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

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

<sup>1</sup>University of Texas at Austin <sup>2</sup>The University of Texas at Austin

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#### Abstract

Hydrologic connectivity controls the lateral exchange of water, solids, and solutes between rivers and floodplains, and is critical to ecosystem function, water treatment, flood attenuation, and geomorphic processes. This connectivity has been well-studied, typically through the lens of fluvial flooding. In regions prone to heavy rainfall, the timing and magnitude of lateral exchange may be altered by pluvial flooding on the floodplain. We collected 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 hydrodynamic model at high resolution for a section of the Trinity River floodplain inspired by the compound flooding of Imelda. We then 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 water to be confined to a narrow corridor on the floodplain, while rainwater residence times were increased from the effect of high river flow. Finally, we analyzed the role of floodplain channels in facilitating hydrologic connectivity by varying model resolution in the floodplain. While the resolution of floodplain channels was important locally, it did not affect as much the overall floodplain behavior. This study demonstrates the complexity of floodplain hydrodynamics under conditions of heavy rainfall, with implications for sediment deposition and nutrient removal during floods.

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 <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

# 9 Key Points:

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10	•	Field data and modeling show distinct signals of pluvial and fluvial flooding
11	•	Floodplain residence times change dramatically between pluvial and fluvial phases of
12		the storm
13	•	Flows between river and floodplain are dominated by the largest channels and levee

Flows between river and floodplain are dominated by the largest channels and levee depressions

<sup>\*</sup>Walter P. Moore, Houston, TX

 $Corresponding \ author: \ Paola \ Passalacqua, \ {\tt paola@austin.utexas.edu}$ 

#### 15 Abstract

Hydrologic connectivity controls the lateral exchange of water, solids, and solutes 16 between rivers and floodplains, and is critical to ecosystem function, water treatment, flood 17 attenuation, and geomorphic processes. This connectivity has been well-studied, typically 18 through the lens of fluvial flooding. In regions prone to heavy rainfall, the timing and 19 magnitude of lateral exchange may be altered by pluvial flooding on the floodplain. We 20 collected measurements of flow depth and velocity in the Trinity River floodplain in coastal 21 Texas (USA) during Tropical Storm Imelda (2019), which produced up to 75 cm of rainfall 22 23 locally. We developed a two-dimensional hydrodynamic model at high resolution for a section of the Trinity River floodplain inspired by the compound flooding of Imelda. We then 24 employed Lagrangian particle routing to quantify how residence times and particle velocities 25 changed as flooding shifted from rainfall-driven to river-driven. Our results show that heavy 26 rainfall initiated lateral exchange before river discharge reached flood levels. The presence 27 of rainwater also reduced floodplain storage, causing river water to be confined to a narrow 28 corridor on the floodplain, while rainwater residence times were increased from the effect of 20 high river flow. Finally, we analyzed the role of floodplain channels in facilitating hydrologic 30 connectivity by varying model resolution in the floodplain. While the resolution of floodplain 31 channels was important locally, it did not affect as much the overall floodplain behavior. 32 This study demonstrates the complexity of floodplain hydrodynamics under conditions of 33 heavy rainfall, with implications for sediment deposition and nutrient removal during floods. 34

## <sup>35</sup> Plain Language Summary

Unaltered river floodplains can support diverse ecosystems, reduce flooding, and re-36 move nutrients from river water. Floodplains near the coast are particularly important, 37 as they typically experience more frequent flooding. Floodplain function relies on a high 38 degree of connectivity with the river, where water can move easily through the floodplain 39 during periods of high river stage. Our study explores the ways in which heavy rainfall on 40 a floodplain impacts this connectivity. We collected flow measurements in the Trinity River 41 floodplain (Texas, USA) during Tropical Storm Imelda in 2019 that showed distinct flood-42 ing patterns between the rainfall and river flooding. We coupled a hydrodynamic model 43 with a particle tracking module to see how particles in the water might move through the 44 floodplain during the transition from rainfall-driven to river-driven flooding. We found that 45 the average time a particle spent in the floodplain changed significantly after the rain in the 46 model stopped. We also noticed that rainwater tended to remain in the floodplain for much 47 longer than river water, especially after the rain stopped. This study describes the various 48 interactions that can occur between local rainfall and river flooding, and moves toward a 49 better understanding of sediment and nutrient transport through floodplains. 50

# 51 **1** Introduction

River floodplains play a fundamental role in flood storage, nutrient cycling, sediment 52 retention, and in general provide support for diverse ecosystems (Ward et al., 1999; Melack 53 & Forsberg, 2001; Kondolf et al., 2006; Roley et al., 2012; Noe et al., 2013; Kufel & 54 Leśniczuk, 2014). Floodplains are complex and heterogeneous, and their structure and 55 function are highly dependent on their degree of connectivity with the river (Hughes et al., 56 2001; Harvey & Gooseff, 2015; Gurnell et al., 2016; Covino, 2017). Topographic controls 57 on hydrologic connectivity, defined here as the degree of surface water movement between 58 rivers and floodplains, have been described for the largest river-floodplain systems using 59 satellite imagery (Lesack & Melack, 1995; Mertes et al., 1995; Mertes, 1997; Alsdorf et 60 al., 2007; Trigg et al., 2012; Lewin & Ashworth, 2014; Park & Latrubesse, 2017), and 61 more recently for medium-size rivers using lidar data and numerical modeling (David et 62 al., 2017; Czuba et al., 2019; Byrne et al., 2019). However, river-floodplain connectivity 63 is poorly understood when hydrodynamics are partially driven by local rainfall. Mixing of 64

rainfall and river floodwaters has been observed and discussed for very large river floodplains 65 (Mertes, 1997; Alsdorf et al., 2007; Day et al., 2008; Rowland et al., 2009; Trigg et al., 2012), 66 where floodplain channels are at a large enough scale to be sensed remotely, and the flood 67 wave occurs over much longer time scales (Junk et al., 1989). But for medium-size rivers, 68 flood waves are less predictable, and floodplain features are often too small to detect using 69 satellite imagery. The goals of this study are to show (a) the impact of local rainfall on 70 floodplain residence times, flow directions, and connectivity within the floodplain, and (b) 71 to determine the role of floodplain channels in facilitating river-floodplain exchange under 72 combined pluvial-fluvial flooding conditions. 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 75 of channel features (Lesack & Melack, 1995; Mertes et al., 1995; Mertes, 1997; Trigg et 76 al., 2012). Flow within floodplain channels can be bidirectional due to the advancing and 77 receding of the flood wave and the timing of rainfall runoff on the floodplain (Alsdorf et 78 al., 2007; Day et al., 2008; Rowland et al., 2009). For smaller river systems as well, sub-79 bankfull discharges can result in floodplain inundation that is limited spatially by the extent 80 of lateral floodplain channels in the system (Kupfer et al., 2015; Czuba et al., 2019). Pluvial 81 flooding may enhance hydrologic 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 differentiate these two flooding modes. 93

The interaction of pluvial and fluvial flooding may have significant impacts on residence 94 time, flow direction, and the overall extent of hydrologic 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 hydrologic connectivity. We present flow depth and velocity 110 measurements collected during Tropical Storm Imelda (2019) in the Trinity River floodplain 111 that show clear and separate signals of pluvial and fluvial flooding. We then develop a two-112 dimensional, depth-averaged numerical model with high mesh resolution inspired by the 113 observed hydrodynamics during the storm. Next, we employ a Lagrangian particle routing 114 tool on the unsteady model flow field to quantify how rainfall and channel processes impact 115 residence time distributions and flow patterns in floodplains. Lastly, we perform a model 116

resolution scaling analysis to determine how flux to and from the floodplain changes asfloodplain channels are smoothed out of the model.

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 124 125 of rainfall on floodplain hydrodynamics. Section 6 provides a discussion of implications for floodplain services and for future modeling studies of river-floodplain connectivity. Lastly, 126 Section 7 summarizes the major findings of the study. 127

# <sup>128</sup> 2 Study Area: The Trinity River

#### 2.1 Site Description

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The Trinity River basin (40,000 km<sup>2</sup>) extends from its outlet in Trinity Bay into northcentral Texas (Figure 1A). The area of investigation spans about 10.5 river kilometers (rkm) of the lower basin in Liberty County, between Liberty and Wallisville, TX. The study area is within the river's backwater reach (BWR), which is recognized by the asymptotic approach of the water surface elevation to the surface elevation of the receiving basin. Under lowdischarge conditions, the BWR begins approximately 15 rkm upstream of the study area (Figure 1B) (Mason & Mohrig, 2018).

As the river transitions from a normal flow regime to the BWR, the channel morpho-137 dynamics respond to the adjustments in flow conditions and the transport of solids. This 138 transition coincides with downstream narrowing and deepening of the channel. Rates of 139 channel-bend migration decrease in the downstream direction, as do the size and shape of 140 point bars, and grain size of bed material (Smith et al., 2020). Similarly, the overbank 141 conditions vary in accordance with the transition to the BWR. For example, upstream of 142 the BWR, the floodplain is geomorphically active (Mason & Mohrig, 2018; Hassenruck-143 Gudipati, 2021) and it largely remains dry during low and moderate flow conditions, but is 144 inundated during floods, when water emerges from channel confinement and spreads across 145 the floodplain. In contrast, the BWR is characterized by a wetland environment due to 146 its relatively low elevation and is prone to inundation by moderate river discharges. Stage 147 change between low and flood flows is smaller in the BWR, with lateral flow spreading play-148 ing a greater role due to the low-gradient environment and reduced freeboard between normal 149 flow water surface elevation and the adjacent floodplains (Smith et al., 2020). Because of 150 these characteristics, surface-water connectivity is greater in the study reach compared to 151 upstream reaches, and floodplain channels are more commonplace and pronounced. This 152 connectivity may cause a degree of "leakiness" in the system, which would be supported by 153 the large decrease in average annual peak discharge between Liberty  $(2,477 \text{ m}^3/\text{s}, \text{USGS})$ 154 08067000) and Wallisville (756 m<sup>3</sup>/s, USGS 08067252) (see Figure 1B for locations). 155

Like many fluvial-deltaic systems worldwide, the Trinity is not free of anthropogenic 156 influences. Within the study area (Figure 1C), there is a single, raised access pathway that 157 traverses the river-right floodplain perpendicular to the main flow direction. The pathway 158 contains several bridges and culverts that pass flow through the larger floodplain channels. 159 Also in the study area are three buried pipeline rights-of-way that are cleared of trees, and 160 currently consist of very tall grasses and shrubs. Just upstream of the study area but within 161 the connected floodplain is another old, raised pathway along a levee that has been eroded 162 significantly due to lack of use and maintenance. Upstream of the study area but south 163 of Liberty, the floodplain contains remnants of oil drilling operations, although this part of 164 the floodplain is disconnected hydraulically from the current study area due to a natural 165 pinch-point along the right bank of the river. There are also two water diversion operations 166

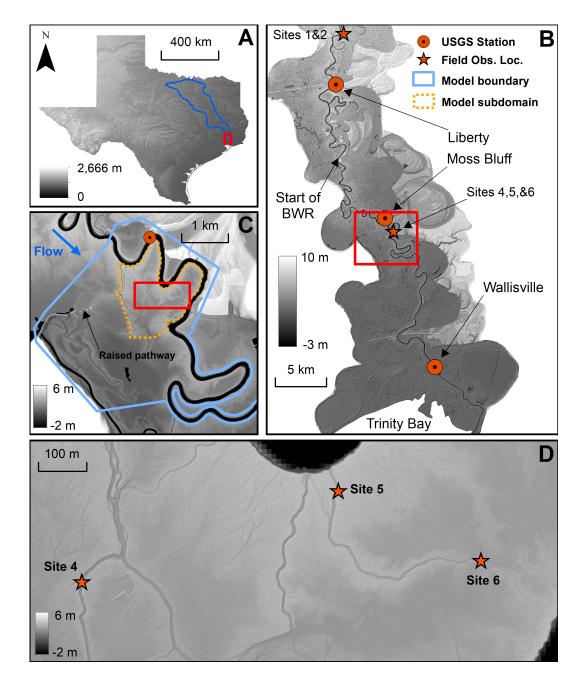


Figure 1. Elevation maps of the Trinity River study area. (A) Location of the Trinity River basin in Texas. (B) Lidar DEM for Trinity River floodplains between Liberty (upstream) and Wallisville (downstream), including locations of field observation sites. (C) Boundary of model domain used in this study. The yellow dashed boundary represents the area in our models with higher mesh resolution. The downstream boundary of the model domain extends to Wallisville, but is not shown here. (D) Location of field sites in the floodplain. Sites 4 and 5 are within channels, while Site 6 is just outside a channel terminus in a floodplain basin.

in the area, both of which are located on perched topography above the floodplain. Perhaps
 most notably, the Livingston Dam (far upstream of the study area) is a run-of-river dam
 that impacts the river geomorphology for the first 50-60 rkm downstream of the structure

(Phillips et al., 2004; Phillips & Slattery, 2007; Smith & Mohrig, 2017). Beyond this
point, sediment mining from the bed and banks of the river re-establishes the bed-material
load (Smith & Mohrig, 2017) and no changes in channel geometry and kinematics have
been observed since reservoir filling. Despite these various human influences, the study
reach is unaffected by significant modifications such as containment levees, wing dykes, and
revetments; thus, the river is able to operate unhindered within its valley.

We chose the model domain boundary (Figure 1C) for three main reasons. First, the 176 domain contains three of the field observation sites (see Section 3.2) that recorded data 177 178 during Tropical Storm Imelda. Second, and related to the first, the floodplain topography in this area features many channels of various sizes that connect the river to the floodplain 179 (Figure 1D). Floodplain channel widths range from small, 1-2 m channels barely detectable 180 in the lidar, to larger, 8-10 m channels with greater depths that most likely play a larger 181 role in river-floodplain exchange. The complex floodplain topography makes this location 182 interesting to study. Third, the domain boundary needed to be limited in space, as the 183 high-resolution modeling needed to resolve the smallest channels requires significant com-184 putational resources. 185

# 2.2 Elevation Data

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

The lidar data were interpolated to a bare-earth digital elevation model (DEM) at 194 1-m resolution. Small voids in the floodplain lidar were interpolated using a second-degree 195 polygon plane fit through the existing data. For larger voids corresponding to floodplain 196 ponds, major channels, and oxbows, bathymetry was approximated by performing the same 197 plane fitting interpolation as above, followed by a 5-m downward shift of the elevation. 198 River bathymetry measurements were taken by the Trinity River Authority in 2017, along 199 four longitudinal profiles at transects spaced every 400 m on average (the river width varies 200 between 80 and 100 m). The bathymetry was interpolated to a 10-m grid, and patched 201 together with the lidar DEM using the Raster to Mosaic tool in ArcGIS. Finally, linear 202 interpolation was performed across the small gaps between the lidar DEM and bathymetry 203 raster. 204

# <sup>205</sup> **3** Tropical Storm Imelda

# 3.1 Storm Background

Tropical Storm Imelda (2019) was a major rainfall event that produced over 75 cm of 207 precipitation across several counties in the area surrounding Houston, TX (Latto & Berg, 208 2020). Imelda made landfall near Freeport, TX (120 km southwest of the study site) on 209 17 September 2019 as a tropical storm, before quickly weakening to a tropical depression 210 as it moved slowly northward through Houston, TX and subsequently across the lower 211 Trinity River watershed. The storm further degenerated to a trough by 19 September, at 212 approximately 160 km north-northeast of Houston, where it continued to dissipate and move 213 northward. The highest recorded rainfall total from Imelda was 112 cm over a three-day 214 period near Fannett, TX, with 79 cm falling within a 31-hour period, which made it the 215 fifth wettest tropical cyclone ever recorded in the contiguous United States. 216

Eastern Texas experienced widespread pluvial flooding during this period. While the 217 lower Trinity River watershed received much of this rainfall, including up to 75 cm at the 218 study site, the river stage at the Liberty USGS station peaked just below the official flood 219 stage on 20 September as defined by the National Weather Service. Likewise, at the Moss 220 Bluff USGS station downstream, the peak stage of 4.0 m was below the adjacent levee 221 crests but over a meter above many of the nearby floodplain channel bottom elevations in 222 the DEM. A sub-bankfull flood event is suitable for analyzing river-floodplain connectivity, 223 as floodplain channels are activated and responsible for any lateral exchange that occurs, 224 and floodplain inundation is heterogeneous (Czuba et al., 2019). The timing mismatch of 225 the pluvial and fluvial flooding peaks, along with the sheer volume of precipitation, created 226 an opportunity for competition between river and floodplain water worth investigating. 227

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# 3.2 Field Data Collection

During August 2019, six outdoor trail cameras, six measuring rods, six water level 229 loggers (Solinst Levelogger, Model 3001), and four tilt current meters (TCM-1 from Lowell 230 Instruments) were installed at various floodplain locations along the Trinity River (Figure 231 1). The loggers recorded water level every six minutes, while the tilt current meters recorded 232 flow speed and direction every minute. The cameras took a photograph of the installation 233 plus measuring rod every five minutes, night and day. The instruments were left in the field 234 until February 2020, and successfully collected data during Tropical Storm Imelda. Sites 235 1, 2, and 3 were located north of Liberty near the bend indicated in Figure 1B (plan view 236 of exact locations of Sites 1 and 2 is shown in Supporting Information Figure S1). Site 1, 237 located in a large floodplain channel, collected water level velocity readings, while Site 2, 238 located in a shallow levee-traversing channel, collected water levels only. Instruments at Site 239 3 were displaced and lost during the storm. Sites 4, 5, and 6 were all located in the study 240 area (Figure 1C and D). Sites 4 and 6 collected both water level and velocity readings, while 241 Site 5 collected water levels only. Site 4 instruments were located in a floodplain channel 242 roughly 930 m from the Trinity River (measured along the channel). Site 5 was located on 243 a different floodplain channel, just 50 m from the river and higher on the river levee. The 244 floodplain slopes downward away from the river, with Sites 4 and 6 at lower elevations than 245 Site 5. Site 6 was located at the terminus of the same channel monitored by Site 5. At Site 246 6 a small internal delta is building out from the mouth of the floodplain channel into the 247 adjacent, small floodplain basin with perennial standing water. Sites 5 and 6 were located on 248 a channel connected to the river bend just downstream from the USGS station at Moss Bluff, 249 while the Site 4 floodplain channel connected to the river immediately upstream of Moss 250 Bluff (Figure 1). The field instrument locations provided a diverse set of topographic and 251 hydrologic conditions for observing the patterns of rainfall and river flooding that occurred 252 during Imelda. 253

Water level and velocity measurements collected during Imelda provided a depiction 254 of hydrodynamics in the floodplain (Figure 2). Each set of water level logger measurements 255 showed a clear distinction between floodplain inundation due to rain (hours 24–72, counting 256 from the start of 17 September 2019) and inundation due to rising river stage (hours 72-257 144, Figure 2A). Since the precipitation was centered over the site, the floodplain response 258 to precipitation always preceded that tied to river stage. Still, flooding patterns varied 259 depending on specifics of the monitored location. During the early hours of 19 September 260 (hour 48), the floodplain channel at Site 5 saw water levels rise and fall with each sequential 261 rain band before the river WSE had risen to the elevation of that channel. Later that day, 262 the river stage had risen to an elevation of 3.17 m, corresponding to the elevation at which 263 river water contributed to flow in the floodplain channel. As Site 5 was located only 50 m 264 265 from the river bank, the peak flow depth at this location tracked river stage closely until the stage fell below the floodplain channel elevation. Water level fluctuations at the Site 266 4 floodplain channel were similar to Site 5, but the 80 cm increase in water levels during 267 the rainfall phase was more substantial than the increase at Site 5, likely due to the larger 268 catchment area of Site 4. River stage reached the elevation of the Site 4 channel at an 269

elevation of 2.95 m. Peak water levels at Site 4 also corresponded to peak river stage, 270 although the rate of drainage during the falling limb was different from those at Site 5 and 271 the main channel, which is also likely related to the longer distance between Site 4 and the 272 river. At Site 6, water level data showed a signal of both rainfall and river flooding, albeit 273 less pronounced than the channelized locations of Sites 4 and 5. Farther upstream, data 274 from Site 2 on a high levee showed a fast increase in water levels during peak rainfall as its 275 local floodplain basin filled up, but unlike the downstream sites, there was no signal of river 276 flooding. In this case, river stage was insufficient to overtop the levee. 277

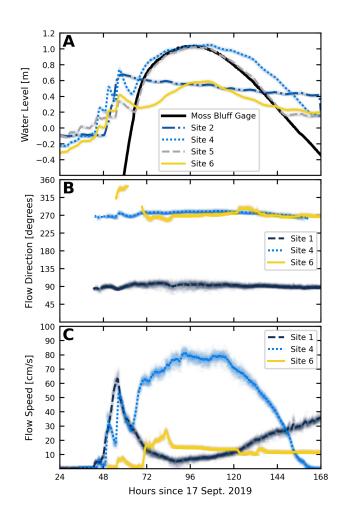


Figure 2. Flow patterns observed in field data in the Trinity River floodplain during Tropical Storm Imelda. Lines represent moving averages of the data, while the raw data are shown underneath in very light colors. (A) Measured water level fluctuations, compared to river water levels at the nearby USGS gage. Note the USGS gage data here are relative, having been translated vertically to show the similar rates of water level change between the river and the Site 5 channel. Also note measured water levels are plotted relative to the initial recorded depths at each site, see Supporting Information Dataset S1 for more information. (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. Flow directions are only plotted for velocities exceeding 5 cm/s and logger depth recordings exceeding 76 cm (based on instrument specifications). (C) Velocity histories at Sites 1, 4, and 6.

Velocity measurements showed diverse flow patterns. At Site 1, water in the floodplain 278 channel flowed out to the river throughout the entire event, with peak velocity tied to peak 279 rainfall, not river stage (Figure 2B and C). Velocities were lower at Site 6 because it is an 280 unchannelized location, situated at the transition between the mouth of a floodplain channel 281 and its connected, small floodplain basin. Interestingly, peak velocity at Site 6 occurred 282 during the period when pluvial flooding drained but prior to arrival of peak river stage. Less 283 than 1 km away in the Site 4 floodplain channel farther from the river, water always flowed 284 away from the river into the floodplain interior (Figure 2B and C). Imagery collected by the 285 time-lapse cameras confirmed these observations, showing a rapid rise in water level soon 286 after the beginning of rainfall, followed by a pattern of drainage consistent with saturated 287 soil conditions throughout the event. The field data collected during the storm showed 288 several distinct patterns of pluvial and fluvial flooding, and provide the inspiration for the 289 modeling efforts in this study. 290

It is important to contextualize these observations with instrument limitations. The 291 water level logger measurements are relative; that is, they are not tied to any datum. 292 During a storm event, it is also possible for loggers to become buried with sediment, and for 293 floodplain geometry to change significantly (Mason & Mohrig, 2018; Hassenruck-Gudipati, 294 2021). For these reasons, it is uncertain how the water levels measured in 2019 relate to 295 elevations and floodplain geometry in the 2017 lidar dataset described in Section 2.2 and 296 used for the modeling in this study. Additionally, the tilt current meters are typically 297 used for deeper-water applications, and have a minimum required depth for accurate results 298 (Lowell et al., 2015). The study of water in floodplains, particularly in the absence of total 299 inundation, involves relatively shallow environments. Therefore, we only present velocity 300 data that meet the minimum depth criterion of 76 cm. 301

# 302 4 Modeling Approach

We employed a numerical model and a Lagrangian particle routing tool to analyze 303 the hydrodynamics of the Trinity River floodplain during Tropical Storm Imelda. First, a 304 numerical model with high-resolution in the floodplain area of interest was developed for 305 the study reach. Simulation results were compared to field measurements of depth in the 306 floodplain. The simulation flow field was used to model passive particle transport and com-307 pute average particle speeds and residence times across the floodplain. A set of additional 308 numerical models was then developed for the same domain, each with progressively lower 309 resolution in the floodplain, and lateral flux was computed and compared between models. 310 Through these methods, we infer the relative impact of pluvial and fluvial flooding, as well 311 as the role of floodplain channel topography, on floodplain hydrodynamics. 312

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# 4.1 ANUGA Model Development

We use the ANUGA hydrodynamic model for numerical modeling in this study. ANUGA 314 is an open-source model developed by researchers at the Australian National University and 315 Geoscience Australia (Roberts et al., 2015). It solves the shallow-water equations using un-316 structured meshes and a finite-volume numerical scheme. Details of the numerical scheme 317 can be found in Nielsen et al. (2005), Mungkasi and Roberts (2011), and Mungkasi and 318 Roberts (2013). ANUGA is the model of choice for several reasons, including: (i) it is open-319 source and therefore easy to control and customize; (ii) the finite-volume method conserves 320 mass and momentum along the wetting-drying front; (iii) it uses unstructured meshes; (iv)321 it scales efficiently in high performance computing environments; and (v) it employs a vari-322 able time step. The flexibility of the unstructured mesh allows for higher model resolution in 323 areas of higher priority, while offering reduced resolution in areas of less concern. This, along 324 with the parallel capabilities and variable time step, reduces the computational resources 325 needed for model simulations, which is important for an application where near-lidar-scale 326 mesh resolution was used. 327

The model domain boundary (shown in Figure 1C) was delineated to incorporate all 328 channel and overbank areas contributing flow to the floodplain area of interest, while using 329 the smallest domain possible for computational reasons. The majority of the model domain 330 consisted of an unstructured mesh with a constant average element edge length of 20 m. The 331 20 m element size is approximately one-fifth of the width of the main channel, which provided 332 a sufficient representation of the channel cross-section geometry. Twenty-m resolution was 333 too coarse to resolve most floodplain channels along the Trinity, and was only able to resolve 334 longer-range elevation changes, such as a floodplain basin or a group of nearby floodplain 335 channels that are averaged collectively into a smooth low area. Within the floodplain region 336 surrounding the three field sites (dashed vellow boundary in Figure 1C), the mesh resolution 337 was increased to 2 m, resulting in a total of 1,308,101 mesh elements. At this resolution, 338 nearly all of the floodplain channels are resolved. The constant, 2-m resolution boundary 339 extends to the edge of the channel, where it transitions to the background spacing of 20 m. 340 As a result, elements in the channel adjacent to the high-resolution boundary are finer than 341 elsewhere in the domain where the general spacing is 20 m. 342

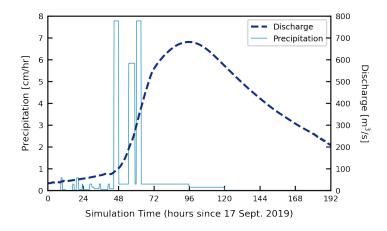
The upstream, left floodplain, and right floodplain boundaries were modeled as no-flow 343 (reflective) boundaries. The downstream domain boundary extended approximately 13.4 344 rkm from the study site to Wallisville, TX. This extension included only the river channel 345 itself, and was appended to the domain to provide a sufficient distance between the study site 346 and the downstream river boundary condition. A constant water surface elevation (WSE) 347 of 0.7 m (NAVD88) was imposed at the downstream boundary, representing the mean WSE 348 measured at the Wallisville USGS station over a 10-day period preceding the storm. We 349 found the model to be insensitive to this boundary condition. The longitudinal boundaries 350 along the channel levees of this extension were modeled as transmissive boundaries to allow 351 for any overbank flow to pass out of the domain. The boundary traversing the river-right 352 floodplain at the downstream end was modeled as a quasi-transmissive boundary. This was 353 a time-varying, zero-momentum boundary with a WSE always equal to 5 cm below the 354 current WSE in the domain adjacent to the boundary. This boundary condition was used 355 as an approximation to the water surface slope moving through the floodplain. 356

The DEM described in Section 2.2 was applied to mesh vertices via a least-squares 357 fit with minimal smoothing. Elevations at mesh element centroids were computed as the 358 average of the three vertices, creating a discontinuous, piecewise-constant elevation surface 359 used by the ANUGA "DE1" flow algorithm (Davies & Roberts, 2015). Friction forcing was 360 applied to the domain as two constant Manning's n values: 0.025 within the main channel 361 and 0.075 in the floodplain. These values were chosen based on guidance from literature 362 (Chow, 1959) and judgment from field visits and site photographs. River floodplains are 363 clearly heterogeneous, with dense forested areas expected to have a higher flow resistance 364 than the channelized portions that are a focus of this study. Although n values are typically 365 suggested at or just over 0.1 for forested areas, we applied the lower value of 0.075 as a 366 compromise between the hydraulic characteristics of channelized and forested areas. 367

The model was run over an 8-day period, beginning at 0000 Central Time on 17 368 September 2019 and lasting through 24 September. The model was forced using a calibrated 369 hydrograph based on discharge data from the Liberty USGS station. The base flow recorded 370 at Liberty at the starting time was 52 m<sup>3</sup>/s, while the peak discharge from Imelda was 371  $793 \text{ m}^3/\text{s}$ , occurring at 1100 on 21 September. The Liberty USGS station hydrograph was 372 calibrated to match the observed WSE in the channel at Moss Bluff (USGS 08067100) where 373 there is no available hydrograph, as it is unknown how much the event discharge changes 374 from Liberty to Moss Bluff. To develop a hydrograph for the model domain, several model 375 simulations were run using the Liberty station as a starting point. With each subsequent 376 simulation, the model stage was compared with the observed stage at Moss Bluff, and the 377 model discharge was adjusted from the previous by the same ratio as the difference in 378 modeled and observed stage. This linear calibration results in a near-perfect match between 379 modeled and observed water levels at Moss Bluff. The calibrated hydrograph used to force 380

the model has an initial discharge of 32 m<sup>3</sup>/s and a peak discharge of 681 m<sup>3</sup>/s (Figure 3). Similar to USGS data in general, discharge values were generated in 15-minute intervals, and these values were applied to the model at each time step using linear interpolation between intervals (see Dataset S2 for the full calibrated hydrograph). This procedure provides the most direct way to force water surface gradients between the river and floodplain to be as close as possible to those observed by the local river gage.

Rainfall data from Imelda in 15-minute intervals were retrieved from the Liberty USGS 387 station, which posts data for a period of 120 days after the storm. These data were not 388 official or quality assured, but the timing and depths of peak rainfall were similar to those 389 reported elsewhere. The data consisted of three distinct passovers of tropical storm bands, 390 corresponding to large hypetograph bars (Figure 3) and spikes in water levels observed on 391 the floodplain (Figure 2A). The data show the first rainfall band arriving at hour 45 (2100 392 CT, 18 September) and the third band ending at hour 63.5. The combined depth of rainfall 393 added to the model from the three bands was 75.3 cm. This depth corresponds to a total 394 volume of  $1.0 \times 10^6$  m<sup>3</sup> added to the domain over an 18.5-hour period. A less intense, 395 background rainfall rate was also added to the model to replicate the steady accumulation 396 of water observed in the field data prior to arrival of the high-intensity tropical storm bands 397 (see Datasets S3 and S4 for the raw and modified hypetograph applied to the model). Rainfall 398 was applied evenly across the entire domain as depths per second. Runoff from outside of 399 the model domain was not considered in this study. 400



**Figure 3.** Hydrograph and hyetograph for Tropical Storm Imelda as applied to the numerical model.

#### 401

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# 4.2 Quantifying Residence Times with dorado

dorado (Hariharan et al., 2020) is an open-source, Lagrangian particle routing package 402 that uses a D-8 random walk algorithm (Pearson, 1905) to simulate passive particle transport 403 through hydrodynamic flow fields on regular grids. Here we provide a brief description of 404 dorado; for more information see Hariharan et al. (2020) and the dorado documentation. 405 The particle walk algorithm is weighted by local flow direction and water depth, in a manner 406 similar to that of the DeltaRCM model (Liang, Voller, & Paola, 2015; Liang, Geleynse, et 407 al., 2015). For a given grid cell, the downstream direction  $F^*$  is computed by a weighted 408 combination of water surface slope  $(F_{\rm sfc})$  and discharge  $(F_{\rm int})$  unit vectors: 409

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

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, with the routing weight of each cell *i* given by:

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

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where  $F^*$  is the local flow direction computed in Equation 1,  $d_i$  is the unit vector pointing 415 to downstream cell i,  $\Delta_i$  is the Euclidian distance to downstream cell i,  $h_i$  is the depth of 416 downstream cell i, and the exponent  $\theta$  is a weighting parameter specified by the user (Liang, 417 Voller, & Paola, 2015; Hariharan et al., 2020). The default value of  $\theta$  is 1.0, which routes 418 particles proportionally based on flow depth under the assumption that deeper cells receive 419 more flow than their shallower neighbors (in the absence of vertical model resolution). The 420 particle routing in this study uses  $\gamma = 0.05$  and  $\theta = 1.0$  (Liang, Voller, & Paola, 2015), where 421 routing weights depend mostly on discharge, and therefore the analysis and discussion that 422 follows can be thought of conceptually as water solute transport. 423

424 dorado tracks individual paths and travel times of particles as they are routed through 425 a flow field. An effective particle travel distance is computed for each iteration, defined by 426 the Euclidian distance traveled to one of the surrounding eight grid cells projected onto 427 the mean flow vector. The particle travel time  $T_{p,i}$  between cell *i* and cell *i*+1 is then 428 back-calculated from the effective travel distance and local flow velocities, with a dispersion 429 coefficient applied that allows  $T_{p,i}$  to vary stochastically up to 10 percent from the mean 430 velocity.

In a steady flow field, a sufficient number of particles initialized at the domain inflow 431 location and routed through the domain can provide a probabilistic, spatial distribution of 432 particle paths. All hydraulically-connected locations in the flow field have some probability 433 of having a particle pass through. The total travel time for each particle can be computed, 434 and the average travel time for all particles passing through a stationary part of the domain 435 can be computed as well. Particle travel paths are limited, however, to the instantaneous 436 WSE gradient and discharge in the steady flow field, which may only be representing a 437 particular snapshot in time. The flow field may show certain areas of the floodplain as 438 connected hydraulically, but the instantaneous directionality of the water fluxes may cause 439 only certain trajectories to be feasible. 440

A flow field that changes through time, due to the rising and falling of the flood wave 441 or unsteady precipitation on the floodplain, creates an environment where potential particle 442 paths are highly dependent on when and where particles enter the floodplain from the river. 443 For example, a particle will not move from river to floodplain until the river stage reaches 444 an elevation higher than the elevation of the deepest floodplain channels. Even then, if 445 the floodplain is already inundated from rainfall, the gradient may not allow river water 446 into the floodplain. Only at a higher river stage might the flow direction change. Routing 447 particles through an unsteady flow field is critical to understanding these river-floodplain 448 interactions. 449

The ANUGA model depth, stage, and momentum outputs were interpolated to a 450 2-m raster grid, and a new particle "cohort" consisting of 1,000 particles was initialized 451 in the domain every 15 minutes of model simulation time. Two classes of particles were 452 analyzed: river particles and floodplain particles. All river particle cohorts were initialized 453 at the inlet of the domain, while floodplain particles were seeded randomly throughout the 454 floodplain in grid cells with depth greater than 20 cm. Separating particles into these two 455 classes is necessary for distinguishing between patterns of rainfall and river flood processes. 456 Floodplain particles were initialized beginning at simulation hour 45 (the onset of intense 457 rainfall, see Figure 3), while river particles were initialized at simulation hour 60, as flow 458 does not move from river to floodplain until sometime after hour 60. All particle cohorts 459 were routed through the model flow field until simulation hour 120. With 1,000 particles per 460

15 minutes, the total number of river particles tracked was 240,000, and the total number
 of floodplain particles was 300,000.

<sup>463</sup> Particle dynamics were quantified in two ways: velocity distributions and residence time distributions. Velocity distributions show the spatial extent of particle paths, as well as the average speed at which particles move through each 2-m grid cell in the model domain. The average time a particle spends in cell (x, y) is calculated as follows:

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

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

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$$V_{avg,xy} = \frac{dx}{t_{avg,xy}} \tag{4}$$

where dx is the cell size. A Gaussian smoothing filter with standard deviation of 0.7 was applied to the  $V_{avg,xy}$  array to reduce noise and enhance visualization.

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

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

where  $N_{p,tot}$  is the total number of particles that enter a control volume,  $dN_p/dt$  is the rate at which particles exit, and at  $t = \infty$ , F(t) = 1. For this study, we define the control volume as the entire river-right floodplain in the model domain. We track individual particle travel times beginning when they enter (or are seeded in) the floodplain, and ending when they leave:

483 
$$t_p = \sum_{i=1}^{N_i} T_{p,i}$$
(6)

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

#### 4.3 Scaling Analysis to Quantify Lateral Flux

We perform a model scaling analysis, with a goal of quantifying lateral flux between 489 river and floodplain as floodplain channels of various sizes are smoothed out of the model. A 490 set of additional model meshes was developed for this task, each with varying resolution in 491 the subdomain area outlined in Figure 1C. In addition to the 2-m model described in Section 492 4.1, mesh resolutions of 5 m, 10 m, and 20 m were evaluated, with total element counts of 493 269,361; 129,051; and 78,752; respectively. Each mesh had the same outer boundary, and 494 the same resolution across the majority of the domain (20 m). All model forcings and other 495 characteristics described above were applied equivalently to each model. 496

The largest floodplain channel in the model domain is close to 50 m wide, and is located 497 at the western edge of the floodplain on the river right (Figure 1C). However, this channel 498 is not directly connected to the river, and instead drains a wetland (Champion Lake) in 499 the floodplain just upstream of the study area. Within the high-resolution subdomain, the 500 largest floodplain channel is about 10-m wide, which can be seen in the DEM (Figure 1C) 501 along the western edge of the subdomain boundary. In general, channel widths range from 502 this upper limit of 10 m down to the scale of 1-m DEM. The channel leading to the Site 503 4 location varies in width, and is mostly between 6 and 8-m wide (Figure 1D). The Site 5 504 location is within a channel that is 4 to 5-m wide. As model resolution is coarsened from 2 505 m, these channel features become smoothed out (Supporting Information Figure S2). 506

Lateral flux is computed from each model by drawing several transects parallel to the river at or near the levee crests, at locations where river-floodplain flow connectivity is significant, and computing the time series of flow through each transect. Transect locations were drawn at locations where *dorado* particles entered the floodplain from the river. This calculation shows which channels are sensitive to model resolution, and the extent to which overall flow into the floodplain changes as these channels are smoothed out of the model.

# 513 5 Results

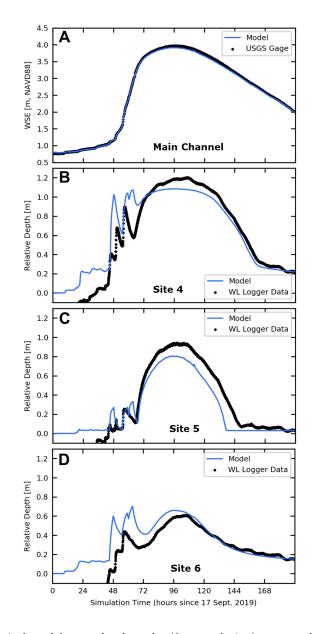
514

#### 5.1 Numerical Model

The numerical model was calibrated to match the observed WSE (relative to NAVD88) 515 at the USGS Moss Bluff gage, and therefore the match between model and observed water 516 level in the main channel is almost exact (Figure 4A). Modeled flow depths in the floodplain 517 were compared with those of the water level loggers at field Sites 4, 5, and 6 (Figure 4B– 518 D). From the beginning of the event (hour 0) to the fluvial peak (hour 96), the range 519 of measured water levels at each site was consistently larger than those from the model. 520 As discussed in Section 3.2, however, vertical positions of the loggers with respect to the 521 floodplain topography are unknown and therefore could not be used as ground-truth depths. 522 For this reason, the measured water levels are most accurately viewed in relative terms. 523

To evaluate model results in the floodplain against measurements, the measured water 524 level time series was set equal to the model at the end of the 8-day simulation (Figure 525 4B-D). At this time, water levels at all three sites began to flatten out toward a steady-526 state condition after river stage dropped below the range of elevations at which river water 527 connects with the floodplain. The model showed similar rates of drainage at each site during 528 this period, and thus simulation day 8 was considered an appropriate point to equate the 529 water level data, as there was less water level change occurring at this time and rainfall and 530 river discharge were no longer actively influencing the sites. 531

The rate of change of water levels seen in the data during the fluvial peak is captured 532 well in the model at all three sites. The sites are located at various distances from the river. 533 but in each case the model was able to move water to various positions in the floodplain 534 at similar rates shown by the data. At Site 4 (Figure 4B), the timing and rate of drainage 535 during the falling limb is particularly aligned with the data. Of note, however, is that the 536 peak depth from rainfall (between hours 48 and 72) is almost exactly equal to the peak 537 depth during fluvial flooding. The maximum flow depth of about 1.1 m corresponds to the 538 depth of the floodplain channel at this location in the lidar, which shows the channel was 539 reaching bankfull flow in the model at each of these times. It would take a significantly 540 greater flow rate in the river to increase this depth, as the floodplain in this area would 541 have to be fully inundated. Instead, it is clear there is a discrepancy between the 2017 lidar 542 and the 2019 channel topography. The measured depth at the time of installation was 0.9543 m, and the logger showed a rise in water level 1.0 m beyond the initial level, implying a 544 total measured channel depth of at least 1.9 m. As such, it is unsurprising that there is a 545 significant vertical offset between modeled and measured water levels at this location. 546



**Figure 4.** Numerical model water level results (2-m resolution) compared to measured values. (A) WSE plot showing the match between stage measured at the USGS Moss Bluff gage and that of the model. (B) Water level comparison in the Site 4 floodplain channel. (C) Water level comparison in the Site 5 floodplain channel. (D) Water level comparison at Site 6, at the terminus of the floodplain channel containing Site 5.

The rainfall signal at Site 5 was less than at the other two sites (Figure 4C). Located 547 on the levee only 50 m from the river, the area draining to Site 5 is much more limited. Due 548 to its proximity to the river, though, the full signal of the flood wave was observed in the 549 data and the shape of the curve was almost identical to the stage curve at the nearby USGS 550 station. The model also showed a flood wave through this channel with a similar shape to 551 the data, peaking at a depth of 0.8 m. By the time the river flood wave receded, the model 552 at this location dried up completely, while the data showed water remaining in the channel 553 (40 cm, based on depth at install). The Site 5 channel slopes gradually down the levee (at 554 roughly 0.08%), so the presence of a near-constant water depth without additional rain or 555

river input suggests the logger may have been in a local depression in the channel deeper than indicated by the lidar.

Modeled water depth at Site 6, located just beyond the terminus of the Site 5 channel, also showed a pattern of drainage similar to measured water levels at times beyond the fluvial peak. However, the peak pluvial and fluvial depths in the model were almost identical (0.7 m) but they were different from each other in the data. Like Site 4, water levels increased dramatically from the heavy rainfall, then decreased slightly, before increasing again beyond the maximum level reached during the rainfall period. It is likely that standing water at this location (initial measured water depth here was 57 cm) causes a disagreement between the lidar data and the true bottom of the floodplain basin.

With each of the three successive rainfall bands, the floodplain became increasingly 566 inundated. Maximum inundation extent occurred at simulation hour 63, corresponding to 567 the end of the third rainfall band, where inundation extent was evaluated over the entire river-right floodplain (excluding isolated areas to the river-left) and included all areas with 569 at least 10 cm of depth. At this time, 65 percent of the floodplain was inundated. The 570 floodplain drained between the end of the last rainfall band and the time of river influence. 571 Peak inundation from river flooding occurred at simulation hour 96, when 55 percent of the 572 floodplain was inundated (see Supporting Information animations for modeled changes in 573 inundation extent during the storm). The differences in inundation extent suggest that, for 574 a sub-bankfull flood event lasting only a couple of days, river water may be limited to a 575 smaller portion of the floodplain based on the number and orientation of floodplain channels 576 facilitating this connectivity. 577

#### 578 5.2 Particle Routing Analysis

Although the water depths in the floodplain did not exactly match the data, the results 579 of the numerical model showed rates of change and overall hydrodynamic patterns similar 580 to the data. The model can be viewed as a realistic representation of the type of conditions 581 in the Trinity River floodplain during Tropical Storm Imelda, where both pluvial and fluvial 582 flooding were major factors. Using the model flow field for particle routing helps describe and 583 quantify the complex interactions that can occur in low-gradient river floodplains during 584 similar events. By continuously seeding passive particles in the river and floodplain, we 585 can observe the differences between water moved by rainfall and river flooding, and how the 586 dominant forcing can change in the floodplain during a storm. Particle velocity distributions 587 show the spatial distribution of particle paths in two dimensions, along with their average 588 velocities. Residence time distributions (RTDs) inform on flow time scales for particles that 589 move through the floodplain. Both show the distinction and interaction between rainfall 590 and river processes. 591

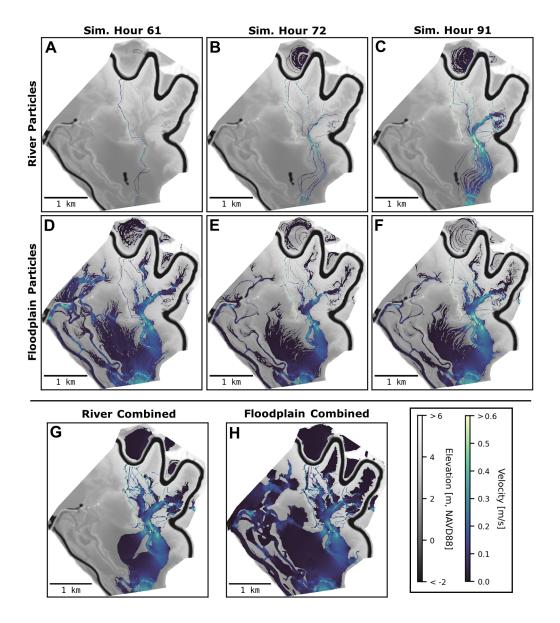
Two particle classes were seeded continuously every 15 minutes throughout the storm 592 event: one in the main channel and one distributed randomly throughout the floodplain (see 593 Supporting Information Movies S1 and S2, respectively, for particle animations). Floodplain 594 particles were only seeded in grid cells where water had accumulated to a depth of 20 cm 595 during the prior time step. Particle velocity distributions show the spatial extent of particle 596 paths in two dimensions (Figure 5). During simulation hour 61 while it was still raining 597 heavily, the gradient was from floodplain to river, and river particles remained confined, even 598 though the floodplain was inundated everywhere except the topographic ridges (Figure 5A, 599 seen in lighter shades of gray). At the same time, floodplain particle paths were widely 600 connected (Figure 5D). At simulation hour 72 (panels B and E) it was no longer raining heavily, but the peak river discharge had not arrived yet. Some of the rainwater had drained 602 from the remote areas of the floodplain, and water pooled in the larger floodplain basins 603 slowed down (darker colors) as it left through the outlet to the south. More river particles 604 began to enter the floodplain, mostly from the counter point bar due south of Site 6, but a 605 few began to enter through the floodplain channels near Sites 4 and 5 as well. Finally, at 606

simulation hour 91 (panels C and F), the river discharge was at its maximum. Floodplain particles (panel F) were limited to the larger floodplain basins, similar to the previous time stamp, but average velocities were slightly higher overall. This is because the river was supplying more water to the floodplain, and thus providing a stronger gradient to the floodplain outlet that was not present at simulation hour 72. As expected for particles originating in the river (panel C), the travel paths were limited to just a fraction of the floodplain, even during peak discharge.

The combined velocity distributions (Figure 5G and H) were computed by taking 614 615 the average velocity for all particles spending time in a given grid cell. The combined distribution for river particles (Figure 5G) shows that river particle paths always remained 616 within the corridor shown in Figure 5 panels B and C. The velocities also show that river 617 water generally spent less time in the floodplain than rainwater, with the exception of the 618 floodplain in the north corner of the domain, which is very deep and highly-connected to the 619 main channel. The large, dark-colored region of lower floodplain particle velocities (Figure 620 5H) was inundated throughout the storm, but river particles never reached it. Instead, river 621 particles seemed to bypass this part of the floodplain entirely, while rainwater spent much 622 more time in this area as it drained slowly to the outlet. 623

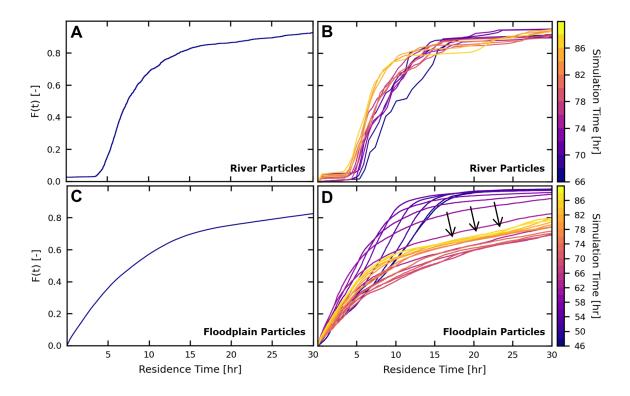
For river particles, residence time distributions (RTDs) were combined for cohorts 624 seeded between simulation hours 66 and 90, as there were not enough particles entering the 625 floodplain at earlier times (Figure 6A). For floodplain particles, the combined time window 626 is between simulation hours 46 and 90 (Figure 6B). The limit at simulation hour 90 was 627 chosen because particles were only tracked up to simulation hour 120, and the residence 628 time window observed was limited to 30 hours (Figure 6, x-axis). Ninety-five percent of 629 river particles spent a minimum of five hours in the floodplain (within the model domain), 630 and about 80 percent of particles had residence times less than 10 hours. The five percent of 631 particles with residence times of less than five hours were those that entered the floodplain 632 briefly before returning to the river. The narrower distribution confirmed what can be 633 seen spatially in the velocity distributions (Figure 5G). Floodplain particles had a wider 634 distribution of residence times. Many particles exited the domain quickly if seeded close to 635 the outlet, but 20 percent of floodplain particles remained in the domain for longer than 30 636 hours, compared to just 10 percent for river particles. Note that the river particle RTDs are 637 composed of less particles by several orders of magnitude, because only a smaller fraction 638 of particles move to the floodplain from the river compared to those that are seeded in the 639 floodplain initially. Also note that the southeastern-most corner of the domain was masked 640 out for the particle analysis because too many river particles were entering the floodplain 641 at this bend and immediately exiting the floodplain due to proximity only, not from faster 642 flow velocities, and this skewed the residence time distributions. 643

Particle RTDs evolved over the course of the storm (Figure 6B and D). At simulation 644 hour 66, only a small number of river particles entered the floodplain, but that number in-645 creased as the storm transitioned to the fluvial phase (Figure 6B). Through this transition, 646 river particles experienced a reduction in minimum residence time as the discharge increased. 647 At higher discharges in the river, more flow moved through the floodplain, increasing veloci-648 ties and reducing residence times. Floodplain particle RTDs show a wider range of behavior, 649 as there may be more competing factors involved in their movement (Figure 6D). The RTD 650 for the earliest group of particle cohorts, representing most of the 8,000 particles seeded 651 between simulation hours 46 and 48, shows that 90 percent of particles left the floodplain 652 after 15 hours. Fifteen hours corresponds to simulation hour 61, when heavy rainfall was 653 still active. Although the rainfall stopped and started twice during this 15-hour period, the 654 overall period of rainfall flushed the floodplain to some degree, and the result was a nearly 655 uniform distribution. Moving forward in time, the sixth group of cohorts (simulation hours 656 56-58) marked a transition in the RTD where a greater fraction of particles left the flood-657 plain faster, but the remaining particles spent longer than those from 10 to 12 simulation 658 hours prior. The transition can be attributed to the period between heavy rainfall and peak 659



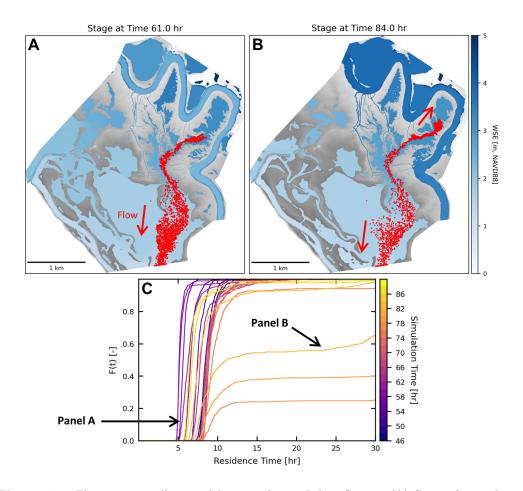
**Figure 5.** Average particle velocities for river particles (top row) and floodplain particles (second row). (A), (B), and (C) Velocity distributions for river particles at simulation hours 61, 72, and 91, respectively. (D), (E), and (F) Velocity distributions for floodplain particles at simulation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

discharge. The longer residence times represent particles stranded in remote areas of the 660 floodplain as rainwater drained, and the shorter residence times were a result of floodwaters 661 accumulating closer to the outlet, where newly seeded particles then had less distance to 662 travel to the outlet. Particle cohorts seeded between simulation hours 56 and 62 began to 663 show an increasingly greater fraction with shorter residence times, due to being seeded *after* 664 the longest pause in rainfall (see Figure 3), but also an increasingly greater fraction with 665 longer residence times, due to the heavy rainfall stopping for good after simulation hour 666 63. After the rainfall phase, the RTD became more consistent through time, as the remote 667 areas of the floodplain drained and inundation became dominated by river water. Average 668 residence times reached a maximum (lowest red curves, Figure 6D) before reducing again 669 during peak discharge when velocities were higher (bright yellow curves). 670



**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 simulation hours, where the darkest purple line is the combined RTD of particles seeded between simulation hours 66 and 68, and the lightest yellow line represents particles seeded between simulation hours 88 and 90. (C) Combined RTD for particles seeded in the floodplain between simulation hours 46 and 90. (D) RTDs for floodplain particles. The darkest purple line represents particles seeded between simulation hours 46 and 90. (D) RTDs for floodplain particles. The darkest purple line represents particles seeded between simulation hours 88 and 90. The black arrows represent the time after rainfall ended, when river stage was increasing.

At field Site 6, located at the terminus of a floodplain channel in a small floodplain 671 basin, a noteworthy reversal of flow occurred in the model that is described well by particle 672 routing (Supporting Information Movie S3). One hundred particles per 15 minutes were 673 seeded at the Site 6 location. During the early rainfall phase, particles flowed directly to 674 the floodplain outlet with a nearly constant residence time of eight hours (Figure 7A and 675 C, purple curves). After the rainfall stopped, flow paths remained similar, and residence 676 times remained nearly constant at five hours for 80–90 percent of particles. Beginning at 677 about simulation hour 78, as more river flow was conveyed to the floodplain from the local 678 floodplain channel and the bend to the south, the small basin began to fill up, causing new 679 particles to become trapped there (Figure 7B and C, orange curves). This reversal lasted 680 until about simulation hour 84, very close in time to the peak discharge, when the floodplain 681 basin water levels equilibrated with the river, and the flow direction reversed again. Particles 682 then returned to the original flow path, with nearly constant residence times of six hours 683 for 90 percent of particles (Figure 7C, yellow curves). The flow reversal lasted for only six 684 hours, but many particles seeded around this time had residence times exceeding 30 hours 685 (Figure 7C, orange curves). This type of flow reversal represents a drastic change in average 686 residence times, and could have significant implications for floodplain processes when scaled 687 to entire floodplain systems. 688



**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 simulation hour 84. Flow reversal due to fluvial flooding pushed particles farther into the local floodplain basin. The flow reversal lasted from simulation hour 78 to 84, after which particles began flowing back toward the floodplain outlet. (C) RTDs for Site 6 particles grouped every two simulation hours, from simulation hour 46 (darkest purple curve) to 90 (lightest yellow curve). Three orange curves at the bottom right of the plot represent six simulation hours of particles that experienced the flow reversal shown in panel B. The black arrows point to RTDs corresponding to figure panels A and B.

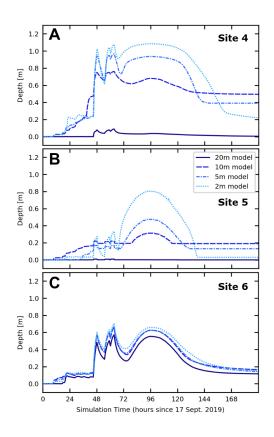
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# 5.3 Role of Floodplain Channels

Changes in numerical model resolution had a varying effect on modeled river-floodplain 690 connectivity. As model resolution was reduced, flow depths through the floodplain channels 691 of Sites 4 and 5 were reduced (Figure 8A and B). These features are completely sub-grid 692 at 20-m model resolution. The flow depth at Site 4 was much greater in the 10-m model 693 than in the 20-m model, although the fluvial signal was dampened compared to the models 694 with further refinement. Because the Site 4 channel is several meters wider than the Site 5 695 channel, the 10-m model resolved the Site 4 channel to a much greater degree (see Supporting 696 Information Figure S2 for channel cross-section geometry at each model resolution). At Site 697 4, the difference in peak depth between the two high-resolution models was 12 cm, or 11 698 percent of the 2-m flow depth, while the difference in peaks at Site 5 was 33 cm, or 41 percent 699 of the 2-m flow depth. This distinction shows that the smaller channel was more dependent 700

on model resolution, although at both sites, capturing the full range of hydrodynamics was
 dependent on resolution finer than the width of the channel.

There was very little distinction between models at the Site 6 floodplain basin (Figure 703 8C). The difference in peak flow depth between the 20-m and 2-m models was only 13 704 cm, while flow depths in the three higher resolution models were virtually identical. It is 705 not surprising that flow depths in the channelized locations were more sensitive to model 706 resolution than depths in the wider, flatter Site 6 location. However, even though Site 6 was 707 located just beyond the terminus of the Site 5 channel, the increased flow quantity delivered 708 709 from Site 5 (Figure 8B) had almost no effect on flow depths at Site 6. Instead, the majority of flow supplied to Site 6 must have originated from sources other than smaller, mesh-scale 710 floodplain channels. 711



**Figure 8.** Model flow depths for model resolutions of 20 m, 10 m, 5 m, and 2 m in the floodplain at (A) Site 4, (B) Site 5, and (C) Site 6.

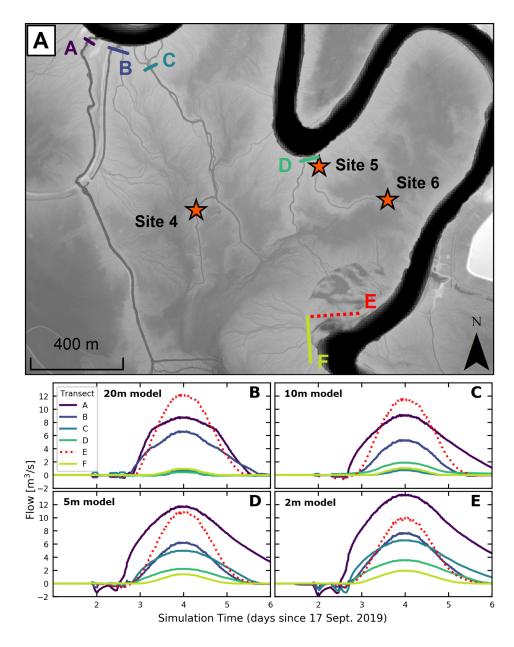
Lateral flux between the river and floodplain changed dramatically at some locations 712 along the levee, and less so at others, with increasing model resolution (Figure 9). Transects 713 in Figure 9 are labeled in increasing order moving downstream, beginning with Transects 714 A and B across the two largest floodplain channels in the domain, Transect C across the 715 channel that leads to field Site 4, transect D across two smaller channels (one of which 716 leads to Sites 5 and 6), Transect E across the wide counter point bar near Site 6 referenced 717 in Section 5.2, and Transect F across a series of smaller levee channels at the same river 718 bend as Transect E. Flow through Transect E is shown as a red dotted line because it is 719 different from the others in that it does not represent a floodplain channel. Flow over this 720 counter point bar into the floodplain was highest in the 20-m model  $(12 \text{ m}^3/\text{s}, \text{Figure 9B})$ 721 and lowest in the 2-m model (10 m<sup>3</sup>/s, Figure 9E), representing a much smaller difference 722 across models than in many of the other floodplain channels. The opposite was true with 723

the large channel at Transect A, where the flow through this channel was highest at 2-m 724 resolution  $(13.5 \text{ m}^3/\text{s})$  and lowest at 20-m resolution  $(8.8 \text{ m}^3/\text{s})$ . The behavior at Transect 725 B was less straightforward, as the peak flow was larger in the two end-member models than 726 in the mid-resolution models, but still the flow here was greater in all models than any of 727 the downstream channelized transects. At Transect C (leading to Site 4), the flow increased 728 from near zero with 20-m resolution to a maximum of  $6.6 \text{ m}^3/\text{s}$  with 2-m resolution, which 729 corresponds to the differences in depth shown in Figure 8A. Similar flow increases were 730 seen with increasing model resolution at Transects D and F, but the total increase was 731 less, as these are smaller channels. Lastly, negative flow through many of the channels, 732 and across the counter point bar, during the period of heavy rainfall represents flow into 733 the channel from the floodplain. The magnitude of reverse flow increased with increasing 734 model resolution, particularly at Transects A and C. This result shows that some floodplain 735 channels can be important conveyors of bidirectional flow between river and floodplain. 736

Despite some of the differences shown in Figure 9, particle dynamics in the overall 737 floodplain were largely unaffected by model resolution. Floodplain residence times were 738 very similar for model resolutions of 2 m, 5 m, and even 10 m (Figure 10A and B). For both 739 particle classes, some differences were observed with the coarser, 10-m resolution, but for 740 the most part the residence time distributions look as they do in Figure 6. This result is 741 supported by the model flow depths at Site 6, where little difference was observed between 742 models of varying resolution (Figure 8C). Although the locations with the highest flow rates 743 entering the floodplain did experience flow changes at different resolutions (Figure 9), the 744 difference in volumes was not as significant as suggested by the model results at Sites 4 745 and 5 (Figure 8A and B). The similarity in floodplain RTDs at different resolutions may 746 be due to the consistent influence of the largest sources of flux (Transects A, B, and E) 747 across all model resolutions, implying that the smaller floodplain channels (e.g., Sites 4 and 748 5, Transects C and D) are less important contributors of flow to the floodplain. 749

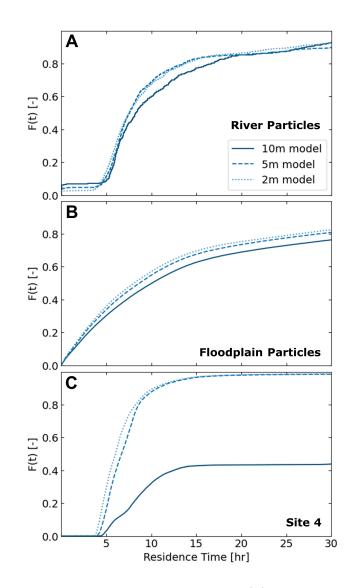
Particles released within the Site 4 floodplain channel reinforced the idea that there 750 can be a significant effect of resolution locally (Figure 10C). RTDs for the 2-m and 5-m 751 resolution models were almost identical, but the 10-m model's failure to resolve the Site 752 4 channel well completely changed the conveyance through the channel. Despite the fact 753 that the majority of the floodplain beyond this local channel had a similar flow field in 754 the model at all mesh resolutions, the lack of resolution here served as a local bottleneck 755 for river water that would have otherwise moved to the floodplain through this particular 756 channel. However, some fraction of particles still moved through the floodplain channel, 757 even at 10-m resolution. Particle animations (Supporting Information Movie S4) show that 758 flow was only conveyed through this channel when rainfall was active, and directly following 759 the peak discharge. During the pauses in rainfall, the flow drained from the channel and 760 particles became stuck, to be flushed out when the rainfall resumed. Particles remained 761 stuck in the channel for the period between heavy rainfall and peak discharge (simulation 762 hours 63-92), after which particles were conveyed through the floodplain due to sufficient 763 river flow. Flow was cut off once again at simulation hour 112. This result aligns with 764 the range of depths at this location in the 10-m model (Figure 8A), and the range of flows 765 (Transect C, Figure 9), where the window of changing depths in the channel was much more 766 limited. In general, the RTDs for Site 4 particles and the model depth curves (Figure 8A 767 and B) show the importance of model resolution on local processes where features near the 768 scale of the mesh resolution are relevant. 769

The impact of model resolution could be seen in the channels close to Site 4 as well. Field observations at Site 4 did not show any flow reversals as the event transitioned from rainfall to river-dominated (Figure 2B), and thus all particles released at Site 4 flowed south into the floodplain (locally, although Figure 9E shows that a flow reversal occurred in the model closer to the river). But particle animations (Supporting Information Movie S5) show that for particles released at Site 4 during the rainfall phase, a portion of flow was siphoned into the larger floodplain channel just west of Site 4 (corresponding to Transect A



**Figure 9.** Modeled lateral flux between river and floodplain. (A) Site map of transect locations where lateral flux is computed. Transect E represents a non-channelized source of exchange. Flow through transects in (B) 20-m model, (C) 10-m model, (D) 5-m model, and (E) 2-m model.

<sup>in Figure 9), where local rainfall was moving water into the river. At the time when rainfall
stopped, the flow reversed, and the gradual rise in river water levels did not allow particles
to move into the river through these channels any longer. Particles only moved in this way
with model resolution of 5 m or finer. The 10-m model could not resolve a deep enough
floodplain channel to convey particles.</sup> 



**Figure 10.** RTDs with different model resolutions, for (A) river particles released between simulation hours 66 and 90, (B) floodplain particles released between simulation hours 46 and 90, and (C) particles released in the Site 4 floodplain channel between simulation hours 46 and 90.

782 6 Discussion

#### 783

# 6.1 Pluvial and Fluvial Flooding Interactions

Field data collected in the Trinity River floodplain during Tropical Storm Imelda 784 showed distinct signals of flooding from the river and from rainfall. The relative timing 785 of rainfall and the peak discharge at the study site created an interesting transition of 786 floodplain hydrodynamics from being pluvial-driven to fluvial-driven. The data showed 787 that with heavy rainfall on the floodplain, river-floodplain connectivity can occur many 788 hours (in this case about 24 hours) prior to the flood wave, and that this connectivity is 789 influenced by floodplain channel topography. Furthermore, the extent of connectivity may 790 be reduced or removed completely if rainfall intensity lessens or stops altogether for a period 791 of time before peak river discharge. The data make clear that pluvial flooding can be an 792 important component of river-floodplain connectivity. 793

Numerical modeling and particle routing analysis reinforced many of the patterns seen 794 in the data relating the timing mismatch of pluvial and fluvial flooding to bidirectional 795 connectivity between the river and floodplain. Studies have shown that river-floodplain 796 connectivity can be established at river stages less than bankfull (Mertes, 1997; Nicholas & Mitchell, 2003; Trigg et al., 2012; Czuba et al., 2019), but the current study showed 798 that connectivity can be established from pluvial flooding at river stages even less than the 799 elevation of the deepest floodplain channels. In fact, model data and measurements from Site 800 1 show that heavy rainfall on a saturated floodplain can provide a competing force against 801 river waters that would otherwise enter the floodplain. And while flow directed toward the 802 river may only occur during a certain phase of a storm, even when a flow reversal occurs 803 river flux into the floodplain may be limited by a reduced gradient from the presence of 804 rainwater. 805

However, many areas of the floodplain can be activated by pluvial flooding that may 806 not otherwise be reached by river water. The large rain bands observed during the storm and 807 applied to the model inundated a majority of the floodplain (Supporting Information Movie 808 S2). In many locations, the rainwater drained rapidly (even between successive bands), while in other locations it collected and slowed down. If given enough time at peak discharge, 810 river water may be able to reach more remote areas of the floodplain. But for events like 811 Imelda where the discharge is sub-bankfull and the flood wave lasts for only a couple of days, 812 river flooding is limited in time and space. In this case, the presence of substantial pluvial 813 flooding on the floodplain reduces the available floodplain storage and increases depths, 814 potentially preventing river water from reaching those areas of the floodplain that are not 815 as directly supplied by flow from floodplain channels. 816

The results discussed here can and should be considered in other river systems with 817 similar characteristics. In lowland systems with the potential for intense precipitation, such 818 as other rivers near the Gulf Coast, pluvial flooding can be a major factor. It is likely, then, 819 that similar patterns of competition between pluvial and fluvial flooding can occur during 820 sub-bankfull discharge. In a different location, even somewhere else along the Trinity River, 821 the results of this study (floodplain residence times and velocities) would likely change to 822 some extent due to topographic differences. However, we anticipate that similar patterns 823 would emerge. The takeaways below of nutrient removal and sediment transport can apply 824 to any system where similar flooding conditions are possible. 825

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

Understanding the mechanisms controlling river-floodplain connectivity is important 827 for understanding how many floodplain processes work. Floodplains, especially those near 828 the coast, are known to act as sinks for nutrients present in river water, such as carbon and 829 nitrogen (Tockner et al., 1999; Aufdenkampe et al., 2011; Noe et al., 2013; Wolf et al., 2013; 830 Cheng & Basu, 2017), and for sediment (Tockner et al., 1999; Verhoeven et al., 2001; Schulz 831 et al., 2003; Day et al., 2008; Juez et al., 2019). In some circumstances, floodplains can be 832 a source of dissolved nutrients (Tockner et al., 1999). We have shown that pluvial flooding 833 has a significant role in river-floodplain connectivity, and the implications for floodplain 834 processes related to sediment retention and nutrient removal are numerous. 835

River-floodplain connectivity is typically studied as a process that is initiated from the 836 river. From this viewpoint, river water spreads into the floodplain over a range of sufficiently 837 high discharges, bringing sediment and solutes to the floodplain. This modeling study 838 showed that there can be a competing gradient between fluvial and pluvial floodwaters, 839 which may reduce the river water that moves into the floodplain, and thus reduce the 840 transport of constituents to the floodplain where they are processed. When the river stage 841 becomes high enough for flow to move into the floodplain, the presence of rainwater still 842 impacts the dynamics. Spatial distributions of velocity (Figure 5) from particle routing 843 analysis show that, for the domain studied, the reach of river water is limited to only a 844

fraction of the floodplain. If there had been no rainwater in the floodplain, the river water 845 and its constituents would have room to spread to a much larger area. Because the path of 846 river particles is restricted to a relatively narrow corridor of the floodplain, residence times 847 for river particles may be less than they otherwise would be. During peak discharge, RTDs 848 showed that 80 percent of river particles move through the floodplain in about five to seven 849 hours (Figure 6C). In contrast, particles randomly sampled in the floodplain during peak 850 discharge show a much wider range of residence times depending on when and where they 851 are seeded (Figure 6D). If we removed from consideration the fraction of sampled floodplain 852 particles seeded close to the outlet, the distribution would be even wider. This result 853 indicates that the active (high-velocity) portion of the floodplain is within the corridor 854 of river particle paths shown in the velocity distributions (Figure 5G), and the water in 855 the remainder of the floodplain that mostly originated as rainwater is slower moving and 856 less active (Figure 5H). So although inundation maps would show water throughout the 857 floodplain, these results show that it is possible for river water and its dissolved nutrients 858 to short-circuit a large portion of the floodplain, potentially bypassing crucial floodplain 859 ecosystem processes. 860

The routing parameters used in the particle analysis assume that each particle moves as 861 a passive tracer. Particles, therefore, more closely represent solutes rather than sediment. 862 However, sediment dynamics may be inferred from flow patterns, average velocities, and 863 residence times in the floodplain. Similar to dissolved nutrients, sediment flux from river to 864 floodplain is dependent on the flow gradient, and it is less likely that floodplain sedimentation 865 will occur if the dominant flow direction is toward the river. Again, floodplains already 866 inundated with rainwater may reach equilibrium with the river more quickly, and reduce 867 the window of time where sediment can be transported out of the river. For sediment that does enter the floodplain, sediment deposition is dependent on flow velocities and 869 residence times, which in turn are dependent on the flow interactions that occur during the 870 storm event. Residence times are shortest during peak rainfall and peak discharge when 871 velocities are higher, but in between they can be significantly longer. So pluvial flooding 872 can reduce overall river flow to the floodplain, but the increase in total floodplain volume 873 due to rainwater may increase residence times. The dynamic nature of pluvial and fluvial 874 compound flooding creates conditions for sediment transport and deposition that can change 875 dramatically over the course of an event. In environments where pluvial flooding can be 876 substantial, sediment dynamics should be considered and modeled within this context. 877

#### 878

#### 6.3 Role of Floodplain Channels and Model Resolution

In floodplain systems where connectivity is truly limited to smaller floodplain channels 879 (during sub-bankfull flow conditions), mesh resolution could be critical for modeling lateral 880 exchange. The model domain used in this study was chosen partly because there were several 881 floodplain channels of various scales present that had been shown by field observations to 882 convey significant flow. Model results showed that for processes in the overall floodplain, 883 resolving those channels was not always important. A large fraction of flow from the river 884 was supplied by a river bend that was connected to the floodplain at low WSE and over a 885 longer length than individual floodplain channels. This river bend consists of a low-lying 886 counter point bar, where the bank-line location is not bounded by a levee (Transect E, 887 Figure 9). Meanwhile, the Site 6 location was located at the end of the Site 5 channel, 888 but as the Site 5 channel was smoothed out by decreasing model resolution, water depths 889 at Site 6 showed little sensitivity to resolution changes. Alternatively, flow over the nearby 890 counter point bar changed much less with model resolution, and we can say that connectivity 891 between the river and this portion of the floodplain near Site 6 is less dependent on nearby 892 893 floodplain channels.

The lateral flux analysis showed that the largest floodplain channels conveyed flow to the floodplain at rates similar to the wide counter point bar, and these flows were much less sensitive to changes in model resolution than the smaller channels of Sites 4 and 5.

River particle RTDs, meanwhile, did not change with model resolution. When we combine 897 these observations, it is apparent that flow to the floodplain and flow patterns within the 898 floodplain are controlled by a combination of the largest floodplain channels and wider 899 depressions in the levee. For the smaller channels such as Site 4, model results showed that 900 changes in flow magnitude and direction occur in these channels only when they are resolved 901 sufficiently (Figure 9 and 10). Model resolution can then be important for understanding 902 local processes, and could even be necessary for processes in the larger floodplain for systems 903 where lateral exchange is completely limited to smaller topographic features. 904

905 For numerical modeling applications in other river-floodplain systems, or even other locations on the Trinity River, running low-cost model simulations prior to detailed inves-906 tigation can provide guidance on the major sources of lateral exchange. In some systems 907 it may be the case that the majority of floodplain connectivity is supplied from a small set 908 of large levee depressions such as the counter point bar described above. While in most 909 systems exchange is also likely to occur via smaller floodplain channels, it may not be on 910 a large enough scale to affect overall flow patterns in the larger floodplain. For systems 911 where it is known that floodplain channels are the main drivers of connectivity, it is neces-912 sary to resolve them with mesh resolution finer than the scale of those channels. In either 913 case, understanding which features are important in a system can allow modelers to shift 914 computational resources to the most important aspects of their model. 915

#### 6.4 Importance of Unsteady Modeling

This study described many ways in which floodplain flow patterns can change during 917 a storm. Floodplains can experience periods of rapid pluvial accumulation, draining, flow 918 reversal, and flow deceleration within the span of several days. The complexity of flow 919 through the Trinity River floodplain during combined pluvial-fluvial events shows that it 920 is critical to model these processes in an unsteady way. For applications where the spatial 921 extent of inundation is of interest for various discharges (e.g., Benke et al., 2000; Czuba et 922 al., 2019), steady modeling of river-floodplain connectivity is appropriate. But for problems 923 related to sediment and solute transport into and out of the floodplain, it is crucial to 924 understand how the spatial extent of inundation, flow time scales, and flow directions change 925 over the course of a storm event. 926

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# 6.5 Limitations and Future Work

The residence times computed in this study were useful for determining how travel 928 times change with different hydrodynamic conditions. But the residence times are relative 929 to the size of the model domain, and cannot be used to assess specific contact times needed 930 for nutrient removal from the water column, for example. It is unclear what happens in 931 the downstream floodplain, and how long water might stay there. It is likely that, for 932 the same flow conditions, residence times change significantly moving down-valley through 933 the lower Trinity River floodplains. It may be worthwhile to increase the model domain 934 to a much larger river-floodplain reach. The domain used in this study needed to be small 935 enough to meet computational constraints for the 2-m simulations, but model results showed 936 that large-scale floodplain processes may not depend on high mesh resolution at the scale of 937 smaller floodplain channels. A less costly numerical mesh that identifies critical topographic 038 features beforehand may be sufficient to perform a similar study on a larger scale. At larger 939 scales, there could be a potential compounding effect of floodplain channels that cannot be 940 seen at the scale of the current model domain. We may also see floodplain flow rejoin the 941 river at points downstream. At these scales, conclusions related to absolute residence times 942 943 can be sought.

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

The lidar data used for numerical modeling was collected in early 2017, and it is 954 955 likely that the floodplain topography changed to some degree between then and field data collection (fall 2019). In fact, an even stronger storm (Hurricane Harvey) passed through the 956 region after lidar was collected. Floodplain topography can change over periods of several 957 years, and significant topographic changes have been observed specifically in the Trinity 958 River lidar data between 2011 and 2015 and also between 2015 and 2017 (Hassenruck-959 Gudipati, 2021). Combined with the possibility of lidar error in the floodplain channels, 960 our model results should be evaluated with this source of error in mind. Still, the model 961 was able to produce flow patterns that generally aligned with the patterns in the field data, 962 and is therefore a useful tool for analyzing hydrodynamics in parts of the floodplain where 963 no data was collected. Even if not an exact replicate of conditions during Tropical Storm 964 Imelda, the relative timing and magnitude of pluvial and fluvial flooding applied to the 965 model created unique conditions related to the competing flooding modes that confirm at a 966 larger spatial extent the observations made from the field data. 967

Lastly, the model results carry some uncertainty related to the calibrated discharge 968 and rainfall inputs and the downstream boundary condition on the floodplain, both of which 969 should be considered when evaluating the results of this study. It is unclear whether the 970 quasi-transmissive boundary condition at the floodplain outlet fully represents the backwater 971 during Imelda, and thus whether the rate of floodplain drainage in the model was accurate. 972 This uncertainty is related to the discussion of larger model domains, where an expanded 973 domain that includes the floodplain farther downstream might reduce the sensitivity of the 974 model and particle analysis to the applied boundary condition. Various boundary conditions 975 were tested during the calibration phase, but this part of the floodplain was too low in 976 elevation to have an impact on WSEs at any of the field sites for confirmation. 977

#### 978 7 Conclusions

This study used field observations, numerical modeling, and Lagrangian particle rout-979 ing to examine river-floodplain connectivity along the Trinity River during Tropical Storm 980 Imelda. Field data and modeling showed the complex hydrodynamic interactions that can 981 result from heavy pluvial flooding occurring in conjunction with high, sub-bankfull river flow. 982 Floodplain residence times and flow directions in the floodplain can be strongly dependent 983 on the dominant mode of flooding, and can change rapidly during a storm. Residence times 984 were shorter during the periods of active rainfall and peak discharge, and flow slowed con-985 siderably in between these phases as flooding transitioned from pluvial to fluvial. Particle 986 routing analysis showed that as river flow moved into a floodplain already inundated from 987 rainwater, the spatial extent of river water was limited to a narrower reach of the floodplain. 988 Without pluvial flooding, river water would likely spread farther into the floodplain where 989 storage is available. Some floodplain channels were shown to facilitate two-way connectivity 990 driven by the timing mismatch between pluvial and fluvial flooding. Although the 5 to 10-m 991 floodplain channels in the study area were shown to be conveyors of lateral exchange, overall 992 processes in the floodplain were unaffected by their resolution in the numerical model, as 993 the majority of lateral exchange came from only a few locations. Variability in how the 994 bank line is constructed also plays an important role in river-floodplain connectivity. 995

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 998 for sediment deposition and nutrient retention are limited to only certain phases of a flood 999 event. Enough pluvial flooding occurring prior to peak discharge may prevent river water 1000 from entering the floodplain altogether, effectively reducing sediment and nutrient fluxes 1001 to the floodplain. Pluvial flooding can also decrease velocities and increase residence times 1002 overall, as deep flow can be achieved sooner with less floodplain storage available for the 1003 peak river discharge. This study challenges the prevailing perspective that river-floodplain 1004 connectivity is dependent only on river discharge, and emphasizes the importance of rainfall 1005 as a driver of that connectivity. 1006

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