


A Satellite-Based Approach to Detecting Small Livestock and Agricultural Ponds to Monitor Change Over Time

Jennifer Jensen, Professor
Garrett Pugh, MS Student
Grayson Wylie, PhD Student

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GEOGRAPHY AND
ENVIRONMENTAL STUDIES

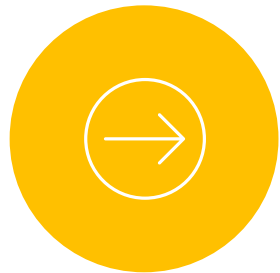
Learning Objectives



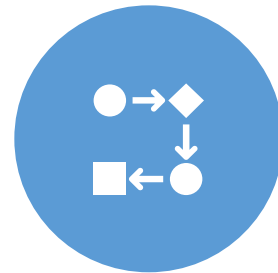
Identify needs and challenges to characterizing and monitoring surface water.



Understand how to use a satellite-based approach to inventory surface water bodies.



Learn how our results compare with the existing National Hydrography Dataset.



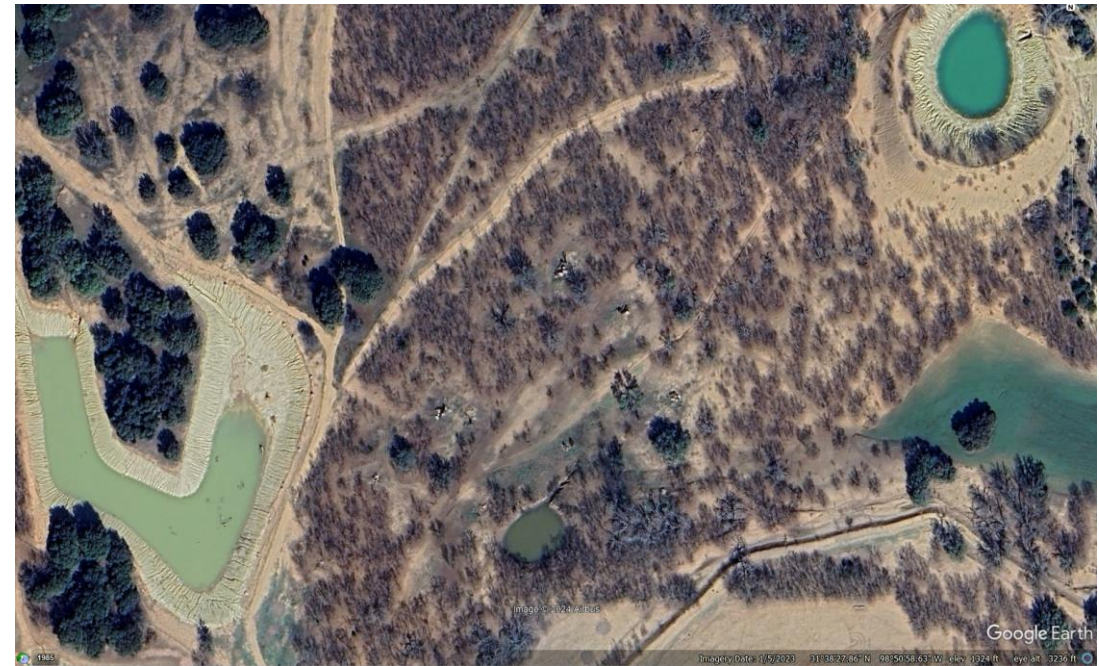
Determine small pond distributions and changes over time.

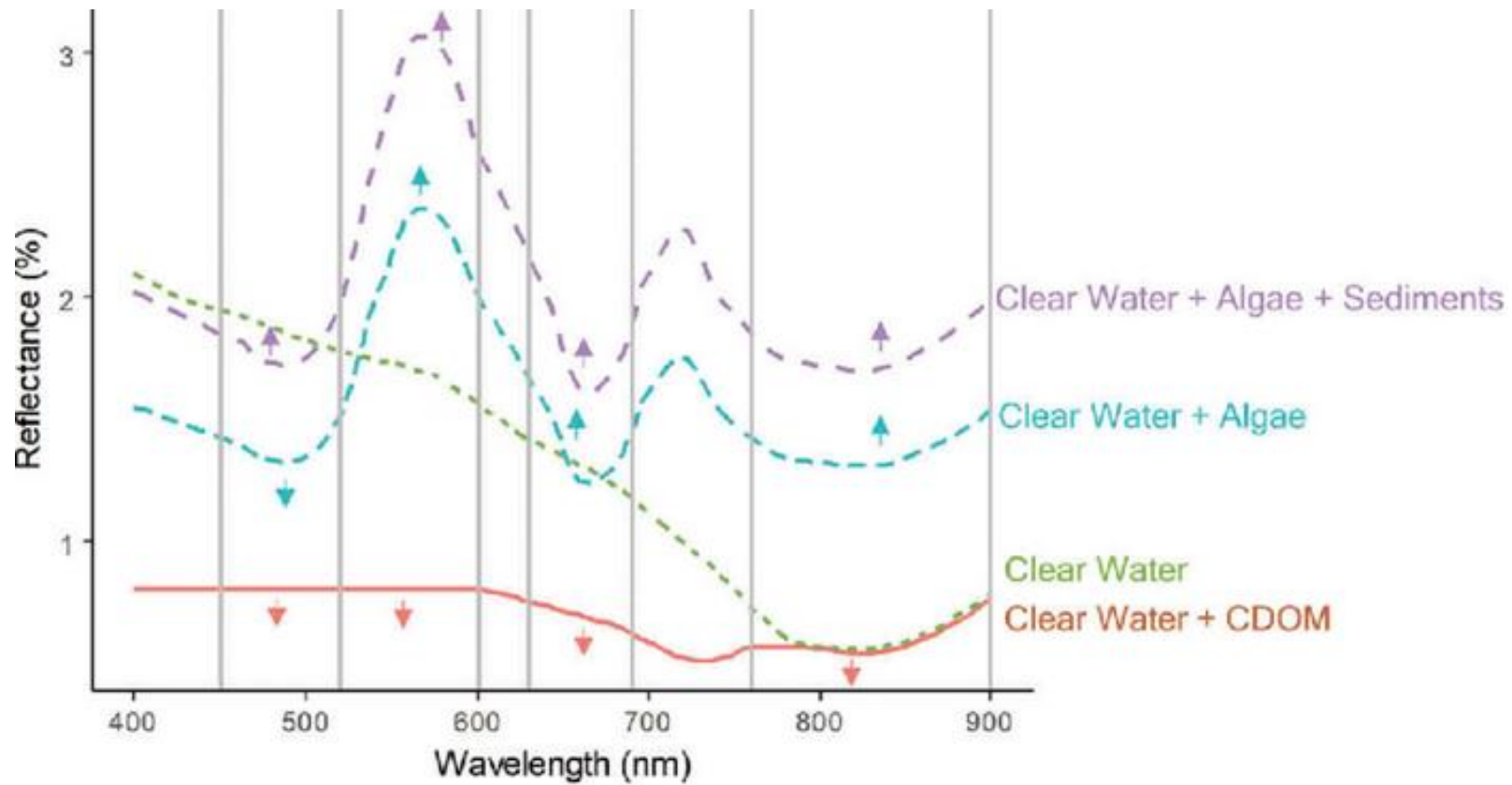
Need for Surface Water Detection and Monitoring

- Critical for studying hydrological and ecological processes and monitoring trajectories of change.
- Over the past 40 years, remote sensing technologies have enabled surface water mapping to be much more efficient and comprehensive compared to traditional manual methods (Bijeesh and Narasimhamurthy, 2020).
- A review of the Web of Knowledge database conducted by Huang et al. (2018) noted a steady increase in the number of studies that used remote sensing to research surface water or flood inundation
 - A three-fold to seven-fold increase in the number of studies since 2008.

Challenges with Surface Water Detection and Monitoring

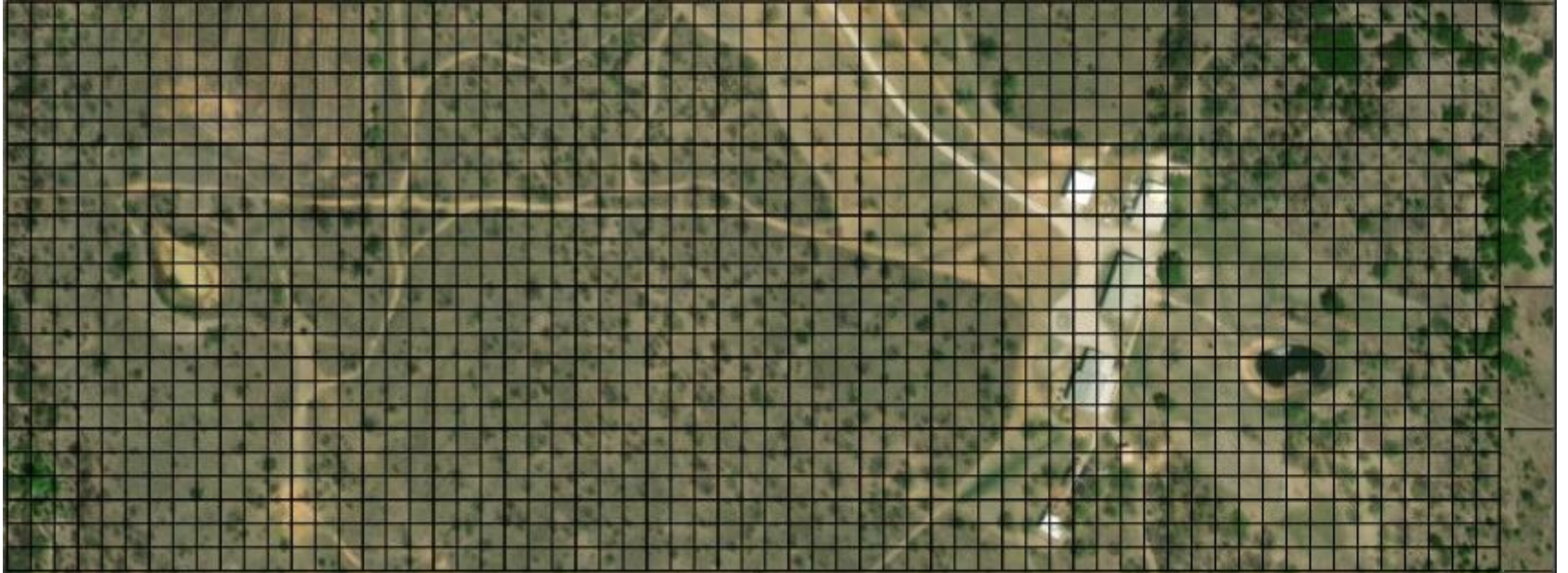
- Pure water reflects shorter wavelengths of energy (i.e., ~400 – 550 nm) and absorbs longer visible, near-infrared (NIR), and shortwave infrared (SWIR) wavelengths (i.e., ~550 nm – 2,900 nm).
- Unfortunately, pure water is the exception to the rule and many surface water bodies water columns include sediments and algae that increase turbidity and alter the basic spectral response.





Challenges
with Surface
Water
Detection and
Monitoring

Challenges with Surface Water Detection and Monitoring

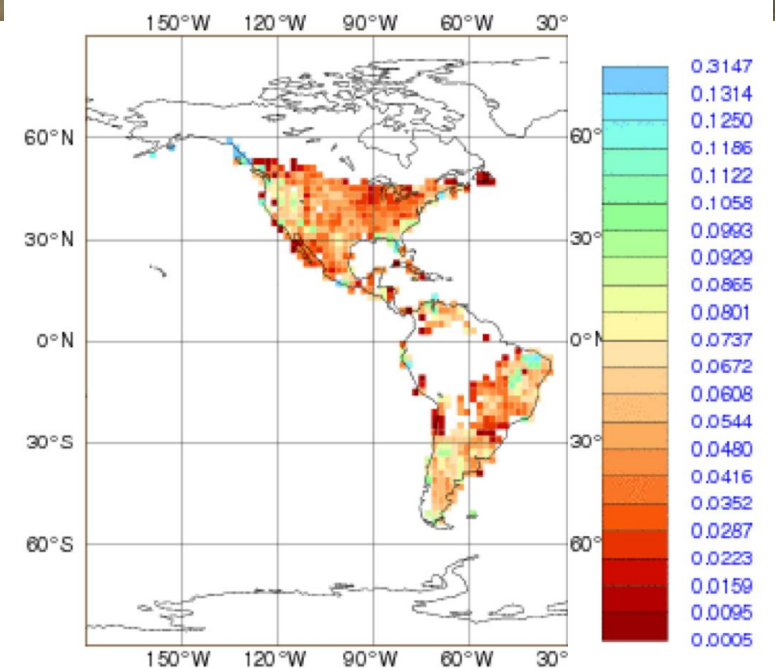
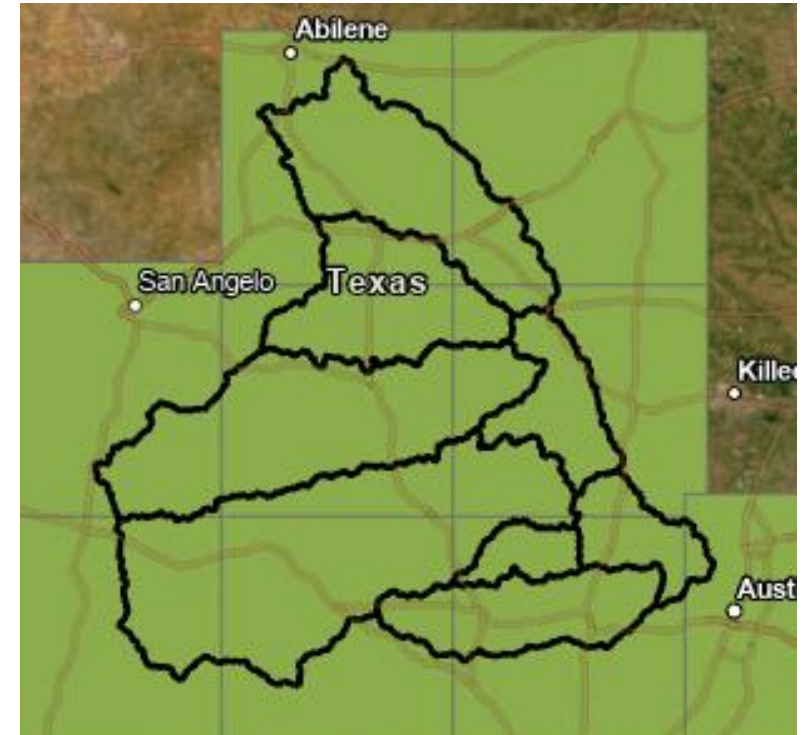


Challenges with Surface Water Detection and Monitoring



Our Study Area, Data, and Time Periods of Analysis

- Highland Lakes watershed/subwatersheds
- Sentinel-2 data (10 m RGB-NIR; 20 m SWIR)
- 10 m Digital Elevation Model to identify sinks
- Soil Moisture Active Passive (SMAP) to determine wettest periods of:
 - 2023 (March)
 - 2019 (February)
 - 2016 (December/January)



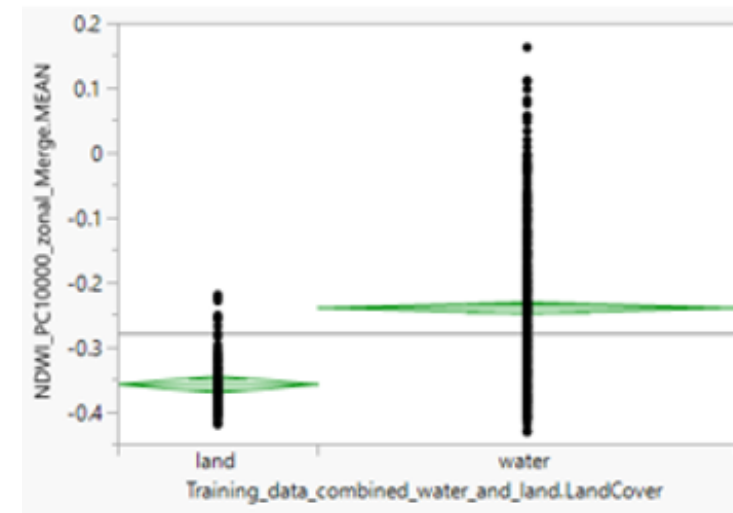
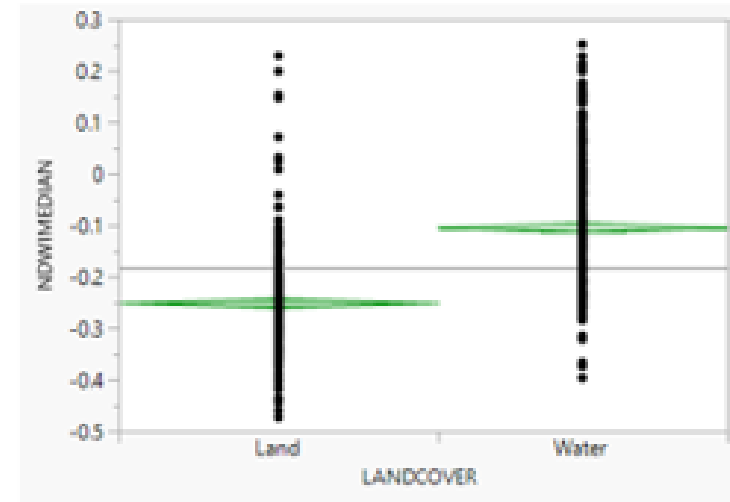
A Spectral Index-Based Approach to Inventory Surface Water Bodies

- A spectral index is a simple equation that uses the reflectance characteristics of a surface to create a new, more easily interpretable output.

Index	Equation	Reference
NDVI	$NDVI = \frac{NIR - Red}{NIR + Red}$	Rouse et al., 1974
NDWI	$NDWI = \frac{Green - NIR}{Green + NIR}$	McFeeters, 1996
MNDWI	$MNDWI = \frac{Green - SWIR}{Green + SWIR}$	Xu, 2006
WNDWI	$WNDWI = \frac{Green - \alpha \times NIR - (1 - \alpha) \times SWIR}{Green + \alpha \times NIR + (1 - \alpha) \times SWIR}; \alpha \in [0,1]$	Guo et al., 2017
AWEI (Non-Shadow)	$AWEI_{nsh} = 4 \times (Green - SWIR) - (0.25 \times NIR + 2.75 \times SWIR)$	Fisher et al., 2016
AWEI (Shadow)	$AWEI_{sh} = Blue + 2.5 \times Green - 1.5 \times (NIR + SWIR) - 0.25 \times SWIR$	Fisher et al., 2016
WRI	$WRI = \frac{Green + Red}{NIR + SWIR}$	Shen and Li, 2010
NDMI	$NDMI = \frac{NIR - SWIR}{NIR + SWIR}$	Gao, 1996
EWI	$EWI_{pc} = \frac{(MNDWI - (PC1 + PC2))}{(MNDWI + (PC1 + PC2))}$	Yang and Du, 2017

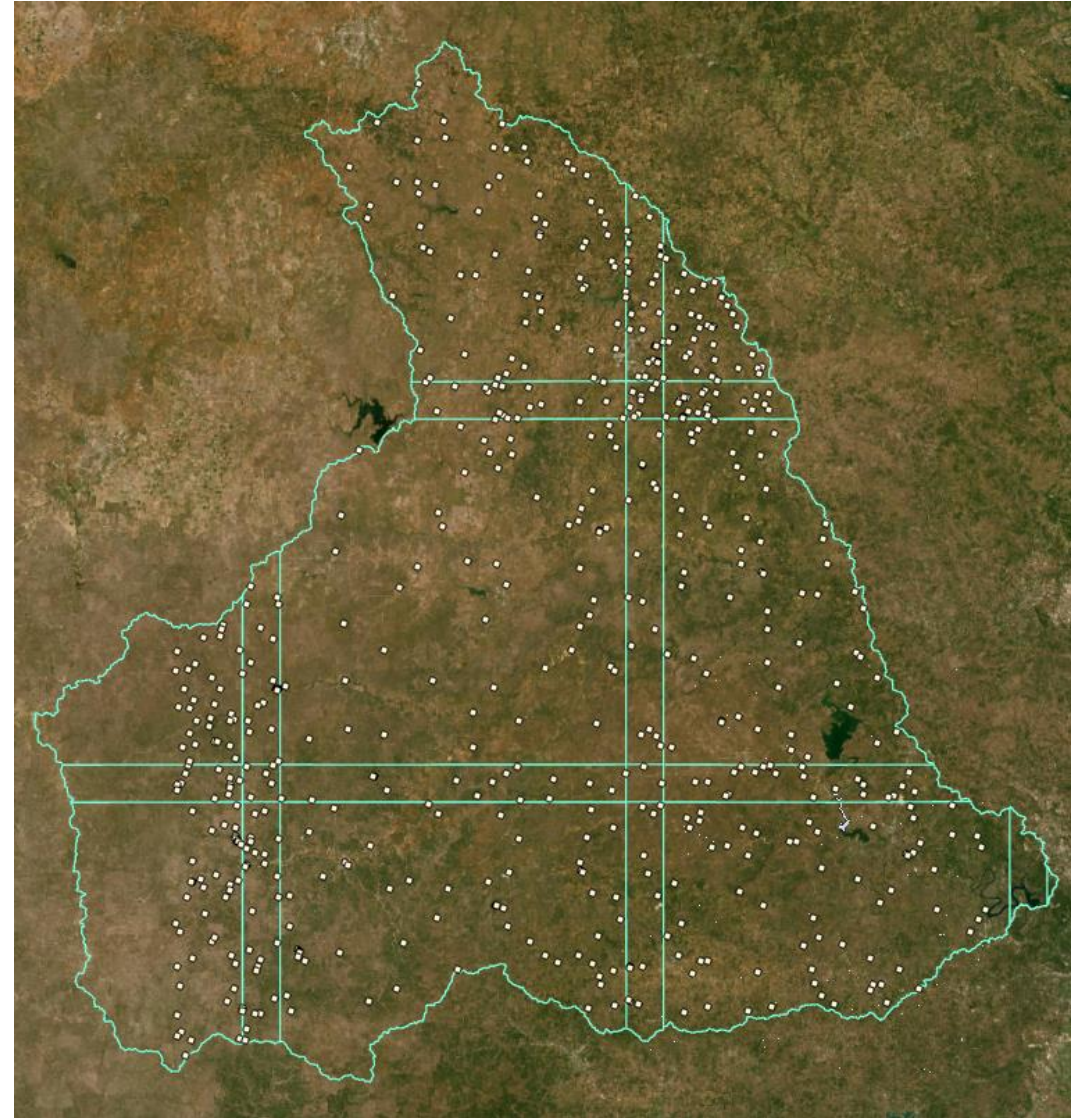
Process Sentinel-2 images to classify water bodies

- Calculated 10 spectral indices and assessed for spectral separability using 399 verified water bodies and 442 verified land features.
- Determined the Normalized Difference Water Index (NDWI) provided best initial results using a -0.1 threshold.
- NDWI misclassified some land pixels as water
 - Removed 'bright' pixels identified from a PCA analysis
- NDWI + PCA misclassified some water pixels as land
 - Incorporated a 10 m DEM to identify sinks
- Removed roads, lower order streams, built-up/developed areas, very large water bodies, etc.



Assess classification accuracy and apply model throughout watershed

- Calculated sample size
 - Expected 95% accuracy and 5% allowable error
 - Needed minimum of 38 AA points; used 80 in a stratified random sample
- Used ArcGIS Pro base maps and Google Earth with historic imagery to conduct accuracy assessment



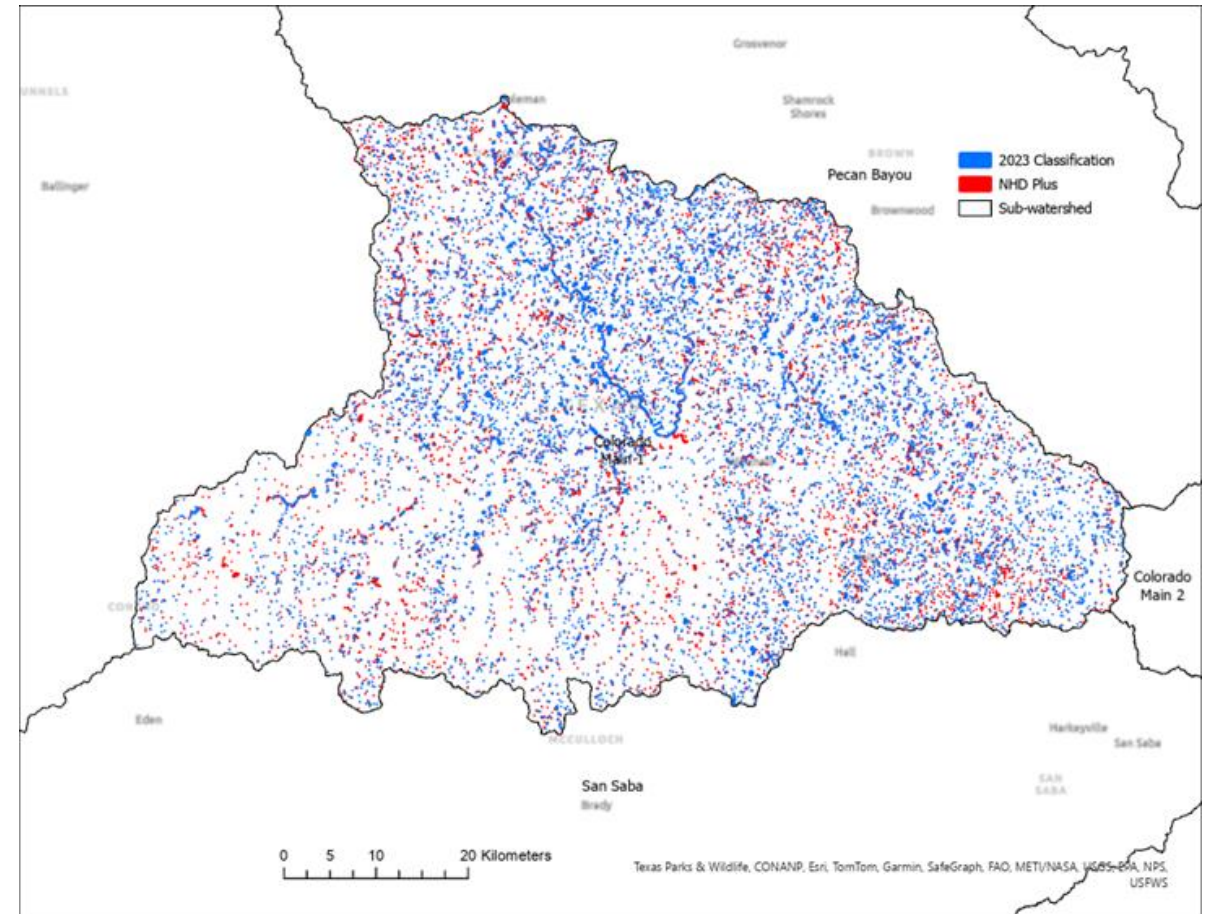
Assess classification accuracy and apply model throughout watershed

- Overall accuracies
 2023: 94.8% (Kappa = 0.896)
 2019: 84.5% (Kappa = 0.891)
 2015: 96.7% (Kappa = 0.934)

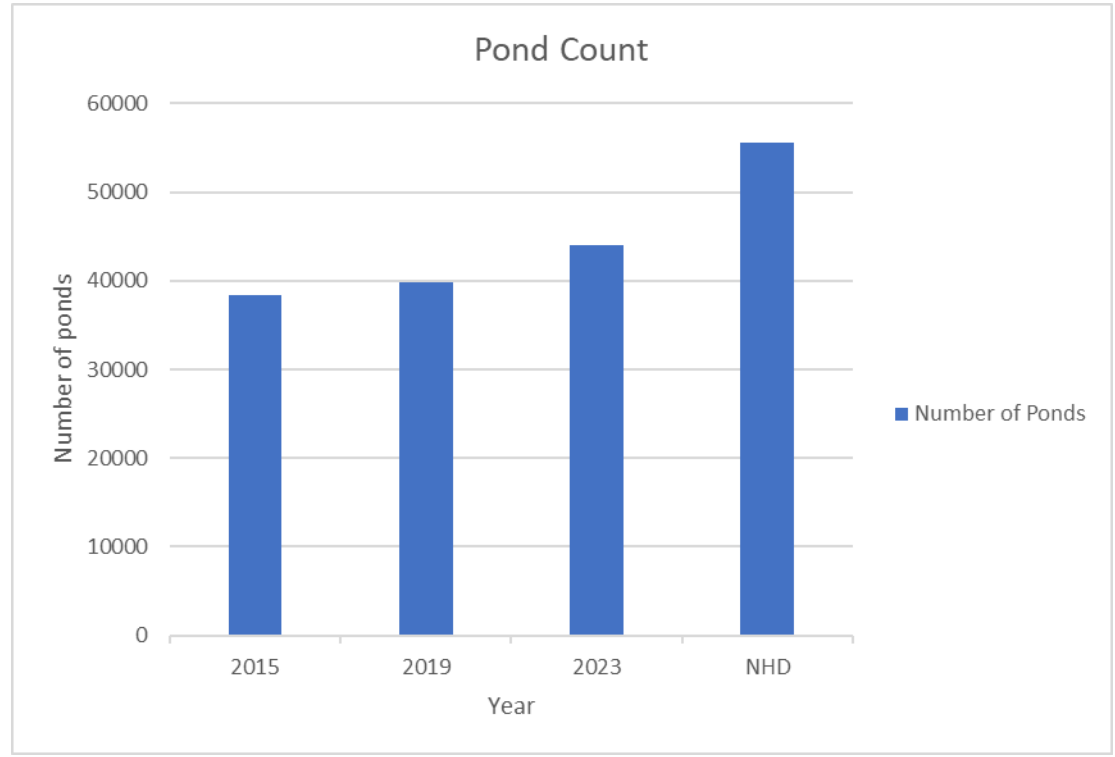
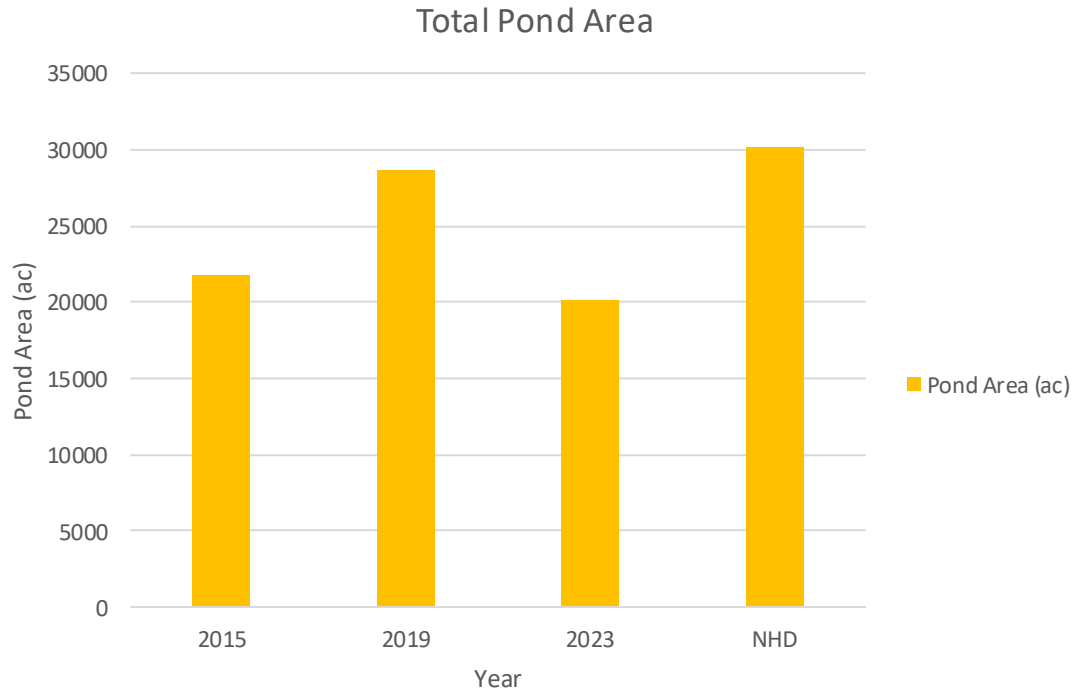
Sentinel-2 Tile	2023			2019			2015		
	Overall Acc (%)	Kappa	User Acc	Overall Acc (%)	Kappa	User Acc	Overall Acc (%)	Kappa	User Acc
T14SNA	97.5	0.95	0.95	97.5	0.95	0.98	98.8	0.98	1
T14SMA	97.5	0.95	0.98	96.3	0.93	1	97.5	0.95	0.98
T14RNV	95.0	0.90	0.93	95.0	0.90	0.98	98.8	0.98	0.98
T14RNU	95.0	0.90	0.90	93.8	0.88	0.90	96.3	0.93	0.95
T14RMV	96.3	0.93	0.93	97.5	0.95	1	96.3	0.93	0.95
T14RMU	100.0	1	1	92.5	0.85	0.90	98.8	0.98	0.98
T14RLV	96.3	0.93	0.93	87.5	0.75	0.78	93.8	0.88	0.93
T14RLU	93.8	0.88	0.88	96.3	0.93	0.98	93.8	0.88	0.88

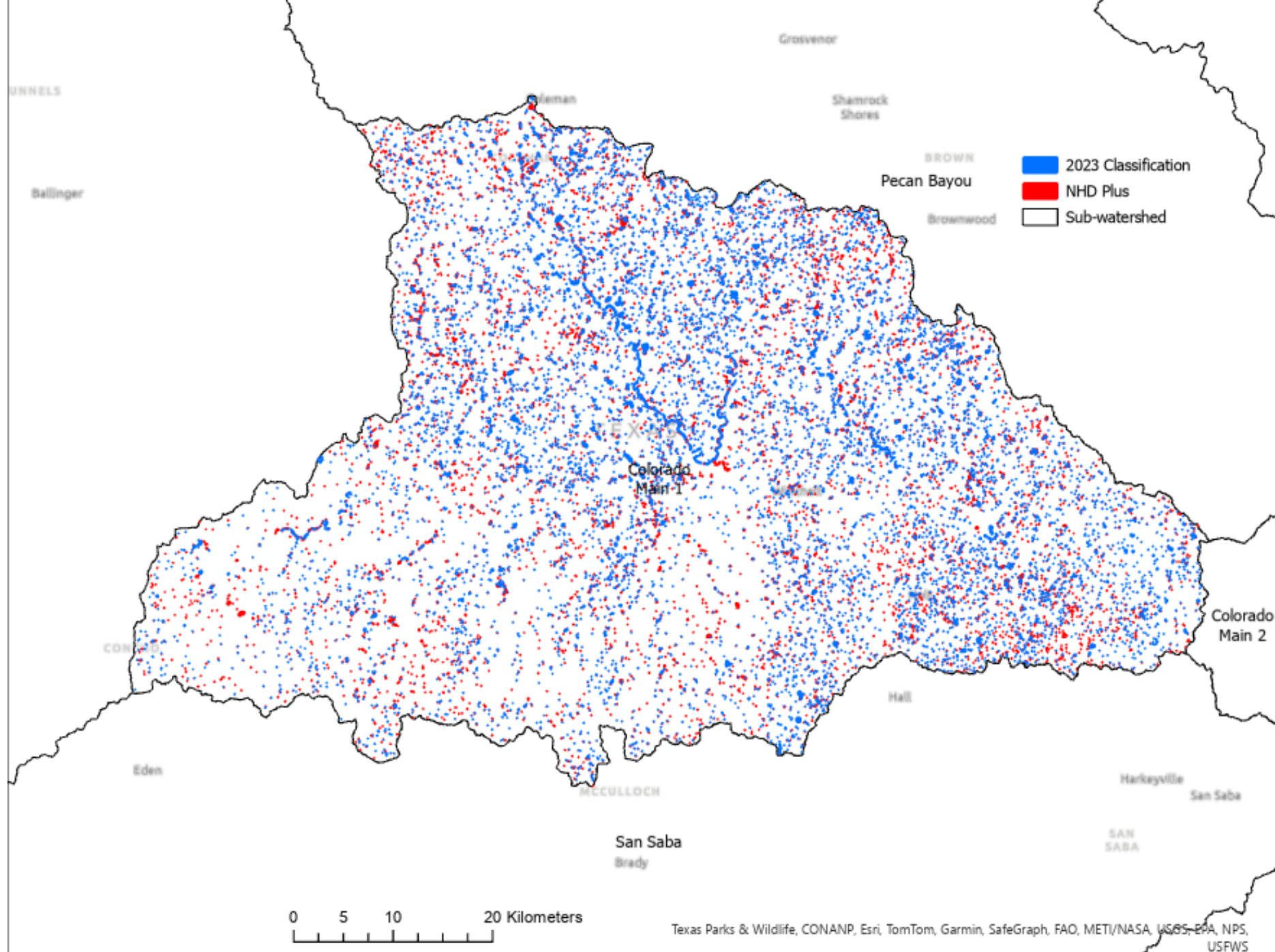
Inventory water bodies within subwatersheds, determine size distributions, and compare to NHD dataset

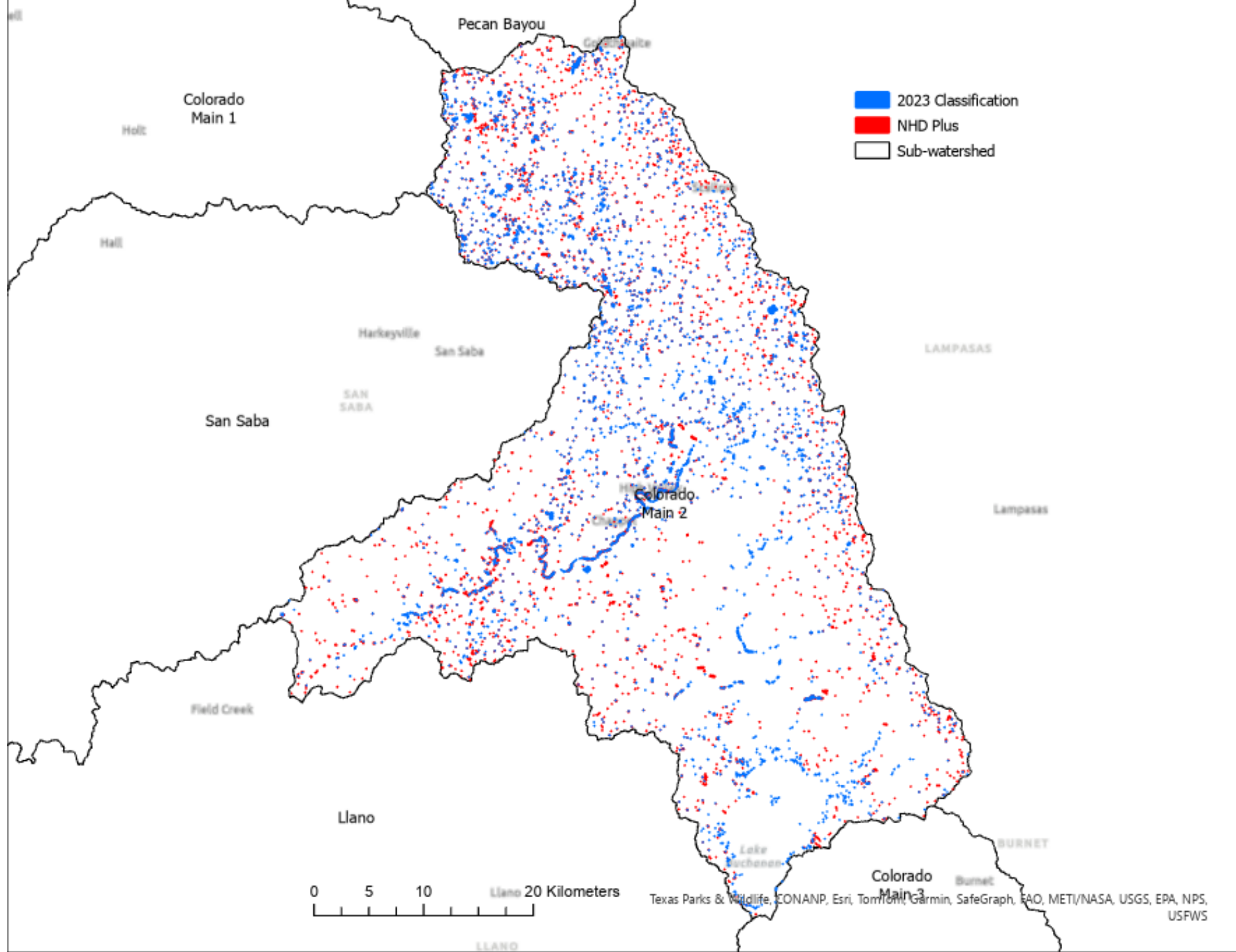
- Merged all modeled tiles and clipped by subwatershed
- Inventoried ponds
- Calculate distributions by size class in 10 bins
- Three classified years and NHD dataset
- Counts and distributions by sub-watershed

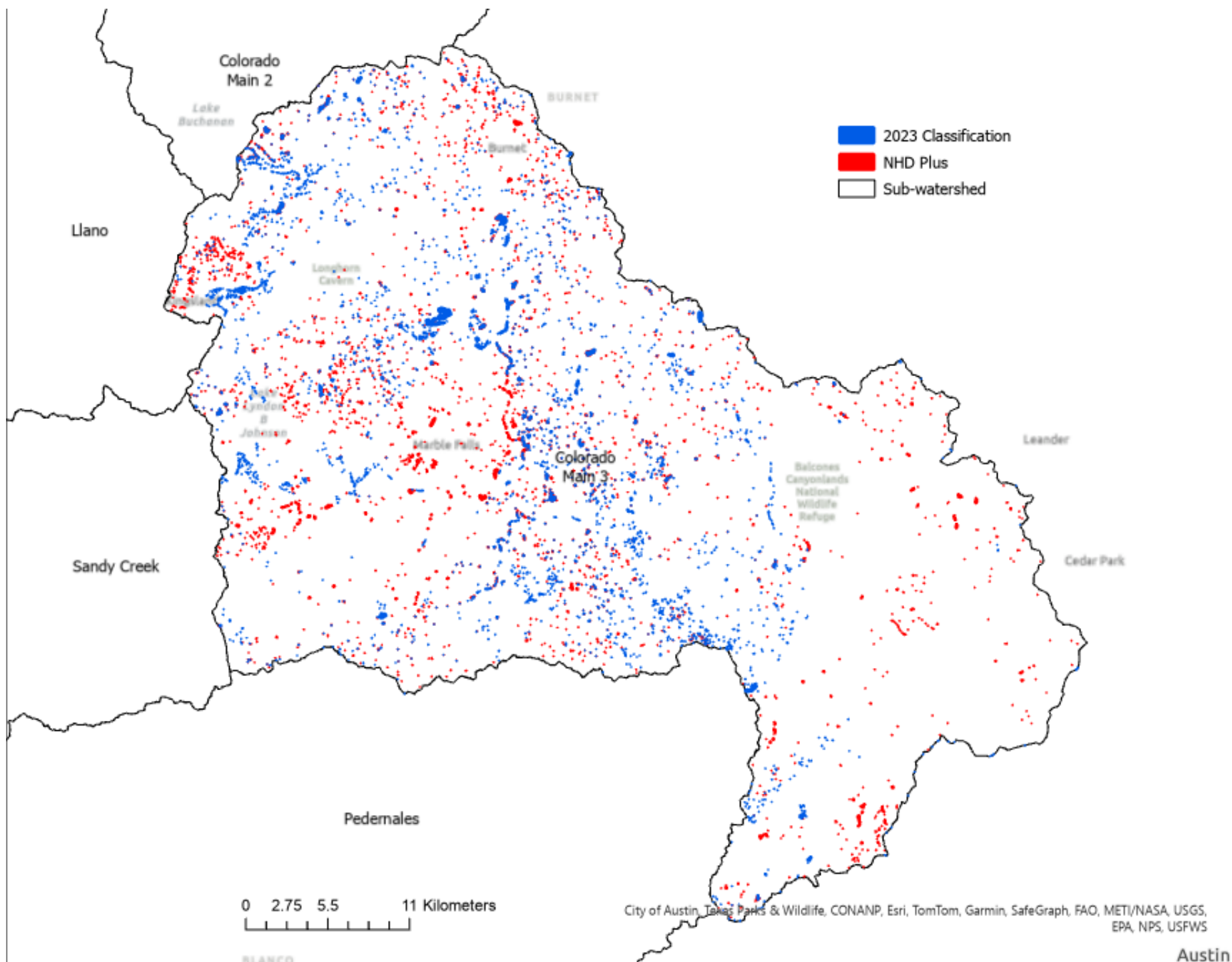


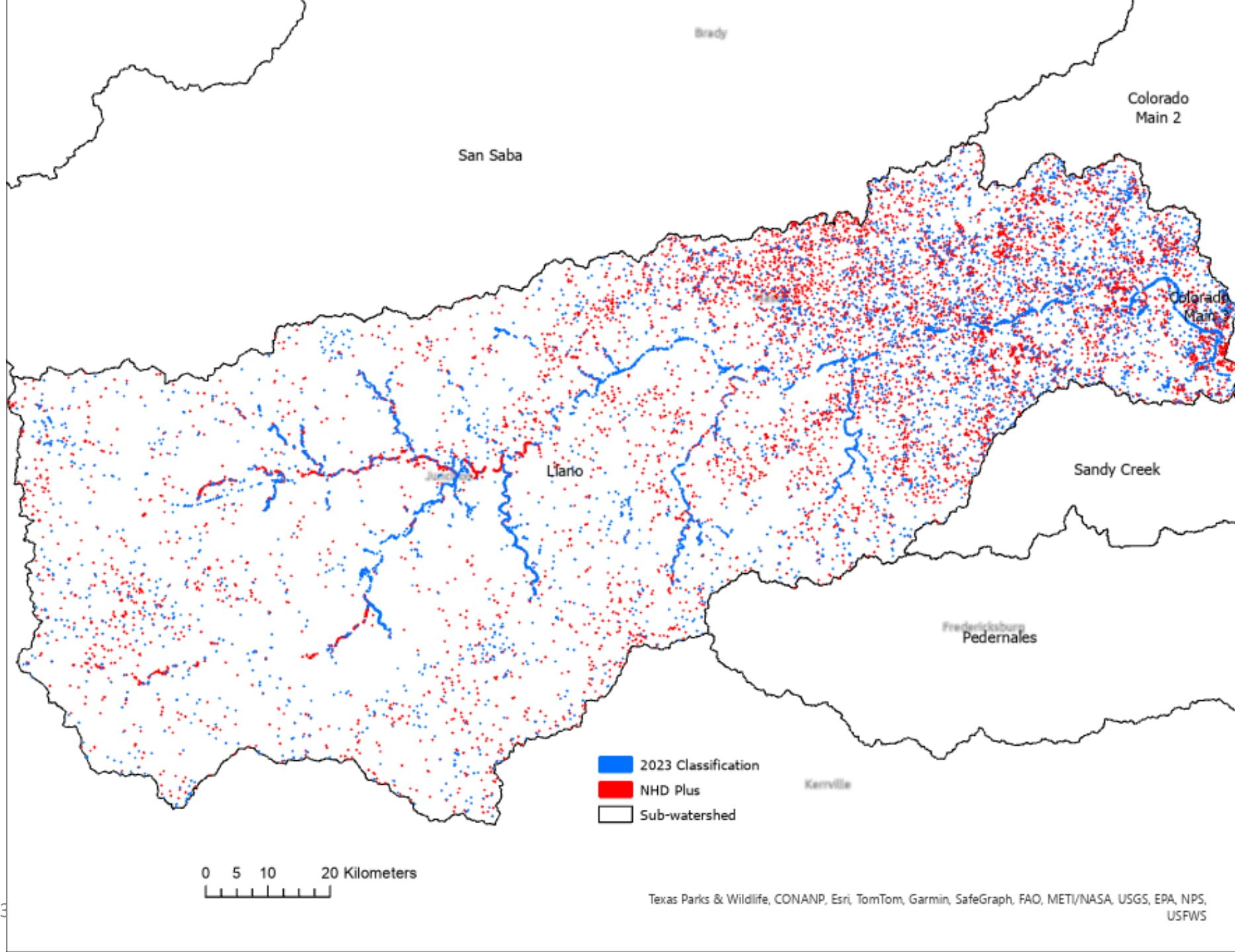
Highland Lakes Watershed Pond Area and Count December 2015 – March 2023

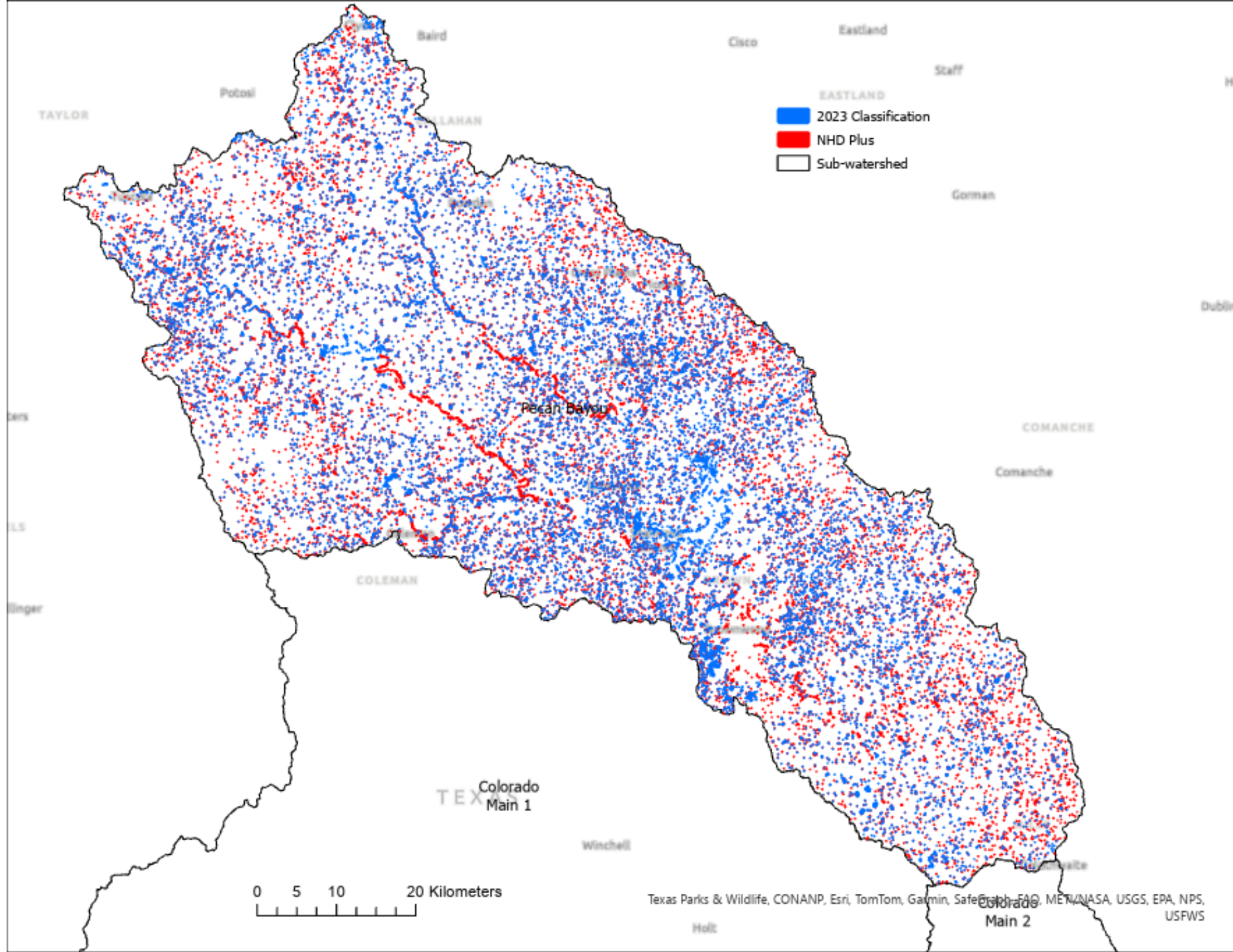


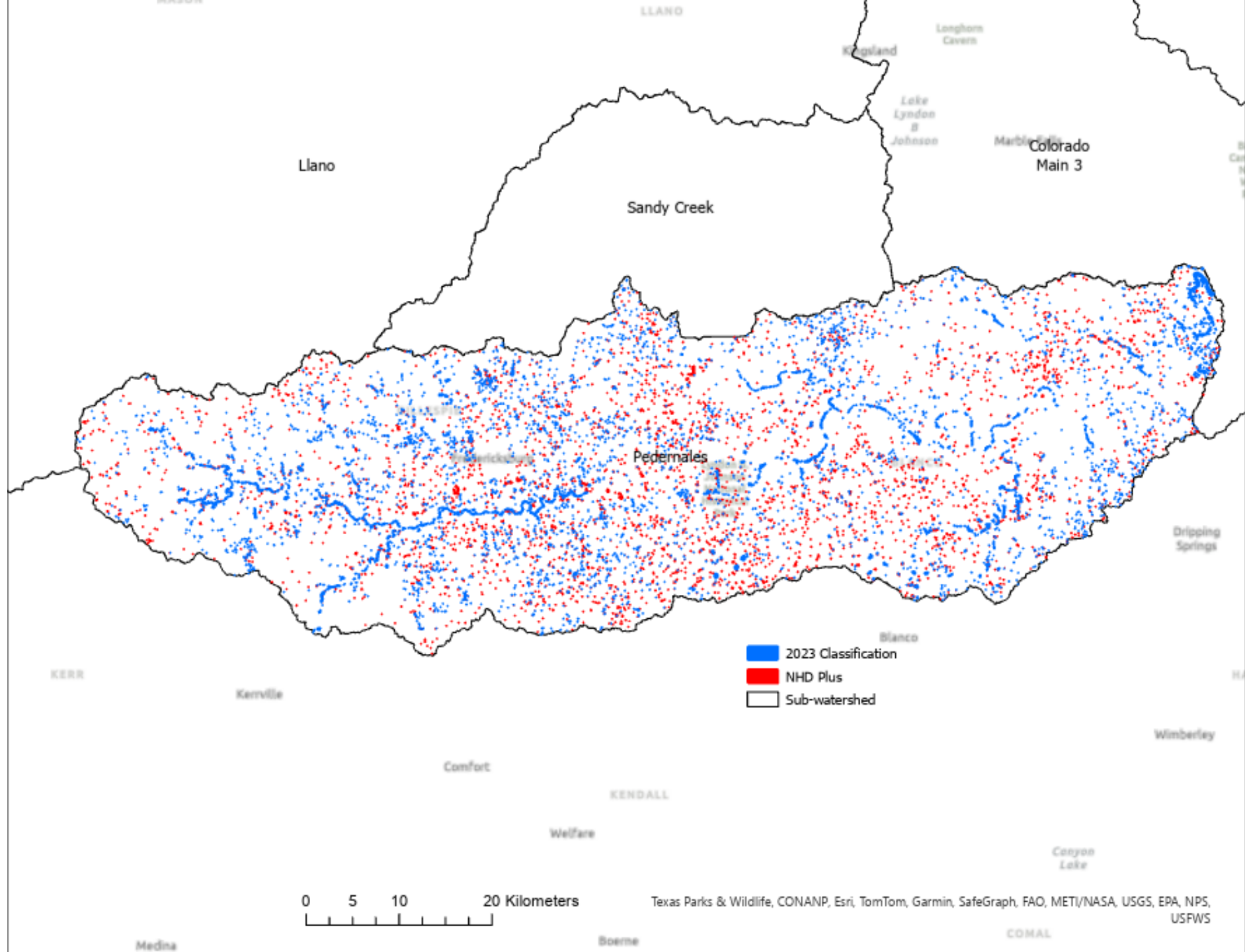


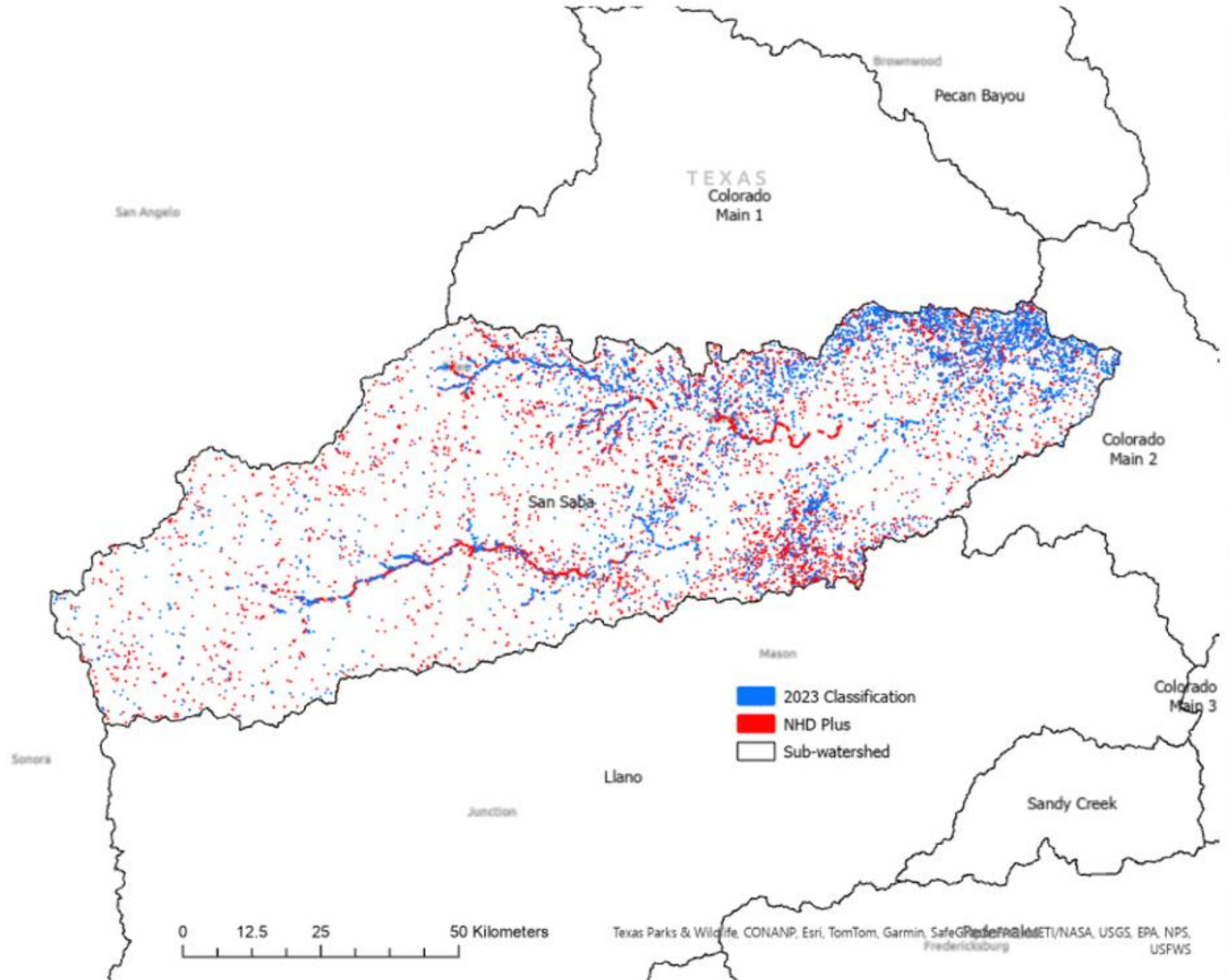


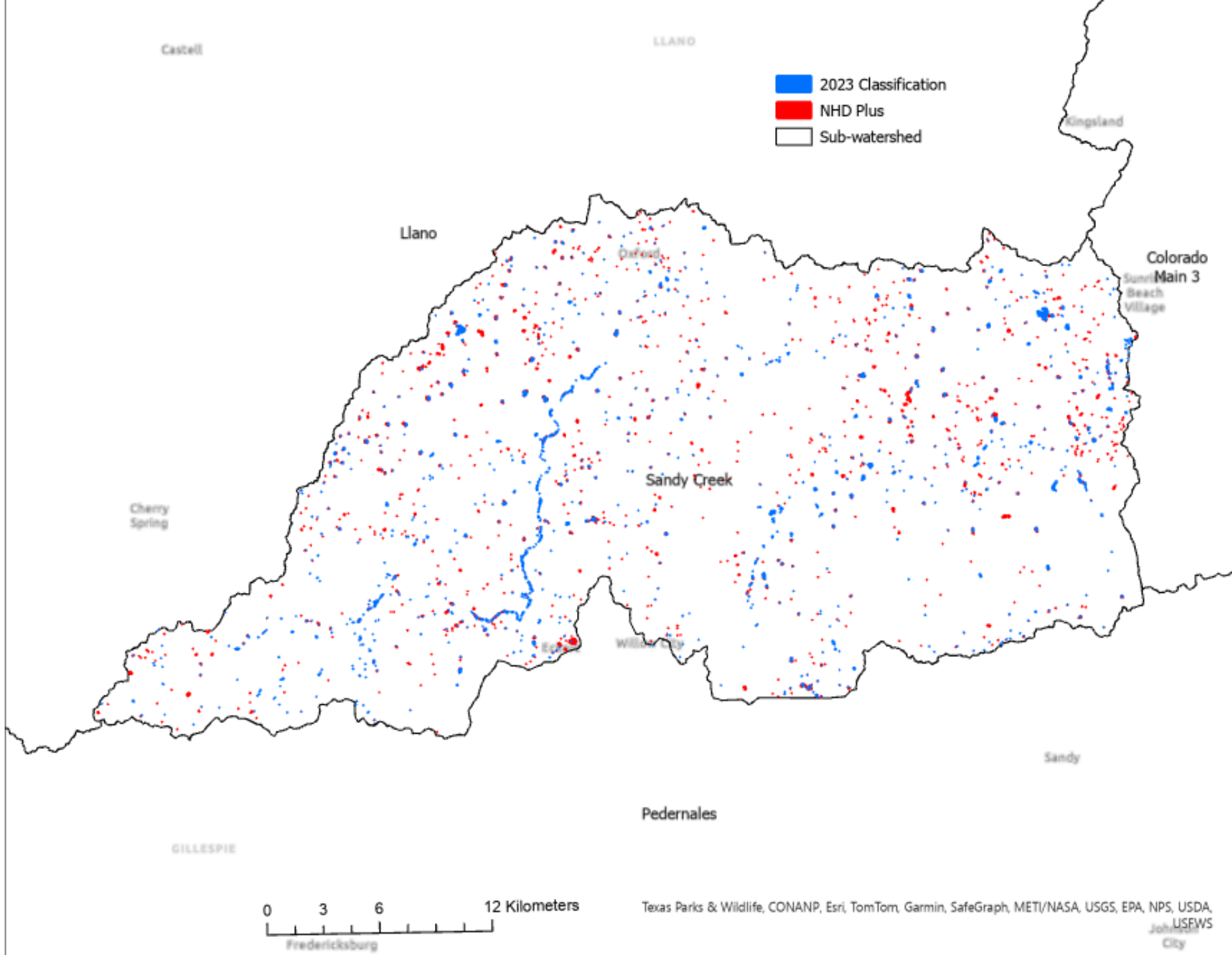


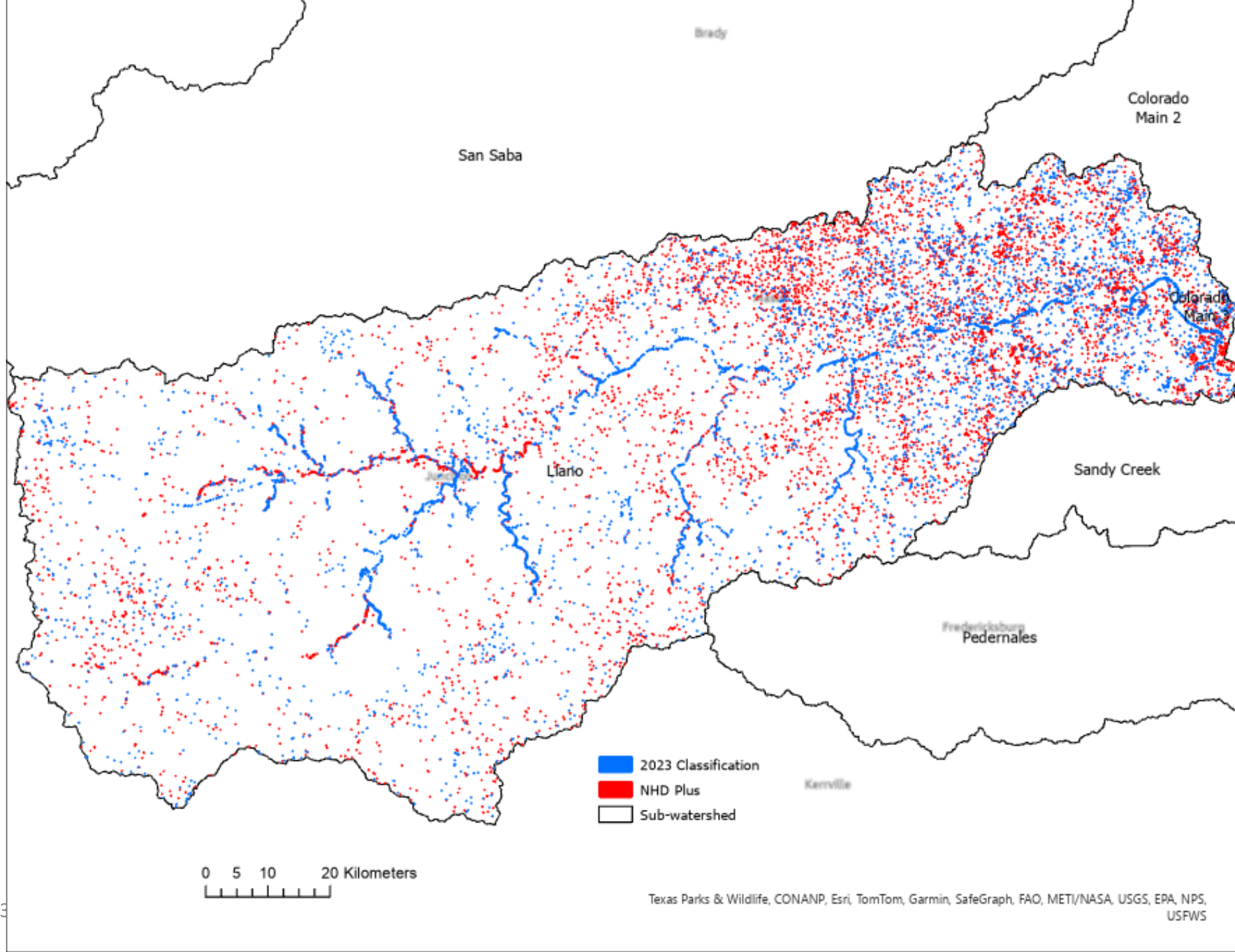




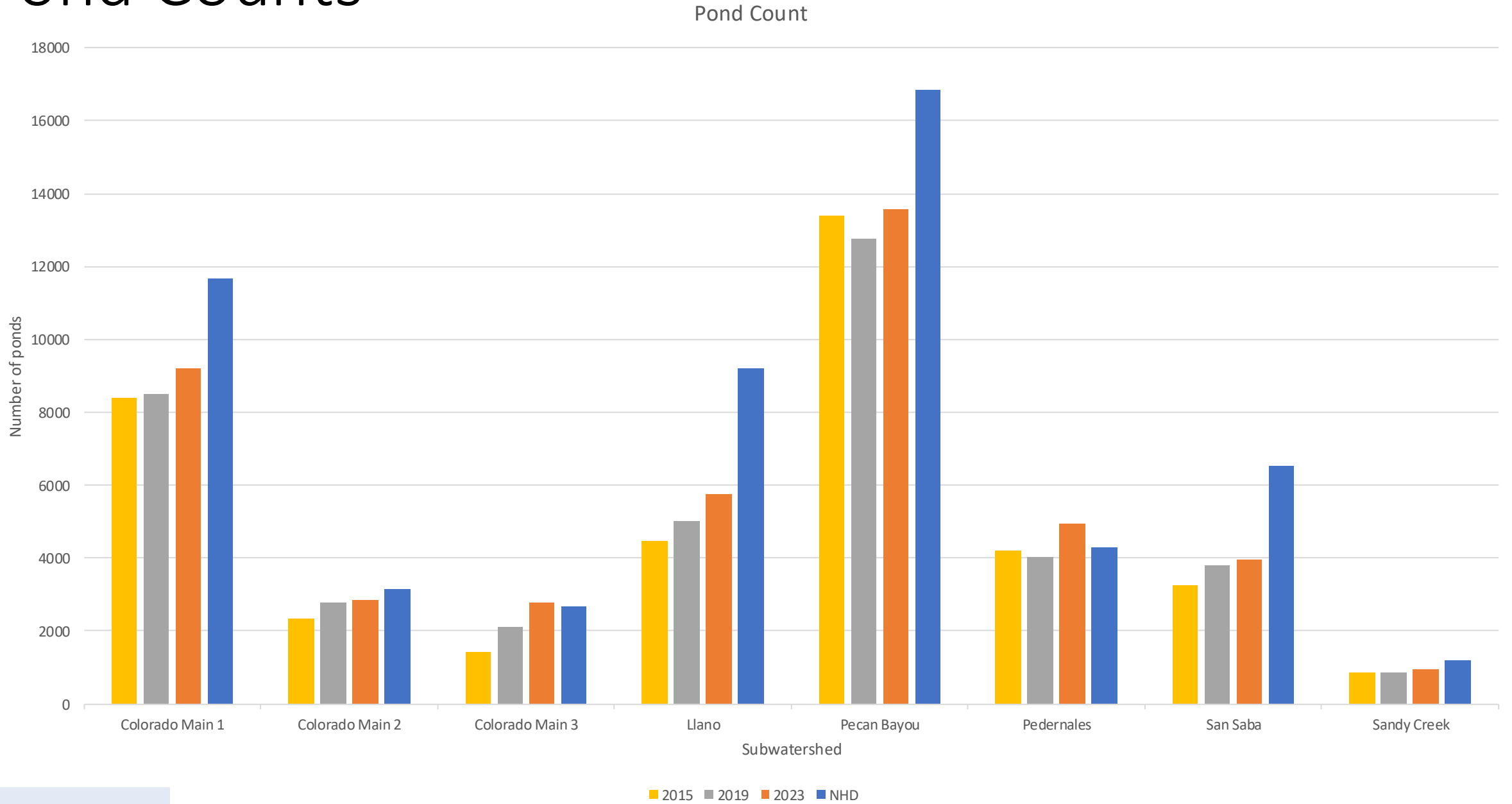






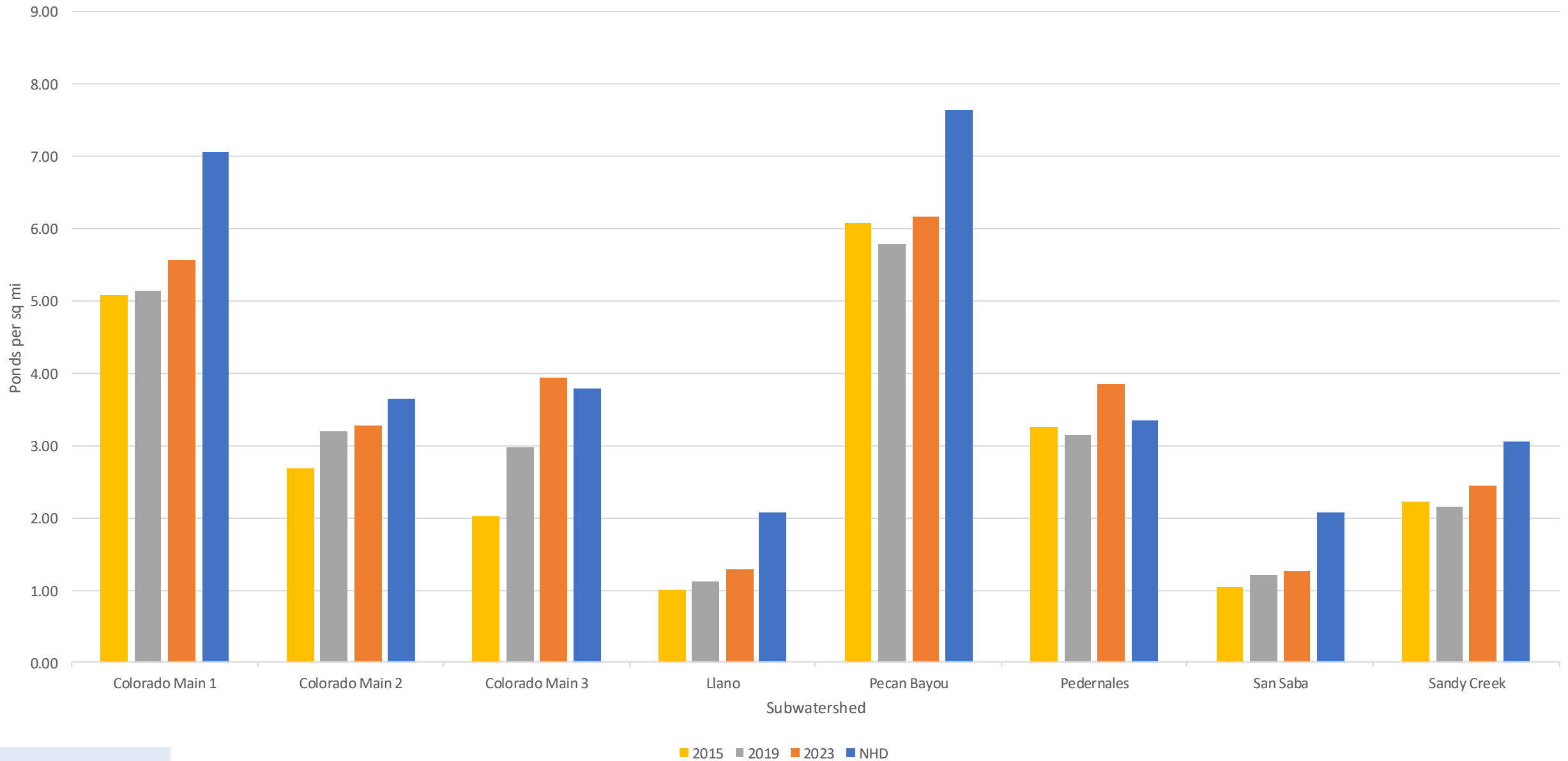


Pond Counts



Pond Density

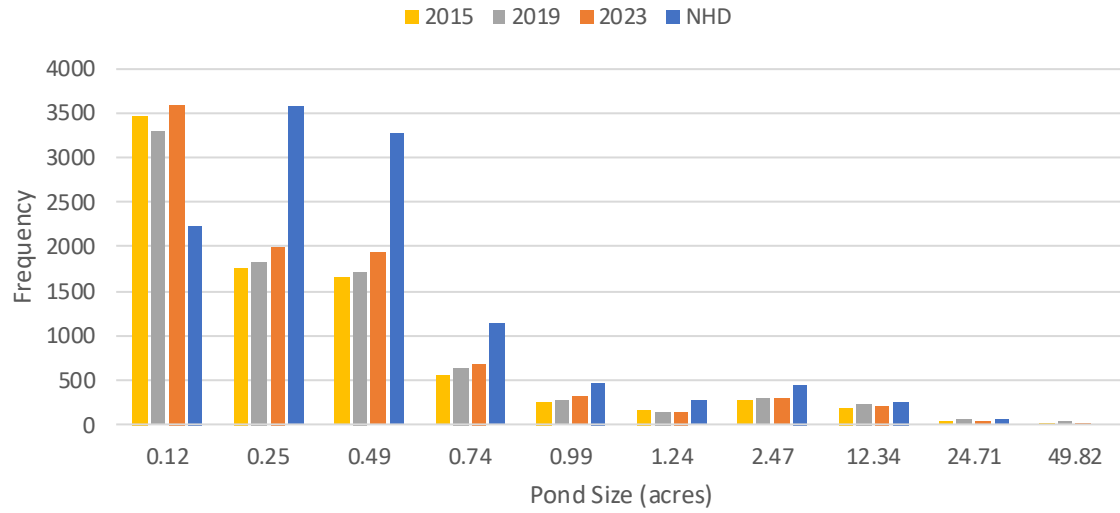
Pond Density



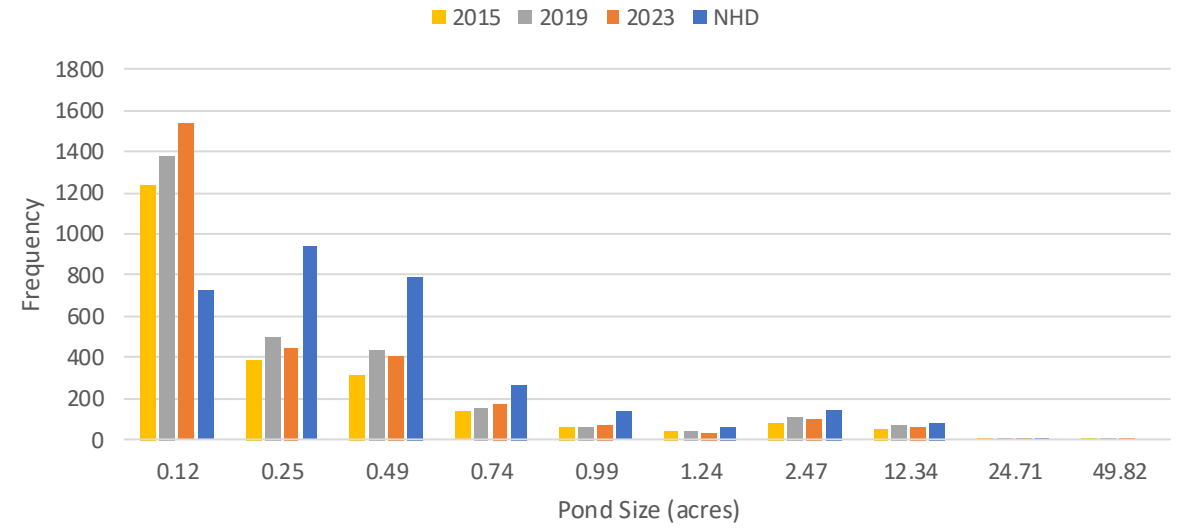
Annualized percent increase in number of ponds for three time intervals for Highland Lakes subwatersheds for all ponds greater than 1,076 ft² (100 m²).

Subwatershed	December 2015- February 2019	February 2019- March 2023	December 2015- March 2023
Colorado Main 1	0.35	2.01	1.28
Colorado Main 2	5.68	0.62	2.80
Colorado Main 3	12.93	7.18	9.65
Llano	3.55	3.47	3.50
Pecan Bayou	-1.54	1.53	0.18
Pedernales	-1.22	5.10	2.29
San Saba	4.91	0.91	2.64
Sandy Creek	-0.78	3.04	1.35

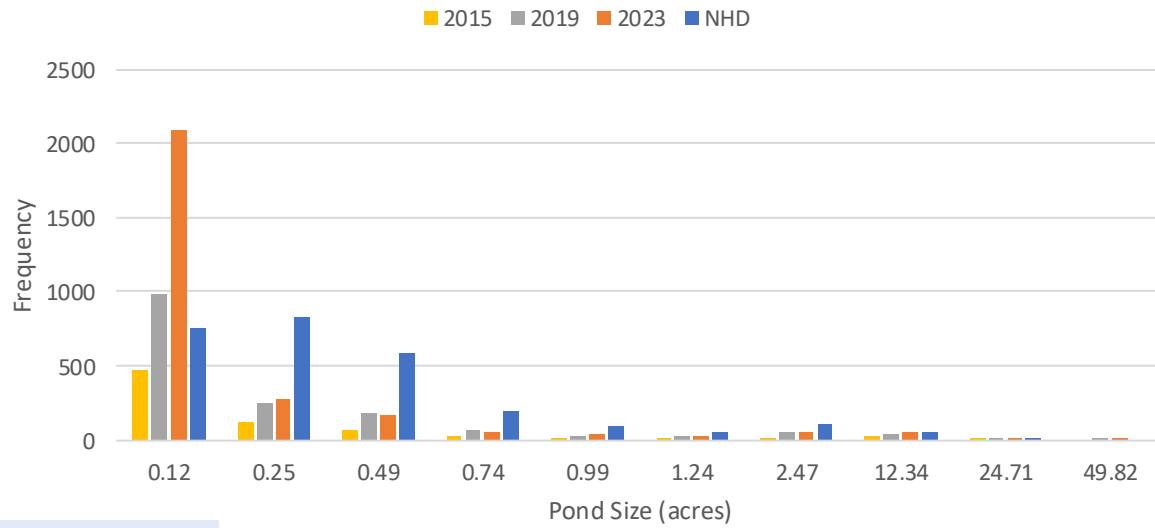
Colorado Main 1



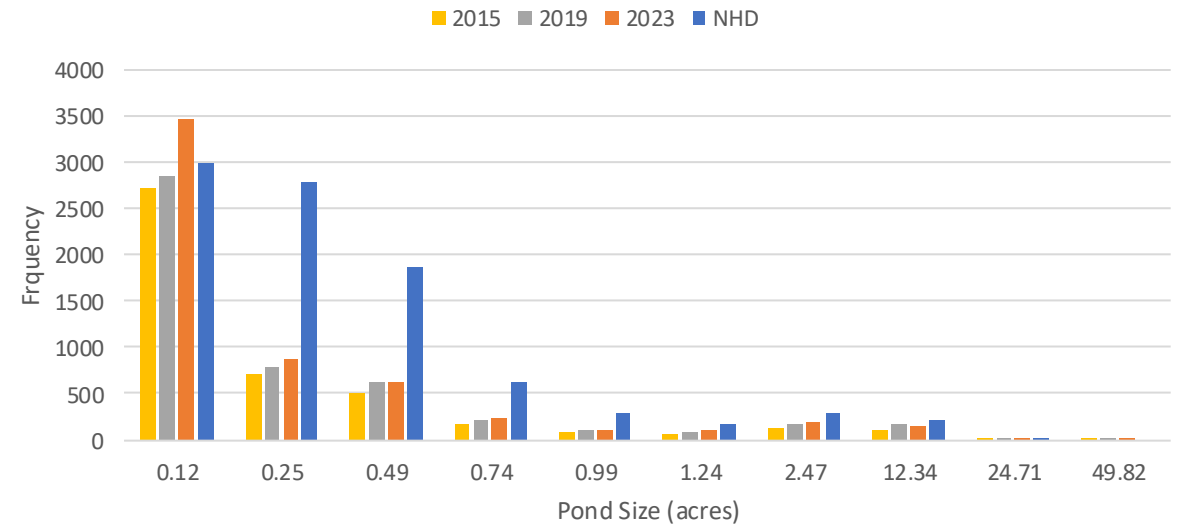
Colorado Main 2



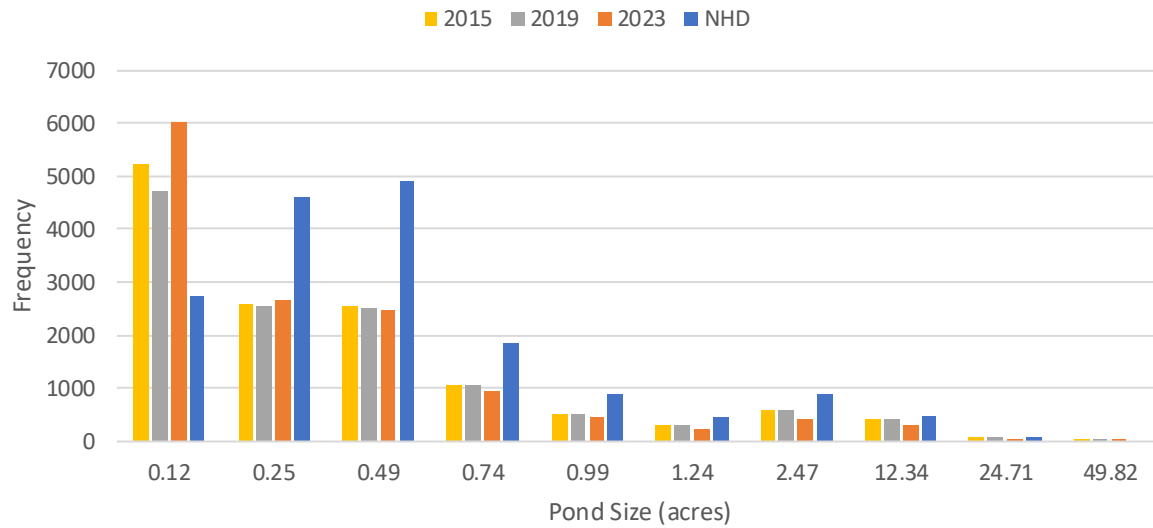
Colorado Main 3



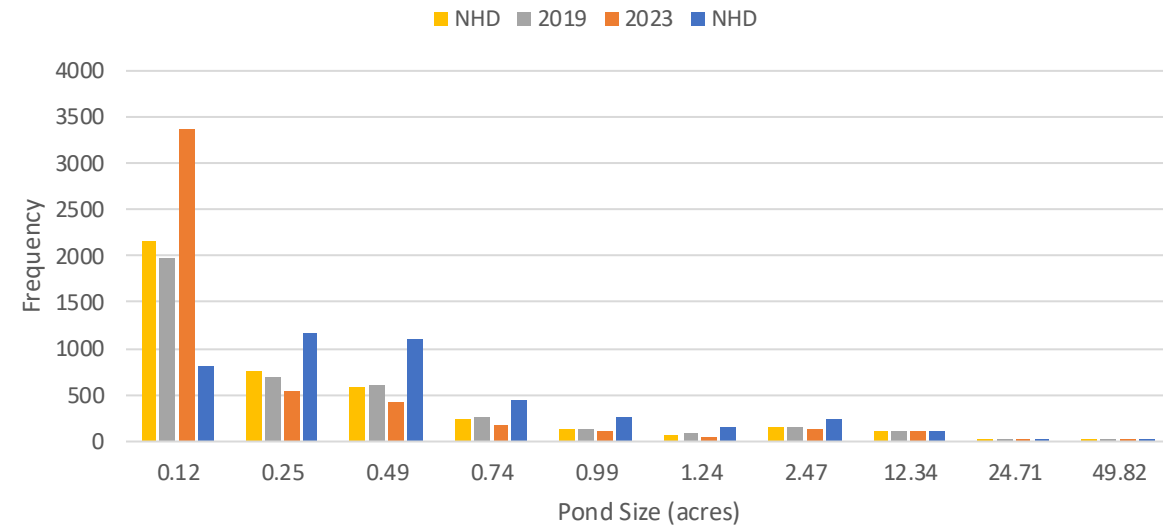
Llano



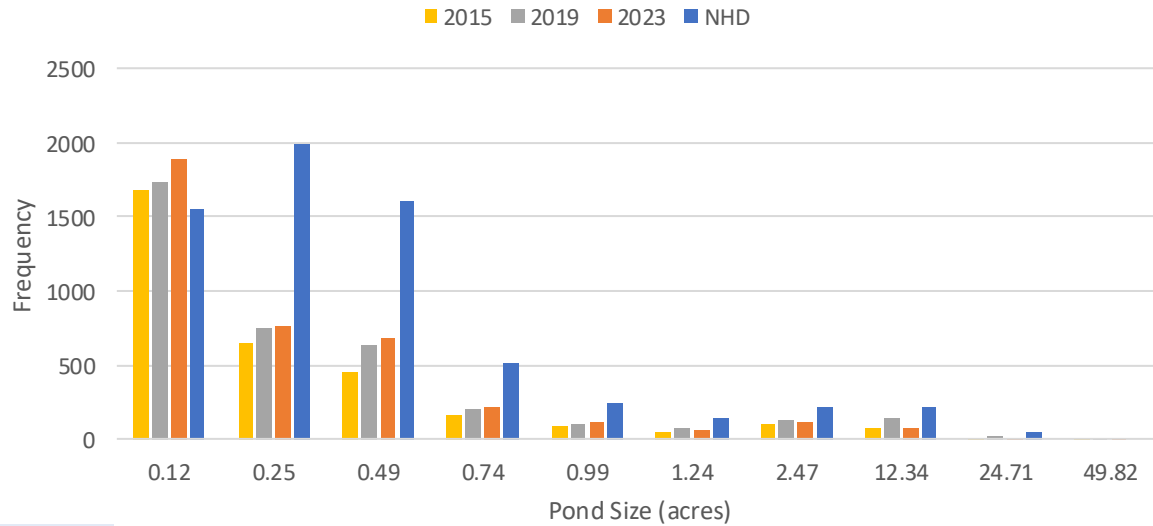
Pecan Bayou



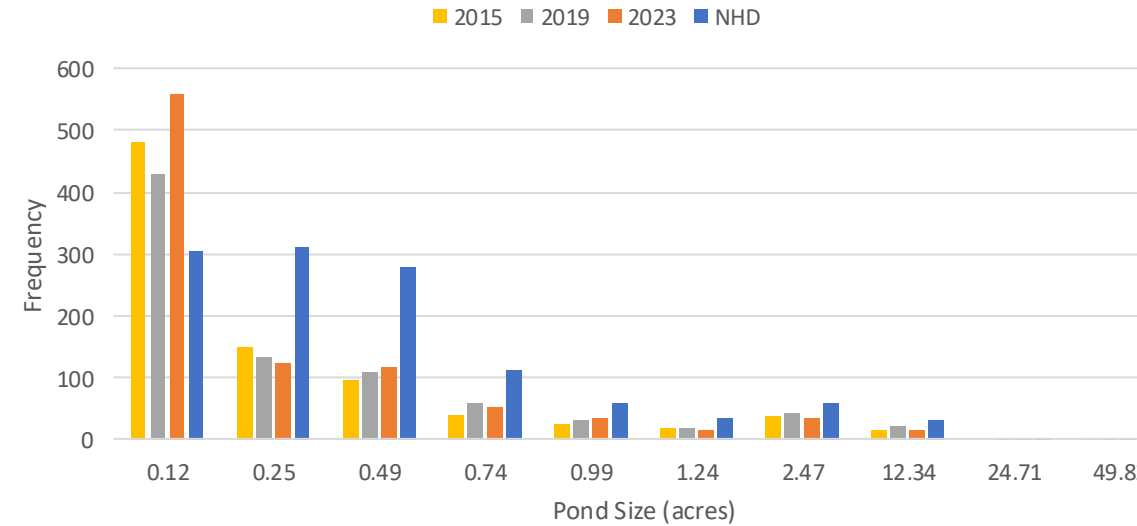
Pedernales



San Saba

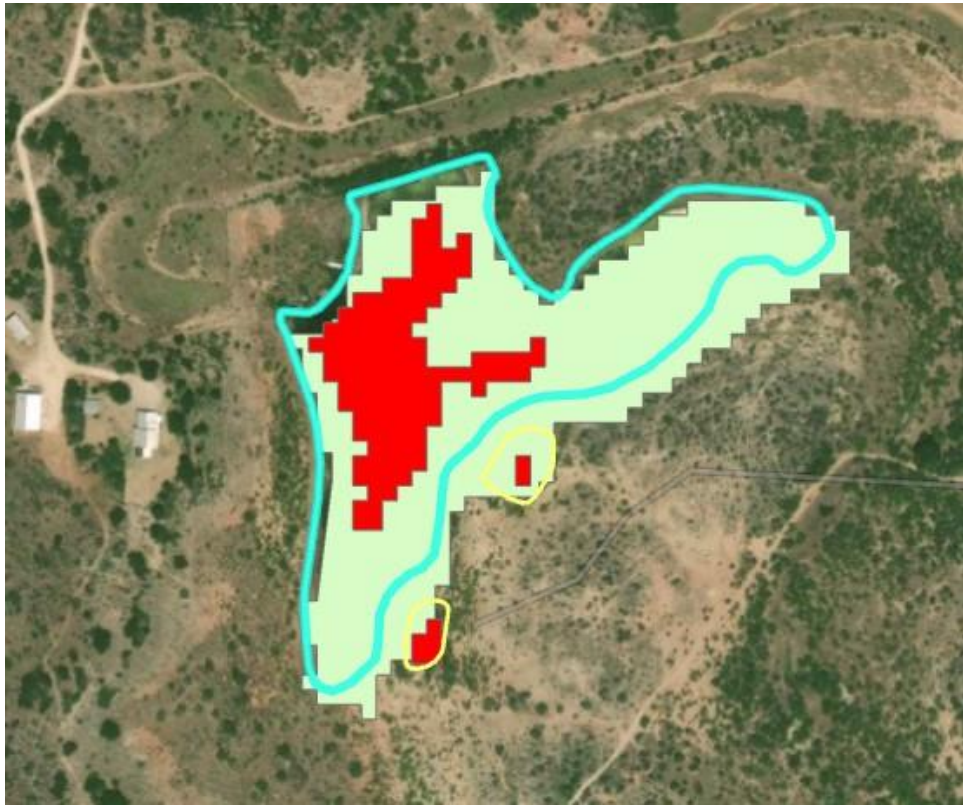


Sandy Creek



Overall Results

- Sentinel-2 + DEM data performed well for pond classification
 - High classification accuracies and relatively low omission errors (2023 highest at 8.5%)
- Analysis over time indicates an increased number of ponds and pond density, though increases vary depending on subwatershed
- Comparison with NHD indicated a large discrepancy between total NHD ponds and our 2023 classification
 - Some 2023 omission errors
 - **Many** NHD pond 'double-counts'



Summary

- Spectral index-based classification of small ponds in the Highland Lakes watershed produced reliable results for detection, inventory, and longer-term monitoring of surface water.
- Simple, customizable procedure using publicly available datasets.
- Excellent potential to extend to additional areas and/or years.

Acknowledgements

- LCRA provided funding for this project through an Inter-Agency Cooperation Agreement (A 2023-0065)
 - Thanks to Ron Anderson, P.E., D. WRE and Hank Zook, Cartographer, Water Engineering Department
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- Garrett Pugh (MS Geography; current PhD student) and Grayson Wylie (PhD candidate)
- Questions?

Thank you! Questions?

- Jennifer Jensen jjensen@txstate.edu
- Garrett Pugh glp54@txstate.edu
- Grayson Wylie nqe1@txstate.edu

