

# Bidirectional River-Floodplain Connectivity During Combined Pluvial-Fluvial Events

Nelson Tull<sup>1</sup>, Paola Passalacqua<sup>1</sup>, Hima J. Hassenruck-Gudipati<sup>2</sup>, Shazzadur  
Rahman<sup>1\*</sup>, Kyle Wright<sup>1</sup>, Jayaram Hariharan<sup>1</sup>, and David Mohrig<sup>2</sup>

<sup>1</sup>Department of Civil, Architectural, and Environmental Engineering, Center for Water and the  
Environment, University of Texas at Austin, Austin, TX, USA

<sup>2</sup>Department of Geosciences, Jackson School of Geosciences, University of Texas at Austin, Austin, TX,  
USA

## Key Points:

- Field data and modeling show distinct signals of pluvial and fluvial flooding
- Floodplain residence times change dramatically between pluvial and fluvial phases of the storm
- Numerical model resolution has a local effect on floodplain channel conveyance, but has little influence on overall residence times

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\*Walter P. Moore, Dallas, TX

Corresponding author: Paola Passalacqua, [paola@austin.utexas.edu](mailto:paola@austin.utexas.edu)

**Abstract**

Hydrological connectivity controls the lateral exchange of water, solids, and solutes between rivers and floodplains, and is critical to ecosystem function, water treatment, flood storage and attenuation, and geomorphic processes. Hydrological connectivity has been studied extensively, typically through the lens of river-driven flooding, but in regions prone to heavy rainfall, the timing and magnitude of lateral exchange may be altered dramatically by local flooding on the floodplain. We collected field measurements of flow depth and velocity in the Trinity River floodplain in coastal Texas (USA) during Tropical Storm Imelda (2019), which produced up to 75 cm of rainfall locally. We developed a two-dimensional, depth-averaged hydrodynamic model at high resolution for a section of the Trinity River to replicate floodplain hydrodynamics and determine the impact of floodplain channels on connectivity. We employed Lagrangian particle routing to quantify how residence times and particle velocities changed as flooding shifted from rainfall-driven to river-driven. Our results show that heavy rainfall initiated lateral exchange before river discharge reached flood levels. The presence of rainwater also reduced floodplain storage, causing river particles to be confined to a narrow corridor on the floodplain, while rainwater residence times were increased from the backwater effect of river flow through the floodplain. Finally, while the resolution of floodplain channels is important locally, it does not affect as much the overall floodplain behavior. This study provides evidence of the added complexity of floodplain hydrodynamics under conditions of heavy rainfall, with implications for sediment deposition and nutrient removal during floods.

**Plain Language Summary**

Unaltered river floodplains can support diverse ecosystems, reduce flooding, and remove nutrients from river water. Floodplains near the coast are particularly important, as they typically experience more frequent flooding. Floodplain function relies on a high degree of connectivity with the river, where water can move easily through the floodplain during high river stages. Our study explores the ways in which heavy rainfall on a floodplain impacts this connectivity. We collected flow measurements in the Trinity River floodplain (Texas, USA) during Tropical Storm Imelda in 2019 that show how flow can change direction as a flood transitions from rainfall-driven to river-driven. We coupled a hydrodynamic model with a particle tracking module to see how particles in the water might move through the floodplain during this transition. We found that the average time a particle spent in the floodplain changed significantly after the rain in the model stopped. We also noticed that rainwater tended to remain in the floodplain for much longer than river water, especially after the rain stopped. This study describes the various interactions that can occur between local rainfall and river flooding, and moves toward a better understanding of sediment and nutrient transport through floodplains.

**1 Introduction**

River floodplains play a fundamental role in flood storage, nutrient cycling, sediment retention, and in general provide support for diverse ecosystems (Ward et al., 1999; Melack & Forsberg, 2001; Kondolf et al., 2006; Roley et al., 2012; Noe et al., 2013; Kufel & Leśniczuk, 2014). Floodplains are complex and heterogeneous, and their structure and function are highly dependent on their degree of connectivity with the river (Hughes et al., 2001; Harvey & Gooseff, 2015; Gurnell et al., 2016; Covino, 2017). Topographic and hydrologic controls on connectivity have been described for the largest river-floodplain systems using satellite imagery (Lesack & Melack, 1995; Mertes et al., 1995; Mertes, 1997; Alsdorf et al., 2007; Trigg et al., 2012; Lewin & Ashworth, 2014; Park & Latrubesse, 2017), and more recently for medium-size rivers using lidar data and numerical modeling (David et al., 2017; Czuba et al., 2019; Byrne et al., 2019). However, river-floodplain connectivity is poorly understood when hydrodynamics are partially driven by local rainfall. Mixing of rainfall and river floodwaters

65 has been observed and discussed for very large river floodplains (Mertes, 1997; Alsdorf et al.,  
66 2007; Day et al., 2008; Rowland et al., 2009; Trigg et al., 2012), where floodplain channels  
67 are at a large enough scale to be sensed remotely, and the flood wave occurs over much  
68 longer time scales (Junk et al., 1989). But for medium-size rivers, flood waves are less  
69 predictable, and floodplain features are often too small to detect using satellite imagery.  
70 The goals of this study are to show (a) the impact of local rainfall on floodplain residence  
71 times, flow directions, and connectivity within the floodplain, and (b) to determine the role  
72 of floodplain channel scale in facilitating river-floodplain exchange.

73 Floodplain topography has been shown as a key control on mixing of local runoff  
74 and river waters in large floodplain systems, where most mixing tends to occur outside of  
75 channel features (Lesack & Melack, 1995; Mertes et al., 1995; Mertes, 1997; Trigg et al.,  
76 2012). Flow within these floodplain channels can be bidirectional due to the advancing  
77 and receding of the flood wave and the timing of rainfall runoff on the floodplain (Alsdorf  
78 et al., 2007; Day et al., 2008; Rowland et al., 2009). For smaller river systems as well,  
79 sub-bankfull discharges can result in floodplain inundation that is limited spatially by the  
80 extent of lateral floodplain channels in the system (Kupfer et al., 2015; Czuba et al., 2019).  
81 Pluvial flooding may enhance connectivity within the floodplain by bringing inundation,  
82 and potentially nutrients and sediment, to areas of the floodplain that would otherwise be  
83 out of reach for river waters. On the other hand, if intense enough, pluvial flooding can  
84 develop a water surface gradient moving from floodplain to channel, which may reduce flux  
85 into the floodplain (Day et al., 2008). It is common that fluvial and pluvial flooding at a site  
86 are not coincident in time, as a result of a storm moving slowly over a watershed, and thus  
87 the interaction between the two flooding modes may be complex in space and time. Recent  
88 studies of river-floodplain connectivity have used unsteady numerical models to show how  
89 floodplain hydrodynamics evolve with the rising and falling of a river flood wave (Byrne  
90 et al., 2019; Chen et al., 2020), but no study has used numerical modeling to analyze the  
91 interaction of fluvial and pluvial flooding. Furthermore, no study to-date has presented field  
92 measurements of floodplain flow that show this interaction.

93 The interaction of pluvial and fluvial flooding may have significant impacts on residence  
94 time, flow direction, and the overall extent of hydrological connectivity in river-floodplain  
95 systems, all of which can be drivers of dissolved nutrient sequestration (Mann & Wetzel,  
96 1995; Tockner et al., 1999; Aufdenkampe et al., 2011; Noe & Hupp, 2005; Noe et al., 2013;  
97 Wolf et al., 2013; Cheng & Basu, 2017) and sediment deposition (Tockner et al., 1999;  
98 Verhoeven et al., 2001; Schulz et al., 2003; Day et al., 2008; Trigg et al., 2012; Juez et al.,  
99 2019) in floodplains. Sediment deposition depends on local availability from the river, as well  
100 as flow velocity distributions across the floodplain to advect the sediment (Marriott, 1992;  
101 Asselman & Middelkoop, 1995), while dissolved nutrients require sufficient contact time to  
102 be removed from floodwaters via biogeochemical processes (Tockner et al., 1999; Noe et al.,  
103 2013; Cheng & Basu, 2017). The depositional environments of lowland river floodplains  
104 are understood to provide conditions conducive to these processes, yet it is unknown how  
105 conditions change when pluvial flooding is substantial.

106 In this study we show how pluvial flooding impacts residence time distributions and  
107 flow patterns in a low-gradient river-floodplain system by using the lower Trinity River  
108 (Texas, USA) as a study site. To our knowledge this is the first modeling study of pluvial  
109 flooding in the context of hydrological connectivity. We present flow depth and velocity  
110 measurements collected during Tropical Storm Imelda (2019) in the Trinity River flood-  
111 plain that show flow reversals as floodplain inundation transitioned from rainfall-driven to  
112 river-driven. We then develop several two-dimensional, depth-averaged numerical models  
113 with varying grid resolution to replicate the observed hydrodynamics during the storm and  
114 determine the role of floodplain channel scale on river-floodplain connectivity. Next, we  
115 employ a Lagrangian particle routing tool on the unsteady model flow fields to quantify  
116 how rainfall and channel processes impact residence time distributions and flow patterns in  
117 floodplains.

118 The outline of the manuscript is as follows. Section 2 describes the characteristics of  
119 the lower Trinity River study site, including a description of the elevation data used for  
120 this study. Section 3 introduces Tropical Storm Imelda, the test-case event, and the hydro-  
121 dynamic data collected in the Trinity River floodplain in 2019 during the storm. Section  
122 4 introduces the ANUGA (Eulerian) and *dorado* (Lagrangian) models, and describes the  
123 modeling approach. Section 5 presents the results of the study, including the unique impacts  
124 of rainfall on floodplain hydrodynamics. Section 6 provides a discussion of implications for  
125 floodplain services and for future modeling studies of river-floodplain connectivity. Lastly,  
126 Section 7 summarizes the major findings of the study.

## 127 2 Study Area: The Trinity River

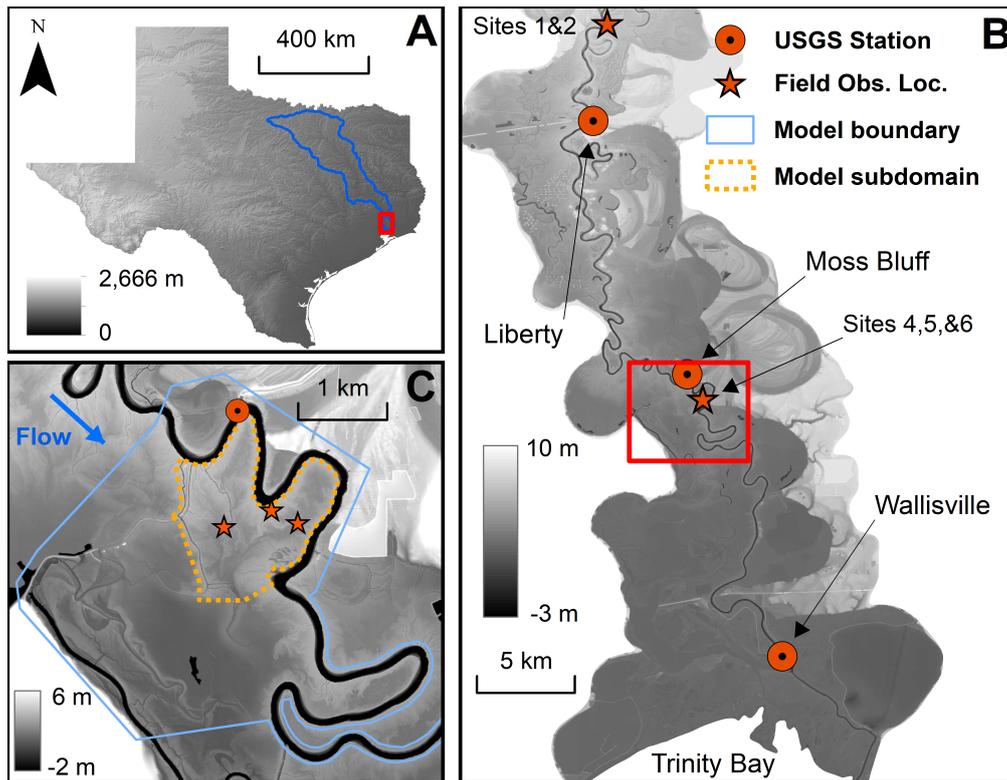
### 128 2.1 Site Description

129 The Trinity River basin (40,000 km<sup>2</sup>) extends from its outlet in Trinity Bay into north-  
130 central Texas (Figure 1A). The area of investigation spans about 10.5 river kilometers (rkm)  
131 of the lower basin in Liberty County, between Liberty and Wallisville, TX. The study area is  
132 within the river’s backwater reach (BWR), which is recognized by the asymptotic approach  
133 of the water surface elevation to the surface elevation of the receiving basin. Under low-  
134 discharge conditions, the BWR begins approximately 15 rkm upstream of the study area  
135 (Figure 1B) (Mason & Mohrig, 2018). As is typical in such systems, stage change between  
136 low and flood flows is smaller in the BWR, with variations in reach-average flow velocity  
137 primarily accommodating water discharge changes.

138 As the river transitions from a normal flow regime to the BWR, the channel mor-  
139 phodynamics responds to the adjustments in flow conditions and the transport of solids.  
140 This transition coincides with downstream narrowing and deepening of the channel. Rates  
141 of channel-bend migration decrease in the downstream direction, as do the size and shape  
142 of point bars, and grain size of bed material (Smith et al., 2020). Similarly, the overbank  
143 conditions vary in accordance with the transition to the BWR. For example, upstream of  
144 the BWR the floodplain is active and largely remains dry during low and moderate flow  
145 conditions, but is inundated during floods, when water emerges from channel confinement  
146 and spreads across the floodplain. In contrast, the BWR is characterized by a wetland  
147 environment due to its relatively low elevation and is prone to inundation by moderate river  
148 discharges. Because of these characteristics, connectivity is greater in the study reach com-  
149 pared to upstream, and floodplain channels are more commonplace and pronounced. This  
150 connectivity causes a “leakiness” in the system, which is supported by the large decrease in  
151 average annual peak discharge from Liberty (2,477 m<sup>3</sup>/s) to Wallisville (756 m<sup>3</sup>/s) observed  
152 between years 2015 and 2020.

153 Like many fluvial-deltaic systems worldwide, the Trinity is not free of anthropogenic  
154 influences. However, the study reach is unaffected by significant modifications such as  
155 containment levees, wing dykes, and revetments; thus, the river is able to operate unhindered  
156 within its valley. The Livingston Dam (upstream of our study area) is a run-of-river dam  
157 that impacts the river geomorphology for the first 50-60 rkm downstream of the structure  
158 (Phillips et al., 2004; Phillips & Slattery, 2007; Smith & Mohrig, 2017). Beyond this  
159 point, sediment mining from the bed and banks of the river re-establishes the bed-material  
160 load (Smith & Mohrig, 2017) and no change in channel geometry and kinematics has been  
161 observed since reservoir filling.

162 The model domain boundary shown in Figure 1C was chosen for three main reasons.  
163 First, the domain contains three of the field observation sites (see Section 3.2) that recorded  
164 data during Tropical Storm Imelda. Second, and related to the first, the floodplain topog-  
165 raphy in this area features many channels of various sizes that connect the river to the  
166 floodplain. Floodplain channel sizes range from small, 1-2 m channels barely detectable  
167 in the lidar, to larger, 8-10 m channels that are lower in elevation and most likely play



**Figure 1.** Elevation maps of the Trinity River study area. (A) Location of the Trinity River basin in Texas. (B) Lidar data for Trinity River floodplains between Liberty (upstream) and Wallisville (downstream), including locations of field observation sites. (C) Boundary of model domain used in this study, along with locations of three field observation sites within the study area (Sites 4, 5, and 6, from left to right). The downstream boundary of the model domain extends to Wallisville, but is not shown here.

168 a larger role in river-floodplain exchange. The complex floodplain topography makes this  
 169 location interesting to study. Third, the domain boundary needed to be limited in space,  
 170 as the high-resolution modeling needed to resolve the smallest channels requires significant  
 171 computational resources.

## 172 2.2 Elevation Data

173 All elevation data and references to elevation in this manuscript are relative to the  
 174 NAVD88 datum. The elevation data shown in Figure 1 were derived from lidar measure-  
 175 ments collected in February and March of 2017 as part of the Texas Strategic Mapping  
 176 Program. Data were acquired and processed by the Sanborn Map Company with third  
 177 party quality assurance and control provided by AECOM. Collection took place during the  
 178 leaf-off season in Texas. The reported horizontal and vertical accuracy of the lidar are 0.25  
 179 and 0.29 m, respectively.

180 The lidar data were interpolated to a bare-earth digital elevation model (DEM) at 1-m  
 181 resolution. Small voids in the floodplain lidar were interpolated using a second-degree poly-  
 182 gon plane fit through the existing data. Larger voids corresponding to floodplain ponds, ma-  
 183 jor channels, and oxbows were interpolated similarly but subtracted by 5 m as a bathymetry  
 184 approximation. River bathymetry measurements were taken by the Trinity River Authority

185 in 2017, along four longitudinal profiles at transects spaced every 400 m on average (the  
186 river width varies between 80 and 100 m). The bathymetry was interpolated to a 10-m  
187 grid, and patched together with the lidar DEM using the Raster to Mosaic tool in ArcGIS.  
188 Finally, the boundary between the datasets was linearly interpolated between the edge of  
189 the lidar data and the closest longitudinal bathymetry profile.

## 190 **3 Tropical Storm Imelda**

### 191 **3.1 Storm Background**

192 Tropical Storm Imelda (2019) was a major rainfall event that produced over 75 cm of  
193 precipitation across several counties in the area surrounding Houston, TX (Latto & Berg,  
194 2020). Imelda made landfall near Freeport, TX on 17 September 2019 as a tropical storm,  
195 before quickly weakening to a tropical depression as it moved slowly northward through  
196 Houston, TX and subsequently across the lower Trinity River watershed. The storm further  
197 degenerated to a trough by 19 September, at approximately 160 km north-northeast of  
198 Houston, where it continued to dissipate and move northward. The highest recorded rainfall  
199 total from Imelda was 112 cm, which made it the fifth wettest tropical cyclone ever recorded  
200 in the contiguous United States.

201 Eastern Texas experienced widespread pluvial flooding during this period. While the  
202 lower Trinity River watershed received much of this rainfall, including up to 75 cm at  
203 the study site, the river stage at the USGS gaging station at Liberty peaked just below the  
204 official flood stage as defined by the National Weather Service. A sub-bankfull flood event is  
205 suitable for analyzing river-floodplain connectivity, as floodplain channels are activated and  
206 responsible for any lateral exchange that occurs, and floodplain inundation is heterogeneous.  
207 The timing mismatch of the pluvial and fluvial flooding peaks, along with the sheer volume  
208 of precipitation, created an opportunity for competition between river and floodplain water  
209 worth investigating.

### 210 **3.2 Field Data Collection**

211 During August 2019, six outdoor trail cameras, six measuring rods, six water level  
212 loggers, and four tilt current meters were installed at various floodplain locations along the  
213 Trinity River (Figure 1). The loggers recorded water level every six minutes, while the  
214 tilt current meters recorded flow speed and direction every minute. The cameras took a  
215 photograph of the installation plus measuring rod every five minutes, night and day. The  
216 instruments were left in the field until February 2020, and successfully collected data during  
217 Tropical Storm Imelda. Sites 1, 2, and 3 were located north of Liberty near the bend  
218 indicated in Figure 1B. Site 1, located in a large floodplain channel, collected water level  
219 velocity readings, while Site 2, located in a shallow levee-traversing channel, collected water  
220 levels only. Instruments at Site 3 were destroyed during the storm. Sites 4, 5, and 6 were all  
221 located in the study area shown in Figure 1C. Sites 4 and 6 collected both water level and  
222 velocity readings, while Site 5 collected water levels only. Site 4 instruments were located  
223 in a floodplain channel roughly 930 m from the Trinity River (measured along the channel).  
224 Site 5 was located on a different floodplain channel, just 50 m from the river. Site 6 was  
225 located at the terminus of the same channel monitored by Site 5. At Site 6 a small internal  
226 delta was building out from the mouth of the floodplain channel into the adjacent, small  
227 floodplain basin with perennial standing water. Sites 5 and 6 were located on a channel  
228 connected to the river bend immediately upstream from the USGS gaging station at Moss  
229 Bluff (Figure 1). The field instrument locations provided a diverse set of topographic and  
230 hydrologic conditions for observing the patterns of rainfall and river flooding that occurred  
231 during Imelda.

## 4 Modeling Approach

This study employed a numerical model and a Lagrangian particle routing tool to analyze the hydrodynamics of the Trinity River floodplain during Tropical Storm Imelda. First, a numerical model of near-constant grid resolution was developed for the study reach. Then a set of additional unstructured meshes was developed for the same domain, where each mesh has increased resolution only within the floodplain region of interest; that is, the floodplain and channel banks surrounding the three field instruments (Figure 1C). Simulations were run on each mesh and results were compared to field measurements of depth in the floodplain. Flow fields from the model simulations were used to model passive particle transport and compute average particle speeds and residence times across the floodplain. Through these methods, we infer the relative impact of pluvial and fluvial flooding, as well as the role of floodplain channel scales, on floodplain hydrodynamics.

### 4.1 ANUGA Model Development

The ANUGA hydrodynamic model was used for numerical modeling in this study. ANUGA is an open-source program developed by researchers at the Australian National University and Geoscience Australia (Roberts et al., 2015). It solves the shallow-water equations using unstructured meshes and a finite-volume numerical scheme. Details of the numerical scheme can be found in Nielsen et al. (2005), Mungkasi and Roberts (2011), and Mungkasi and Roberts (2013). ANUGA was the model of choice for several reasons, including: (i) it is open-source and therefore easy to control and customize; (ii) the finite-volume method conserves mass and momentum along the wetting-drying front; (iii) it uses unstructured meshes; (iv) it scales efficiently in high-performance computing environments; and (v) it employs a variable time step. The flexibility of the unstructured mesh allows for higher model resolution in areas of higher priority, while offering reduced resolution in areas of less concern. This, along with the parallel capabilities and variable time step, reduces the computational resources needed for model simulations, which is important for an application where near-lidar-scale grid resolution was used.

The model domain boundary (shown in Figure 1C) was delineated to incorporate all channel and overbank areas contributing flow to the floodplain area of interest, while using the smallest domain possible for computational reasons. The “control” domain for this study consisted of an unstructured mesh with a constant average element edge length of 20 m, and 78,752 total elements. The 20 m element size is approximately one-fifth of the width of the main channel, which provided a sufficient representation of the channel cross-section geometry. 20-m resolution was too coarse to resolve most floodplain channels along the Trinity, and was only able to resolve longer-range elevation changes, such as a floodplain basin or a group of nearby floodplain channels that are averaged collectively into a smooth low area.

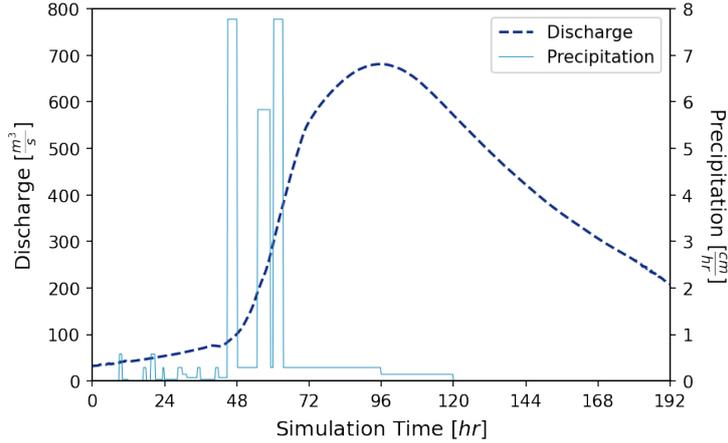
Following development of the 20-m mesh, in which all mesh elements were of a constant size at 20 m per side, three additional meshes were developed. Each mesh had the same outer boundary, and the same resolution across the majority of the domain. The difference lies in the subdomain area delineated in Figure 1C, where the bends and floodplain channel networks contributing to the three field data sites were represented with higher mesh resolution. The three meshes had mean resolutions within the subdomain of 10 m, 5 m, and 2 m, with total element counts of 129,051; 269,361; and 1,308,101; respectively. For each mesh, elements were uniform in size within the floodplain subdomain, but a buffer region was present where the mesh elements transitioned in size between the subdomain and the base, 20-m resolution. 10 m was chosen as the upper resolution bound because it was approximately the scale of the largest floodplain channels observed in the lidar. Ideally, the highest resolution mesh would reach the scale of the 1-m DEM, and provide the best representation of floodplain channel geometry. However, 2 m was found to be the computational limit. The time step required by such small elements, especially those located along

283 cut-banks near the main channel thalweg, was very small. Further increasing the resolution  
284 to 1 m in these areas would result in numerical instability or prohibitive computation time.  
285 Although there are visible features in the lidar less than 2 m in width, the majority of these  
286 channels are located on top of levees, where only overbank flow would activate them.

287 The upstream and river-left boundaries were modeled as no-flow (reflective) bound-  
288 aries. The downstream domain boundary extended approximately 13.4 rkm from the study  
289 site to Wallisville, TX. This extension included only the river channel itself, and was ap-  
290 pended to the domain to provide a sufficient distance between the study site and the  
291 downstream river boundary condition. A constant water surface elevation (WSE) of 0.7  
292 m (NAVD88) was imposed at the downstream boundary, representing the mean WSE mea-  
293 sured at the Wallisville USGS gaging station in the days preceding the storm. The longitu-  
294 dinal boundaries along the channel levees of this extension were transmissive boundaries to  
295 allow for any overbank flow to pass out of the domain. The boundary passing through the  
296 river-right floodplain at the downstream end was modeled as a quasi-transmissive boundary.  
297 This was a time-varying, zero-momentum boundary with a WSE always equal to 5 cm be-  
298 low the current WSE in the domain adjacent to the boundary. Several boundary conditions  
299 were applied at this location, including a completely transmissive boundary and a constant  
300 WSE boundary, but these conditions overestimated and underestimated drainage from the  
301 floodplain, respectively. The quasi-transmissive boundary was used as an approximation to  
302 the water surface slope moving through the floodplain.

303 The DEM described in Section 2.2 was applied to mesh vertices via a least-squares  
304 fit with minimal smoothing. Elevations at mesh element centroids were computed as the  
305 average of the three vertices, creating a discontinuous, piecewise-constant elevation surface  
306 used by the ANUGA “DE1” flow algorithm (Davies & Roberts, 2015). Friction forcing was  
307 applied to the domain as two constant Manning’s  $n$  values: 0.025 within the main channel,  
308 and 0.075 in the floodplain. These values were chosen based on guidance from literature  
309 (Chow, 1959), intuition from field visits and site photographs, and model calibration. River  
310 floodplains are clearly heterogeneous, with dense forested areas expected to have a higher  
311 flow resistance than the channelized portions that are the focus of this study. However,  
312 analyzing frictional heterogeneity in the floodplain (and channel) is beyond the scope of this  
313 study. Future work may involve a more complex distribution of friction based on analysis  
314 of aerial images, specifically vegetation color and density, and topographic images, where  
315 surface curvature may be correlated to flow resistance.

316 The model was run over a 10-day period, beginning at 0000 Central Time on 17  
317 September 2019 and lasting through 26 September. The model was forced using discharge  
318 and rainfall data collected at the Liberty USGS gage during Tropical Storm Imelda. The  
319 base flow recorded at Liberty at the starting time was  $52 \text{ m}^3/\text{s}$ , while the peak discharge from  
320 Imelda was  $793 \text{ m}^3/\text{s}$ , occurring at 1100 on 21 September. The Liberty gage hydrograph  
321 was calibrated to match the observed WSE in the channel, as the rating curve changes  
322 significantly in the BWR and applying a dynamic downstream boundary condition carries  
323 too much uncertainty. The hydrograph used to force the model is shown in Figure 2, with  
324 an initial discharge of  $32 \text{ m}^3/\text{s}$  and a peak discharge of  $681 \text{ m}^3/\text{s}$ . Rainfall measurements at  
325 Liberty consisted of three distinct passovers of tropical storm bands, corresponding to the  
326 spikes in water levels on the floodplain seen in Figure 2. The data show the first rainfall band  
327 arriving at 2100 CT on 18 September and the third band ending at 1530 on 19 September.  
328 The combined depth of rainfall added to the model from the three bands was 75.3 cm. This  
329 depth corresponds to a total volume of  $1.0 \times 10^6 \text{ m}^3$  added to the domain over an 18.5-hour  
330 period. A less intense, background rainfall rate was also added to the model to replicate the  
331 steady accumulation of water observed in the field data prior to the tropical storm bands.



**Figure 2.** Hydrograph and hyetograph for Tropical Storm Imelda as applied to the numerical model. The period of intense rainfall occurring from storm bands preceded peak river flooding by about 30 hours.

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#### 4.2 Quantifying Residence Times with *dorado*

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*dorado* (Hariharan et al., 2020) is an open-source, Lagrangian particle routing package that uses a D-8 walk algorithm (Pearson, 1905) to simulate passive particle transport through hydrodynamic flow fields on regular grids. Here we provide a brief description of *dorado*; for more information see Hariharan et al. (2020) and the *dorado* documentation. The particle walk algorithm is weighted by local flow direction and water depth, in a manner similar to that of the DeltaRCM model (Liang, Voller, & Paola, 2015; Liang, Geleynse, et al., 2015). For a given grid cell, the downstream direction  $F^*$  is computed by a weighted combination of water surface slope ( $F_{sfc}$ ) and discharge ( $F_{int}$ ) unit vectors:

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$$F^* = \gamma F_{sfc} + (1 - \gamma) F_{int} \quad (1)$$

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where the parameter  $\gamma$  is specified by the user depending on the nature of transport. Particles are then routed based on orientation to the mean flow direction and the depth in each cell:

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$$w_i = \frac{h_i^\theta \max(0, F^* \cdot d_i)}{\Delta_i} \quad (2)$$

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where  $F^*$  is the local flow direction computed in Equation 1,  $d_i$  is the unit vector pointing to downstream cell  $i$ ,  $\Delta_i$  is the Euclidian distance to downstream cell  $i$ ,  $h_i$  is the depth of downstream cell  $i$ , and the exponent  $\theta$  is a weighting parameter specified by the user (Liang, Voller, & Paola, 2015; Hariharan et al., 2020). The default value of  $\theta$  is 1.0, which routes particles proportionally based on flow depth under the assumption that deeper cells receive more flow than their shallower neighbors (in the absence of vertical model resolution). The particle routing in this study uses  $\gamma = 0.05$  and  $\theta = 1.0$ , where routing weights depend mostly on discharge, and therefore the analysis and discussion that follows can be thought of conceptually as water solute transport.

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*dorado* tracks individual paths and travel times of particles as they are routed through a flow field. An effective particle travel distance is computed for each iteration, defined by the Euclidian distance traveled to one of the surrounding eight grid cells projected onto the

358 mean flow vector. The particle travel time  $T_{p,i}$  between two cells is then back-calculated from  
 359 the effective travel distance and local flow velocities, with a dispersion coefficient applied  
 360 that allows  $T_{p,i}$  to vary stochastically up to 10 percent from the mean velocity.

361 In a steady flow field, a sufficient number of particles initialized at the domain inflow  
 362 location and routed through the domain can provide a probabilistic, spatial distribution of  
 363 particle paths. All hydraulically-connected locations in the flow field have some probability  
 364 of having a particle pass through. The total travel time for each particle can be computed,  
 365 and the average travel time for all particles passing through a stationary part of the domain  
 366 can be computed as well. Particle travel paths are limited, however, to the instantaneous  
 367 WSE gradient and discharge in the steady flow field, which may be only representing a  
 368 particular snapshot in time. The flow field may show certain areas of the floodplain as con-  
 369 nected hydraulically, but the instantaneous directionality of the water surface may prevent  
 370 particles from reaching those areas.

371 A flow field that changes through time, due to the rising and falling of the flood wave  
 372 or unsteady precipitation on the floodplain, creates an environment where potential particle  
 373 paths are highly dependent on when and where particles enter the floodplain from the river.  
 374 For example, a particle will not move from river to floodplain until the river stage reaches  
 375 an elevation higher than the elevation of the deepest floodplain channels. Even then, if the  
 376 floodplain is already inundated from rainfall the gradient may not allow river flow into the  
 377 floodplain. Only at a higher river stage might the flow direction change. Routing particles  
 378 through an unsteady flow field is critical to understanding these river-floodplain interactions.

379 The ANUGA model depth, stage, and momentum outputs were interpolated to a  
 380 2-m raster grid, and a new particle “cohort” consisting of 1,000 particles was initialized  
 381 in the domain every 15 minutes of model simulation time. Two classes of particles were  
 382 analyzed: river particles and floodplain particles. All river particle cohorts were initialized  
 383 at the inlet of the domain, while floodplain particles were seeded randomly throughout the  
 384 floodplain in grid cells with depth greater than 20 cm. Separating particles into these two  
 385 classes is necessary for distinguishing between patterns of rainfall and river flood processes.  
 386 Floodplain particles were initialized beginning at simulation hour 45 (the onset of intense  
 387 rainfall, see Figure 2), while river particles were initialized at simulation hour 60, as flow  
 388 does not move from river to floodplain until sometime after hour 60. All particle cohorts  
 389 were routed through the model flow field until simulation hour 120. With 1,000 particles per  
 390 15 minutes, the total number of river particles tracked was 240,000, and the total number  
 391 of floodplain particles was 300,000.

392 Particle dynamics were quantified in two ways: velocity distributions and residence  
 393 time distributions. Velocity distributions show the spatial extent of particle paths, as well  
 394 as the average speed at which particles move through each 2-m grid cell in the model domain.  
 395 The average time a particle spends in cell  $(x, y)$  is calculated as follows:

$$396 \quad t_{avg,xy} = \sum_{p=1}^{N_p} \frac{0.5 \times (T_{p,i,xy} + T_{p,i+1,xy})}{N_{p,i,xy}} \quad (3)$$

397 where  $N_p$  is the total number of particles, the numerator is the average of travel times for  
 398 particle  $p$  as it entered ( $i$ ) and as it left ( $i+1$ ) cell  $(x, y)$ , and  $N_{p,i,xy}$  is the number of times  
 399 a particle entered cell  $(x, y)$ . The array is masked for  $N_{p,i,xy} = 0$ . Then the average flow  
 400 speed  $V_{avg,xy}$  is:

$$401 \quad V_{avg,xy} = \frac{dx}{t_{avg,xy}} \quad (4)$$

402 where  $dx$  is the cell size. A Gaussian smoothing filter with standard deviation of 0.7 was  
 403 applied to the  $V_{avg,xy}$  array to account for cases of strongly unequal particle visits between  
 404 nearby cells due to the stochasticity involved in modeling a finite number of particles.

405 Particle residence time distributions are calculated in the form of the cumulative exit  
 406 age distribution  $F(t)$  (Benjamin & Lawler, 2013):

$$407 \quad F(t) = \int_0^t \frac{dN_p/dt}{N_{p,tot}} dt \quad (5)$$

408 where  $N_{p,tot}$  is the total number of particles that enter a control volume,  $dN_p/dt$  is the rate  
 409 at which particles exit, and at  $t = \infty$ ,  $F(t) = 1$ . For this study, we define the control volume  
 410 as the entire river-right floodplain in the model domain. Rather than observe the rate at  
 411 which particles leave the floodplain, since we do not necessarily know when particles entered  
 412 the floodplain, we instead track individual particle travel times beginning when they enter  
 413 (or are seeded in) the floodplain, and ending when they leave:

$$414 \quad t_p = \sum_{i=1}^{N_i} T_{p,i} \quad (6)$$

415 where  $t_p$  is the total travel time for particle  $p$  within the floodplain boundary,  $N_i$  is the  
 416 number of iterations performed while within the boundary, and  $T_{p,i}$  is the travel time for  
 417 each iteration. All values of  $t_p$  are sorted in ascending order, and then  $F(t_p)$  is simply the  
 418 cumulative fraction of particles that spent less than  $t_p$  in the domain.

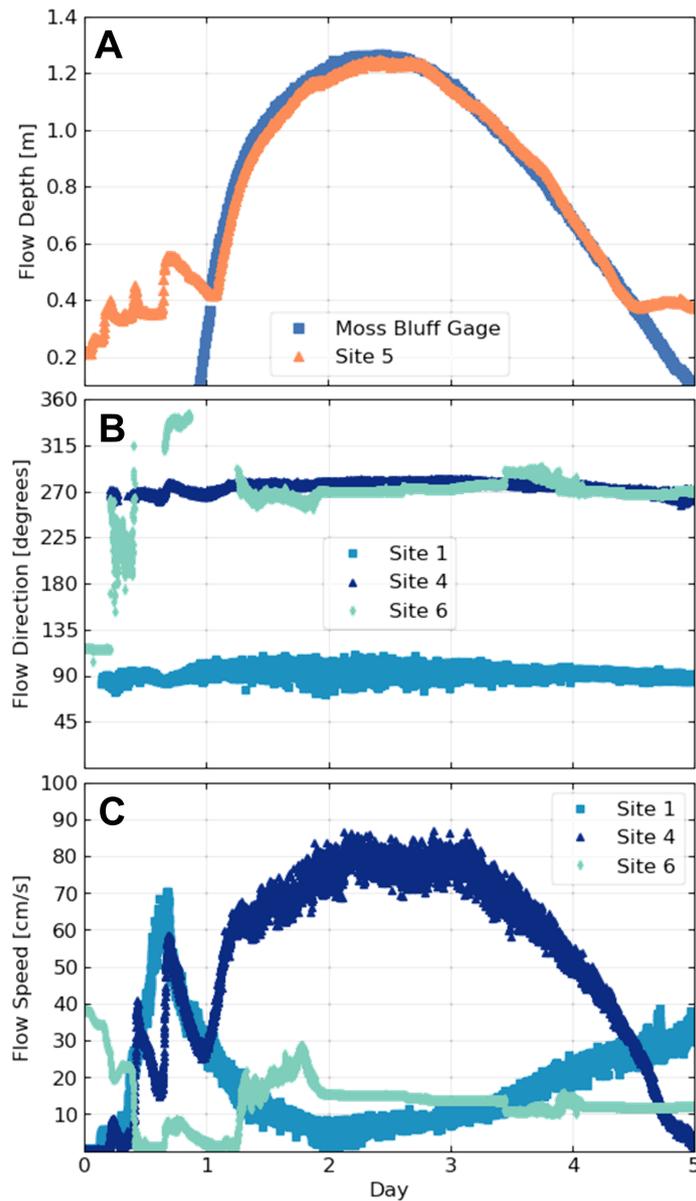
## 419 5 Results

### 420 5.1 Field Observations of Water Level and Velocity

421 Data collected during Tropical Storm Imelda showed a clear separation of floodplain  
 422 inundation due to rain and inundation due to rising river stage (Site 5, Figure 3A). Since  
 423 the precipitation was centered over the site, the floodplain response to precipitation always  
 424 preceded that tied to river stage. Still, flooding patterns varied depending on specifics of the  
 425 monitored location. At Site 5, the first day of the storm saw water levels rise and fall with  
 426 each sequential rain band before the river WSE had risen to the elevation of the floodplain  
 427 channel (Figure 3A). At the start of the second day, the river stage had risen to the point  
 428 that river water contributed to flow in the floodplain channel. For the following three days  
 429 peak flow depth tracked river stage until stage fell below water levels on the floodplain.

430 The other field sites showed diverse flow patterns. At Site 1, water in the floodplain  
 431 channel flowed out to the river throughout the entire event, with peak velocity tied to peak  
 432 rainfall, not river stage (Figure 3B and C). Data collected at Site 2 on a high levee showed  
 433 that water always flowed down the levee and away from the river, but this flow was driven  
 434 entirely by rainfall, as river stage was insufficient to overtop the levee. At Site 6 in the  
 435 study area, water flowed out to the river through the nearby floodplain channel during peak  
 436 rainfall and then reversed course, flowing back onto the floodplain with peak velocity tied to  
 437 peak river stage (Figure 3B and C). Velocities were lower at Site 6 because it was situated at  
 438 the transition between the mouth of a floodplain channel and its connected, small floodplain  
 439 basin, and flow directions were also sensitive to rainwater draining to the site from several  
 440 directions. Less than 1 km away in the Site 4 floodplain channel farther from the river,  
 441 water always flowed away from the river into the floodplain interior (Figure 3B and C).

442 Imagery collected by the time-lapse cameras confirmed these observations, showing a  
 443 rapid rise in water level soon after the beginning of rainfall, followed by a pattern of drainage  
 444 consistent with saturated soil conditions throughout the event. Saturated soils are common



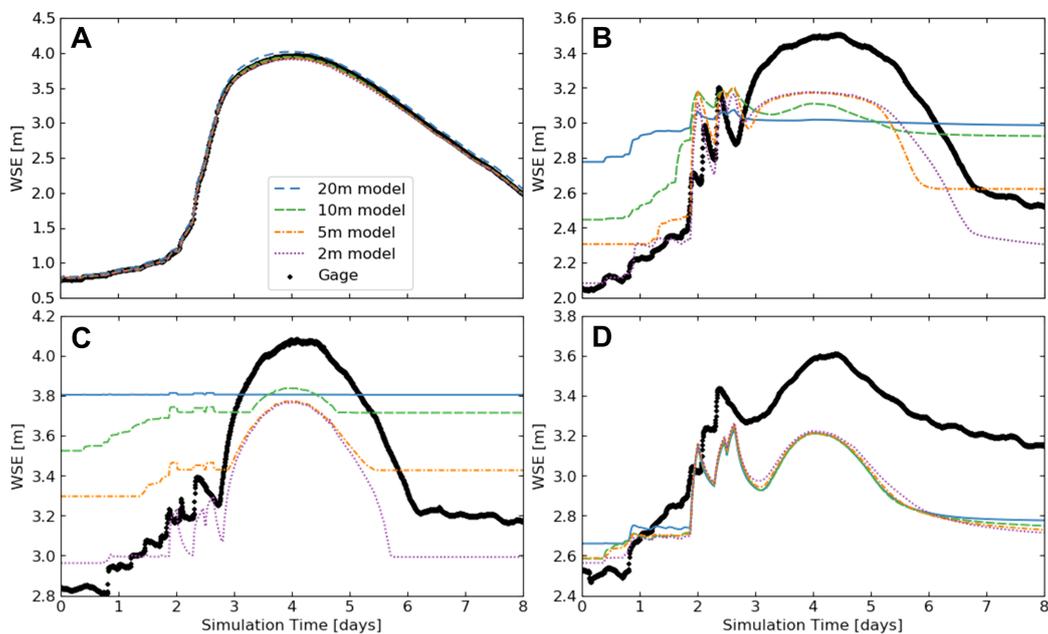
**Figure 3.** Flow patterns observed in field data in the Trinity River floodplain during Tropical Storm Imelda. (A) Water depths in the Site 5 floodplain channel, compared to river water levels at the nearby USGS gage (relative to the floodplain channel bottom). (B) Flow direction histories for three floodplain channels. Data are oriented so that 90 degrees aligns with flow out of channel and into river, and 270 degrees aligns with flow out of river and into floodplain. The channel at Site 1 always flowed out to the river, flow at Site 4 always flowed out onto the floodplain, and flow at Site 6 reversed direction during the rainfall phase. Flow directions are only plotted for velocities exceeding 5 cm/s. (C) Velocity histories at Sites 1, 4, and 6 were tied to competing water-level histories on the floodplain versus the river. At peak river stage, flow velocities toward the river at Site 1 are at a minimum, while flow velocities away from the river at Site 4 are maximum.

445 at this site due to proximity to the river and low elevation gradients; with an average annual  
 446 precipitation exceeding 1.3 m/yr, the water table on the floodplain remains high, within 1

447 m of the surface as confirmed by persistent standing water within local pools of floodplain  
 448 channels. The saturated soil simplifies the analysis of pluvial flooding as any infiltration is  
 449 small enough to be ignored without accruing measurable error. The complex flow patterns  
 450 observed along the Trinity River during Imelda provided motivation to investigate the effect  
 451 of rainfall on connectivity with numerical modeling and Lagrangian particle routing.

## 452 5.2 Numerical Modeling

453 The control model (20-m resolution) was calibrated to match the observed WSE at  
 454 the USGS Moss Bluff gage. All models yielded similar results for WSE in the main channel  
 455 at Moss Bluff (Figure 4A). The small differences in WSE at peak discharge are a product  
 456 of the differences in lateral exchange that occur across models with different resolution.  
 457 The high-resolution models have slightly lower WSE in the main channel than those with  
 458 lower resolution, due to the increased exchange that occurs with more floodplain channels  
 459 resolved.



**Figure 4.** Numerical model results for models with local mesh resolution of 20 m, 10 m, 5 m, and 2 m. (A) WSE plot showing the match of each model to the stage measured at the USGS Moss Bluff gage. (B) Comparison of WSE measurements in the Site 4 floodplain channel to modeled WSE. (C) WSE comparison in the Site 5 floodplain channel. (D) WSE comparison at Site 6, at the terminus of the floodplain channel containing Site 5.

460 Depth measurements at Sites 4, 5 and 6 were converted to WSE by adding the depth  
 461 to the lidar elevation at the GPS location of the instrument site. Model WSEs were then  
 462 compared to measurements at the three sites (Figure 4B, C, and D). The flow patterns  
 463 during the rainfall bands (hours 45-63) match the observations well, especially at Site 4  
 464 (Figure 4B), which is a wider and deeper channel than Site 5 (8-9 m compared to 4-5 m).  
 465 However, all models show a mismatch in WSE during the river flooding phase. At Site 4,  
 466 the shape of the WSE curve is modeled better with increasing resolution, but there appears  
 467 to be an upper limit on depth in the floodplain channel. An elevation of 3.2 m is reached  
 468 during the rainfall phase, followed by a brief drainage period, then another rising phase  
 469 up to the same elevation of 3.2 m. The depth of this channel in the lidar is about 1 m, and the

470 WSE range for the 2-m model is also 1 m (Figure 4B), indicating that model flow depth  
471 is limited to the channel bankfull depth. Since the range of water levels recorded by the  
472 Site 4 instrument is close to 1.5 m, there is evidence that the actual depth of the channel is  
473 greater than indicated by the lidar.

474 Results are similar at Site 5 (Figure 4C). The shape of the fluvial signal is best matched  
475 by the 2-m model, but the measured depth in this floodplain channel is greater than in any  
476 of the models. Like Site 4, the effect of model resolution is most evident at the start of the  
477 simulation, where the 2-m model resolves the floodplain channel well compared to the lidar,  
478 and the 20-m model averages over the channel entirely. Unlike Site 4, where the 5-m model  
479 starting elevation is comparable to the 2-m model and lidar, the smaller channel width at  
480 Site 5 means the 5-m model does not resolve it as well. Site 5 results are also distinct in that  
481 higher resolution seems to reduce water levels by 5-10 cm in the floodplain channel. In the  
482 10-m model, the channel is barely resolved, and there may be a bottleneck downstream in  
483 the floodplain channel from this lack of resolution. In the 2-m and 5-m models the channel  
484 may be resolved enough to convey flow more consistently.

485 There is essentially no difference among models at Site 6 (Figure 4D). The floodplain  
486 topography is mostly flat at this location, as it is located just beyond the terminus of a  
487 floodplain channel where it enters a small floodplain basin. But the nearly identical water  
488 levels may still describe the system of floodplain channels in some way, albeit indirectly.  
489 Rainfall accumulation in the floodplain is less dependent on the scales of connectivity along  
490 the river levees, but the fact that the same quantities of water reached this location in  
491 each model during the fluvial phase shows that, at this location and for this storm, river-  
492 floodplain connectivity is not dependent on channel size. In this case, the water is mostly  
493 supplied by the low-freeboard river bend located 500 m south-southwest of Site 6 (Figure  
494 1C), which overtops across the entire bend beginning at about simulation hour 72.

495 Because the recorded water depths at the field sites were tied to the lidar rather than  
496 a datum, any error present in the lidar was passed to the measured data. This includes  
497 the 29-cm vertical accuracy of the lidar, which is similar to the WSE differences seen for  
498 Sites 4 and 5. It is possible that the differences are partially explained by lidar error,  
499 but considering that the model results are consistently lower by close to 30 cm, it is also  
500 possible that the sites had a combination of standing water and woody debris during the  
501 lidar flyovers. While the former is more likely, the presence of either element would alter  
502 the channel bottom elevations in the lidar and would artificially raise the measured depths  
503 plotted in Figure 4.

### 504 **5.3 Particle Routing Analysis**

505 Although the water depths in the floodplain did not exactly match the data, the flow  
506 patterns achieved in the numerical model followed many of the same trends, particularly  
507 for the high-resolution models. Using the numerical model flow fields for particle routing  
508 can help describe and quantify the complex interactions occurring in the floodplain during  
509 Tropical Storm Imelda. By continuously seeding passive particles in the river and floodplain,  
510 we can observe the differences between water moved by rainfall and river flooding, and how  
511 the dominant forcing changes in the floodplain during the storm. The analysis that follows  
512 is based on the ANUGA model with 5-m resolution unless otherwise noted, as the modeling  
513 results show that higher resolution improves the match of flow patterns to observations  
514 in the floodplain channels, but the improvement is not substantial between 5-m and 2-m  
515 resolution (Figure 4). More importantly, particle tracking results changed very little between  
516 flow grids derived from the 5-m and 2-m models.

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### 5.3.1 Particle Dynamics in the Floodplain

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Two particle classes were seeded continuously every 15 minutes throughout the storm event: one in the main channel and one distributed randomly throughout the floodplain. Floodplain particles were only seeded in grid cells where water had accumulated to a depth of 20 cm during the prior time step. Particle velocity distributions show the spatial extent of particle paths in two dimensions (Figure 5). During simulation hour 61 while it is still raining heavily, the gradient is from floodplain to river, and river particles remain confined, even though the floodplain is inundated everywhere except the topographic ridges (Figure 5A, seen in lighter shades of gray). At the same time, floodplain particle paths are widely connected (Figure 5D). At hour 72 (panels B and E) it is no longer raining heavily, but the peak river discharge has not arrived yet. Some of the rainwater has drained from the remote areas of the floodplain, and water pooled in the larger floodplain basins slows down (darker red colors) as it leaves through the outlet to the south. More river particles begin to enter the floodplain, mostly from the river bend due south of Site 6, but a few begin to enter through the floodplain channels near Site 4 as well. Finally, at hour 91 (panels C and F), the river discharge is at its maximum. Floodplain particles (panel F) are limited to the larger floodplain basins, similar to the previous time stamp, but average velocities are slightly higher overall. This is because the river is supplying more water to the floodplain, and thus providing a stronger gradient to the floodplain outlet that was not present at hour 72. As expected for particles originating in the river (panel C), the travel paths are limited to just a fraction of the floodplain, even during peak discharge. The combined average velocities across the entire duration (Figure 5G) show that river particle paths remain within the corridor shown in panels B and C throughout the simulation. The velocities show that river water generally spends less time in the floodplain than rainwater, with the exception of the floodplain in the north corner of the domain, which is highly-connected to the main channel and very deep (Figure 5G). The large, warm-colored region of lower floodplain particle velocities (Figure 5H) is inundated throughout the storm, but river particles never reach it. Instead, river particles seem to bypass this part of the floodplain entirely, while rainwater spends much more time in this area as it drains slowly to the outlet.

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For river particles, residence time distributions (RTDs) were combined for cohorts seeded between hours 66 and 90, as there were not enough particles entering the floodplain at earlier times (Figure 6A). For floodplain particles, the combined time window is between hours 46 and 90 (Figure 6B). The limit at hour 90 was chosen because particles were only tracked up to hour 120, and the residence time window observed was limited to 30 hours. Ninety-five percent of river particles spent a minimum of five hours in the floodplain (within the model domain), and about 80 percent of particles had residence times less than 10 hours. The five percent of particles with residence times of less than five hours were those that entered the floodplain briefly before returning to the river. The narrower distribution confirms what can be seen spatially in the velocity distributions (Figure 5G). Floodplain particles had a wider distribution of residence times. Many particles exited the domain quickly if seeded close to the outlet, but 20 percent of floodplain particles remained in the domain for longer than 30 hours, compared to just 10 percent for river particles. Note that the river particle RTDs are composed of less particles by several orders of magnitude. Also note that the southeastern-most corner of the domain was masked out for the particle analysis because too many river particles were entering the floodplain at this bend and immediately exiting the floodplain due to proximity only, and this skewed the distributions.

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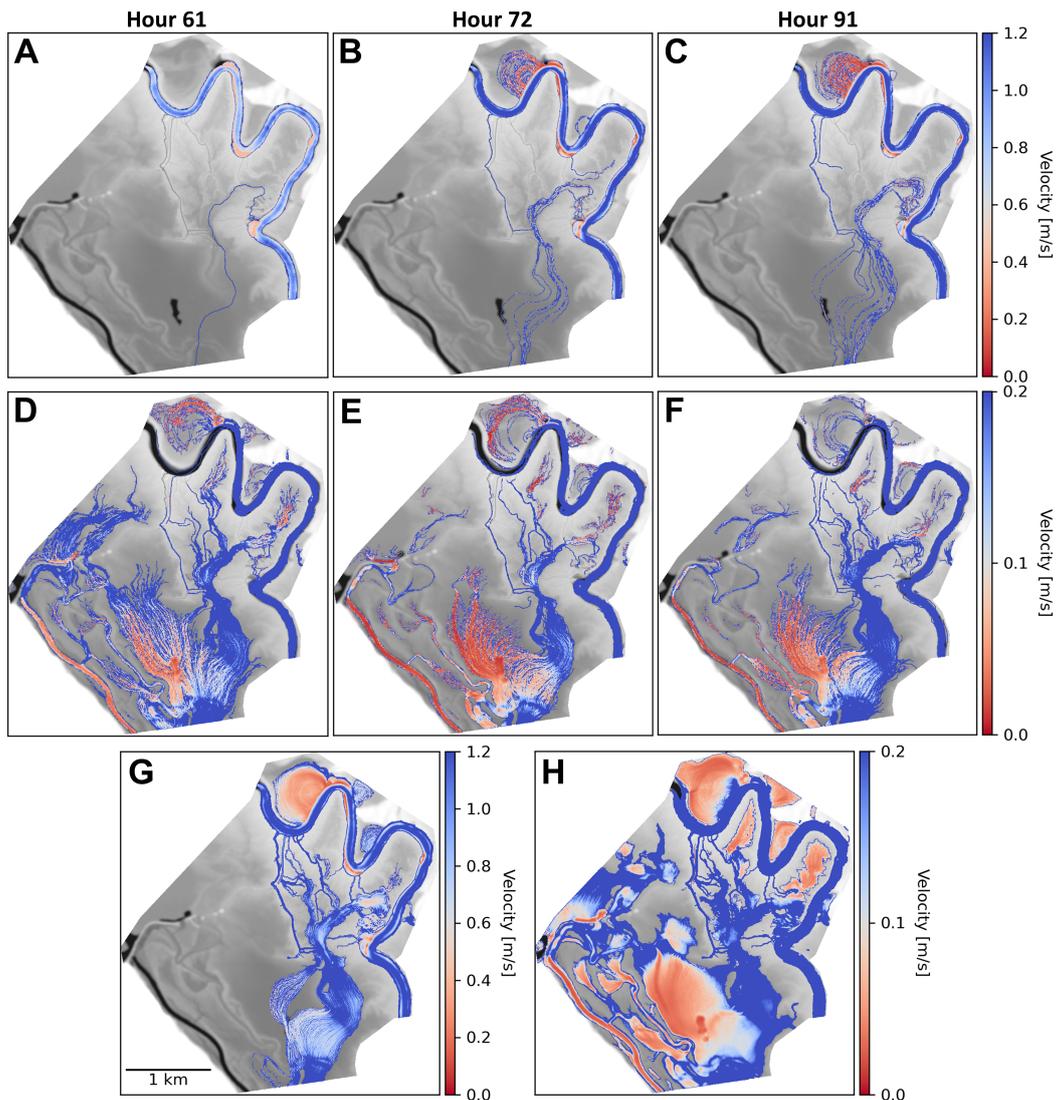
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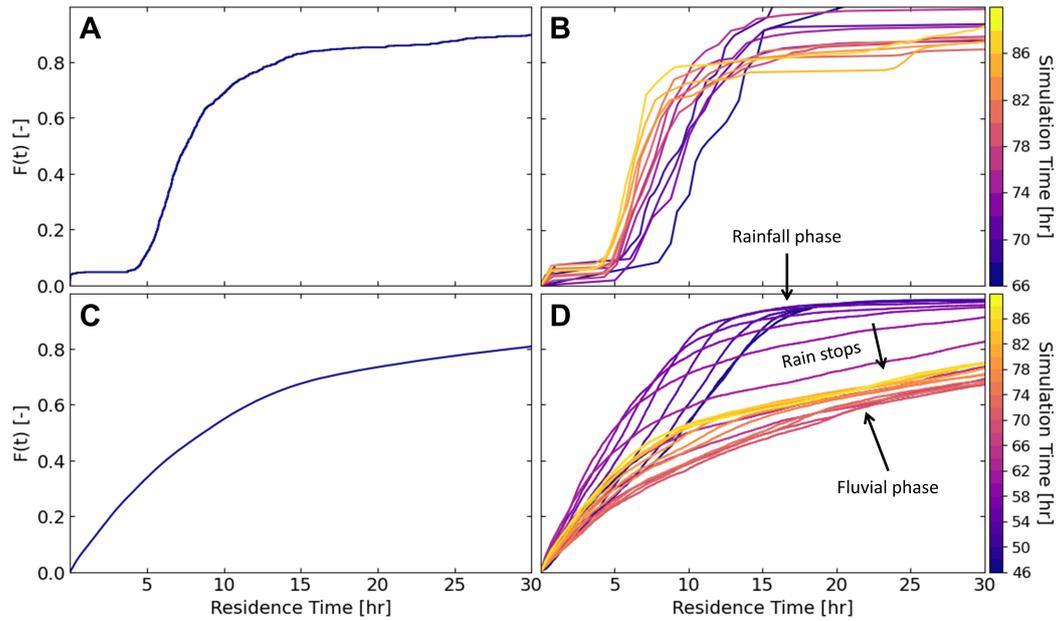
Particle RTDs evolved over the course of the storm (Figure 6B and D). At hour 66, only a very small number of river particles entered the floodplain, but that number increased as the storm transitioned to the fluvial phase (Figure 6B). Through this transition, river particles experienced a reduction in minimum residence time and an increase in maximum residence time as the discharge increased. At higher discharges in the river, more flow moved through the floodplain, increasing velocities and reducing residence times. Conversely, activation of the floodplain channels near Site 4 at higher discharges brought particles into



**Figure 5.** Average particle velocities for river particles (top row) and floodplain particles (second row). (A), (B), and (C) Velocity distributions for river particles at simulation hours 61, 72, and 91, respectively. (D), (E), and (F) Velocity distributions for floodplain particles at simulation hours 61, 72, and 91, respectively. Combined velocity distributions for (G) river and (H) floodplain particles.

570 the channels, creating longer flow paths than those that entered earlier from the bend near  
 571 Site 6. The yellow and orange curves in Panel B imply that this was the case for about 20  
 572 percent of river particles.

573 Floodplain particle RTDs show a wider range of behavior, as there may be more  
 574 competing factors involved in their movement (Figure 6D). The RTD for the earliest group  
 575 of particle cohorts, representing most of the 8,000 particles seeded between hours 46 and  
 576 48, shows that 90 percent of particles left the floodplain after 15 hours. Fifteen hours  
 577 corresponds to hour 61, when heavy rainfall was still active. Although the rainfall stops and  
 578 starts twice during this 15-hour period, the overall period of rainfall flushed the floodplain  
 579 to some degree, and the result is a nearly uniform distribution. Moving forward in time,  
 580 the sixth group of cohorts (hours 56-58) marks a transition in the RTD where a greater

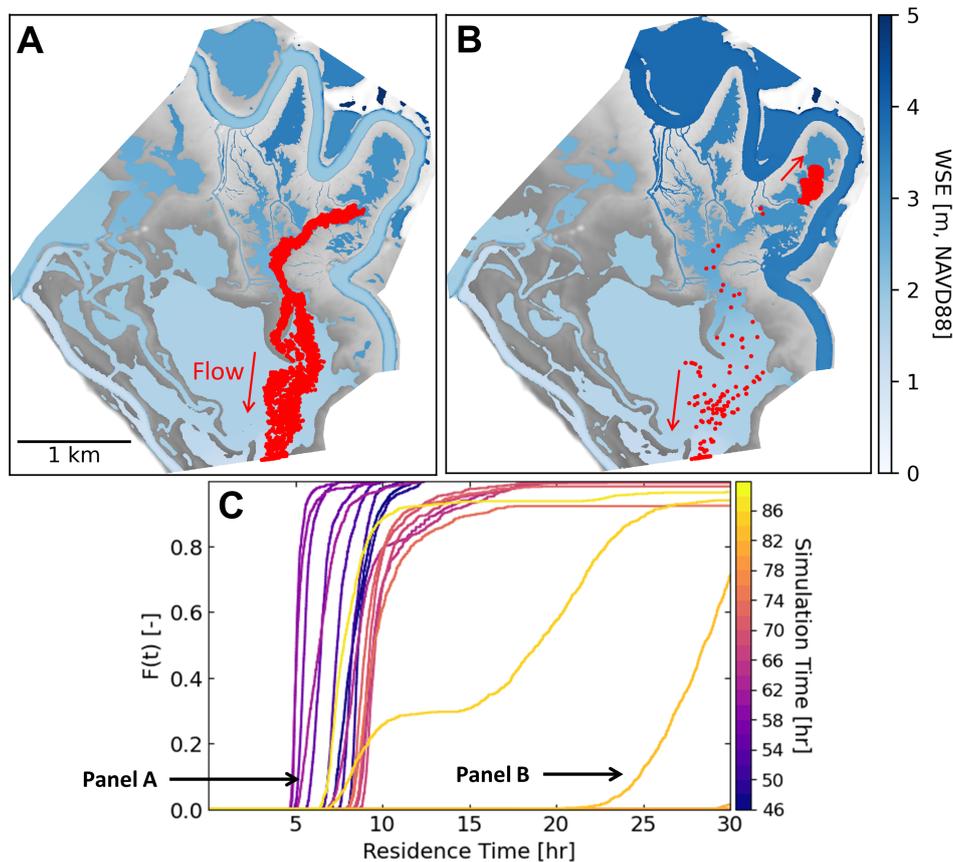


**Figure 6.** Cumulative Residence Time Distributions (RTDs) for river (top row) and floodplain (bottom row) particles. (A) Combined RTD for particles seeded in the river between simulation hours 66 and 90. (B) RTDs for river particles, grouped in intervals of two hours, where the darkest purple line is the combined RTD of particles seeded between hours 66 and 68, and the lightest yellow line represents particles seeded between hours 88 and 90. (C) Combined RTD for particles seeded in the floodplain between hours 46 and 90. (D) RTDs for floodplain particles. The darkest purple line represents particles seeded between hours 46 and 48, and the lightest yellow line represents particles seeded between hours 88 and 90.

581 fraction of particles left the floodplain faster, but the remaining particles spent longer than  
 582 those from 10 to 12 hours prior. The transition can be attributed to the period between  
 583 heavy rainfall and peak discharge. The longer residence times represent particles stranded  
 584 in remote areas of the floodplain as rainwater drained, and the shorter residence times are  
 585 a result of floodwaters accumulating closer to the outlet, where newly seeded particles then  
 586 had less distance to travel to the outlet. Particle cohorts seeded between hours 56 and 62  
 587 began to show an increasingly greater fraction with shorter residence times, due to being  
 588 seeded *after* the longest pause in rainfall (see Figure 2), but also an increasingly greater  
 589 fraction with longer residence times, due to the heavy rainfall stopping for good after hour  
 590 63. After the rainfall phase, the RTD becomes more consistent through time, as the remote  
 591 areas of the floodplain drain and inundation becomes dominated by river water. Average  
 592 residence times reach a maximum (lowest red curves, Figure 6D) before reducing again  
 593 during peak discharge when velocities are higher (bright yellow curves).

594 At field Site 6, located at the terminus of a floodplain channel in a small floodplain  
 595 basin, a noteworthy reversal of flow occurred that was captured by the tilt current meter  
 596 at the site (Figure 3B). A flow reversal at Site 6 was also observed in the model, albeit one  
 597 with a slightly different pattern. The reversal can be well-described with particle routing.  
 598 One hundred particles per 15 minutes were seeded at the Site 6 location. During the  
 599 early rainfall phase, particles flowed directly to the floodplain outlet with a nearly constant  
 600 residence time of five hours (Figure 7A and C, purple curves). After the rainfall stopped,  
 601 flow paths remained similar, and residence times remained nearly constant at eight hours for  
 602 80-90 percent of particles. Beginning at about hour 76, as more river flow was conveyed to

603 the floodplain from the local floodplain channel and the bend to the south, the small basin  
 604 began to fill up, causing new particles to become trapped there (Figure 7B and C, orange  
 605 curves). This reversal lasted until about hour 88, very close in time to the peak discharge,  
 606 when the floodplain basin water levels equilibrated with the river, and the flow direction  
 607 reversed. Particles then returned to the original flow path, with nearly constant residence  
 608 times of six hours for 90 percent of particles (Figure 7C, yellow curve). The flow reversal  
 609 lasted for only 12 hours, but many particles seeded around this time had residence times  
 610 exceeding 30 hours. In fact, the four groups of cohorts seeded between hours 78 and 86  
 611 had no particles with residence times less than 30 hours, while two particular RTD curves  
 612 show the transition on either side of the flow reversal (Figure 7C, orange curves). Although only  
 613 one local observation of flow reversal, it represents a drastic change in average residence  
 614 times, and could have significant implications for floodplain processes when scaled to entire  
 615 floodplain systems.



**Figure 7.** Flow patterns illustrated by particles seeded at Site 6. (A) State of particles at simulation hour 61. Heavy rainfall on the floodplain moved all particles toward the floodplain outlet. (B) State of particles at hour 89.5. Flow reversal due to fluvial flooding pushed particles farther into the scroll bar. The flow reversal lasted from hour 76 to hour 88, after which particles began flowing back toward the floodplain outlet. (C) RTDs for Site 6 particles grouped every two hours, from hour 46 (darkest purple curve) to hour 90 (lightest yellow curve). Six orange curves at the bottom right of the plot represent 12 hours of particles that experienced the flow reversal shown in panel B, four of which are completely flat.

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### 5.3.2 Effect of Model Resolution

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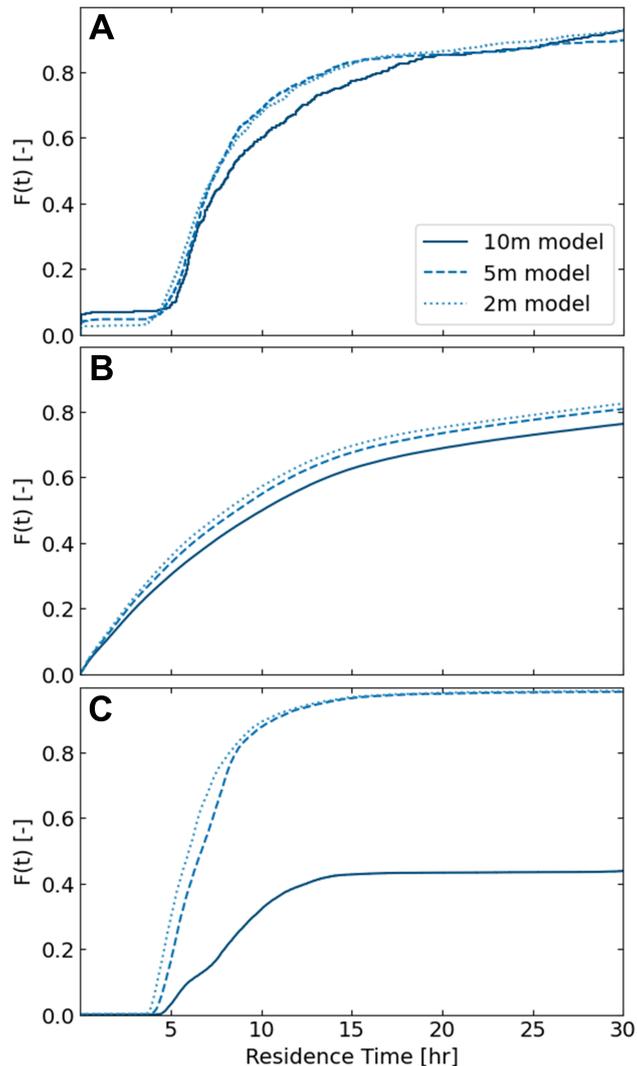
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Hydrodynamics in the overall floodplain were largely unaffected by model resolution. The modeled WSEs at Site 6 (Figure 4D) showed that outside of floodplain channels there was little difference with higher resolution. There are locations in the model domain, such as the river bend near Site 6, where large amounts of water can move from river to floodplain on scales that are wider than the grid resolutions analyzed. The result is that floodplain residence times were very similar for model resolutions of 2 m, 5 m, and even 10 m (Figure 8A and B). For both particle classes, there was some difference observed with the coarser, 10-m resolution, but the general flow patterns were the same, and the major sources of river water to the floodplain were not dependent on changes in grid resolution below 10 m.



**Figure 8.** RTDs with different model grid resolutions, for (A) river particles released between hours 66 and 90, (B) floodplain particles released between hours 46 and 90, and (C) particles released in the Site 4 floodplain channel between hours 46 and 90. Flow processes in the overall floodplain were largely independent of grid resolution, but modeling local processes in floodplain channels requires a minimum resolution.

626 Particles released in the Site 4 floodplain channel showed that there can be a local  
627 resolution effect (Figure 8C). RTDs for the 2-m and 5-m resolution models were almost  
628 identical, but the 10-m model's failure to resolve the channel well completely changed the  
629 conveyance through the channel. Despite the fact that the majority of the floodplain beyond  
630 this local channel had a similar flow field in the model regardless of grid resolution, the  
631 under-resolved channel served as a local bottleneck for river flow. However, some fraction of  
632 particles still moved through the channel, even at 10-m resolution. Particle animations (see  
633 Supporting Information) showed that flow was only conveyed through this channel when  
634 rainfall was active, and directly following the peak discharge. During the pauses in rainfall,  
635 the flow drained from the channel and particles became stuck, to be flushed out when the  
636 rainfall resumed. Particles remained stuck in the channel for the period between heavy  
637 rainfall and peak discharge (hours 63-92), after which particles were conveyed through the  
638 floodplain due to sufficient river flow. Flow was cut off once again at hour 112. This result  
639 aligns with the range of WSE at this location in the 10-m model (Figure 4B), where the  
640 window of activity in the channel is much more limited. In general, the RTDs for Site 4  
641 particles and the modeled WSE curves (Figure 4B and C) show the importance of model  
642 resolution on local processes where features near the scale of grid resolution are relevant.

643 The impact of model resolution can be seen in the channels close by to Site 4 as well.  
644 Field observations at Site 4 did not show any flow reversals as the event transitioned from  
645 rainfall to river-dominated (Figure 3B), and thus all particles released at Site 4 flowed south  
646 into the floodplain. But particle animations (see Supporting Information) showed that for  
647 particles released at Site 4 during the rainfall phase, a portion of flow was siphoned into the  
648 larger floodplain channel just west of Site 4, where local rainfall was moving water into the  
649 river. At the time when rainfall stopped, the flow reversed, and the gradual rise in river  
650 water levels did not allow particles to move into the river through these channels any longer.  
651 Particles only moved in this way with model resolution of 5 m or finer. The 10-m model  
652 could not resolve a deep enough floodplain channel to convey particles. Grid resolution is  
653 thus important for modeling flow through floodplain channels.

## 654 **6 Discussion**

### 655 **6.1 Pluvial and Fluvial Flooding Interactions**

656 Field data collected in the Trinity River floodplain during Tropical Storm Imelda pro-  
657 vided strong evidence for flow bidirectionality between rivers and floodplains. The relative  
658 timing of rainfall and the peak discharge at the study site created an interesting transition of  
659 floodplain hydrodynamics from being pluvial-driven to fluvial-driven. The data showed that  
660 with heavy rainfall on the floodplain, river-floodplain connectivity can occur many hours (in  
661 this case about 24 hours) prior to the flood wave, and that this connectivity is influenced  
662 by floodplain channel topography. Furthermore, the extent of connectivity may be reduced  
663 or removed completely if rainfall intensity lessens or stops altogether for a period of time  
664 before peak river discharge. The data make clear that pluvial flooding can be an important  
665 component of river-floodplain connectivity.

666 For backwater systems such as the Trinity River, river discharges high enough to cause  
667 floodplain connectivity are more frequent than in upstream environments, although not as  
668 frequent as in deltas where lateral exchange can occur at all discharges (Hiatt & Passalacqua,  
669 2017). Similarly, backwater floodplains are likely associated with high groundwater tables  
670 and saturated soil conditions, providing requisite conditions for pluvial flooding. The fre-  
671 quency of floodplain inundation, whether from rainfall or river discharge or both, indicates  
672 that a variety of floodplain processes are dependent on this connectivity, and unraveling  
673 some of the complexity can improve our understanding of the important hydrodynamic  
674 processes in floodplain ecosystems.

675 Numerical modeling and particle routing analysis reinforced many of the patterns seen  
676 in the data relating the timing mismatch of pluvial and fluvial flooding to bidirectional  
677 connectivity between the river and floodplain. Studies have shown that river-floodplain  
678 connectivity can be established at river stages less than bankfull (Mertes, 1997; Nicholas &  
679 Mitchell, 2003; Trigg et al., 2012; Czuba et al., 2019), but the current study showed that  
680 connectivity can be established from pluvial flooding at stages even less than the elevation  
681 of the deepest floodplain channels. In fact, model data and measurements from Site 1 show  
682 that heavy rainfall on a saturated floodplain can provide a competing force against river  
683 waters that would otherwise enter the floodplain. And while flow directed toward the river  
684 may only occur during a certain phase of a storm, even when a flow reversal occurs river flux  
685 into the floodplain may be limited by a reduced gradient from the presence of rainwater.

686 However, many areas of the floodplain can be activated by pluvial flooding that may  
687 not otherwise be reached by river water. If given enough time at peak discharge, river water  
688 may be able to reach more remote areas of the floodplain. But for events like Imelda where  
689 the discharge is sub-bankfull and the flood wave lasts for only a couple of days, river flooding  
690 is limited in time and space. Pluvial water may allow a hydraulic connection to form across  
691 a wider fraction of the floodplain, causing both a reduction in overall flow velocity in the  
692 basin from the deeper water and an opportunity for river water and its constituents to  
693 diffuse to more remote areas of the floodplain.

## 694 **6.2 Implications for Nutrient Removal and Sediment Transport**

695 Understanding the mechanisms controlling river-floodplain connectivity is important  
696 for understanding how many floodplain processes work. Floodplains, especially those near  
697 the coast, are known to act as sinks for nutrients present in river water, such as carbon and  
698 nitrogen (Tockner et al., 1999; Aufdenkampe et al., 2011; Noe et al., 2013; Wolf et al., 2013;  
699 Cheng & Basu, 2017), and for sediment (Tockner et al., 1999; Verhoeven et al., 2001; Schulz  
700 et al., 2003; Day et al., 2008; Juez et al., 2019). In some circumstances, floodplains can be  
701 a source of dissolved nutrients (Tockner et al., 1999). We have shown that pluvial flooding  
702 has a significant role in river-floodplain connectivity, and the implications for floodplain  
703 processes related to sediment retention and nutrient removal are numerous.

704 River-floodplain connectivity is typically studied as a process that is initiated from the  
705 river. In this framework, river water spreads into the floodplain over a range of sufficiently  
706 high discharges, bringing sediment and solutes to the floodplain. This study showed that  
707 there can be a competing gradient between fluvial and pluvial floodwaters, which may reduce  
708 the river water that moves into the floodplain, and thus reduce the transport of constituents  
709 to the floodplain where they are processed. When the river stage becomes high enough  
710 for flow to move into the floodplain, the presence of rainwater still impacts the dynamics.  
711 Velocity distributions (Figure 5) from particle routing analysis show that, for the domain  
712 studied, the reach of river water is limited to only a fraction of the floodplain. If there had  
713 been no rainwater in the floodplain, the river water and its constituents would have room to  
714 spread to a much larger area. Because the path of river particles is restricted to a relatively  
715 narrow corridor of the floodplain, residence times for river particles may be less than they  
716 otherwise would be. During peak discharge, RTDs showed that 80 percent of river particles  
717 move through the floodplain in about five to seven hours (Figure 6C). In contrast, particles  
718 randomly sampled in the floodplain during peak discharge show a much wider range of  
719 residence times depending on when and where they are seeded (Figure 6D). If we removed  
720 from consideration the fraction of sampled floodplain particles seeded close to the outlet,  
721 the distribution would be even wider. This result indicates that the active portion of the  
722 floodplain is within the corridor of river particle paths shown in the velocity distributions  
723 (Figure 5G), and the water in the remainder of the floodplain that mostly originated as  
724 rainwater is slower moving and less active (Figure 5H). So although inundation maps would  
725 show water throughout the floodplain, these results show that it is possible for river water

726 and its dissolved nutrients to short-circuit a large portion of the floodplain, potentially  
727 bypassing crucial floodplain ecosystem processes.

728 The routing parameters used in the particle analysis assume that each particle moves as  
729 a passive tracer. Particles, therefore, more closely represent solutes rather than sediment.  
730 However, sediment dynamics may be inferred from flow patterns, average velocities, and  
731 residence times in the floodplain. Similar to dissolved nutrients, sediment flux from river  
732 to floodplain is entirely dependent on the flow gradient, and it is less likely that floodplain  
733 sedimentation will occur if the dominant flow direction is toward the river. Again, floodplains  
734 already inundated with rainwater may reach equilibrium with the river more quickly, and  
735 reduce the window of time where sediment can be transported out of the river. For sediment  
736 that does enter the floodplain, sediment deposition is dependent on flow velocities and  
737 residence times, which in turn are dependent on the flow interactions that occur during the  
738 storm event. Residence times are shortest during peak rainfall and peak discharge when  
739 velocities are higher, but in between they can be significantly longer. So pluvial flooding  
740 can reduce overall river flow to the floodplain, but the increase in total floodplain volume  
741 due to rainwater may increase residence times. The dynamic nature of pluvial and fluvial  
742 compound flooding creates conditions for sediment transport and deposition that can change  
743 dramatically over the course of an event. In environments where pluvial flooding can be  
744 substantial, sediment dynamics should be considered and modeled within this context.

### 745 **6.3 Floodplain Channel Scales and Model Resolution**

746 In floodplain systems where connectivity is truly limited to smaller floodplain channels  
747 (during sub-bankfull flow conditions), grid resolution could be critical for modeling lateral  
748 exchange. The model domain used in this study was chosen partly because there were several  
749 floodplain channels of various scales present that had been shown by field observations to  
750 convey water in both directions. Model results showed that for processes in the overall  
751 floodplain, resolving those channels was not important. A large fraction of flow from the  
752 river was supplied by two river bends that were connected to the floodplain at low WSE  
753 and over length scales much larger than individual floodplain channels. Model results also  
754 showed that changes in flow magnitude and direction occur in these channels only when they  
755 are resolved sufficiently. Model resolution can then be important for understanding local  
756 processes, and could even be necessary for processes in the larger floodplain for systems  
757 where lateral exchange is completely limited to smaller topographic features.

758 For numerical modeling applications in other river-floodplain systems, or even other  
759 locations on the Trinity River, running low-cost model simulations prior to detailed inves-  
760 tigation can provide guidance on the major sources of lateral exchange. In some systems it  
761 may be the case that the majority of floodplain connectivity is supplied from a small set of  
762 large breaches. While in most systems exchange is also likely to occur via smaller floodplain  
763 channels, it may not be on a large enough scale to affect overall flow patterns in the larger  
764 floodplain. For systems where it is known that floodplain channels are the main drivers of  
765 connectivity, it is necessary to resolve them with grid resolution finer than the scale of those  
766 channels. In either case, understanding which features are important in a system can allow  
767 modelers to shift computational resources to the most important aspects of their model.

### 768 **6.4 Importance of Unsteady Modeling**

769 This study described many ways in which floodplain flow patterns can change during  
770 a storm. Floodplains can experience periods of rapid pluvial accumulation, draining, flow  
771 reversal, and flow deceleration within the span of several days. The complexity of flow  
772 through the Trinity River floodplain during Tropical Storm Imelda shows that it is critical  
773 to model these processes in an unsteady way. For applications where the spatial extent of  
774 inundation is of interest for various discharges (e.g., Benke et al., 2000; Czuba et al., 2019),  
775 steady modeling of river-floodplain connectivity is appropriate. But for problems related

776 to sediment and solute transport into and out of the floodplain, it is crucial to understand  
777 how the spatial extent of inundation, flow time scales, and flow directions change over the  
778 course of a storm event.

## 779 **6.5 Limitations and Future Work**

780 The residence times computed in this study were useful for determining how travel  
781 times change with different hydrodynamic conditions. But the residence times are relative  
782 to the size of the model domain, and cannot be used to assess specific contact times needed  
783 for nutrient removal from the water column, for example. It is unclear what happens in the  
784 downstream floodplain, and how long water might stay there. It is likely that, for the same  
785 flow conditions, residence times change significantly moving down-valley through the lower  
786 Trinity River floodplains. It may be worthwhile to increase the model domain to a much  
787 larger river-floodplain reach. The domain used in this study needed to be small enough  
788 to meet computational constraints for the 2-m and 5-m simulations, but model results  
789 showed that large-scale floodplain processes may not depend on grid resolution at the scale  
790 of floodplain channels. A less costly numerical mesh that identifies critical topographic  
791 features beforehand may be sufficient to perform a similar study on a larger scale. At larger  
792 scales, we would expect to see floodplain flow rejoin the river at points downstream, and  
793 conclusions related to absolute residence times can be sought.

794 In addition to being limited in space, the particle analysis was also limited in time to  
795 just after the passing of the flood wave. The phase of the storm and associated floodplain  
796 dynamics related to the falling hydrograph limb and drainage from the floodplain was not  
797 analyzed here. We saw that particles in the floodplain slowed down after the period of  
798 intense rainfall ended (Figure 5D and E) and the floodplain began to drain. We also saw  
799 that residence times decreased during peak discharge as the total flow in the floodplain  
800 increased. It is expected that, following peak discharge, floodplain flow would slow down  
801 again as the forcing from the river decreases. This is an additional hydrodynamic phase  
802 not captured by the particle analysis, but one that could have implications for sediment  
803 deposition and nutrient retention.

804 The lidar data used for numerical modeling was collected in early 2017, and it is  
805 likely that the floodplain topography changed to some degree between then and field data  
806 collection (fall 2019). In fact, an even stronger storm (Hurricane Harvey) passed through the  
807 region after lidar was collected. Floodplain topography can change significantly over several  
808 years, which has been observed in Trinity River lidar data dating back to 2011 (Hassenruck-  
809 Gudipati, 2021). Combined with the possibility of lidar error in the floodplain channels, our  
810 numerical model results should be evaluated with these sources of error in mind. Still, the  
811 model was able to produce flow patterns that generally aligned with the patterns in the field  
812 data, and is therefore a useful tool for analyzing hydrodynamics in parts of the floodplain  
813 where no data was collected. Even if not an exact replicate of conditions during Tropical  
814 Storm Imelda, the relative timing and magnitude of pluvial and fluvial flooding applied to  
815 the model created unique conditions related to the competing flood modes that confirm at  
816 a larger spatial extent the observations made from the field data.

817 It is possible that the boundary condition imposed at the floodplain outlet is not  
818 representative of the conditions in the larger floodplain during Imelda. We imposed a  
819 dynamic, zero-momentum boundary with stage equal to the adjacent model floodplain stage  
820 during the previous time step. It is unclear whether this location saw more or less backwater  
821 during Imelda, and thus whether the rate of floodplain drainage in the model was accurate.  
822 Various boundary conditions were tested during the calibration phase (see Section 4.1), but  
823 this part of the floodplain was too low in elevation to have an impact on WSEs at any of  
824 the field sites for confirmation.

## 7 Conclusions

This study used field observations, numerical modeling, and Lagrangian particle routing to examine river-floodplain connectivity along the Trinity River during Tropical Storm Imelda. Field data and modeling showed the complex hydrodynamic interactions that can result from heavy pluvial flooding occurring in conjunction with high, sub-bankfull river flow. Floodplain residence times and flow directions in the floodplain can be strongly dependent on the dominant mode of flooding, and can change rapidly during a storm. Residence times were shorter during the periods of active rainfall and peak discharge, and flow slowed considerably in between these phases as flooding transitioned from pluvial to fluvial. Particle routing analysis showed that as river flow moved into a floodplain already inundated from rainwater, the spatial extent of river water was limited to a narrower reach of the floodplain. Without pluvial flooding, river water would likely spread farther into the floodplain where storage is available. Some floodplain channels were shown to facilitate two-way connectivity driven by the timing mismatch between pluvial and fluvial flooding. Although the 5 to 10-m floodplain channels in the study area were shown to be conveyors of lateral exchange, overall processes in the floodplain were unaffected by their resolution in the numerical model, as the majority of lateral exchange came from only a few locations. It is interesting to note that in this study, these bank-line locations are not bounded by levees, but rather are the sites of a previous cutoff of a river bend and the low-lying deposits of a counter point bar. Variability in how the bank line is constructed also plays an important role in river-floodplain connectivity.

The dynamic environment of competing pluvial and fluvial flooding during a storm has many implications for sediment and nutrient exchange between rivers and floodplains. The extent to which residence times and flow directions change indicates that optimal conditions for sediment deposition and nutrient retention are limited to only certain phases of a flood event. Enough pluvial flooding occurring prior to peak discharge may prevent river water from entering the floodplain altogether, effectively reducing sediment and nutrient fluxes to the floodplain. Pluvial flooding can also decrease velocities and increase residence times overall, as deep flow can be achieved sooner with less floodplain storage available for the peak river discharge. This study challenges the prevailing perspective that river-floodplain connectivity is dependent only on river discharge, and emphasizes the importance of rainfall as a driver of that connectivity.

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