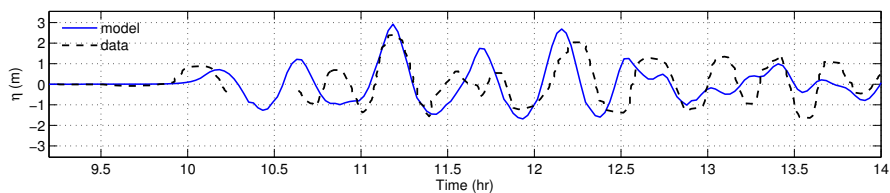


Fig. 17. Same case as Fig. 15. Comparison between model result and measured data inside the Crescent City harbor.



479 **5 Conclusions**

480 The main goal of this study was to develop a multigrid nesting interface for the Boussi-
481 nesq wave model, FUNWAVE-TVD, which can be used as a MASTER program to manage
482 time sequencing and nesting processes, and make it both easier and more accurate and effi-
483 cient performing multi-scale tsunami simulations. The background model couples a series of
484 submodels with grid refinement in a hierarchical manner. Unlike other AMR-type models,
485 the new modeling framework does not alter the original solver, and hence FUNWAVE-TVD
486 can still be used as a stand-alone program for each individual grid.

487 The nesting algorithm performs a two-way coupling between the parent and child grids.
488 The child grid is driven by the boundary conditions provided by the parent grid. Linear
489 interpolators are performed both in time and space at the ghost cells of nesting boundaries.
490 The parent grid is updated with results from the child grids using a linear restriction oper-
491 ator. No correction of mass and momentum is needed during the nesting process because
492 of the use of conservative forms of mass and momentum equations.

493 Workload balance is handled by an equal-load scheme, which performs the same domain-
494 decomposition algorithm on all grid-levels using the same number of processors, guaranteeing
495 equal CPU-load over the entire computation. Communication between the parent and child
496 grids is direct without a data-gathering process. The parent-child proximity is pre-calculated
497 at the beginning of the model run and, hence, does not cause additional computational cost.
498 A strategy of shared array allocations is used in data management. Grids at all levels share
499 the same memory allocations, and no additional memory allocation is required, allowing for
500 a large number of nesting levels to share the same memory allocation.

501 The nested grid model was verified on four applications, three of which are standard
502 benchmarks and one is a tsunami case study. The numerical test of wave evolution from
503 a rectangular hump examined the consistency and general performance of the nesting al-
504 gorithm. The simulation of Berkhoff et al.'s (1982) experiment showed that the model is
505 capable of simulating surface waves and their transformation in shallow water, which in-
506 volves dispersive and nonlinear effects. The simulation of experiments for solitary wave
507 runup on a shelf with an island was used to assess the accuracy of the wetting and drying
508 processes in the nested grid system. The last application, the simulation of the Tohoku-Oki
509 2011 tsunami and its effects on Crescent City Harbor, CA, demonstrated the robustness of



510 the two-way nesting model for the multi-scale modeling of ransoceanic tsunamis and their
511 coastal effects.

512 Future work will include the development of an interface for the GPU version of
513 FUNWAVE-TVD and of an adaptive mesh refinement algorithm for the nesting framework.

514 *Code availability and data availability.* The computer code, all examples illustrated in the
515 paper, MATLAB post-processing scripts, and data used in this research are archived at
516 (<http://doi.org/10.5281/zenodo.4735599>).

517 *Author contributions.* The conceptualization and methodology of the study were developed
518 by Shi and Choi. The code was implemented by Choi. Code validation was conducted by
519 Malej. Smith, Kirby, and Grilli performed result analysis and paper revision.

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

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