# STILL WATER INUNDATION MODELING WITH HYDROLOGICAL CONNECTIVITY

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### STILL WATER INUNDATION MODELING WITH HYDROLOGICAL CONNECTIVITY

Keith C. Horen<sup>1</sup>

#### **ABSTRACT**

This publication details a hydrologically connected inundation modeling method developed by the Alaska Division of Geological & Geophysical Surveys (DGGS) using geographic information system (GIS) processing tools. This method improves upon a widely used, GIS-based inundation modeling method commonly called a bathtub model (BTM; Poulter and Halpin, 2008). DGGS's method considers hydrological barriers and connections during the modeling process to map the extent of inundation more accurately (fig. 1). Leveraging Esri's ArcGIS Pro Raster Function tools, DGGS takes advantage of in-memory processing to improve the speed and efficiency of our models. This method specifically models still water inundation at a community-specific level and is not comparable with coastal total water level models, nor is it appropriate for regional or larger areas. With the increasing prevalence of high-resolution digital elevation models (DEM), DGGS's still water inundation model (SWIM) method is an effective, locally replicable alternative to larger, more complex modeling methods, making our approach accessible to municipal and community analysts.



**Figure 1.** Comparison of a simple bathtub inundation model (left) and a hydrologically connected bathtub model (right) in Teller, Alaska. The area in blue is hydrologically connected to the inundation source, whereas the areas in red are protected by barriers and incorrectly depicted as inundated in the simple bathtub model.

### **INTRODUCTION**

Inundation modeling is an invaluable tool for public safety, and accurately modeling inundation extents is a critical step for hazard assessment and forecasting. Various methods have been developed to model inundation extents, the most common and accessible of which is the GIS-

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based, simple BTM described by Poulter and Halpin (2008), which delineates inundation extent (Moorhead and Brinson, 1995; Titus and Richman, 2001) by identifying DEM raster cells with values below a given elevation representing the modeled still water height. The most significant drawback to using a simple BTM is that all areas below the water height input, even those that are disconnected from the flood source by physical barriers, will be shown as inundated, creating unrealistic inundation extents (fig. 1). In contrast, complex hydrodynamic models that rely on specialized software, large computational resources, and expert knowledge may account for hydrological connectivity, but these models are often inaccessible to the average GIS user, especially in Alaska, where necessary data coverage and resources are often lacking (Overbeck, 2017).

Multiple solutions have been suggested to address the hydrological connectivity issue presented by simple BTM. Li and others (2014) incorporated Esri's Cost Distance geoprocessing tool into their BTM method to limit modeled extents to only those raster cells that are connected to the flood source, informing the models of Sekovski and others (2015). Alternately, Poulter and Halpin (2008) applied raster cell connectivity definitions that are equivalent to those found in Esri's Region Group geoprocessing tool used by the National Oceanic and Atmospheric Administration (NOAA, 2017). Either of these geoprocessing tools can be used to assess raster surface connectivity, but the Cost Distance tool is needlessly computationally (and time) intensive unless a cost variable such as slope is incorporated, as demonstrated in the total water level inundation models produced by Perini and others (2016). The Region Group tool also identifies connections between raster cells but, crucially, does not eliminate unconnected groups of cells from the final output the way the Cost Distance tool does, instead categorizing each group of connected cells with a unique identifier. A user can then assess these categorized groups of cells to discriminate those that should be retained from those that should be removed from the final inundation extent. Using the Region Group tool allows users to identify areas blocked by hydrological barriers for removal while maintaining areas that are hydrologically connected by culverts and other subsurface avenues of connectivity that may not be represented within a DEM. As noted by NOAA, this step is necessary because most DEMs do not include "...detailed pipe network analysis or engineering-grade hydrologic [sic] analysis..." (NOAA, 2017).

The SWIM method developed by DGGS primarily differs from previous methods by taking advantage of the built-in, in-memory processing made possible by Esri's ArcGIS Pro Raster Function tools. While in-memory processing was possible in previous iterations of Esri software using Python scripting and utilized by the Cost Distance and Region Group methodologies, the previous GIS-based modeling methods investigated for this paper generated multiple intermediate datasets in excess of their final output (Li and others, 2014; NOAA, 2017; Perini and others, 2016; Sekovski and others, 2015). The SWIM method requires at most two input datasets, and while several temporary raster datasets are created in-memory, only one intermediate table is created as an output, significantly reducing permanent memory storage requirements. An additional, major point of difference between DGGS's modeling method and other GIS-based methods is that hydrological connectivity is addressed without the need for DEM editing, time-consuming visual selection of unconnected regions, or complex modeling software.

#### **METHOD**

DGGS developed the SWIM method in Esri's ArcGIS Pro version 3.2.0 software, though it is advisable to update to the most recent software version available to ensure the Raster Function tool performs as expected. Multiple steps require an ArcGIS Pro Spatial Analyst extension license. Additionally, a DEM in GeoTIFF format is required and, where available and necessary, a line or polygon feature class representing hydrological connections such as culverts. All data should be projected in the same coordinate system, using the same units. This method is compatible with both digital terrain model (DTM) and digital surface model (DSM) rasters, but it should be noted that man-made hydrological barriers may be absent from DTM, while DSM may include man-made features that could act as erroneous barriers, such as bridges. If a feature class of hydrological connections is needed but unavailable, modelers may create and populate one, provided these connective features are known or identifiable within imagery.

The following provides details on the processing steps required to generate a SWIM with hydrological connectivity using DGGS's method. A visual schematic of these steps is provided in figure 2. The location of the tools described in the following steps are provided in parentheticals for reference and unique user input values are denoted by brackets.

## **Processing Steps**

- 1. <u>Reclassify DEM cells below the modeled still water height using a simple BTM.</u> This step outputs an integer raster with values of 1 for inundated and NoData for non-inundated cells, regardless of connectivity.
  - 1.1. Open the Remap tool (Raster Function pane > Reclass > Remap) and set the following:
    - 1.1.1. General
      - 1.1.1.1. Name = Extents
      - 1.1.1.2. Output Pixel Type = 8 Bit Unsigned
    - 1.1.2. Parameters
      - 1.1.2.1. Raster = [DEM]
      - 1.1.2.2. Remap Definition type = List
      - 1.1.2.3. Minimum = [minimum elevation of DEM]
      - 1.1.2.4. Maximum = [modeled still water height]
      - 1.1.2.5. Output = 1
      - 1.1.2.6. Change missing values to NoData = CHECKED
- 2. <u>Group simple BTM cells by spatial connectivity.</u> This step outputs an integer raster with values between 1 and 65,535, each unique value representing a contiguous region of cells. As discussed by Poulter and Halpin (2008), the number of neighboring cells included in a

connectivity assessment is critical to the final extent results. Using a four-cell neighborhood for this step only assesses cells that are orthogonal to the cell being evaluated, while an eight-cell neighborhood includes orthogonal and diagonal cells. An eight-cell neighborhood is recommended for this analysis.

- 2.1. Open the Region Group tool (Raster Function pane > Data Management > Region Group) and set the following:
  - 2.1.1. General
    - 2.1.1.1. Name = Regioned
    - 2.1.1.2. Output Pixel Type = 16 Bit Unsigned
  - 2.1.2. Parameters
    - 2.1.2.1. Zone Raster = Extents
    - 2.1.2.2. Number of Neighbor Cells = Eight
    - 2.1.2.3. Zone Connectivity = Within
    - 2.1.2.4. Add Link = No Link
    - 2.1.2.5. Exclude Value = 0
- 3. <u>Assess hydrological connectivity of grouped cells.</u> This step outputs either a single integer value representing a region that is connected to the inundation source or a table of features and corresponding integer values representing the regions those features connect.
  - 3.1. If a feature class of hydrological connection features is NOT being used, use the Explore tool (Map tab > Navigate panel > Explore) and do the following:
    - 3.1.1. Click on the region(s) of cells that is connected to the inundation source.
    - 3.1.2. Note the Pixel Value provided in the Pop-up pane, then skip to Step 6.
  - 3.2. If a feature class of hydrological connection features (line or polygon) is being used, open the Tabulate Area tool (Geoprocessing pane > Spatial Analyst > Zonal > Tabulate Area) and set the following:
    - 3.2.1. Parameters
      - 3.2.1.1. Input raster or feature zone data = [hydrological connections feature class]
      - 3.2.1.2. Zone field = [numeric attribute field containing unique values] (e.g. OBJECTID)
      - 3.2.1.3. Input raster or feature class data = Regioned\_Extents
      - 3.2.1.4. Class field = Value
      - 3.2.1.5. Output table = Connections\_Table

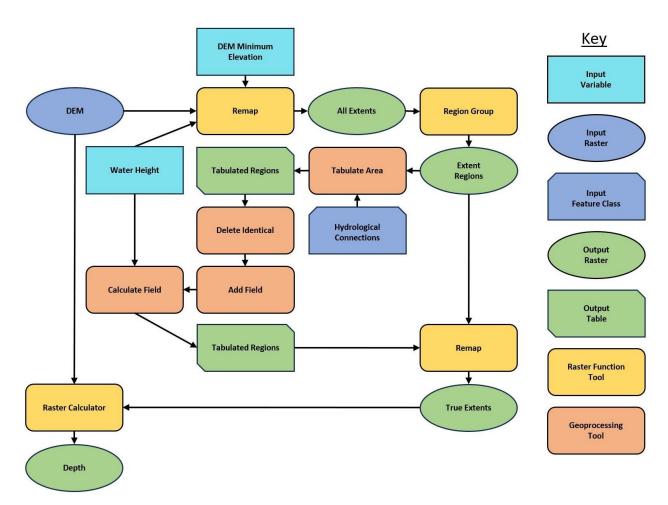
- 3.2.1.6. Classes as rows in output table = CHECKED
- 3.2.1.7. Processing cell size = [DEM resolution]
- 4. Remove duplicate region values from the hydrological connections table. Multiple features may provide connections to a single region. This step ensures only one instance of each region value is present within the hydrological connections table and is necessary for future steps to function correctly.
  - 4.1. Open the Delete Identical tool (Geoprocessing pane > Data Management > General > Delete Identical) and set the following:
    - 4.1.1. Parameters
      - 4.1.1.1. Input Dataset = Connections\_Table
      - 4.1.1.2. Field(s) = Value
- 5. Add and populate a numeric attribute field in the hydrological connections table. This step assigns the modeled still water height as the output value for each of the connected regions and is necessary for future steps to function correctly. Value must use the same elevation units (e.g., feet, meters) as the input DEM.
  - 5.1. Open the Add Field tool (Geoprocessing pane > Data Management > Fields > Add Field) and set the following:
    - 5.1.1. Parameters
      - 5.1.1.1. Input Table = Connections\_Table
      - 5.1.1.2. Field Name = Output
      - 5.1.1.3. Field Type = Double (64-bit floating point)
  - 5.2. Open the Calculate Field tool (Geoprocessing pane > Data Management > Fields > Calculate Field) and set the following:
    - 5.2.1. Parameters
      - 5.2.1.1. Input Table = Connections\_Table
      - 5.2.1.2. Field Name = Output
      - 5.2.1.3. Expression Type = Python
      - 5.2.1.4. Output = [modeled still water height]
- 6. Reclassify only the hydrologically connected cells below the modeled still water height to create a single elevation surface. The output of this step is a floating-point raster depicting the modeled extents with cell values equal to the modeled still water height.
  - 6.1. Open the Remap tool (Raster Functions pane > Reclass > Remap) and set the following:

- 6.1.1. General
  - 6.1.1.1. Name = Connected
  - 6.1.1.2. Output Pixel Type = 32 Bit Float
- 6.1.2. Parameters
  - 6.1.2.1. Raster = Regioned\_Extents
  - 6.1.2.2. If following from Step 3.1.2 (no hydrological connection feature class used), repeat sub-steps 6.1.2.2.2 and 6.1.2.2.3 to add a row for each value corresponding to all regions connected to the inundation source(s). Multiple region value inputs may be required if more than one inundation source is present or a single inundation source is functionally disconnected from itself due to a lack of coverage within the DEM. Skip to Step 7 after running the Remap tool with the following parameters:
    - 6.1.2.2.1. Remap Definition = List
    - 6.1.2.2.2. Minimum = [region value]
    - 6.1.2.2.3. Maximum = [region value + 1]
    - 6.1.2.2.4. Output = [modeled still water height]
  - 6.1.2.3. If following from Step 5.2.1.4 (hydrological connection feature class used).
    - 6.1.2.3.1. Remap Definition = Table
    - 6.1.2.3.2. Remap Table = Connections\_Table
    - 6.1.2.3.3. Input Field = Value
    - 6.1.2.3.4. Output Field = Output
    - 6.1.2.3.5. Input Max Field = (leave blank)
    - 6.1.2.3.6. Remap Table Type = Simple
    - 6.1.2.3.7. Change missing values to NoData = CHECKED
- 7. Generate inundation depth raster. The output of this step is the final product of this model, a 32-bit floating point raster that depicts the inundation depth at the modeled still water height for all hydrologically connected extents.
  - 7.1. Open the Calculator tool (Raster Functions pane > Math > Calculator) and set the following:
    - 7.1.1. General
      - 7.1.1.1. Name = Inundation\_Depth

- 7.1.1.2. Output Pixel Type = 32 Bit Float
- 7.1.1.3. Match Variables = CHECKED
- 7.1.1.4. Union Dimensions = UNCHECKED

### 7.1.2. Parameters

- 7.1.2.1. Raster Variables
  - 7.1.2.1.1.  $c = Connected\_Regioned\_Extents$
  - 7.1.2.1.2. d = [DEM]
- 7.1.2.2. Expression
  - 7.1.2.2.1. Con(c-d > 0, c-d, 0)



**Figure 2.** Schematic of still water inundation modeling with hydrological connectivity processing steps.

#### **PROCESSING EXAMPLES**

In September 2022, DGGS collected high-water marks (HWM) in several communities impacted by flooding caused by Extra-tropical Cyclone Merbok (Horen and others, 2022). We applied this modeling method to two locations: Hooper Bay, where the model could be applied without the use of a hydrological connection feature class (Christian and others, 2024), and Teller, where hydrological connection features were necessary. All data in the following examples are reported in meters in the NAVD88 (GEOID12B) vertical datum.

## **Hooper Bay**

DGGS collected 121 high-quality HWM in Hooper Bay on September 20, 2022, using Global Navigation Satellite System (GNSS) equipment (Horen and others, 2022). We divided these HWM data into two subsets using a random number generator, with one subset (A) used to estimate the flood height and the other (B) used to evaluate that estimate. Subsets A and B consisted of 60 and 61 HWM observations, respectively (appendix A). Using the upper-lower bounds method of uncertainty propagation (University of North Carolina [UNC], 2018), the estimate derived from subset A was  $4.899 \pm 0.644$  m. An individual comparison to this estimate was calculated for each HWM in subset B, resulting in an average agreement of +0.015 m and a root-mean-square error (RMSE) of 0.277 m.

DGGS produced both a simple BTM and a hydrologically connected SWIM using a 1-meter resolution lidar DTM collected in 2016 by the United States Geological Survey (USGS; OCM Partners, 2024). DGGS used the processing steps for applying our method without a hydrological connection feature class to create the SWIM and generated the simple BTM following the processing instructions detailed in Step 1 of the SWIM method. Screenshots of the tools and settings used during processing, as well as visualizations of the resultant outputs, are provided in appendix B.

To compare the extents of these two models, DGGS evaluated the total area of modeled inundation within a 4 km² region centered on the community of Hooper Bay (table 1; fig. B1). Within this region, the simple BTM method modeled an area of inundation covering 3.417 km², while the SWIM method modeled an area of inundation covering 3.370 km². This comparison demonstrates that within this study area the simple BTM produced a seven-percent overestimation of inundation extents in uninundated areas. Field observations corroborated the inundation extents modeled by the SWIM method (Christian and others, 2024; Horen and others, 2022).

Table 1. Simpl	le BTM and	SWIM method	comparison results.
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Method	Total Area (km²)	Inundated Area (km²)	Uninundated Area (km²)
Simple BTM	4.000	3.417	0.583
SWIM	4.000	3.370	0.630

#### **Teller**

DGGS collected 79 high-quality HWM in Teller on September 21, 2022, using Global Navigation Satellite System (GNSS) equipment (Horen and others, 2022). We divided these HWM data into two subsets using a random number generator, with one subset (A) used to estimate the flood height and the other (B) used to evaluate that estimate. Subsets A and B consisted of 39 and 40 HWM observations, respectively (appendix C). Using the upper-lower bounds method of uncertainty propagation (UNC, 2018), the estimate derived from subset A was  $2.813 \pm 0.326$  m. An individual comparison to this estimate was calculated for each HWM in subset B, resulting in an average agreement of +0.033 m and a root-mean-square error (RMSE) of 0.158 m.

DGGS produced two hydrologically connected models using the SWIM method and the Cost-Distance method. We applied each method to a 1-meter resolution topographic-bathymetric (topobathy) lidar DTM collected in 2019 by the U.S. Army Corps of Engineers (USACE; OCM Partners, 2023); this DEM was not edited to reflect hydrological connections. DGGS used the processing steps for applying our method with a hydrological connection feature class to create the SWIM and followed the steps described by Sekovski and others (2015) and Perini and others (2016) to create the Cost-Distance model. The attenuation artifice (Sekovski, 2015) applied during the Cost-Distance method was 0.004; this value acts "as a proxy for bed friction and infiltration over distance from the shoreline" (Perini, 2016) during modeling. Screenshots of the tools and settings used during processing, as well as visualizations of the resultant outputs, are provided in appendix D.

To compare the extents of these two models, DGGS evaluated the total area of modeled inundation within an 8-km² region centered on the community of Teller (table 2; fig. D1). Within this region, the Cost-Distance method modeled an area of inundation covering 5.155 km², while the SWIM method modeled an area of inundation covering 5.800 km². This comparison demonstrates that within this study area the Cost-Distance method produced an eight-percent underestimation of the total inundation extents. Field observations corroborated the inundation extents modeled by the SWIM method (Horen and other, 2022).

**Table 2.** Cost-Distance and SWIM method comparison results.

Method	Total Area (km²)	Inundated Area (km²)	Uninundated Area (km²)
Cost-Distance	8.000	5.155	2.845
SWIM	8.000	5.800	2.700

#### **DISCUSSION**

DGGS's method is not unique in displaying hydrological connectivity, but it can more accurately depict inundation extents by not only accounting for barriers within a DEM but also connections that might otherwise be overlooked in other workflows. Furthermore, DGGS's method utilizes the built-in, inmemory processing made possible by Esri's ArcGIS Pro Raster Function tools, limiting extraneous file outputs and improving processing time. However, DGGS's hydrologically connected inundation modeling method does have limitations.

This method is intended to map the extent of still water height inundation, which does not include variable inundation associated with wave setup and runup. Total water level inundation models (tide + surge + wave setup + runup) require complex input data that are currently difficult to collect or absent from much of Alaska (Overbeck, 2017). Despite this, it may be possible to use proxy variables with other models to emulate total water level inundation. For example, classifying a DEM with Manning's roughness coefficients (Arcement and Schneider, 1989) could provide a cost surface for the Cost Distance tool to be used in conjunction with predicted tide and surge levels to estimate variable inundation extents. Additionally, the hydrological connectivity depicted using DGGS's method does not factor in the capacity and flow rates of connecting features. For the purposes of community-level modeling, it is reasonable to assume still water heights will achieve hydrostatic equilibrium on either side of connecting features, but it is worth noting that fluid dynamic variables may need to be incorporated for more complex pipe networks.

By developing this still water inundation model with hydrological connectivity, DGGS seeks to provide access to accurate inundation modeling at the community scale. Alaska's coastline is more than 75,000 km (46,600 miles) long, accounting for more than half of the coast of the continental U.S. (NOAA, 2022), yet national hydrodynamic models are not currently able to provide adequate coverage for most of the state (Overbeck, 2018). The method presented here is meant to fill this modeling gap with an easy-to-use, user-friendly approach for local officials and GIS users to assess inundation risk for their communities, a vital and necessary step for public safety and community planning.

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## **APPENDIX A: HOOPER BAY HIGH-WATER MARK DATA SUBSETS**

For a full description of these data, including coordinate, accuracy, and description information, reference DGGS Raw Data File 2022-14 (Horen and others, 2022). HWM data are listed by point identification number (PID).

SUBSET A		
PID	PID	
3008	3059	
3009	3060	
3011	3074	
3013	3076	
3014	3077	
3019	3078	
3020	3079	
3023	3083	
3024	3086	
3025	3088	
3026	3091	
3029	3092	
3031	3093	
3032	3096	
3034	3097	
3035	3098	
3036	3106	
3038	3107	
3039	3109	
3040	3111	
3041	3113	
3044	3114	
3047	3115	
3048	3116	
3050	3119	
3051	3120	
3055	3121	
3056	3122	
3057	3125	
3058	3001B	

SUBSET B		
PID	PID	
3002	3070	
3003	3071	
3004	3072	
3005	3073	
3006	3075	
3007	3080	
3010	3081	
3012	3082	
3018	3084	
3021	3085	
3022	3087	
3027	3089	
3028	3090	
3030	3094	
3033	3095	
3037	3099	
3042	3100	
3043	3101	
3045	3102	
3046	3103	
3049	3104	
3052	3105	
3053	3108	
3054	3110	
3061	3112	
3062	3117	
3063	3118	
3064	3123	
3065	3124	
3066	3001A	
3069		

#### APPENDIX B: HOOPER BAY INUNDATION MODEL PROCESSING

The following is a step-by-step example of how DGGS modeled the still water inundation in Hooper Bay resulting from Extra-tropical Cyclone Merbok. We used HWM data (Horen and others, 2022) to estimate the flood height of this event as  $4.899 \pm 0.644$  m NAVD88 (GEOID12B). Using this estimate, we applied both a simple BTM and SWIM to a 1-meter resolution DTM (fig. B1; OCM Partners, 2024).



**Figure B1.** Hooper Bay study area (red outline) with 2016 USGS lidar DTM hillshade basemap (OCM Partners, 2024).

## **Simple BTM Processing**

DGGS followed the processing instructions detailed in Step 1 of the SWIM method (page 3) to produce the simple BTM.

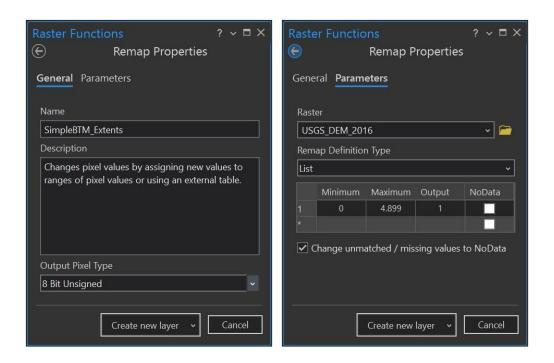


Figure B2. Screenshots of Remap Raster Function tool with settings used to create simple BTM.



Figure B3. Simple BTM inundation extents (blue) at 4.899 m.

## **SWIM Processing**

Step 1 (page 3). Reclassify DEM cells below the modeled still water height using a simple BTM.

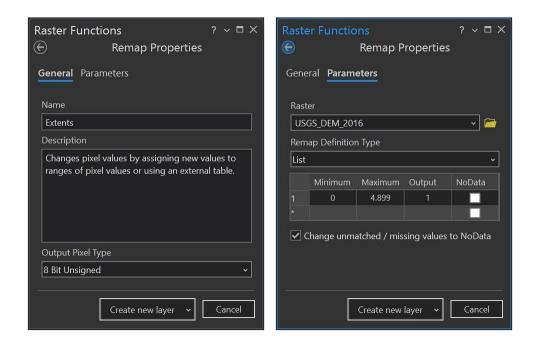


Figure B4. Screenshots of Remap tool with settings used to create Step 1 output.



Figure B5. Step 1 output.

Step 2 (page 3). Group simple BTM cells by spatial connectivity.

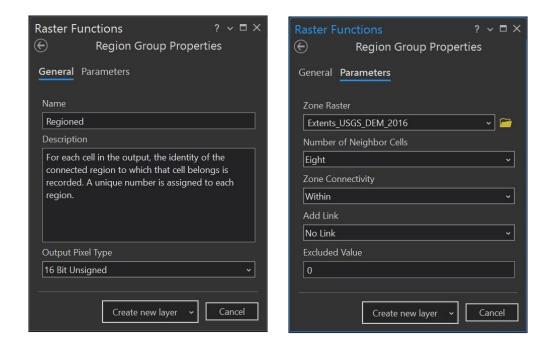


Figure B6. Screenshots of Region Group tool with settings used to create Step 2 output.

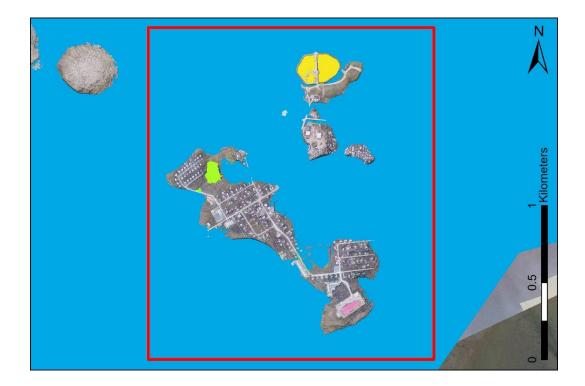


Figure B7. Step 2 output.

Step 3 (page 4). <u>Assess hydrological connectivity of grouped cells.</u> Since a hydrological connection feature class was not used, we followed the instructions of Step 3.1., skipping to Step 6 after noting the Pixel Value of the selected region.

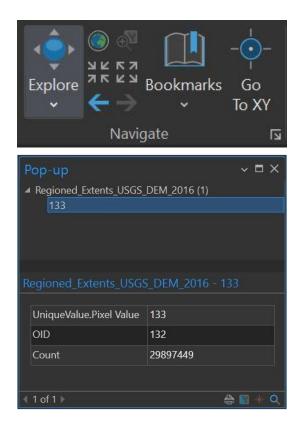


Figure B8. Step 3 Explore tool (top) and Pop-up results (bottom).

Step 6 (page 5). <u>Reclassify only the hydrologically connected cells below the modeled still water height to create a single elevation surface.</u> Since a hydrological connection feature class was not used, we followed the instructions of Step 6.1.2.2.

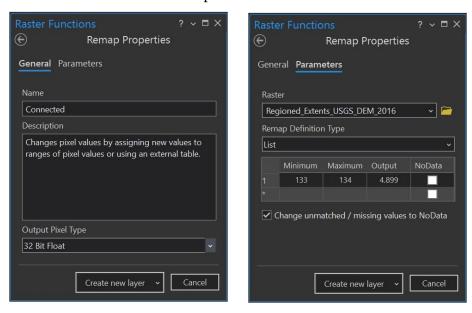


Figure B9. Screenshots of Remap tool with settings used to create Step 6 output.

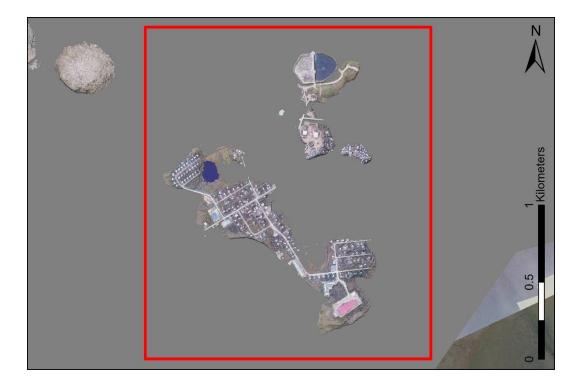


Figure B10. Step 6 output.

Step 7 (page 6). Generate inundation depth raster.

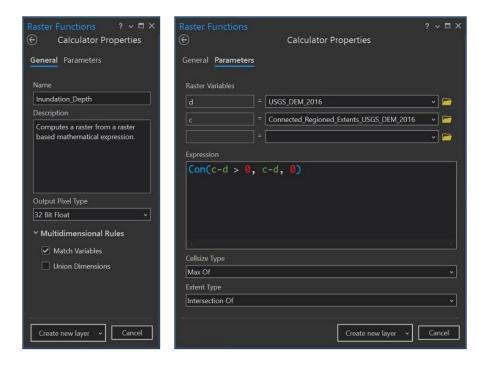


Figure B11. Screenshots of Calculator tool with settings used to create Step 7 output.



Figure B12. Step 7 output.

## **APPENDIX C: TELLER HIGH-WATER MARK DATA SUBSETS**

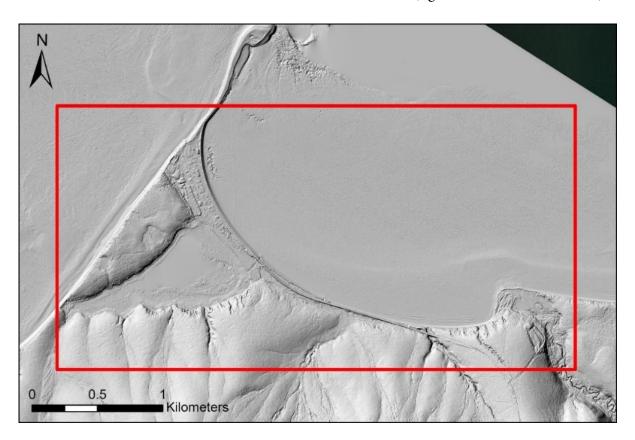
For a full description of these data, including coordinate, accuracy, and description information, reference DGGS Raw Data File 2022-14 (Horen and others, 2022). HWM data are listed by point identification number (PID).

SUBSET A		
PID	PID	
2000	2042	
2002	2044	
2003	2045	
2005	2050	
2006	2052	
2011	2054	
2015	2055	
2018	2057	
2019	2059	
2021	2060	
2023	2061	
2024	2062	
2025	2066	
2030	2067	
2031	2068	
2033	2079	
2036	2082	
2037	2083	
2038	2084	
2039		

SUBSET B		
PID	PID	
2001	2040	
2004	2041	
2007	2043	
2008	2046	
2009	2047	
2010	2048	
2012	2049	
2013	2051	
2014	2053	
2016	2056	
2017	2058	
2020	2063	
2022	2064	
2026	2065	
2027	2069	
2028	2070	
2029	2071	
2032	2072	
2034	2080	
2035	2081	

#### APPENDIX D: TELLER INUNDATION MODEL PROCESSING

The following is a step-by-step example of how DGGS modeled the still water inundation in Teller resulting from Extra-tropical Cyclone Merbok. We used HWM data to estimate the flood height of this event as  $2.813 \pm 0.326$  m NAVD88 (GEOID12B). Using this estimate, we applied both a Cost-Distance model and SWIM to a 1-meter resolution DTM (fig. D1; OCM Partners, 2023).



**Figure D1.** Teller study area (red outline) with USACE 2019 topobathy lidar DTM hillshade basemap (OCM Partners, 2023).

## **Cost-Distance Model Processing**

DGGS followed the processing instructions detailed by Sekovski and others (2015) to produce the Cost-Distance method model.

Step 0. <u>Create a source raster.</u> DGGS used the Remap tool to generate a source raster consisting of cells from the DTM with values less than 0 m NAVD88 (GEOID12B).

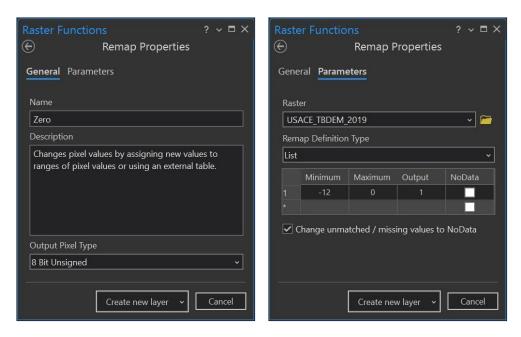


Figure D2. Screenshots of Remap tool with settings used to create Step 0 output.

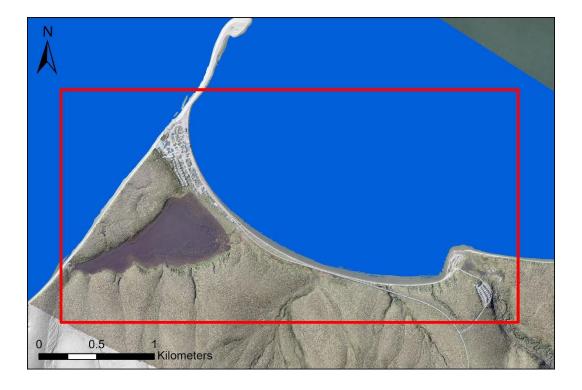
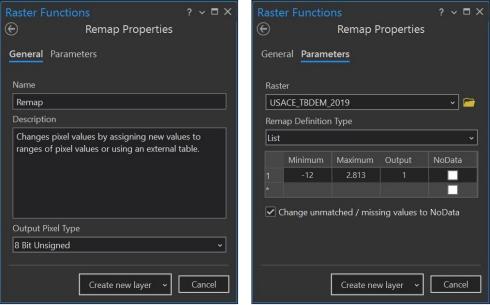


Figure D3. Step 0 output.

Step 1. Reclassify to exclude all elements that have an elevation greater than the modeled water level.



**Figure D4.** Screenshots of Remap tool with settings used to create Step 1 output.

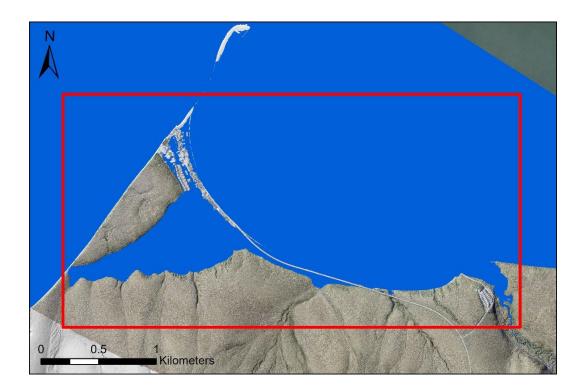


Figure D5. Step 1 output.

Step 2. Apply Cost-Distance tool to generate cost value raster.

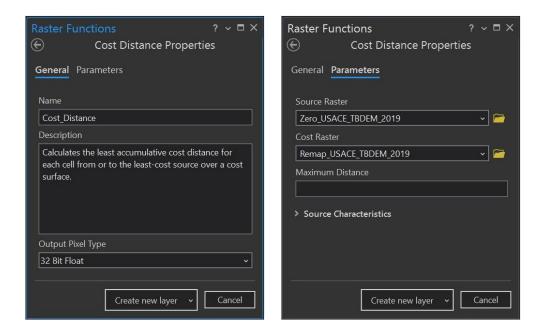


Figure D6. Screenshots of Cost-Distance tool with settings used to create Step 2 output.

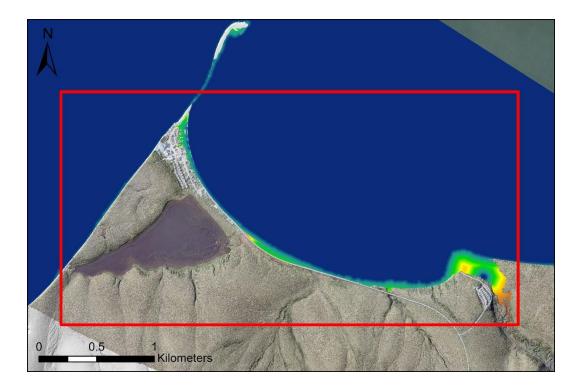
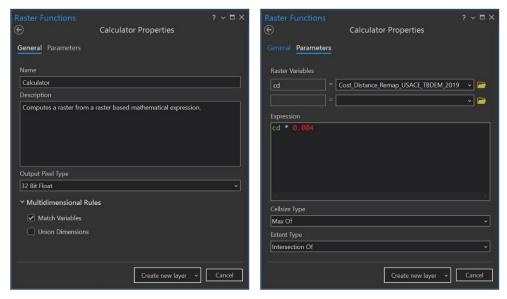


Figure D7. Step 2 output.

Step 3. <u>Multiply the cost value raster by the attenuation artifice to generate a proxy elevation raster.</u>



**Figure D8.** Screenshots of Calculator tool with settings used to create Step 3 output.

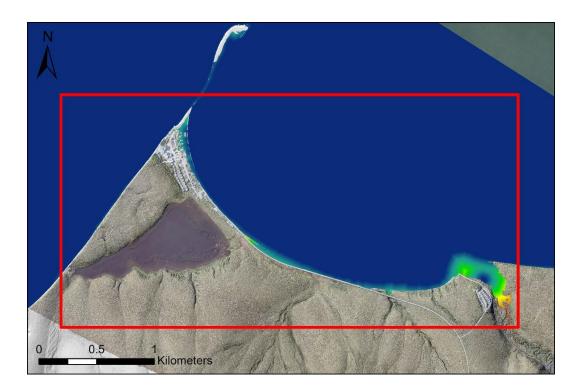


Figure D9. Step 3 output.

Step 4. Reclassify proxy elevation raster to exclude all elements that have an elevation greater than the modeled water level.

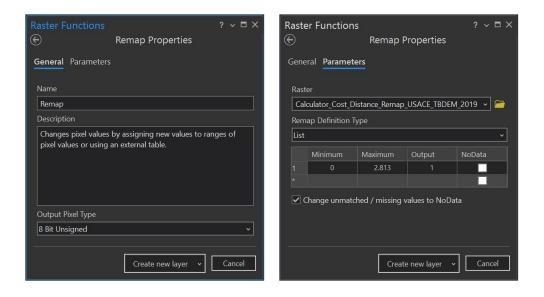


Figure D10. Screenshots of Calculator tool with settings used to create Step 4 output.

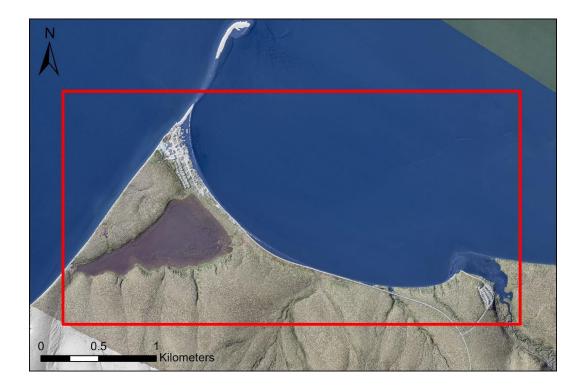


Figure D11. Step 4 output.

# **SWIM Processing**

Step 1 (page 3). Reclassify DEM cells below the modeled still water height using a simple BTM.

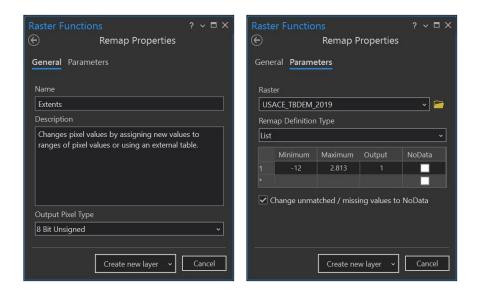


Figure D12. Screenshots of Calculator tool with settings used to create Step 4 output.

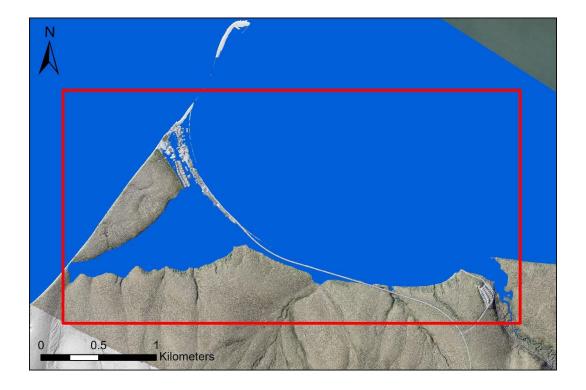


Figure D13. Step 4 output.

Step 2 (page 3). Group simple BTM cells by spatial connectivity.

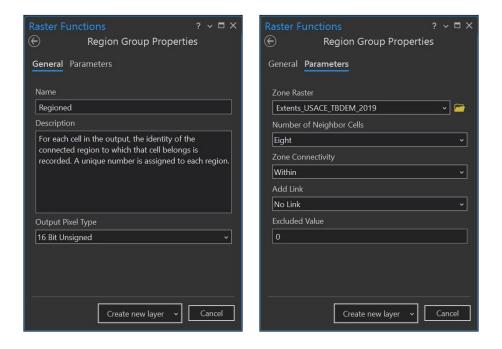


Figure D14. Screenshots of Region Group tool with settings used to create Step 2 output.

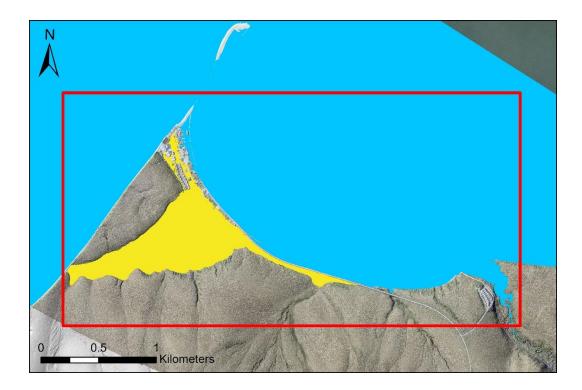


Figure D15. Step 2 output.

Step 3 (page 4). <u>Assess hydrological connectivity of grouped cells.</u> Since we applied a hydrological connection feature class, we followed the instructions of Step 3.2.



**Figure D16.** Example of connectivity features (culverts) used in Step 3 and feature class attribute table.

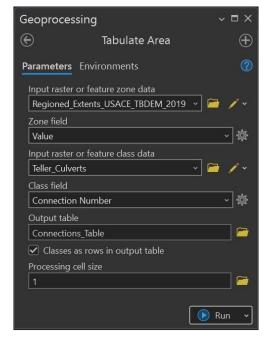


Figure D17. Screenshot of Tabulate Area tool with settings used.

Step 4 (page 5). Remove duplicate region values from the hydrological connections table.

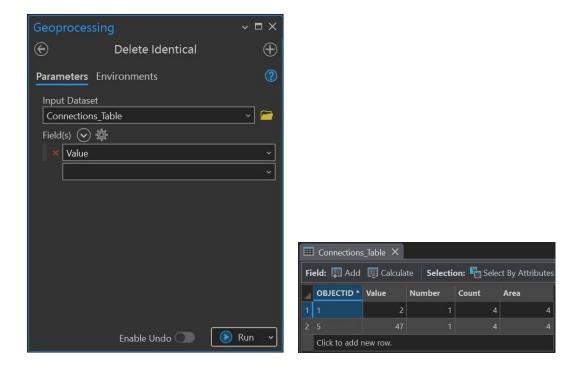


Figure D18. Screenshots of Delete Identical tool with settings used and Step 4 output.

Step 5 (page 5). Add and populate a numeric attribute field in the hydrological connections table.

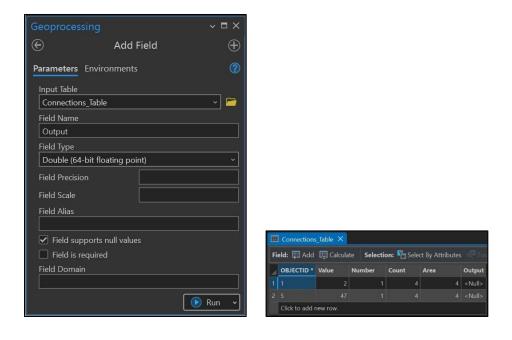


Figure D19. Screenshots of Add Field tool with settings used and Step 5.1. output.

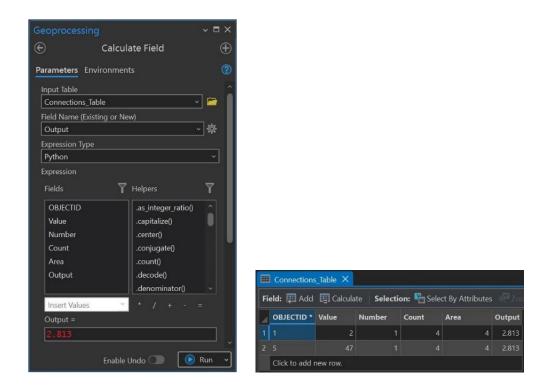


Figure D20. Screenshots of Calculate Field tool with settings used and Step 5.2. output.

Step 6 (page 5). Reclassify only the hydrologically connected cells below the modeled still water height to create a single elevation surface. Since we used a hydrological connection feature class, we followed the instructions of Step 6.1.2.3.

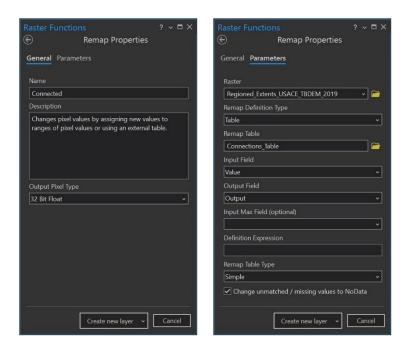


Figure D21. Screenshots of Remap tool with settings used to create Step 6 output.

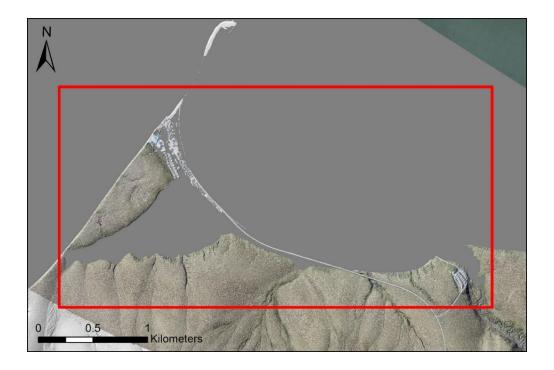


Figure D22. Step 6 output.

Step 7 (page 6). Generate inundation depth raster.

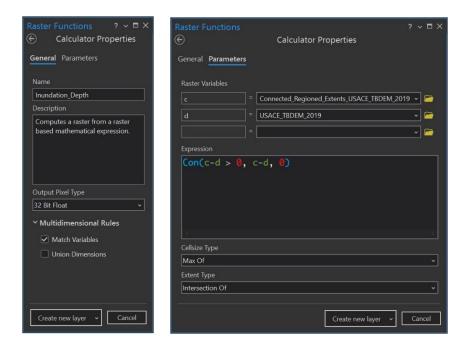


Figure D23. Screenshots of Calculator tool with settings used to create Step 7 output.

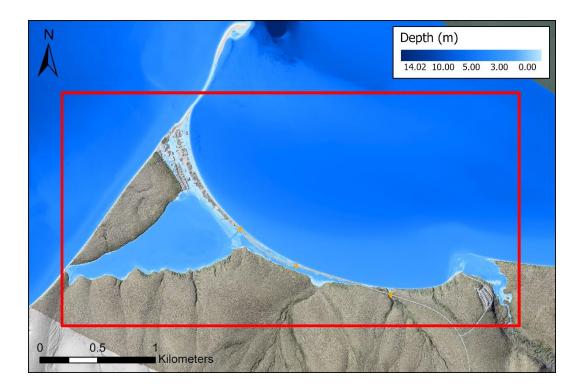


Figure D24. Step 7 output.