





Ethan Duke

MRBO - Co-Founder / Co-Director

Originally from Western NY

Studied birds in:

New York

Michigan

Tennessee

Arkansas

Wyoming

Missouri

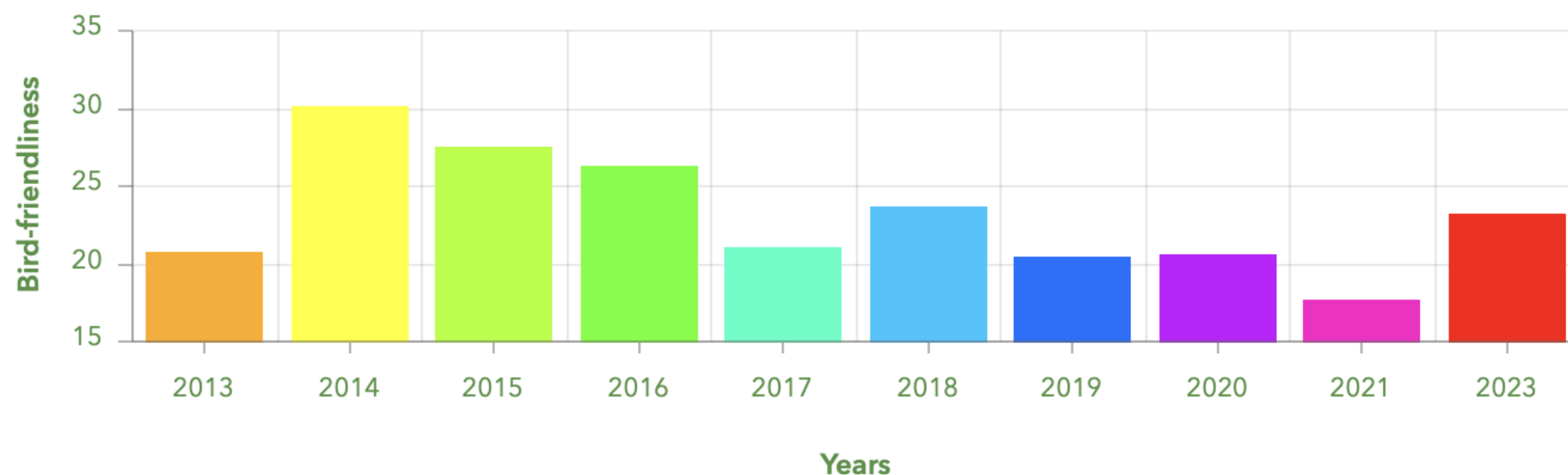


Pounds of CAFO Waste Per Acre in MO Senate Districts

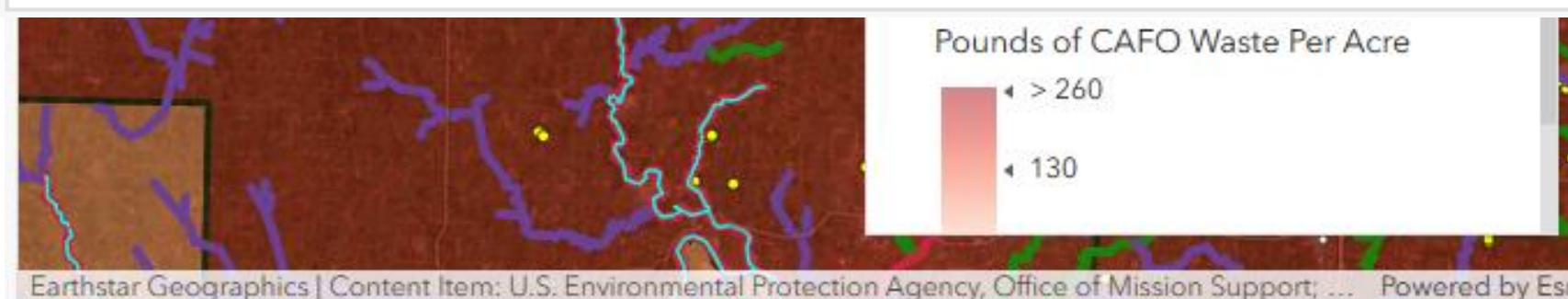
Tons of Waste by Known Sources (CAFO and Human)

105k

Bird-friendliness Scores by Year Since 2012: Best used for site-level, year-to-year comparisons.



↓ Last update: 14 seconds ago



Rusty Black  
Sandy Crawfor  
Denny Hoskin  
Holly Thompson Rehde  
Curtis Tren  
Nick Schroe

Nest Plant Species: SUMA

OBSERVATIONS



# ACKNOWLEDGEMENTS

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## THE BIRDS!

*People that see the vision for  
more effective conservation  
through tech*

A HUGE LIST OF MISSOURI  
PARTNERS  
MRBO SUPPORTERS,  
DONORS, AND BOARD

DANA RIPPER

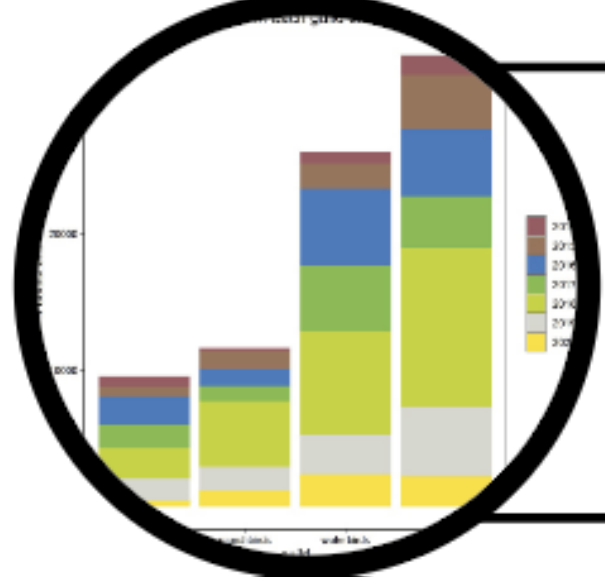


# Our Mission



## Conservation

To contribute to the conservation of Missouri's migratory and resident birds through scientific research, education, and conservation policy advocacy.



## Science

To gather information about avian communities and habitat use that will assist state, federal, and private natural resource managers in their efforts to implement conservation programs.



## Education & Outreach

To provide opportunities for people of all ages to learn about species and their habitats.



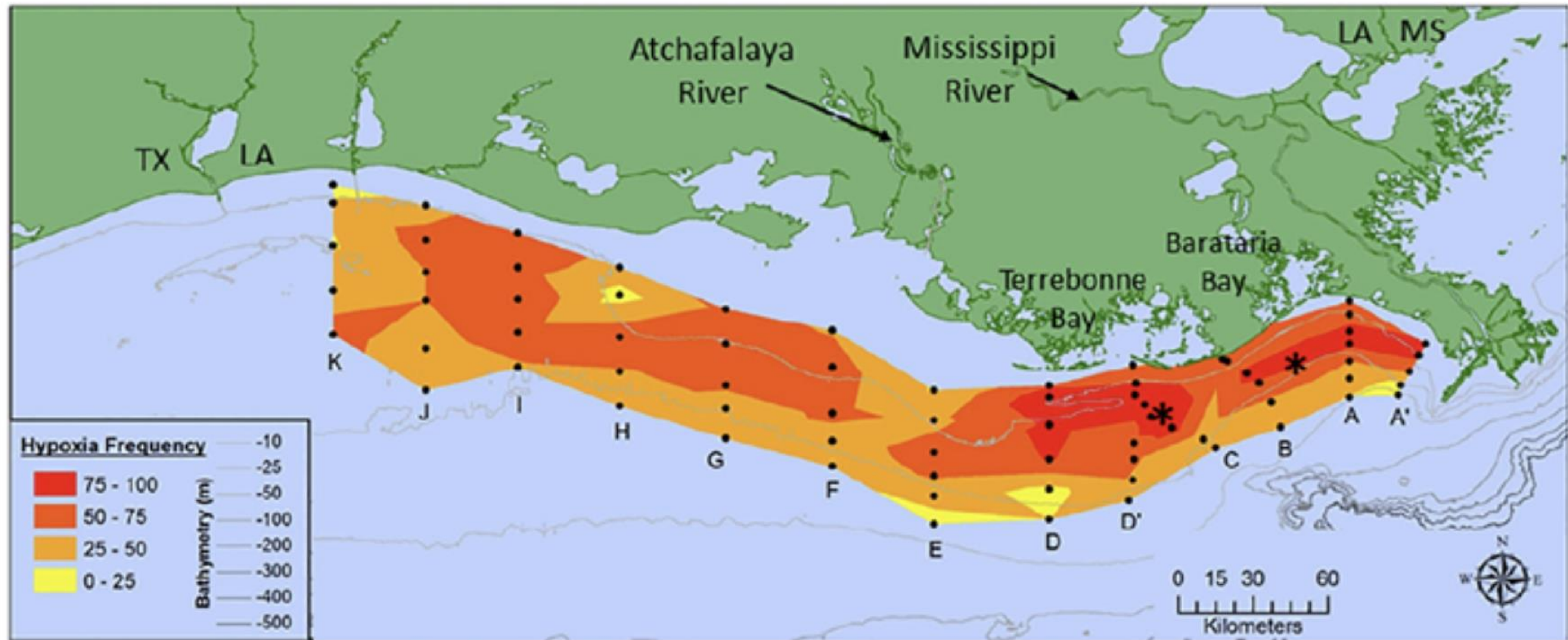
## Advocacy

To advocate for sound, science-based conservation policies that benefit birds, other wildlife and environmental quality.

# SOLUTIONS TO MATCH THE CHALLENGES



Acres Lost

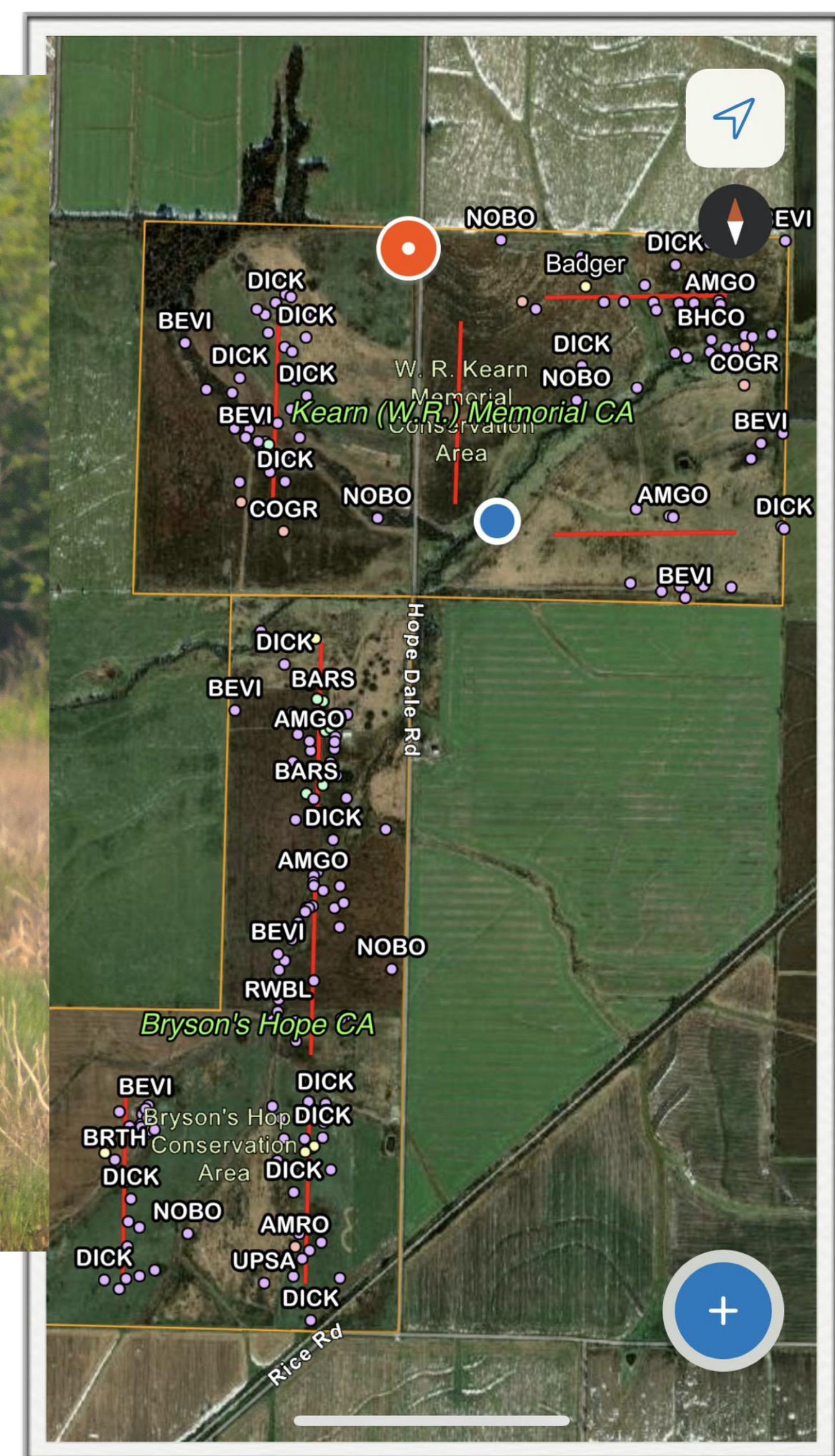


**FIG. 1.** The frequency of bottom-water hypoxia from shelf-wide hypoxia mapping (1985–2014) (updated from Rabalais et al. (2007b); frequency is determined from stations for which there are data for at least half of all cruises. Asterisks (\*) indicate locations of near-bottom oxygen meters; transects C and F identified. Data source: N. N. Rabalais and R. E. Turner.

# *“A Look Into The Future”*

*Northern Bobwhite Quail Photo by Mark Ramsey*





Modern Data Collection and workflows



# SOLUTIONS TO MATCH THE CHALLENGES

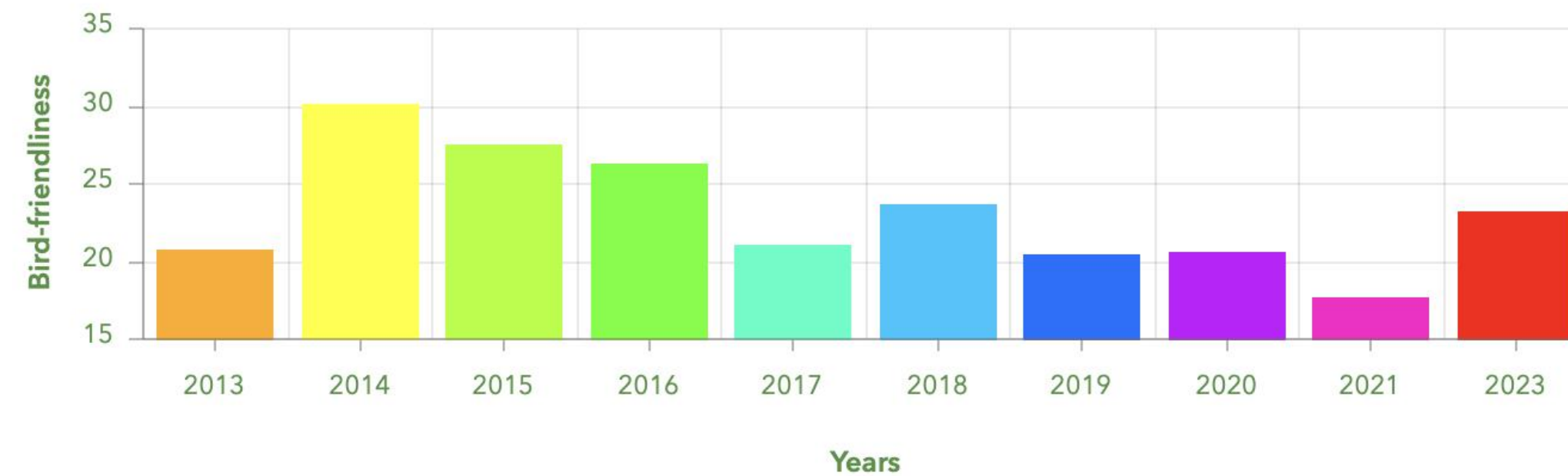


## Bird-friendliness Scores on Public, Missouri Prairie Foundation, and The Nature Conservancy Lands

Combined grassland-obligate species densities, conservation concern values, and diversity metrics.

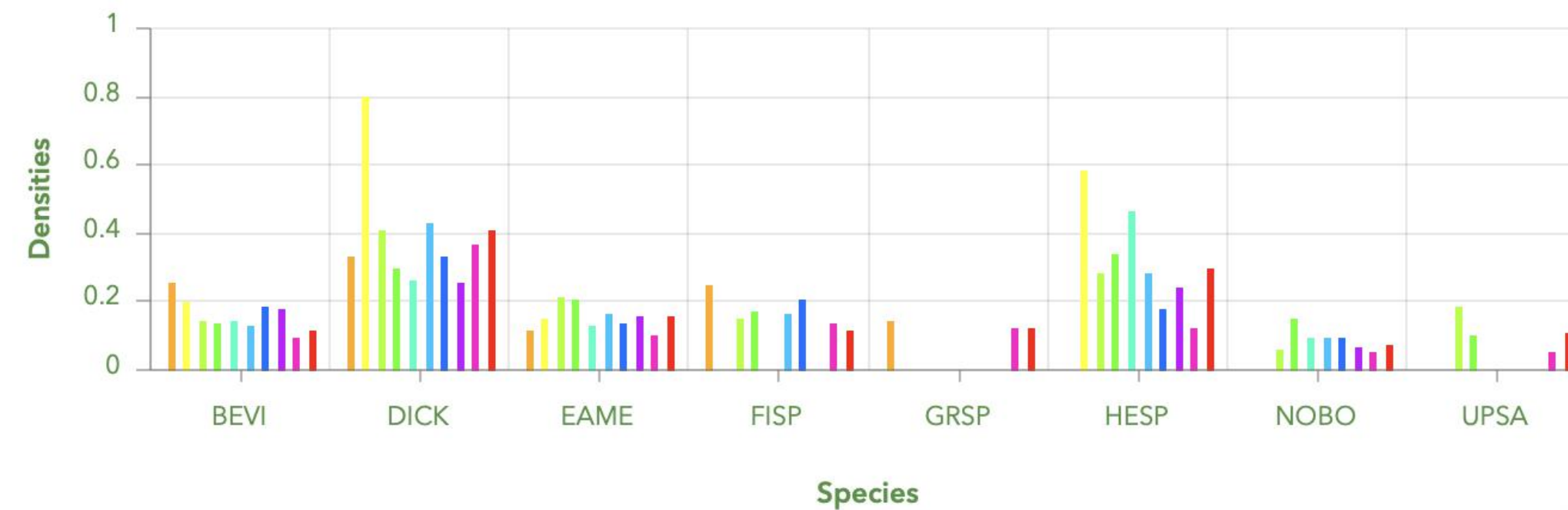


Bird-friendliness Scores by Year Since 2012: Best used for site-level, year-to-year comparisons.



↓ Last update: 14 seconds ago

Grassland-obligate Species Densities: Densities are sums, so examine this data on site-level only.



↓ Last update: 14 seconds ago







# Grassland Bird Survey Results on Public Lands

as well as partner-owned lands such as the Missouri Prairie Foundation and The Nature Conservancy

## Grassland Birds on Public Lands 2023

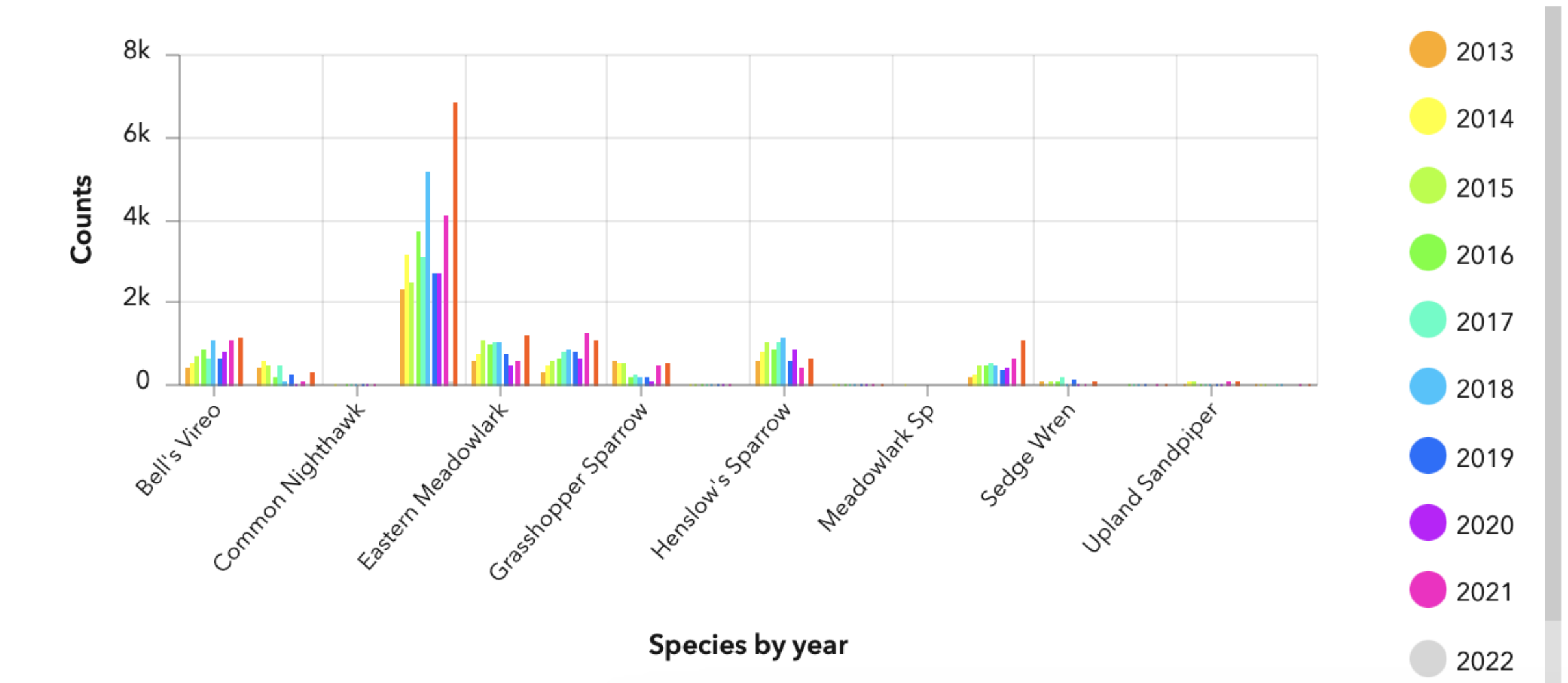


## Counts of bird species detected in 2023

Species	Counts
	7
Acadian Flycatcher	16
Alder Flycatcher	12
American Bittern	1
American Coot	1
<b>Total</b>	<b>23,530</b>

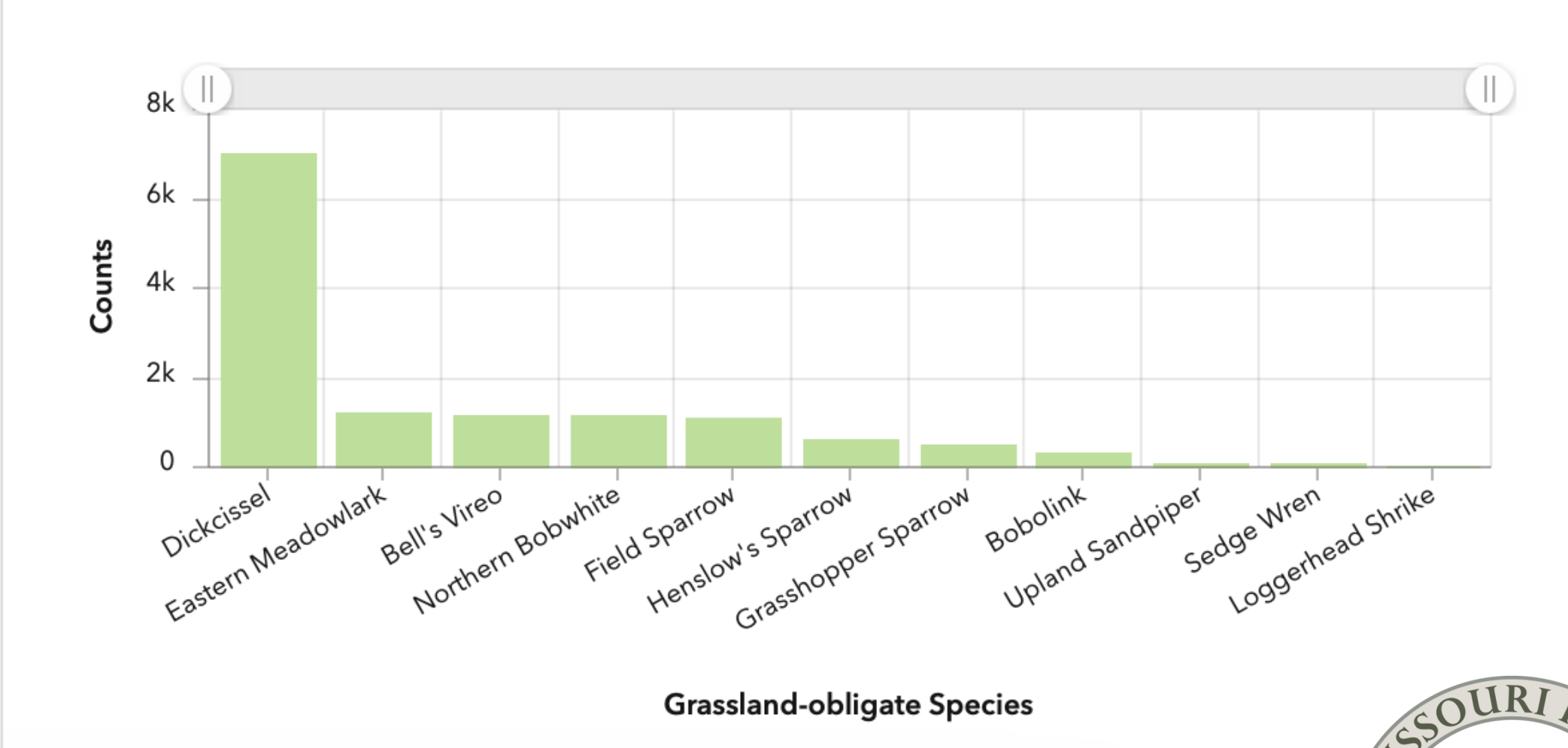
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## Counts of grassland-obligates by year since 2012



↓ Last update: 14 seconds ago

## Grassland-obligate species detected in 2023



↓ Last update: 14 seconds ago

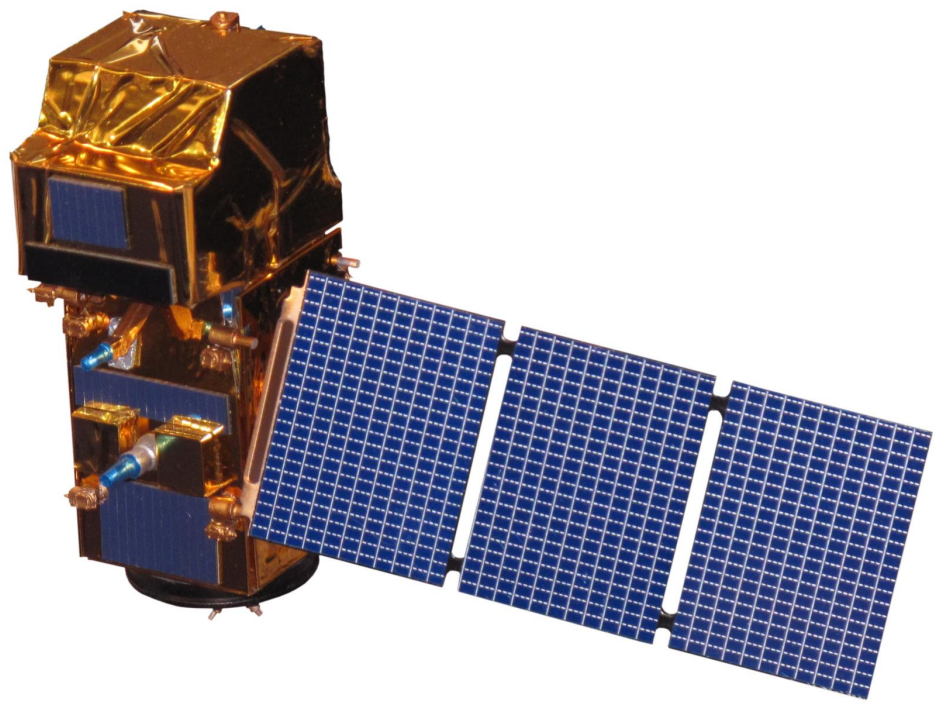
Automated deliverables



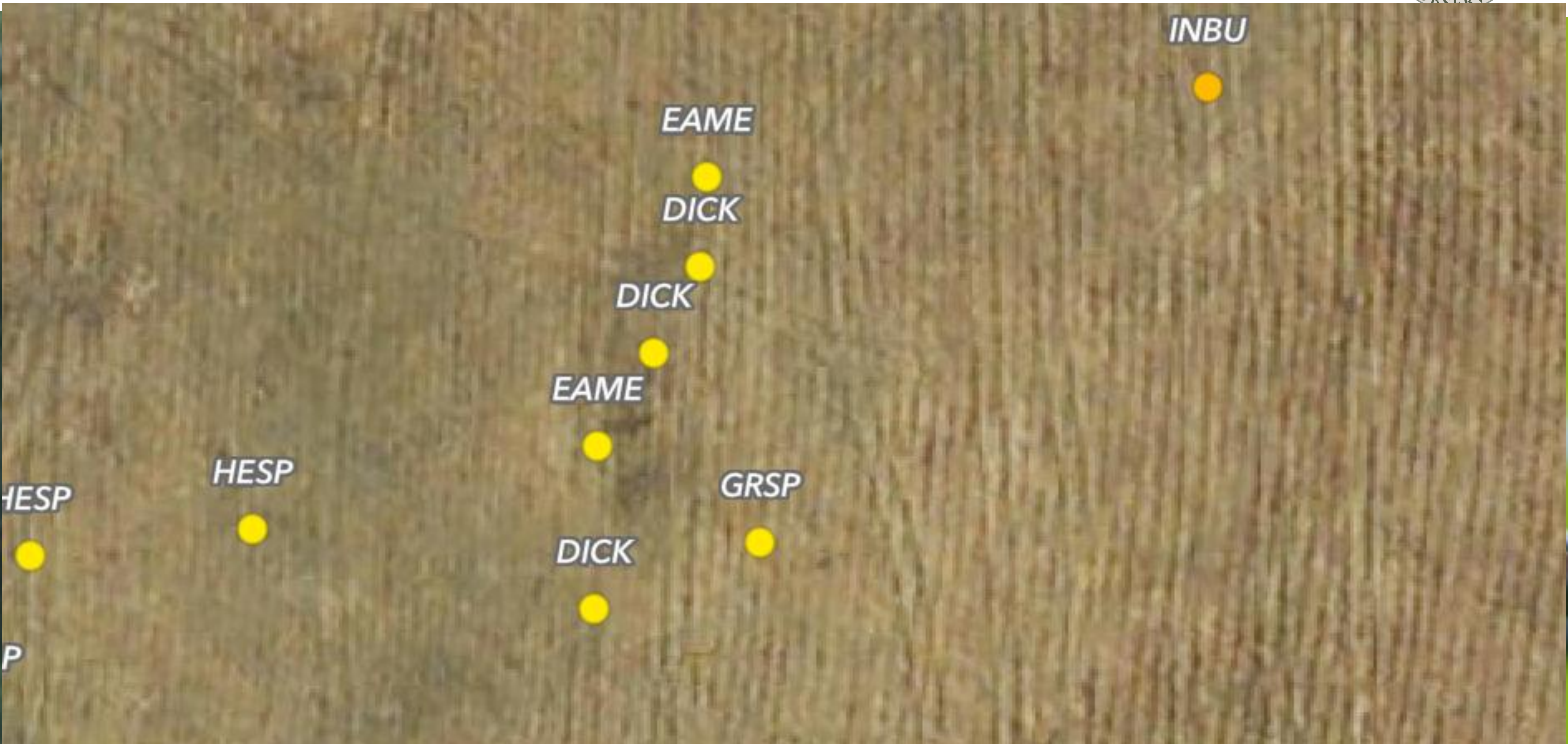


# REMOTE SENSING: FUTURE (AND CURRENT) POSSIBILITIES FOR GRASSLAND CONSERVATION

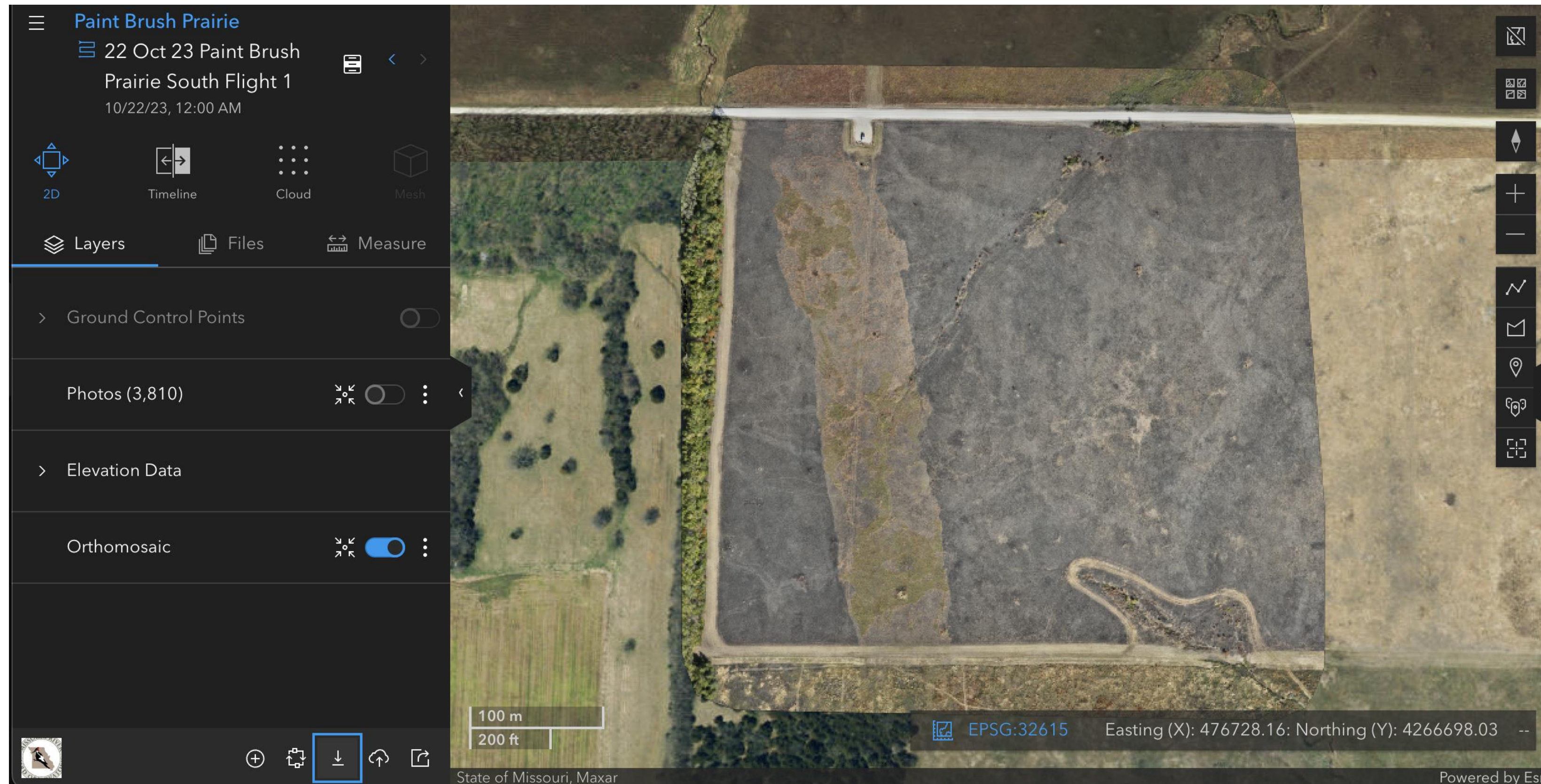
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# SAVING TIME AND MONEY.... AND HAVING BETTER DATA



# STANDARD DELIVERABLES



## RASTER DATA

Orthomosaic (.tiff)

Orthomosaic preview (.png)

DSM (.tiff)

DTM (.tiff)

Multispectral Ortho (.tiff)

## CONTOURS

Contour (.shp.zip) -DTM

Contour (.dxf) -DTM

Contour (.shp.zip) -DSM

Contour (.dxf) -DSM

## POINT CLOUD

Point Cloud (.las.zip)

Point Cloud (.laz)

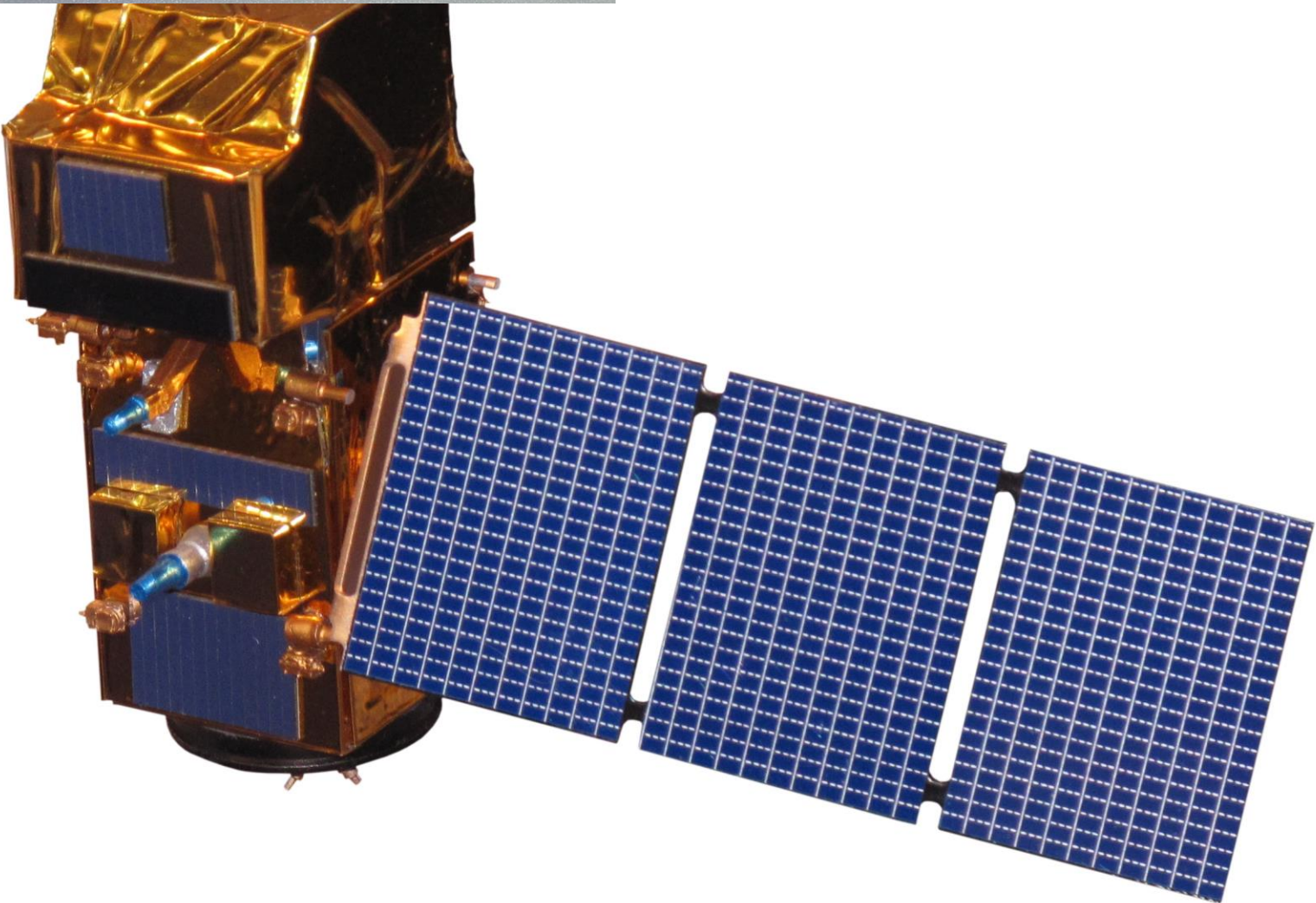
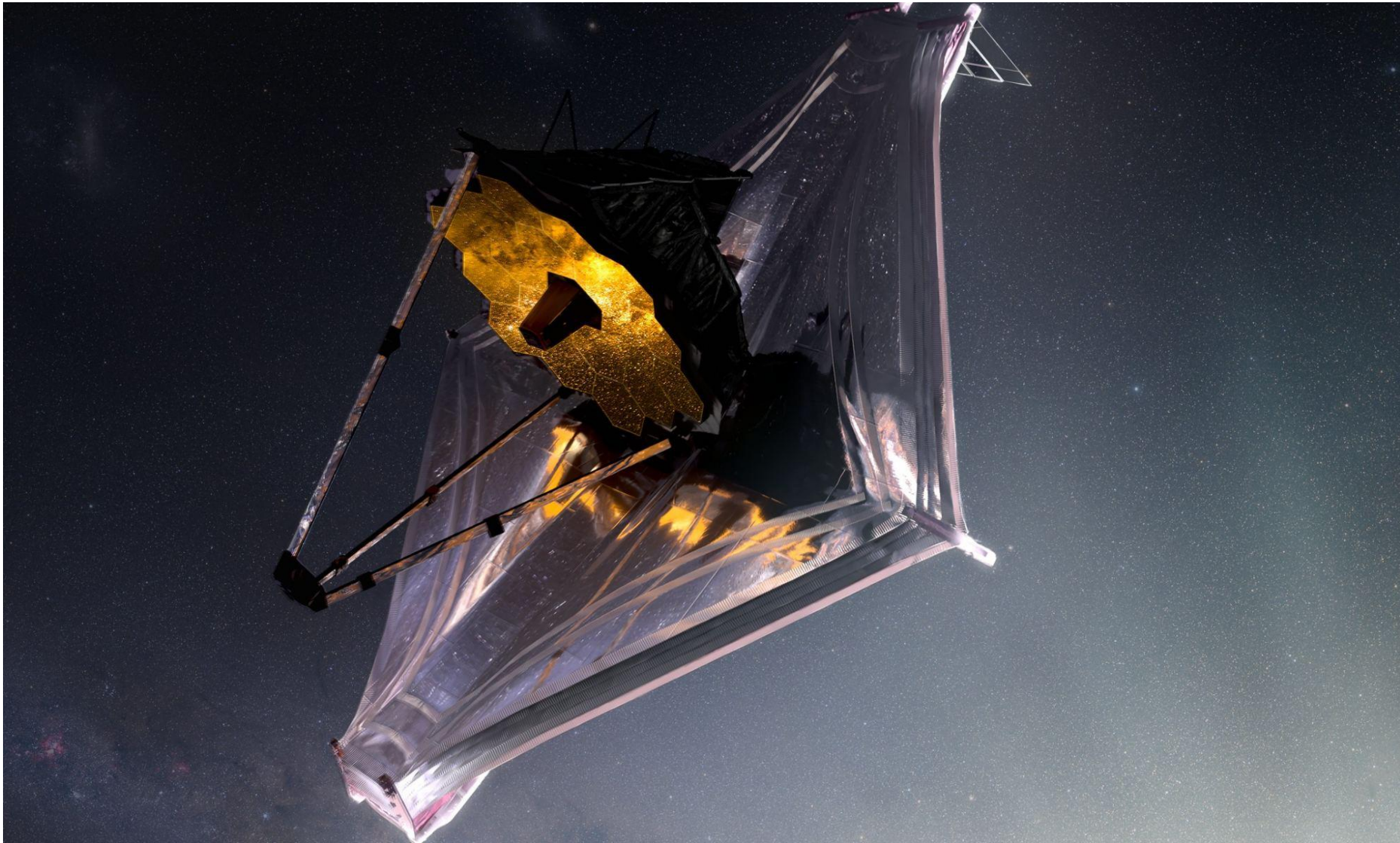
## PROCESSING REPORT

Processing Report (.pdf)

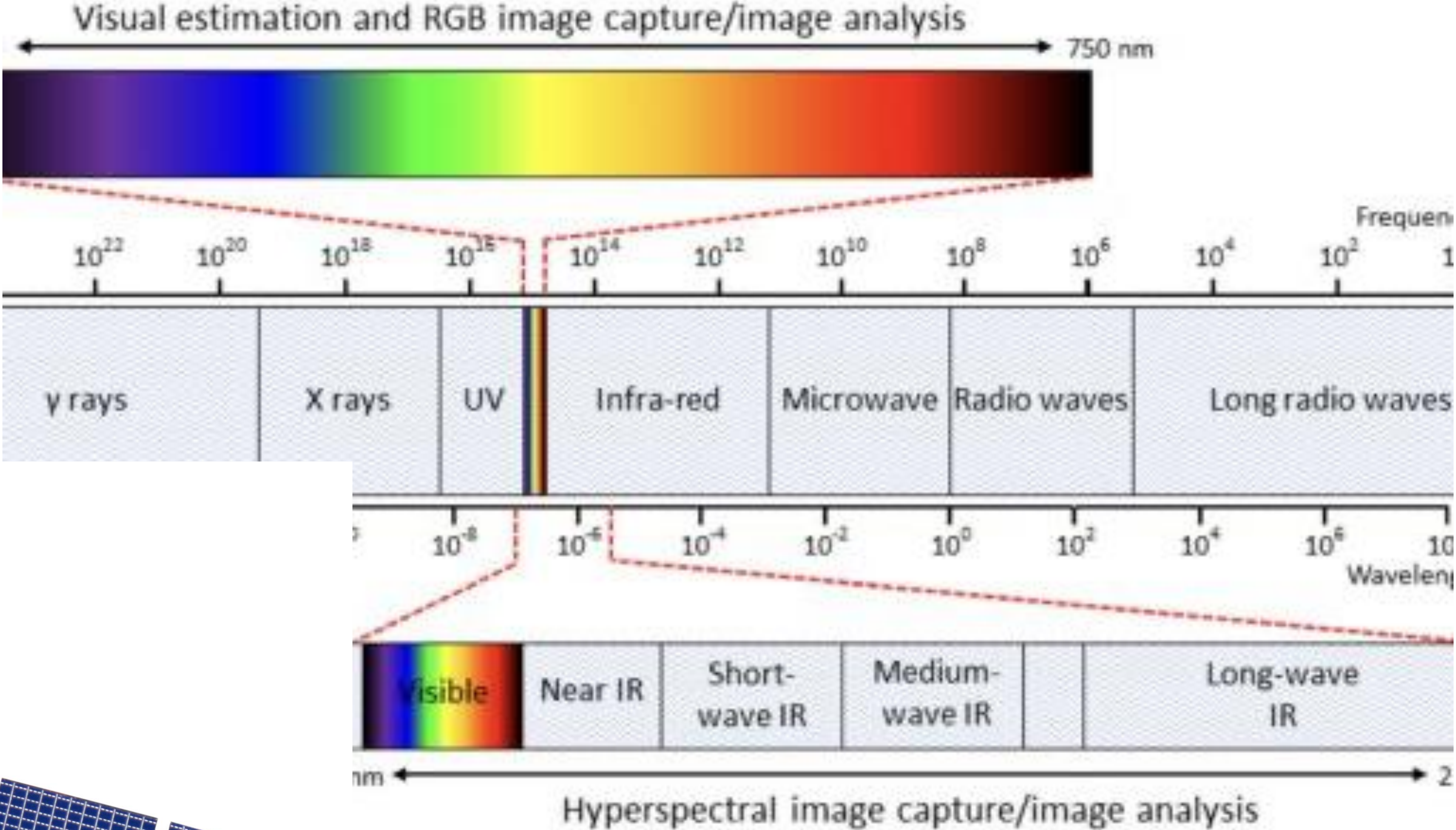
All Photos

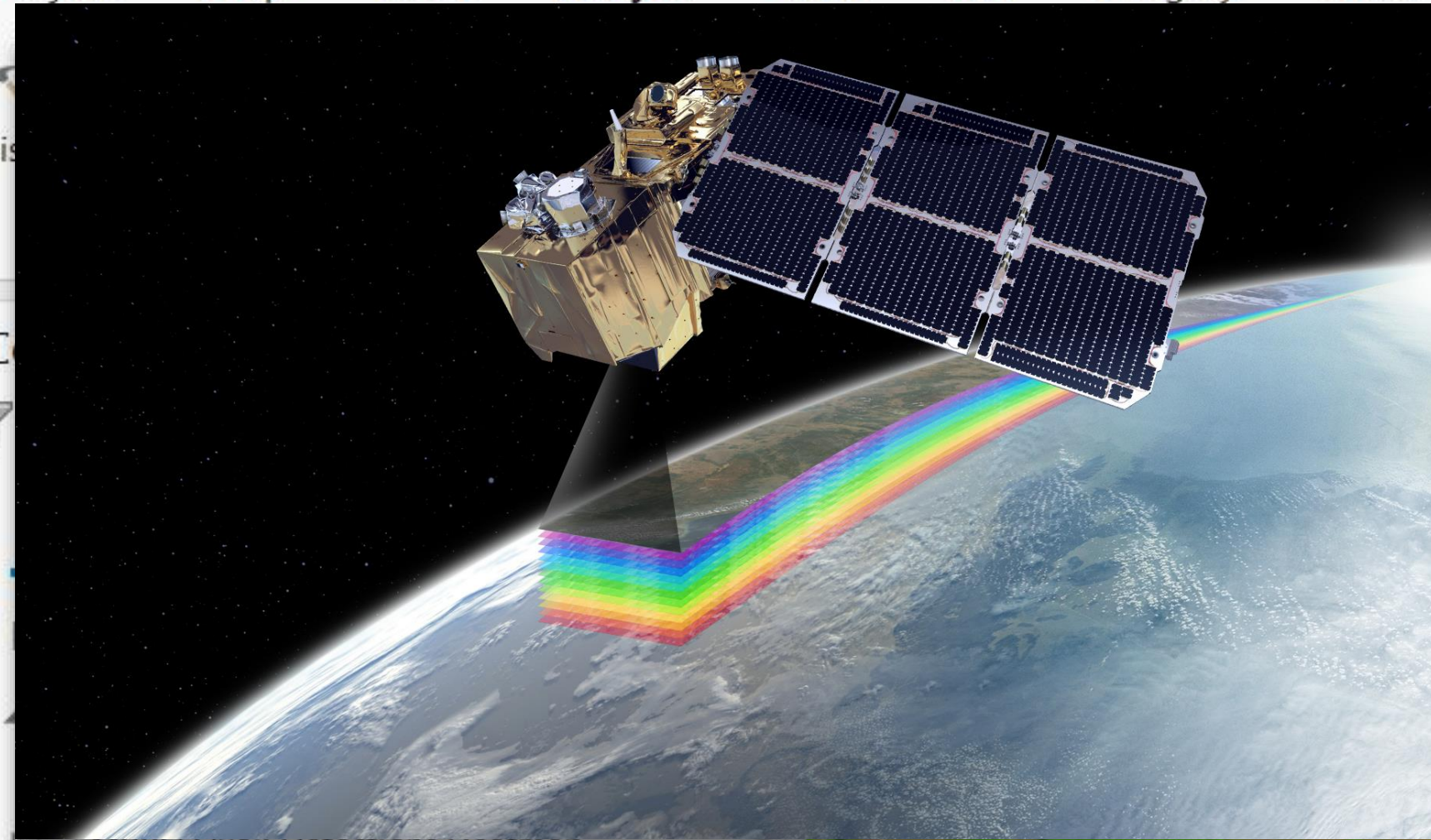
Photos (.zip)

# NOT SO STANDARD DELIVERABLES



➤ Multispectral





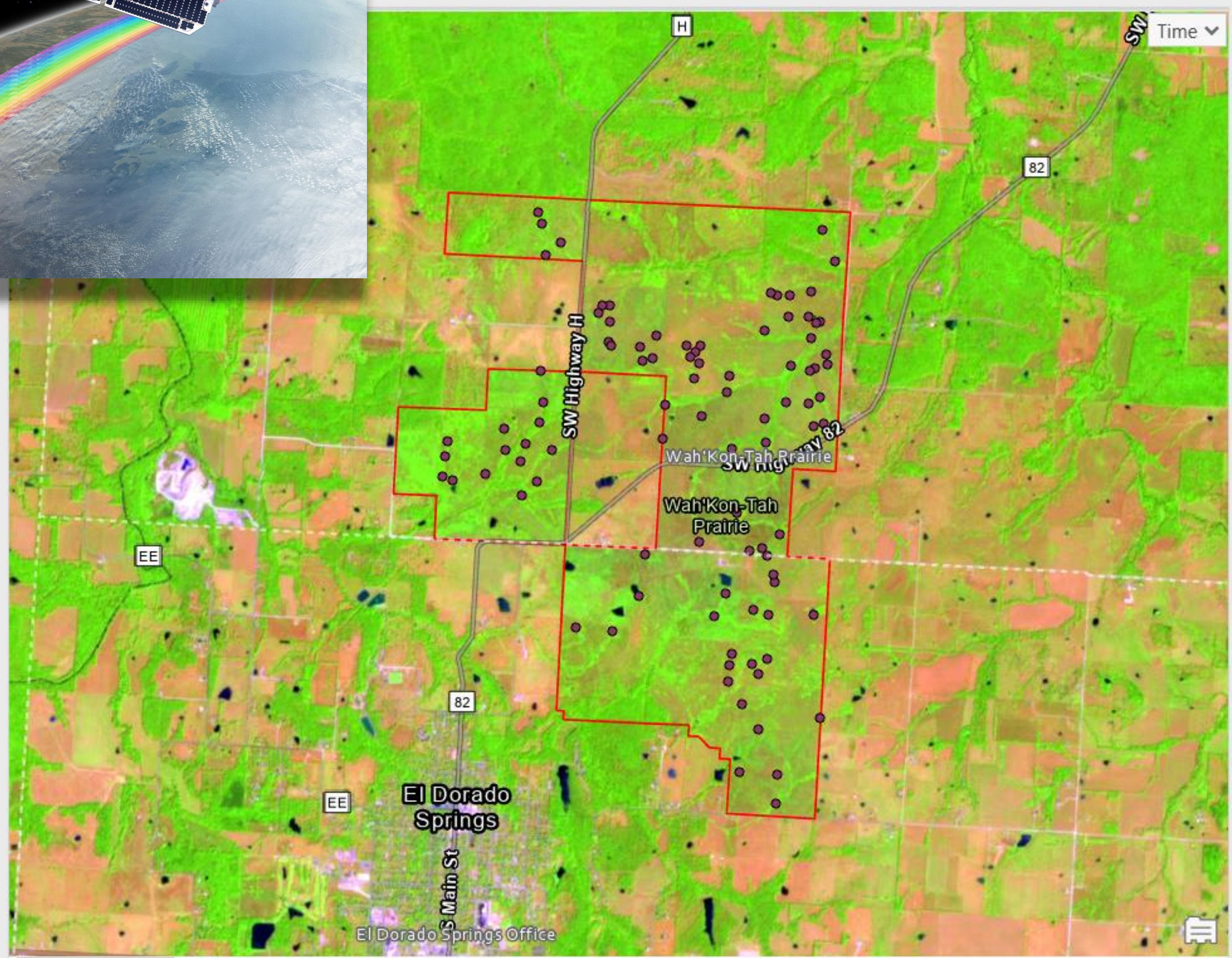
Pairwise Clip Optimized Hot Spot Analysis

Feature Analysis Raster Analysis

Data Engineering Suitability Modeler Visibility Analysis Exploratory 3D Analysis Network Analysis Geostatistical Wizard Business Analysis Data Interop

Raster Functions Function Editor

- MRBO Quail Detections Heatmap
- MRBO Quail Detections
- MRBO Quail Detections
- Priority\_Geographies
- MDC Lands
- Tiered\_Approach\_to\_Natural\_Community\_an
- NAIPCLipWKT
- easements\imagery2021
- USA NAIP Imagery: NDVI
- USA NAIP Imagery: Natural Color
- Soil Depth, root depth of all components
- USA NLCD Land Cover
- Sentinel-2 Views
  - RGB
  - Red: B11\_ShortWaveInfraRed
  - Green: B8\_NearInfraRed
  - Blue: B2\_Blue
- Sentinel layer - April 26, 2022 - Vegetation I
- World Imagery



Catalog

Project Portal Favorites

Living Atlas

Search Results

- Sentinel-2 Views
- Sentinel-2 10m Land Use/Land Cover Time Series
- Sentinel-2 Level-2A
- Sentinel-2 10m Land Use/Land Cover Change from 2018 to 2021
- India: Sentinel-2 10m Land Use/Land Cover 2021
- Tratamiento de imágenes Sentinel-2 con ArcGIS
- Sentinel-5P - Nitrogen Dioxide (NO2)
- European Space Agency WorldCover 2020 Land Cover
- Sentinel-5P - Sulfur Dioxide (SO2)
- Cloud Mask Generation (Sentinel-2)
- Land Cover Classification (Sentinel-2)
- Corine Land Cover 2018
- Human Settlements Classification (Sentinel-2)
- Sentinel-2 10-Meter Land Use/Land Cover Time Series
- Well Pad Detection - Permian Basin
- Solar Photovoltaic Park Classification - C
- Sentinel 5P - Methane (CH4)

Find more items...

Missouri River Bird Observatory logo

NOT SO STANDARD DELIVERABLES

# 5 MP Multispectral Camera

**Near-infrared  
(NIR)**  
860 nm ± 26 nm

**Red (R)**  
650 nm ± 16 nm

**Red edge (RE)**  
730 nm ± 16 nm

**Green (G)**  
560 nm ± 16 nm



**Conservation Drone**

**dji**





Ghostbird



Conservation Drone

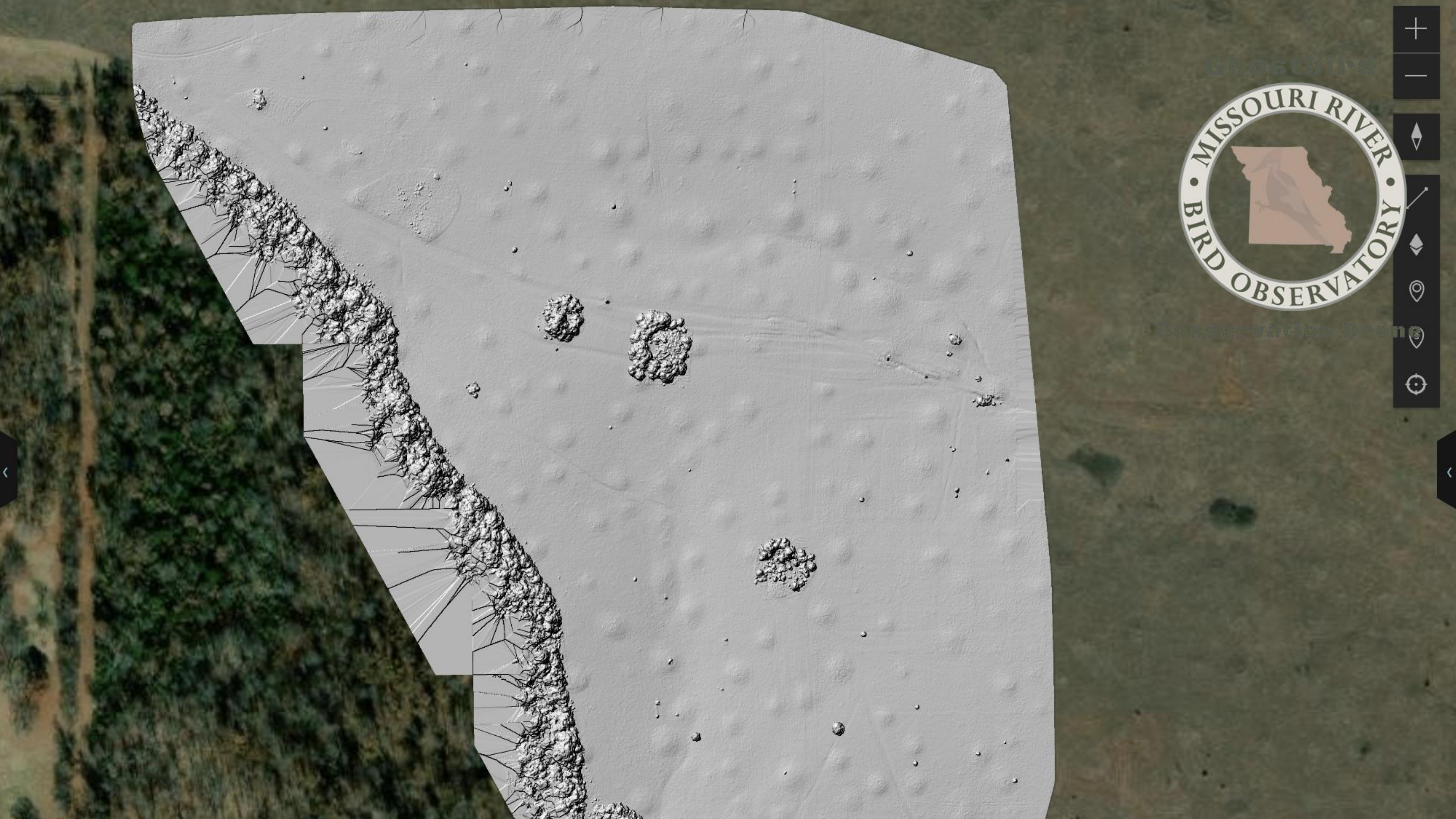


**Ghostbird**



**Conservation Drone**

Sources: Esri, Airbus DS, USGS, NGA, NASA, CGIAR, N Robinson, NCEAS, NLS, OS, NIMA, Geo  
Rijkswaterstaat, GSA, Geoland, FEMA, Intermap and the GIS user community, Esri Community Maps  
Saline County, Missouri Dept. of Conservation, Missouri DNR, © OpenStreetMap, Microsoft,  
Garmin, SafeGraph, GeoTechnologies, Inc. METI/NASA, USGS, EPA, NPS, US Census Bureau, U



Ghostbird

Conservation Drone

Ghostbird



Conservation Biogeography



**Ghostbird**



**Conservation Drone**

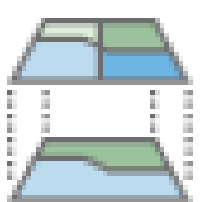
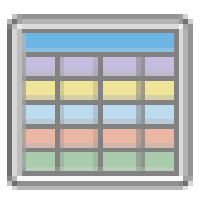
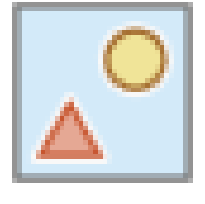
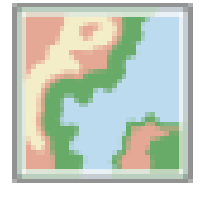
Sources: Esri, Airbus DS, USGS, MGA, NASA, CGIAR, N Robinson, NCEAS, NLS, OS, NMA, Geodatastyreisen, Rijkswaterstaat, GSA, Geoland, FEMA, Intermap and the GIS user community, Esri Community Maps Contributors, Johnson County MO, Missouri Dept. of Conservation, Missouri DNR, © OpenStreetMap, Microsoft, Esri, TomTom, Garmin, SafeGraph, GeoTechnologies, Inc. MET/NASA, USGS, EPA, NPS, US Census Bureau, USDA, USFWS

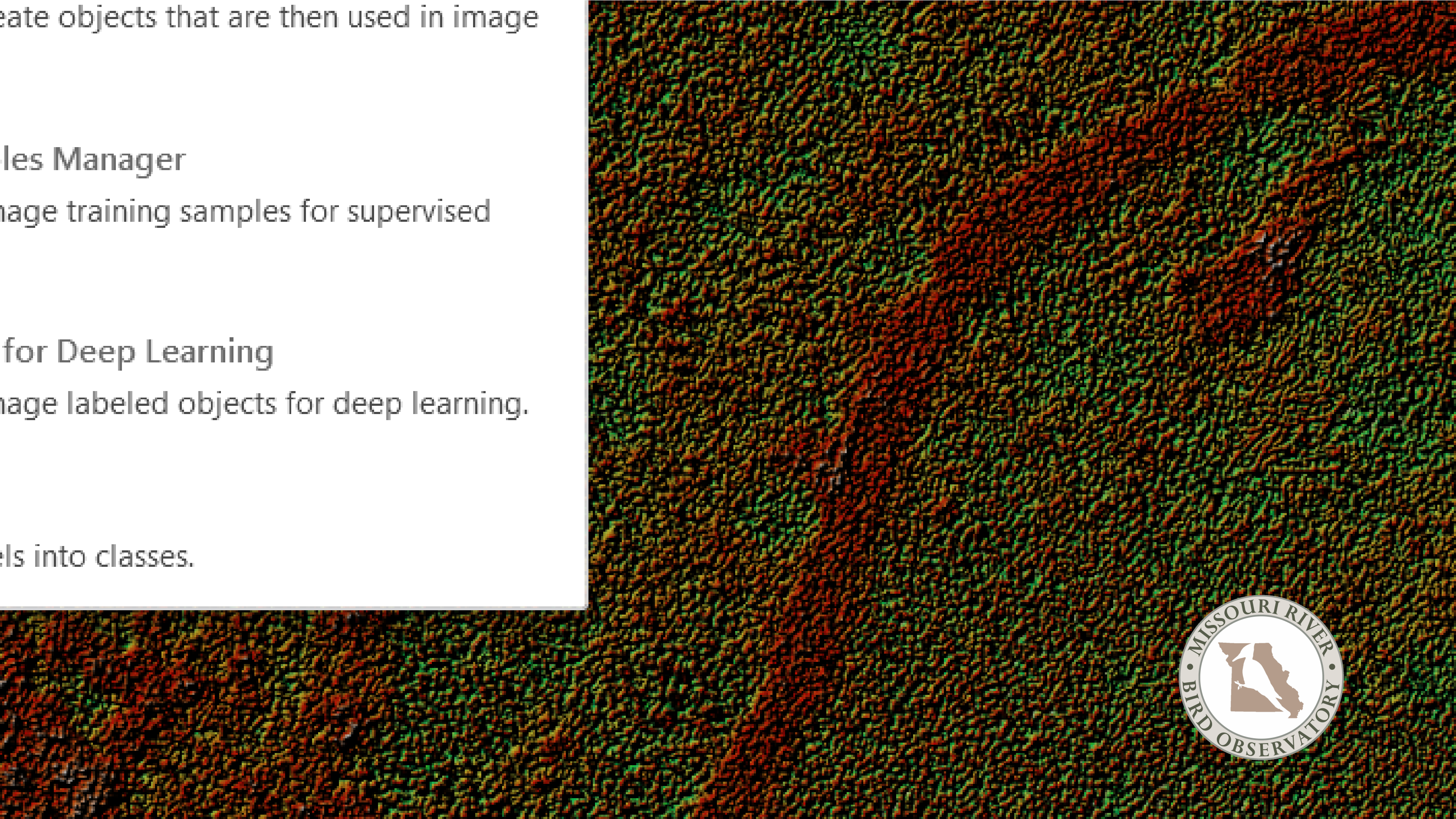
**Contents**

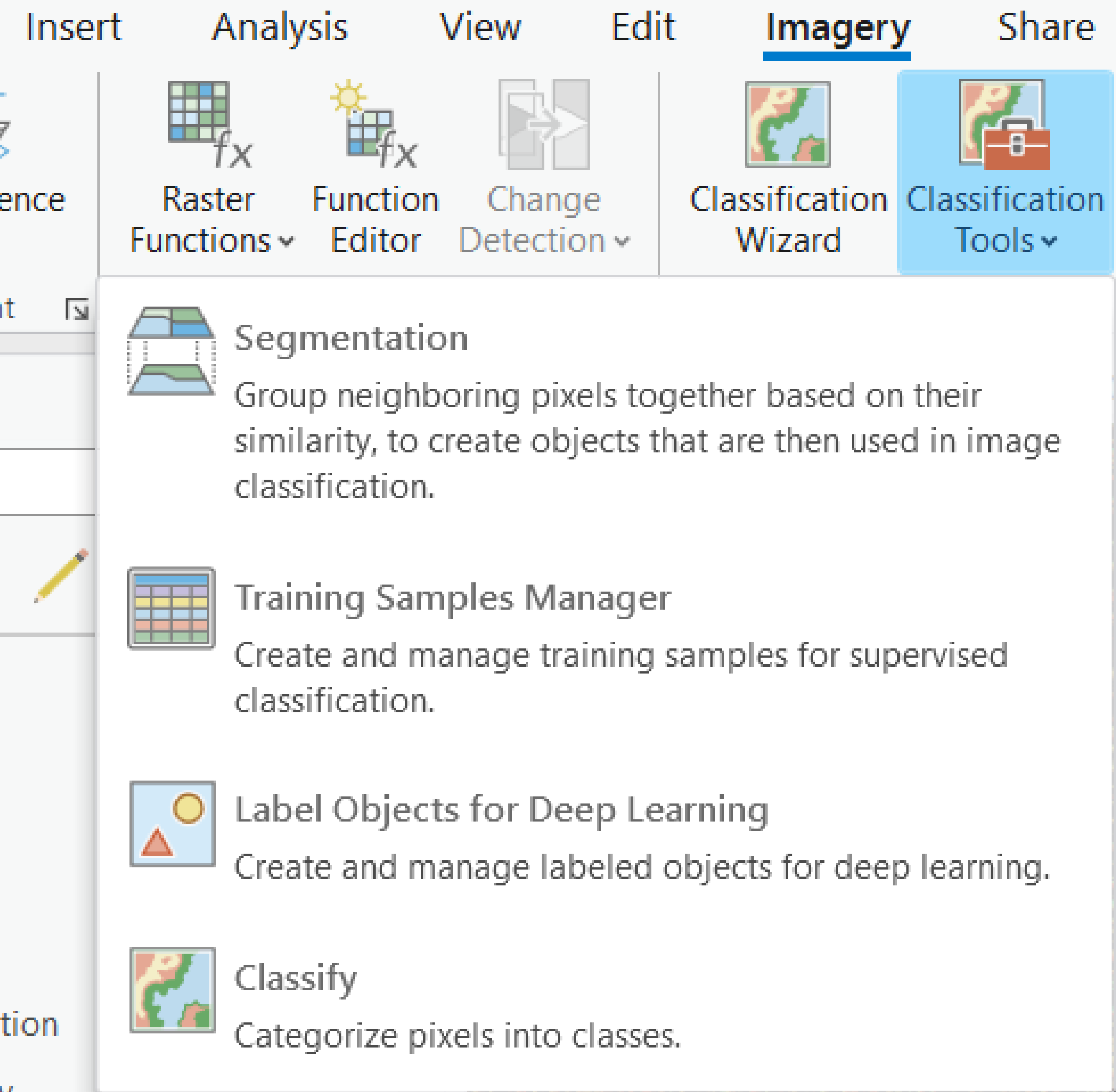
Search

Ortho Mapping

- location
- Flight Path
- path
- Source Data
  - Image Collection
    - Boundary
    - Footprint

-  **Segmentation**  
Group neighboring pixels together based on their similarity, to create objects that are then used in image classification.
-  **Training Samples Manager**  
Create and manage training samples for supervised classification.
-  **Label Objects for Deep Learning**  
Create and manage labeled objects for deep learning.
-  **Classify**  
Categorize pixels into classes.





## USES OF DEEP LEARNING

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- Quantify Native Plant Area
- Quantify Invasive Species Area
- Quantify Pollen Availability



# OTHER NOT SO STANDARD DELIVERABLES

## Remote Sensing in Ecology and Conservation

Open Access

Research Article

 Open Access



## Automating sandhill crane counts from nocturnal aerial imagery using deep learning

Emilio Luz-Ricca , Kyle Landolt, Bradley A. Pickens, Mark Koneff

First published: 18 October 2022 | <https://doi.org/10.1002/rse2.301>

Editor: Temuulen Sankey

Associate Editor: Francesco Rovero

 SECTIONS

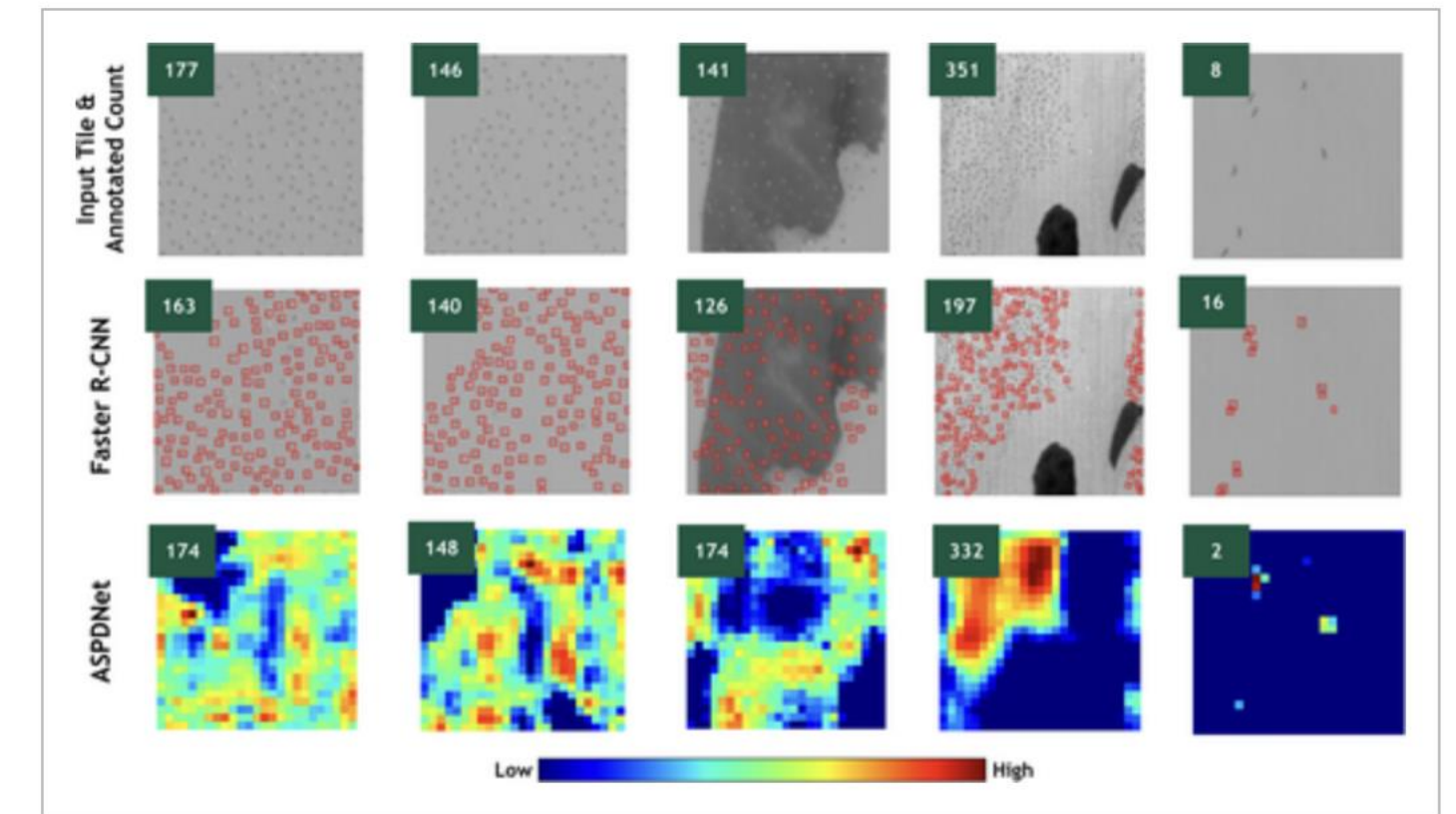


Figure 4

[Open in figure viewer](#) | [Download PowerPoint](#)

A sampling of  $200 \times 200$  pixel input tiles derived from our test imagery with each column also containing the annotated count and predictions for both deep learning models. Counts are included in the top left corner for each cell of the grid. Depicted are both high-quality predictions (predicted count is very close to annotated count) and tiles with high prediction error (large difference between predicted and annotated). In predicted density maps, densities range from low (dark blue) to high (dark red), but the scale differs between tiles.





# APPLICATIONS AND IDEAS

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## ***More Quantifying Stuff:***

- ▶ True Extent of burns and management
- ▶ Cattle impacts on woodies
- ▶ Invasive species
- ▶ Natives Species
- ▶ Pollen availability
- ▶ Elevation profiles (i.e. beaver dam and catchment holding capacity)
- ▶ Erosion
- ▶ Crawfish burrows
- ▶ Carbon Sequestration
- ▶ Stream Temperatures



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Rangeland Ecology & Management  
 ELSEVIER Volume 72, Issue 6, November 2019, Pages 916-922

# Diagnostic Feed Values of Natural Grasslands Based on Multispectral Images Acquired by Small Unmanned Aerial Vehicle ☆

Rui Gao<sup>a</sup>, Qingming Kong<sup>a</sup>, Hongguang Wang<sup>b</sup>, Zhongbin Su<sup>a</sup> ✉

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<https://doi.org/10.1016/j.rama.2019.06.005> Get rights and content

## Abstract

Grasslands are the largest renewable source of terrestrial chlorophytes. Furthermore, grasslands can be both fiber sources and the primary metabolizable energy source for ruminants. Therefore, rapid, accurate, and large-scale monitoring of grassland ecosystems is important to provide spatial information on forage quality control and rangeland management. In this experiment, 100 grassland sites were

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Journals / Drones / Volume 6 / Issue 3 / 10.3390/drones6030073

Open Access Editor's Choice Article

### Quantification of Grassland Biomass and Nitrogen Content through UAV Hyperspectral Imagery—Active Sample Selection for Model Transfer

by Marston H. D. Franceschini<sup>1,2</sup>, Rolf Becker<sup>3</sup>, Florian Wichern<sup>2,\*</sup>, Lammert Kooistra<sup>1,\*</sup>

- Laboratory of Geo-Information Science and Remote Sensing, Wageningen University and Research, P.O. Box 47, 6700 AA Wageningen, The Netherlands
- Faculty of Life Sciences, Rhine-Waal University of Applied Sciences, Marie-Curie-Str. 1, 47533 Kleve, Germany
- Faculty of Communication and Environment, Rhine-Waal University of Applied Sciences, 47475 Kamp-Lintfort, Germany

\* Authors to whom correspondence should be addressed.

Academic Editor: Diego González-Aguilera

*Drones* **2022**, *6*(3), 73; <https://doi.org/10.3390/drones6030073>

Received: 20 January 2022 / Revised: 2 March 2022 / Accepted: 7 March 2022 / Published: 11 March 2022

(This article belongs to the Section *Drones in Agriculture and Forestry*)

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## Abstract

Accurate retrieval of grassland traits is important to support management of pasture production and phenotyping studies. In general, conventional methods used to measure forage yield and quality rely on costly destructive sampling and laboratory analysis, which is often not viable in practical applications. Optical imaging systems carried as payload in Unmanned Aerial Vehicles (UAV) platforms have increasingly been proposed as alternative non-destructive solutions for crop characterization and monitoring. The vegetation spectral response in the visible and near-infrared wavelengths provides information on many aspects of its composition and structure. Combining

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Volume 54, 2021 - Issue sup1: Special Issue: Digital Earth Observation

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Review Article

## An open source workflow for weed mapping in native grassland using unmanned aerial vehicle: using *Rumex obtusifolius* as a case study

Olee Hoi Ying Lam, Marcel Dogotari, Moritz Prüm, Hemang Narendra Vithlani, Corinna Roers, Bethany Melville, ... show all

Pages 71-88 | Received 31 Oct 2019, Accepted 05 Jul 2020, Published online: 04 Aug 2020

Download citation <https://doi.org/10.1080/22797254.2020.1793687> Check for updates

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## ABSTRACT

Weed control is one of the biggest challenges in organic farms or nature reserve areas where mass spraying is prohibited. Recent advancements in remote sensing and airborne technologies provide a fast and efficient means to support environmental monitoring and management, allowing early detection of invasive species. However, in order to perform weed classification, current studies mostly relied on object-based image analysis (OBIA) and proprietary software which required substantial human inputs. This paper proposes an open source workflow for automated weed mapping using a commercially available unmanned aerial vehicle (UAV). The UAV was flown at a low altitude between 10 m and 20 m, and collected RGB imagery



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55 Results



## Car Detection - USA

Deep Learning Package By [esri\\_analytics](#)

Deep learning model to detect cars in high resolution imagery.

Authoritative



## Car Detection - New Zealand

Deep Learning Package By [eaglegis](#)

New Zealand car detection Deep Learning Package to detect cars in high resolution imagery.

Authoritative

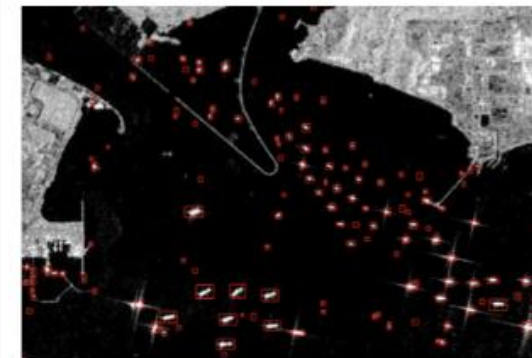


## Tree Detection

Deep Learning Package By [esri\\_analytics](#)

Deep learning model to detect trees in high resolution imagery.

Authoritative



## Ship Detection (SAR)

Deep Learning Package By [esri\\_analytics](#)

Deep learning model to detect ships from Sentinel-1 C band SAR data.

Authoritative

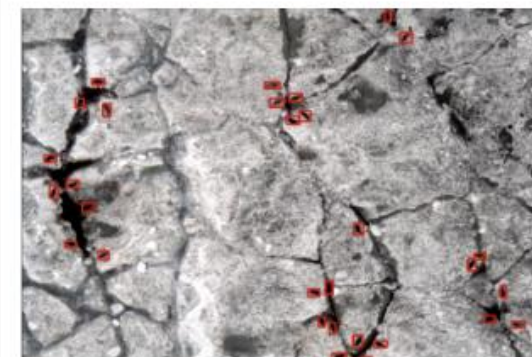


## Pool Detection - USA

Deep Learning Package By [esri\\_analytics](#)

Deep learning model to detect swimming pools in high-resolution aerial or satellite imagery.

Authoritative



## Arctic Seal Detection

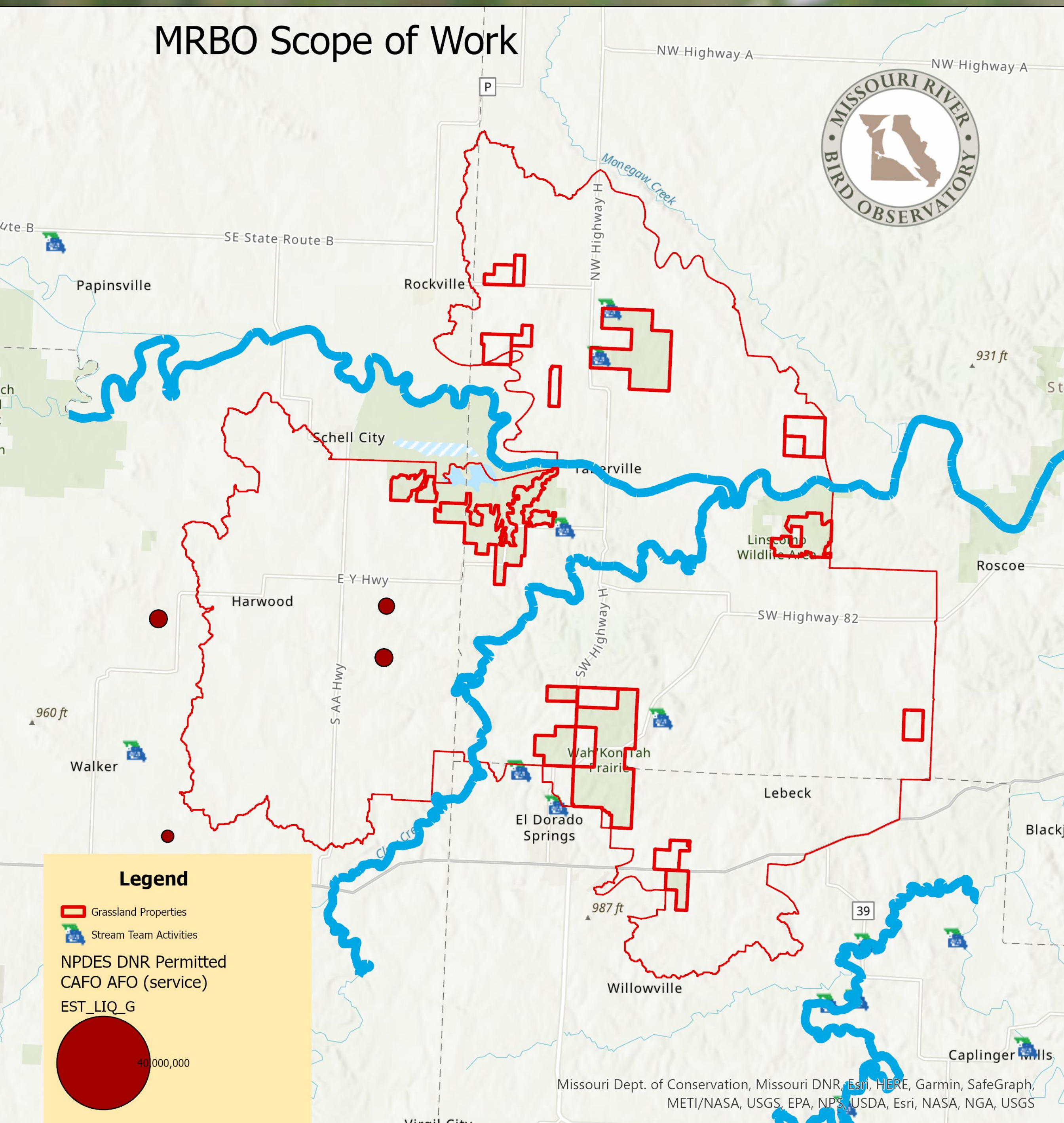
Deep Learning Package By [esri\\_analytics](#)

Deep learning model to detect Arctic seals using drone imagery.

Authoritative

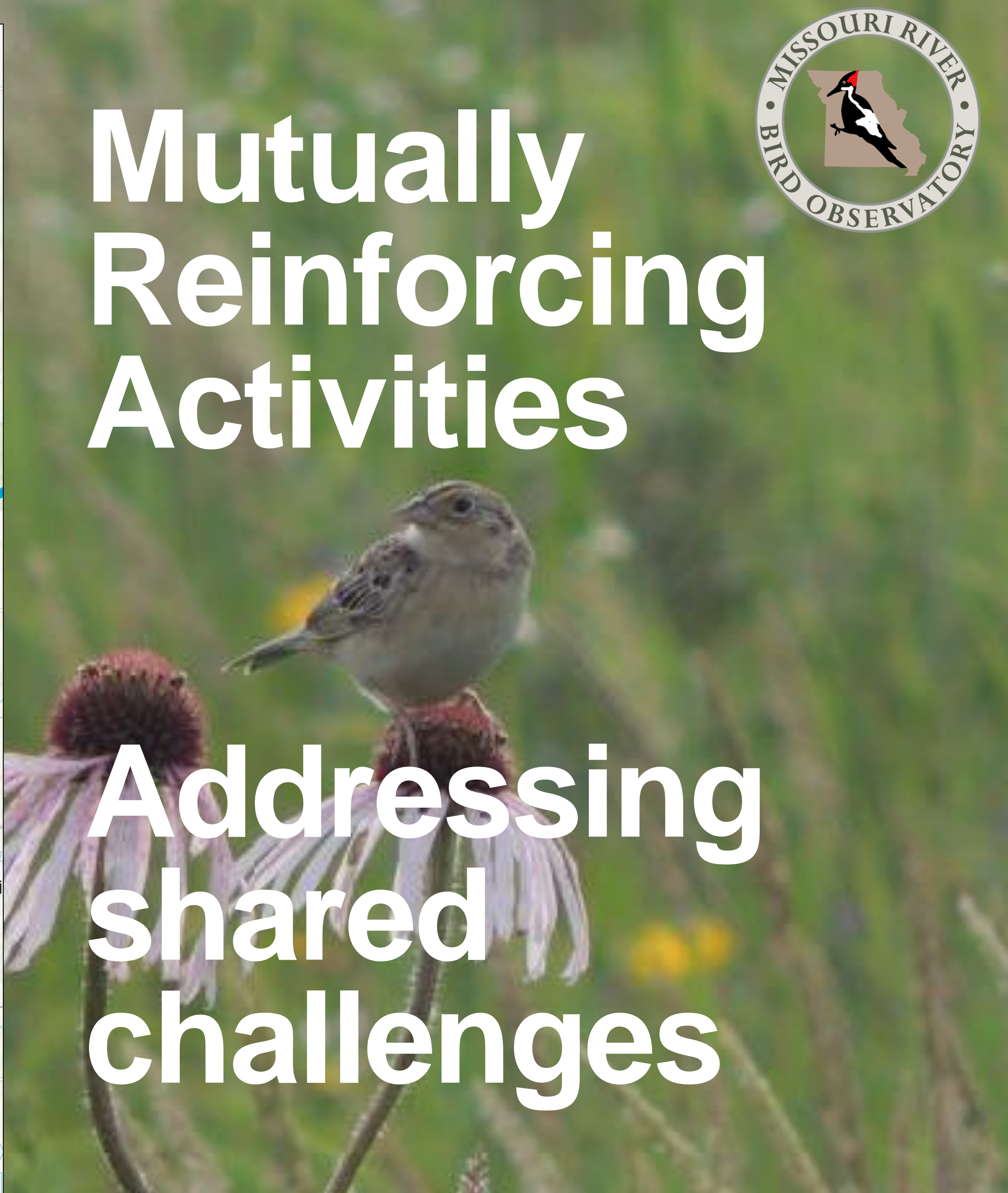


# MRBO Scope of Work



# Mutually Reinforcing Activities

# Addressing shared challenges



IDEAS?

QUESTIONS?

Thank you

Ethan.Duke@mrbo.org

