Development of a Stage-Specific Floodplain Inundation Model to Predict Suitable Spawning Habitat Availability for Assessing Alligator Gar Recruitment

Clint Robertson and Karim Aziz – TPWD, River Studies Program Nolan Raphelt – Texas Water Development Board Dave Buckmeier and Nate Smith – TPWD, HOHFSC

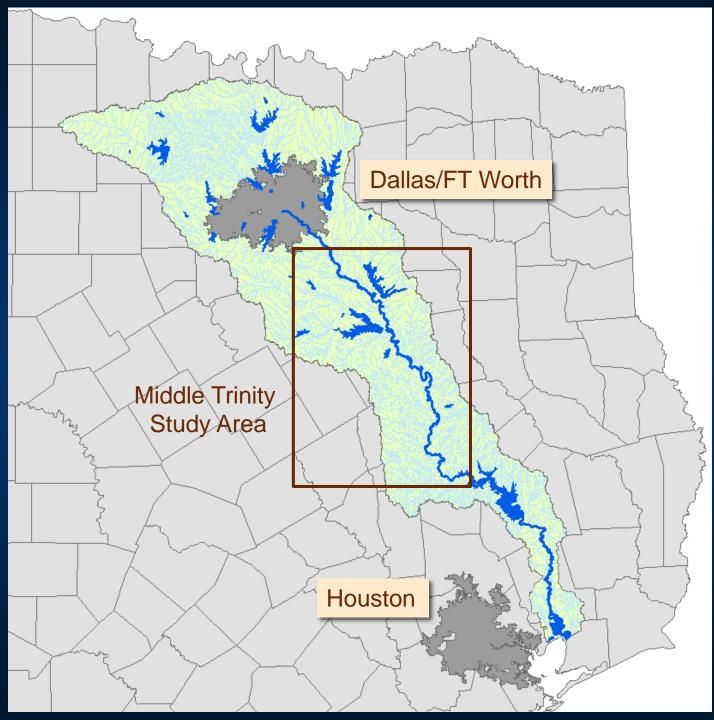


Alligator Gar

- Large floodplain river species that is dependent on floodplain connectivity for successful reproduction
- Most large floodplain rivers throughout their range have been severely altered, and from which they have been extirpated.

Floodplain Inundation Modeling

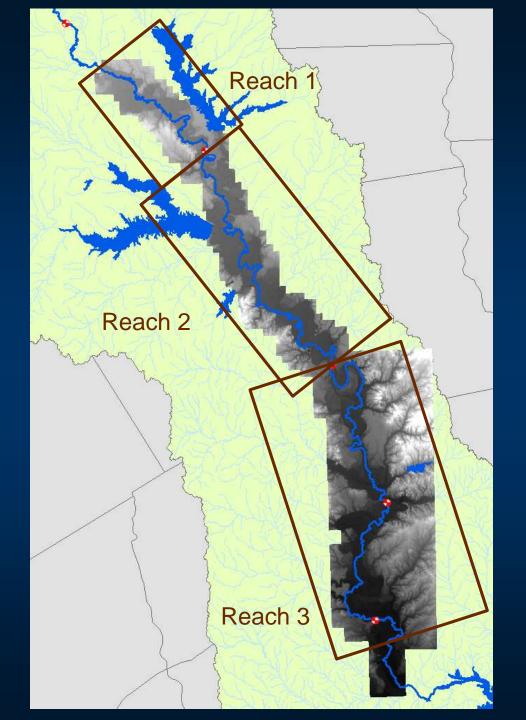
- Successful conservation of Alligator Gar depends on defining the specific characteristics of flood pulse events that lead to successful recruitment.
- The purpose of this study is to develop a river basin scale floodplain inundation model to predict Alligator Gar spawning habitat availability to assess the flood pulse characteristics that correlate with successful Alligator Gar recruitment.



Trinity River Basin

Inundation Model Development

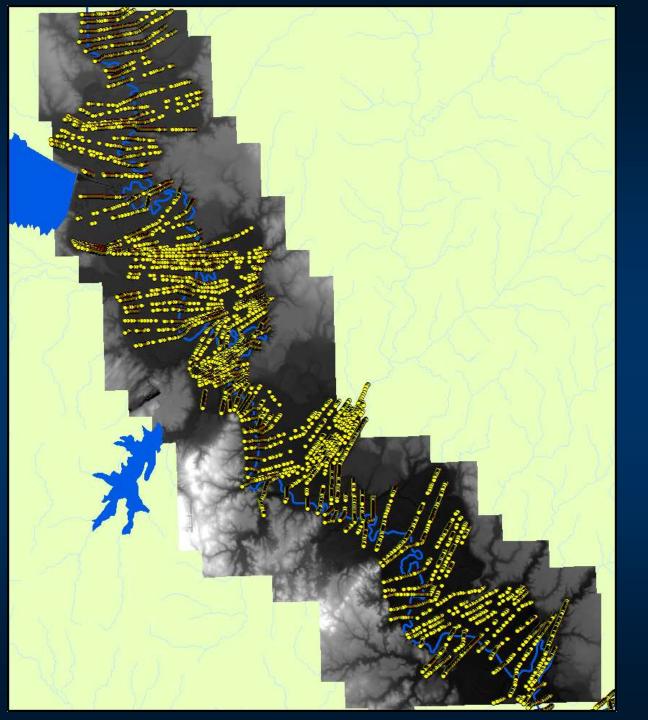
- 1 meter DEMs (~1500 tiles)
- HEC-RAS model from Tarrant Regional Water District
- ArcMap and Erdas Imagine



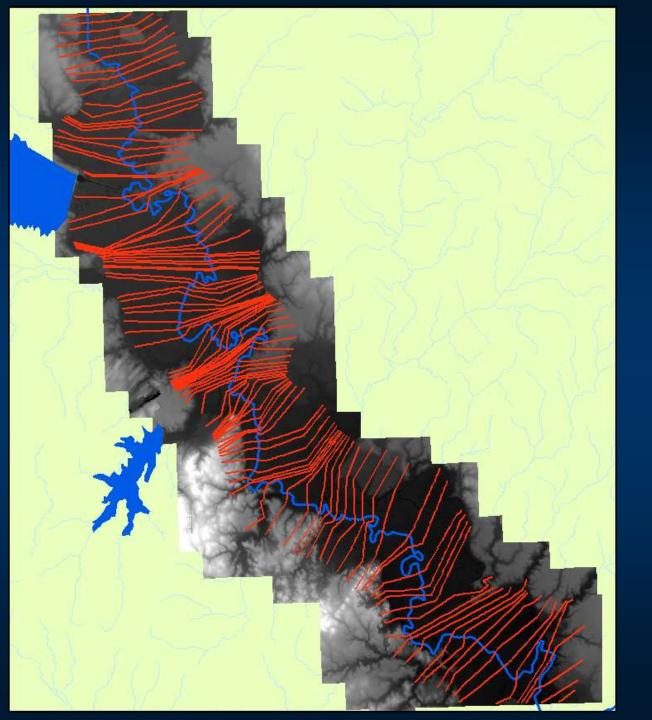
Merge DEMs into three reaches for data management purpose

Developing Modeled Water Surfaces:

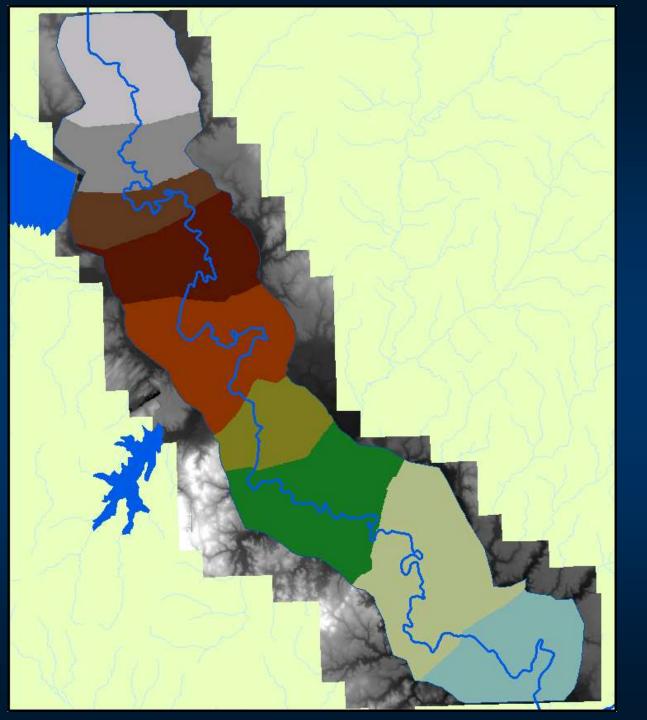
- Using HEC-RAS model developed by Tarrant Regional Water District, cross sections were produced for flows of: 5, 10, 15, 20, 25, 30, 35, 40, 50, 60, 70 and 80K cfs
- Water surface elevations generated for each discharge



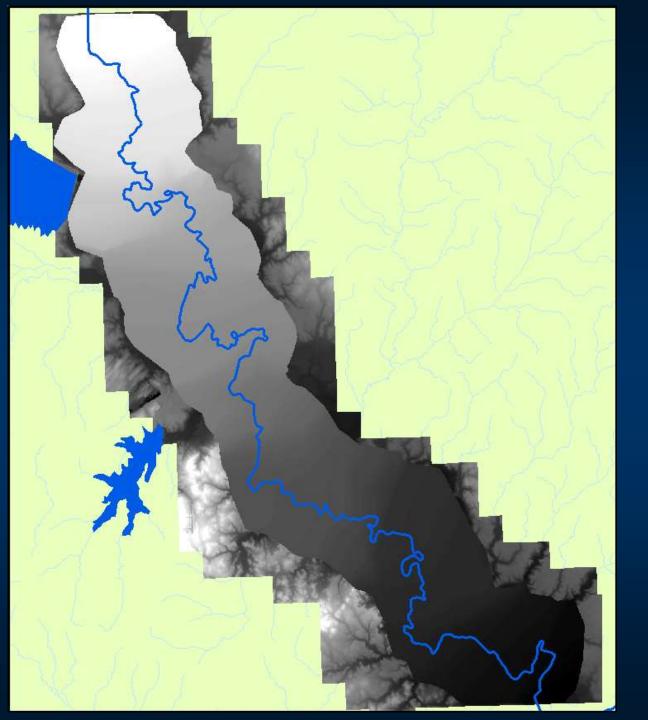
Cross section data imported from HEC RAS



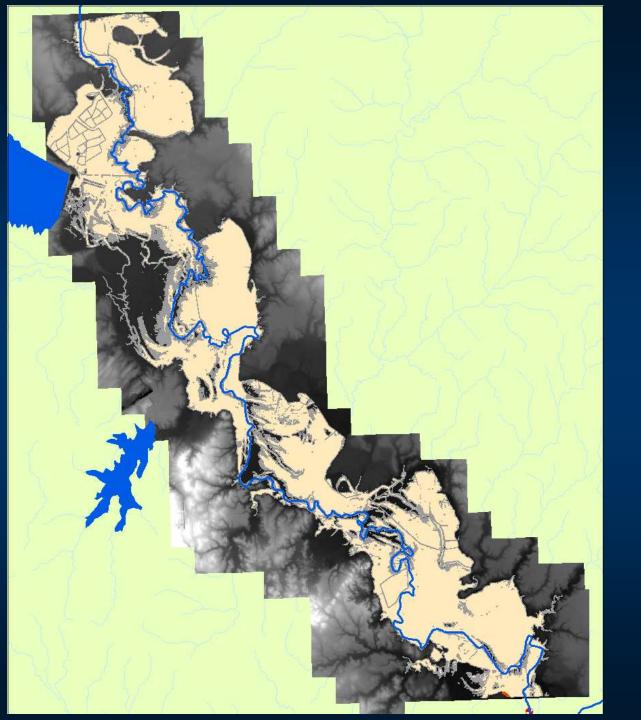
WSE from points spatially joined to transect lines for TIN development



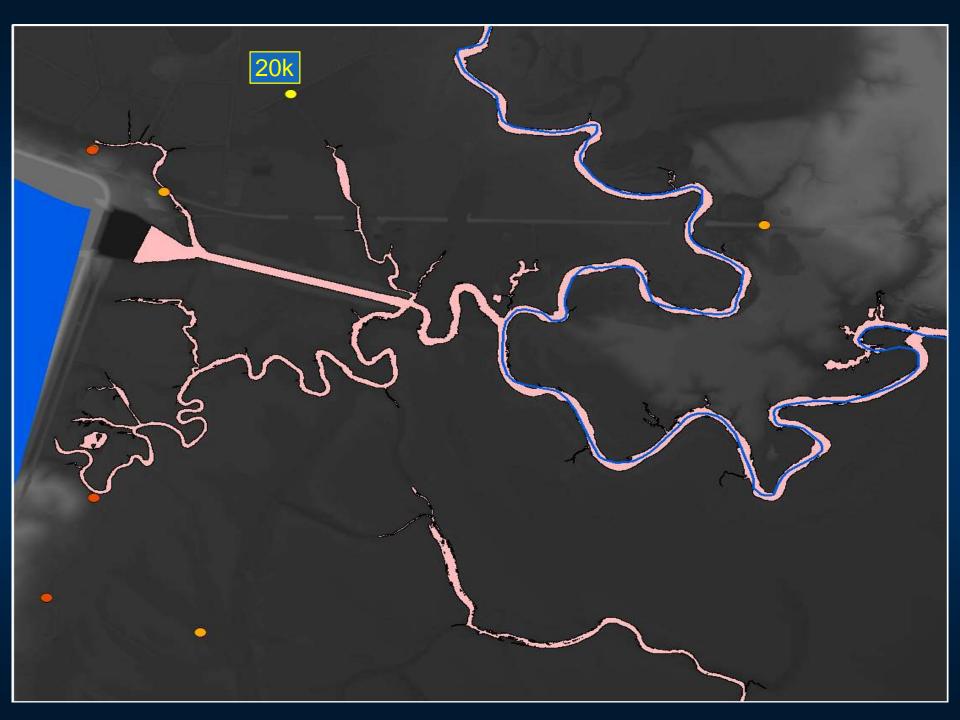
TIN generated to produce a smooth water surface and converted to raster



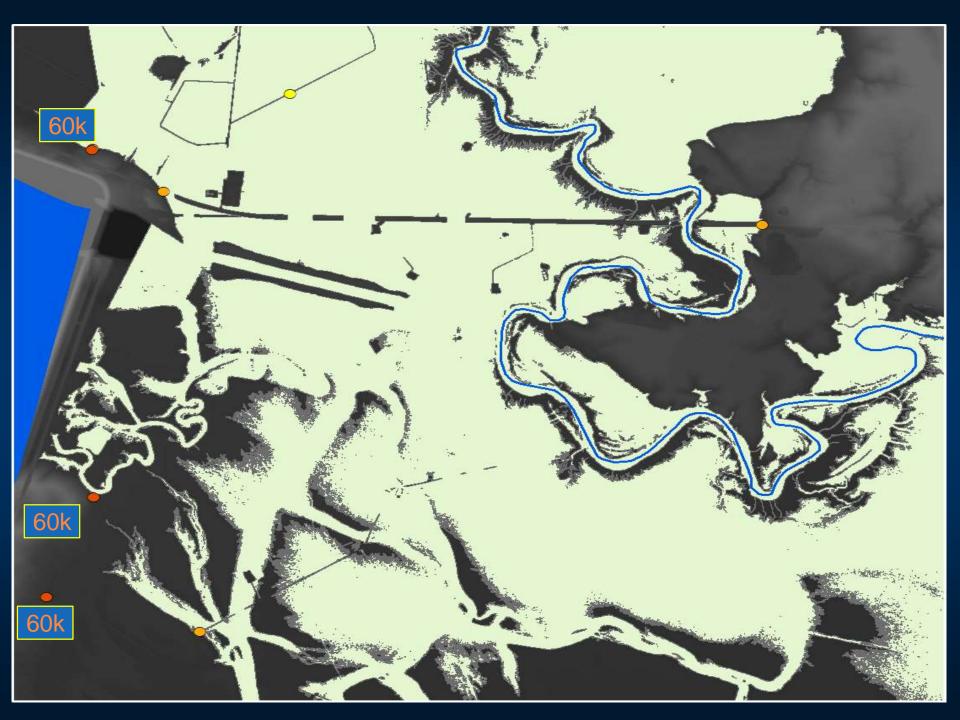
Water surface raster used to query DEMs and calculate inundation extent using ERDAS Imagine



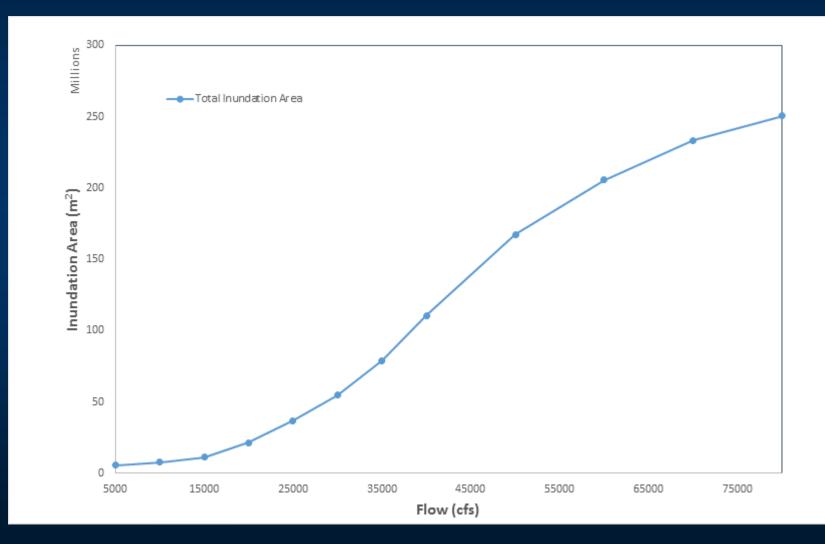
Total calculated inundated area.







Discharge Relationship to Area Inundated



Alligator Gar Spawning Habitat

- Previous work in LA¹ and ARK² show preferred habitat for spawning gar as open canopy vegetation types:
 - Examples:
 - Flooded herbaceous vegetation
 - Flooded shrubs
- Spawning typically occurs in shallow areas

¹Allen, Y.C., Kimmel, and G.C. Constant 2014. Using remote sensing to assess alligator gar spawning habitat suitability in the lower Mississippi River. U.S. Fish and Wildlife Service. Baton Rouge Fish and Wildlife Conservation Office report.

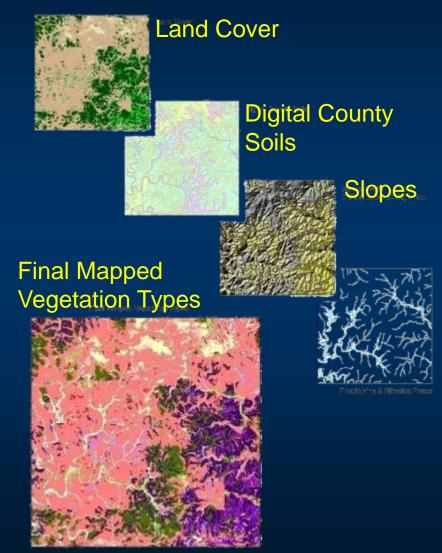
²Inebit, T.E. 2009. Aspects of the reproductive and juvenile ecology of Alligator Gar in the Fourche LaFave River, Arkansas. M.S. Thesis, University of Central Arkansas, Conway, Arkansas.

Inundated Habitat Data

- Determine types and areas of habitats being inundated
 - Vegetation classes
- Clip inundated vegetation classes to depths from 0.5-2m

Ecological Mapping Systems Data-Inputs

- Statewide **current** vegetation data
- Spatial Resolution
 - 10meters
- Thematic Resolution
 - 398 mapped habitat types
 - Abiotic variables (Enduring Features)
- Ground verified
 - >14,000 field data points
- Accuracy
 - 74% to 90%
- Anthropogenic effects
 - 19 Invasive types mapped

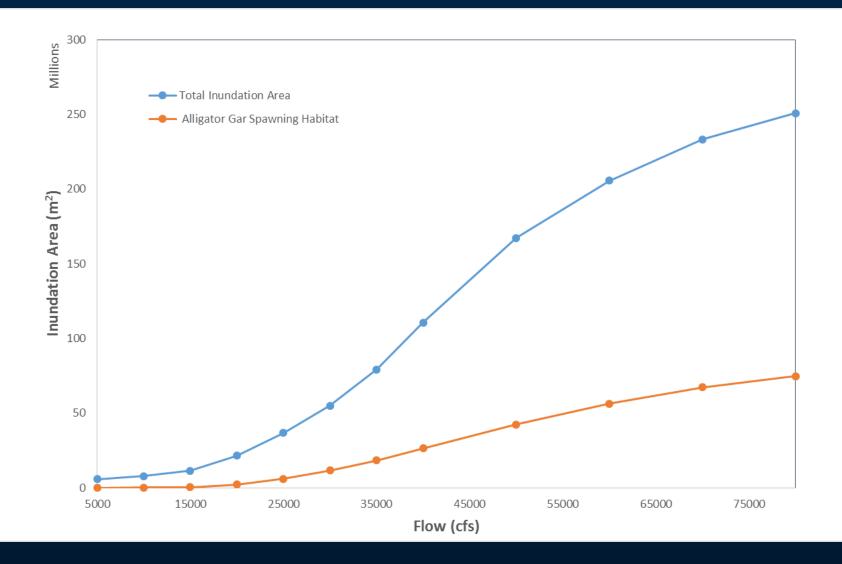


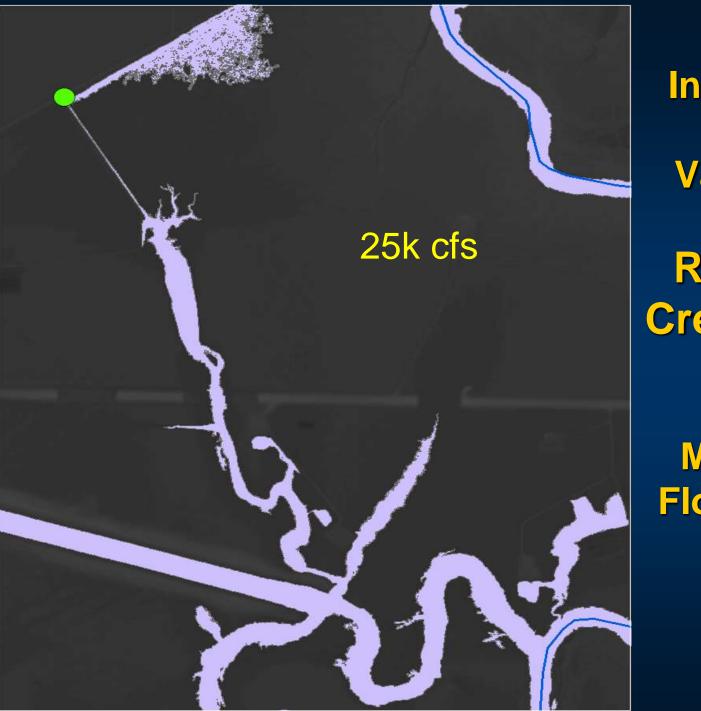


EMS clipped to spawning depth criteria

This was done for each of the target flows

Alligator Gar Spawning Habitat





Inundation Model Validation

Richland Creek WMA

May 2015 Flood Pulse



Selected Vegetation Classes

100.00

Floodplain Hardwood Forest Floodplain Deciduous Shrubland Floodplain Herbaceous Vegetation Floodplain Seasonally Flooded Hardwood Forest

 Correlate hydrology (e.g. spawning habitat availability) and environmental factors with known successful year classes to determine which factors are important for successful recruitment.

Year	Obs	Exp	Deviation	Obs/Exp	Year	Obs	Ехр	Deviation	Obs/Exp	
1986	2	1.09	0.91	1.84	2003	0	4.91	-4.91	0	
1987	2	1.19	0.81	1.69	2004	2	5.37	-3.37	0.37	
1988	0	1.30	-1.30	0	2005	0	5.87	-5.87	0	
1989	10	1.42	8.58	7.06	2006	5	6.42	-1.42	0.78	
1990	6	1.55	4.45	3.87	2007	43	7.01	35.99	6.13	
1991	6	1.69	4.31	3.54	2008	5	7.66	-2.66	0.65	
1992	1	1.85	-0.85	0.54	2009	5	8.37	-3.37	0.60	
1993	1	2.02	-1.02	0.49	2010	1	9.15	-8.15	0.11	
1994	3	2.21	0.79	1.36	•	120	Alliaa	tor Car		
1995	0	2.41	-2.41	0	 120 Alligator Gar 					
1996	1	2.64	-1.64	0.38	 Expected YCS was calculated from assumed 					
1997	0	2.88	-2.88	0						
1998	1	3.15	-2.15	0.32	constant annual					
1999	0	3.45	-3.45	0	recruitment and mortality					
2000	1	3.77	-2.77	0.27		rate	S			
2001	0	4.11	-4.11	0	Buckmei	er, D.L., <u>N.G</u>	. Smith, D <u>.J. [</u>	Daugherty, and D.L. B	ennett. <i>In Revie<u>w.</u></i>	
2002	1	4.50	-3.50	0.22	Reproductive ecology of Alligator Gar: identification of environmental driver for recruitment success.					

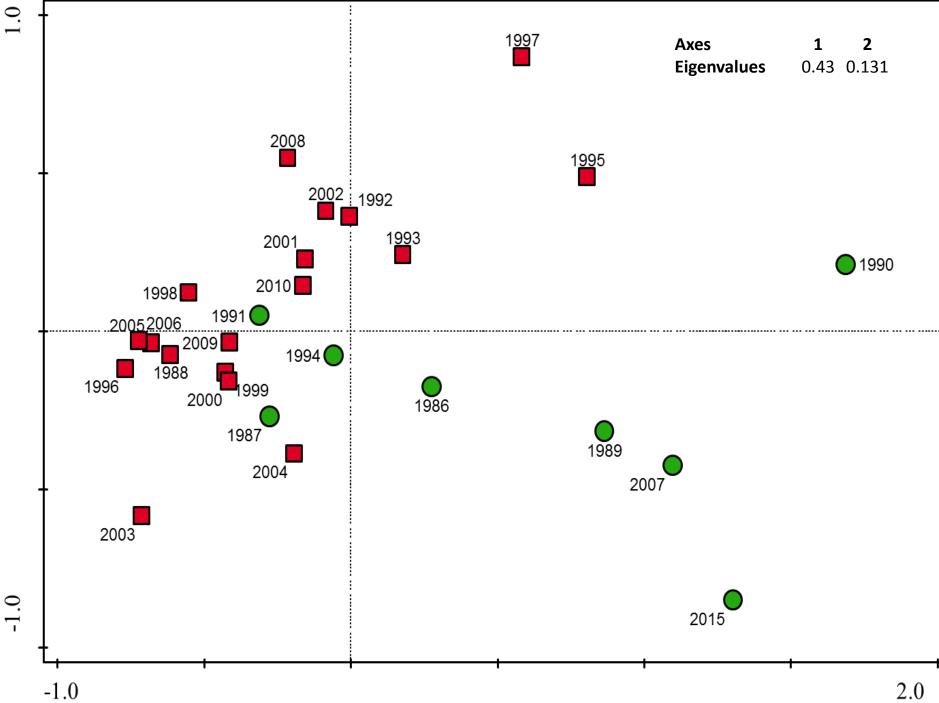
Hydrologic/Environmental Variables

- All variables assessed only during the spawning season (April-July)
- 42 total variables compiled from 1986-2010 (2015 also included)
- Temperature
 - Average Water Temp (monthly)
 - Cumulative Degree Days above 20°C (monthly)

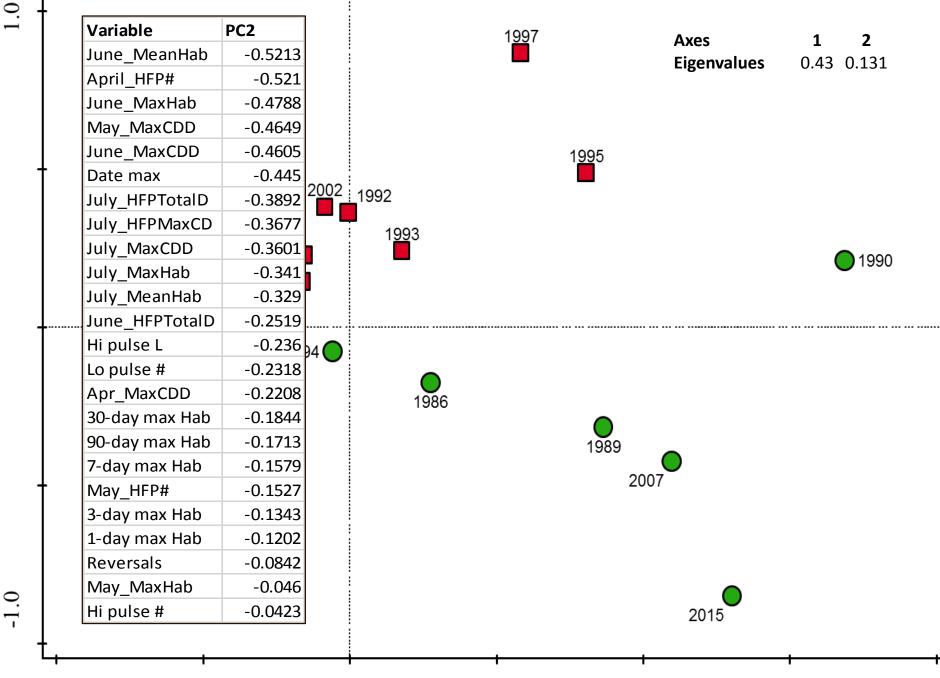
Hydrologic/Environmental Variables

- Utilized variables derived from Indicators of Hydrologic Alteration (IHA) analysis constrained to the spawning months
 - Mean/Max monthly flows (converted to available spawning habitat)
 - Monthly pulse count and pulse duration
 - Seasonal pulse count and pulse duration

AvgT_Apr	1-day max	April_HFP#
AvgT_May	3-day max	April_HFPMaxCD
AvgT_June	7-day max	April_HFPTotalD
Avgt_July	30-day max	May_HFP#
Apr_MaxCDD	90-day max	May_HFPMaxCD
May_MaxCDD	Date max	May_HFPTotalD
June_MaxCDD	Lo pulse #	June_HFP#
July_MaxCDD	Lo pulse L	June_HFPMaxCD
April_MeanQ	Hi pulse #	June_HFPTotalD
May_MeanQ		July_HFP#
June_MeanQ	Fall rate	July_HFPMaxCD
July_MeanQ	Reversals	July_HFPTotalD
April_MeanHab	April_MaxQ	Spawn Season HFP_totalD
May_MeanHab	May_MaxQ	Spawn Season HFP MaxCD
June_MeanHab	June_MaxQ	
July_MeanHab	July_MaxQ	
1-day max Hab		
3-day max Hab		
7-day max Hab		
30-day max Hab		
90-day max Hab		
April_MaxHab		
May_MaxHab		
June_MaxHab		
July_MaxHab		



3- 7-	-day max Hab B-day max Hab	0.9378							
7-	B-day max Hab			1997 1992 1993	1997		Axes	1	2
		0.9357				1 <u>99</u> 5	Eigenvalues	0.43 0.131	—
S	/-day max Hab	0.9335							
	pawn Season HFP_totalD	0.9211							
90	0-day max Hab	0.8975							
н	li pulse #	0.8916							
- H	li pulse L	0.8879							
30	80-day max Hab	0.8586	02		 1989 20				
\mathbb{N}	∕lay_MaxHab	0.8339							
Ju	uly_HFPMaxCD	0.8102							
Ju	uly_HFPTotalD	0.8074						0 1990	
Ju	une_HFPMaxCD	0.7932							
S	pawn Season HFP MaxCD	0.7708	ullet						
- №	May_HFPTotalD	0.7703							
N	∕lay_MeanHab	0.679							
Ju	une_HFPTotalD	0.638							
\mathbb{N}	May_HFPMaxCD	0.6041		1986					
Ju	une_MaxHab	0.594							
Ju	une_MeanHab	0.584				1989			
А	April_HFPTotalD	0.5579							
- A	April_MeanHab	0.5568				200	07		
А	April_MaxHab	0.5448							
Ju	uly_MaxHab	0.4443							
Ju	uly_MeanHab	0.4357							
Ju	uly_HFP#	0.43							
О. D	Date max	0.3135							
ΓA	April_HFPMaxCD	0.2298					2015		
- A	April_HFP#	0.0109							



-1.0

2.0

- Utilizing the results of the PCA analysis, 42 variables were reduced to 21 for correlation analysis.
- Spearman rank correlation analysis ran on the 21 variables and the Obs/Exp YCS values to determine the important variables that reflect strong YCS.

	Spearman's	
	Rank Corr.	
30-day max Hab	0.447	0.0252
90-day max Hab	0.414	0.0396
July_HFPMaxCD	0.407	0.0433
June_MeanHab	0.397	0.0487
May_MaxHab	0.394	0.0506
Hi pulse L	0.386	0.0561
3-day max Hab	0.386	0.0561
July_HFPTotalD	0.378	0.0621
7-day max Hab	0.377	0.0627
1-day max Hab	0.357	0.0783
Spawn Season HFP_totalD	0.345	0.0905
May_HFPTotalD	0.341	0.0943
May_HFPMaxCD	0.326	0.11
Hi pulse #	0.318	0.12
Spawn Season HFP MaxCD	0.256	0.215
June_HFPMaxCD	0.212	0.305
April_MeanHab	0.111	0.592
April_HFPTotalD	0.0808	0.698
Reversals	-0.215	0.297
Fall rate	-0.34	0.0951
Lo pulse #	-0.459	0.0211

Year	July_HFPMaxCD	June_MeanHab	30-day max Hab	90-day max Hab	May_MaxHab
1986	62	5,577,364	5,810,430	2,902,645	5,193,686
1989	90	10,970,836	12,135,607	6,570,011	47,887,662
1990	130	5,669,539	63,658,867	18,574,340	75,722,432
1994	10	1,528,282	5,466,580	2,120,645	8,577,706
2007	67	5,217,529	26,235,891	7,815,945	8,067,695
2015	83	36,908,278	49,939,644	14,704,057	70,517,223
Avg	74	10,978,638	27,207,836	8,781,274	35,994,401
FLOW (cfs)		24,983	39,107	24,182	51,617

Next Steps

- Assess new HEC-RAS model to develop inundated habitat criteria for all three reaches
- Correlate hydrologic indices from all three reaches to adult year class data to develop high flow pulse recommendations for recruitment
- Flow recommendations will be included in Texas Instream Flow Program instream flow study for the middle Trinity River





Questions?

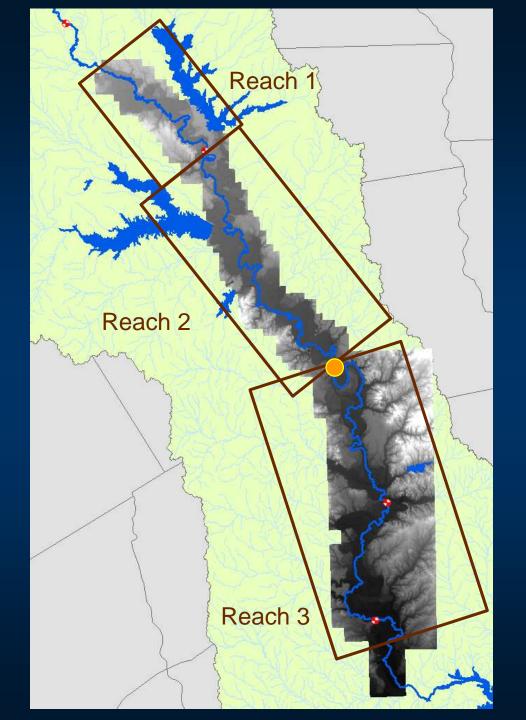
Clint Robertson – clint.robertson@tpwd.texas.gov **Acknowledgements** Alice Godbey – TRWD Dr. Thom Hardy – TSU **Duane German– TPWD** Dan Bennett – TPWD TEXAS Dan Daugherty – TPWD PARKS &



WILDLIFE

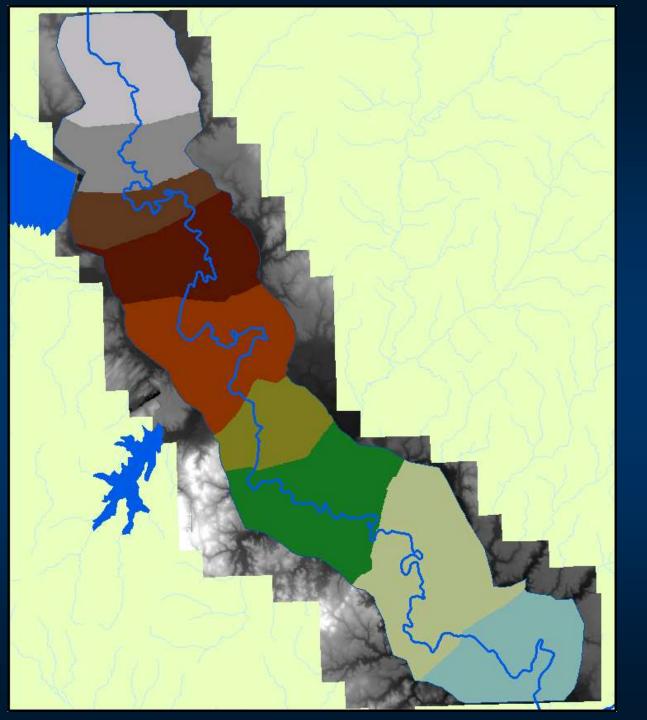
Developing Modeled Water Surfaces: Using USGS gage data

- Utilized gage at downstream extent of Reach 2
- Used USGS published rating curve
- Kept same flood flows of: 5, 10, 15, 20, 25, 30, 35, 40, 50, 60, 70 and 80k cfs
- Water surface elevations generated for each discharge

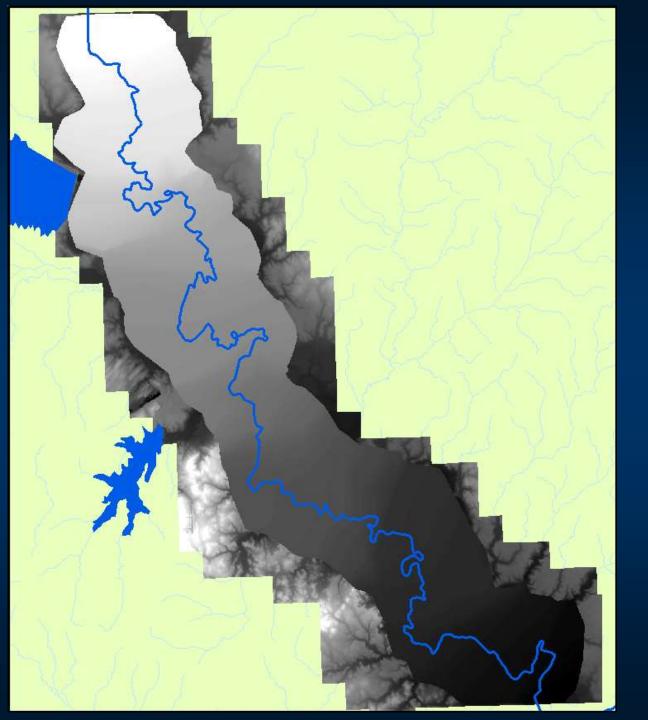


Gage at the downstream extent of Reach 2

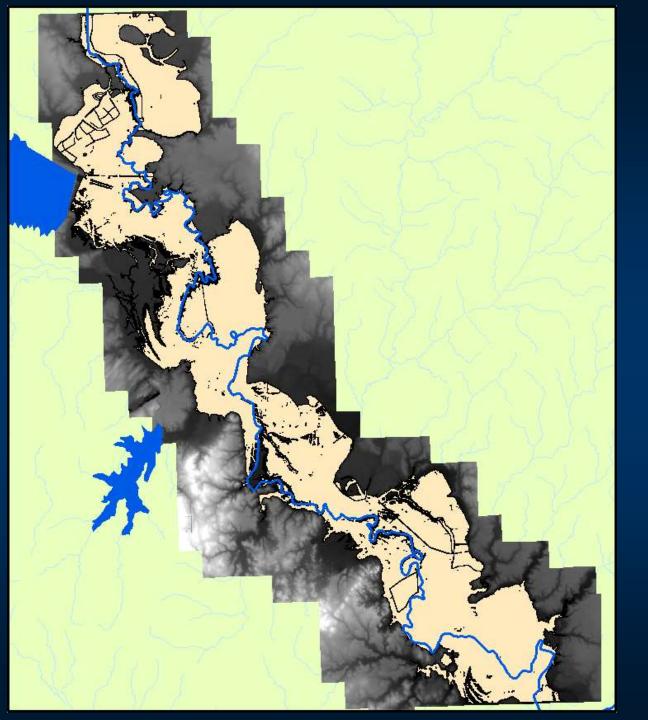
M . 10		K A L L L L			
	14-Feb	5.92	1200	LIDAR Flight Dates	LIDAR flight
	15-Feb	5.83	1170		
					to establish
	CFS	gage ht ft	gage ht M	hght add to WSE points M	lationship
	5000	11.08	3.377184	1.59	een LIDAR
	10000	18.72	5.705856		
	15000	25.895	7.892796	6.10	
	20000	31.505	9.602724	7.81	ge heights
b	25000	35.46	10.808208	9.02	
	30000	38.56	11.753088	9.96	
	35000	41.055	12.513564	10.72	
	40000	42.71	13.018008	11.23	
	50000	44.75	13.6398	11.85	
A	60000	46.09	14.048232	12.26	
X	70000	47.08	14.349984	12.56	
	80000	47.86	14.587728	12.80	
11	A VALANA HI				



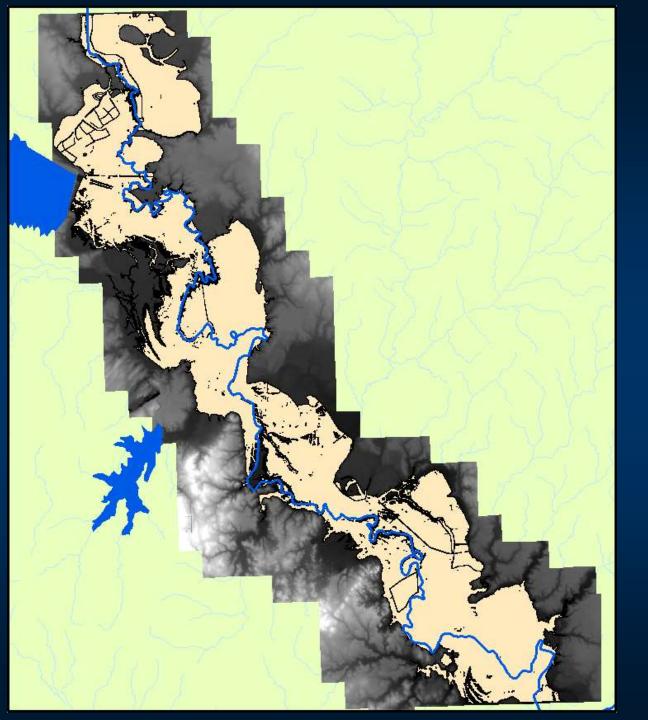
TIN generated to produce a smooth water surface and converted to raster



Water surface raster used to query DEMs and calculate inundation extent using ERDAS Imagine



Total calculated inundated area.



Reach 2 80K example

Total calculated inundated area.

Landsat Imagery

- Using USGS gage data, searched for dates of targeted flood events
- Ran unsupervised classification on 15 classes, recoded into three classes (water, no water, and mixed)

