## UPDATING THE MINNESOTA NATIONAL WETLAND INVENTORY

An Integrated Approach Using Object-Oriented Image Analysis,

Human Air-Photo Interpretation and Machine Learning

# ANALYTICS INC

#### AARON SMITH EQUINOX ANALYTICS INC.

FUNDING PROVIDED BY THE MINNESOTA ENVIRONMENT AND NATURAL RESOURCES TRUST FUND



#### ENVIRONMENT AND NATURAL RESOURCES TRUST FUND

AS RECOMMENDED BY THE LEGISLATIVE-CITIZEN COMMISSION ON MINNESOTA RESOURCES (LCCMR)

THE TRUST FUND IS A PERMANENT FUND CONSTITUTIONALLY ESTABLISHED BY THE CITIZENS OF MINNESOTA TO ASSIST IN THE PROTECTION, CONSERVATION, PRESERVATION, AND ENHANCEMENT OF THE STATE'S AIR, WATER, LAND, FISH, WILDLIFE, AND OTHER NATURAL RESOURCES.

#### ROBB MACLEOD DUCKS UNLIMITED INC.

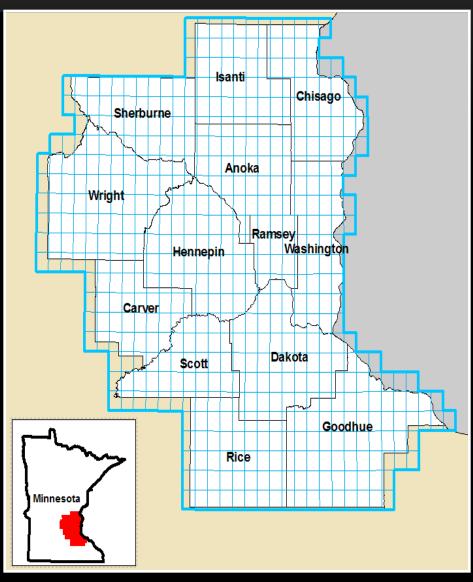
STEVE KLOIBER MINNESOTA DEPARTMENT OF NATURAL RESOURCES

DUCKS UNLIMITED

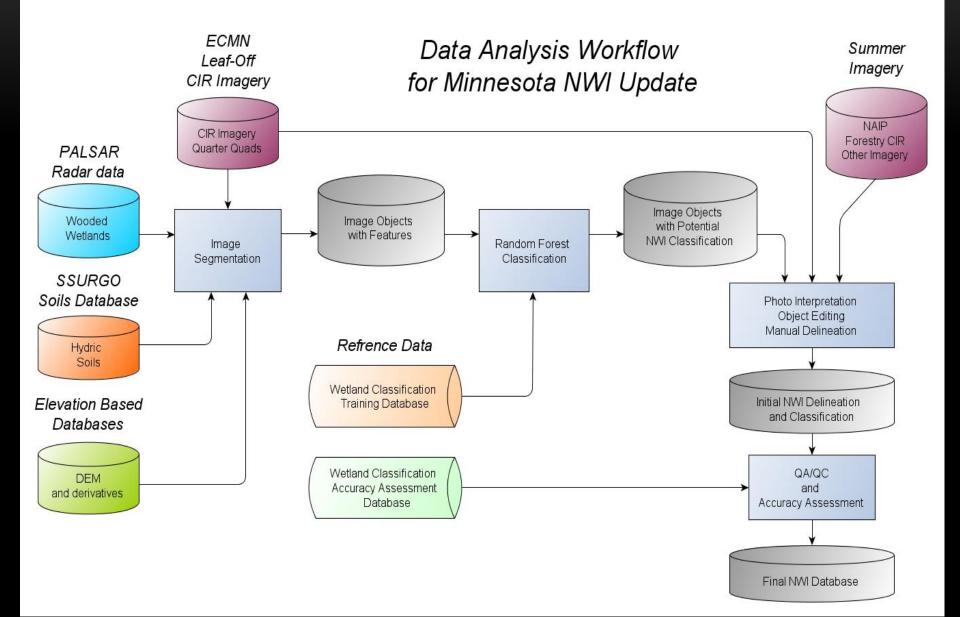


# EAST CENTRAL MN NWI UPDATE PHASE II

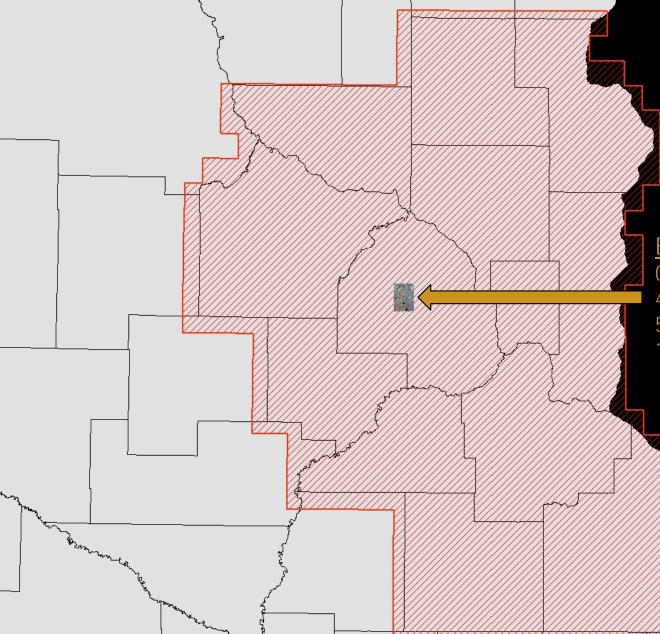
- MN NWI update program was initiated in 2008 by MN DNR
- Funding provided by: The Minnesota Environment and Natural Resources Trust Fund
- Production mapping for East Central Region (Phase II) began in 2010
- Expected completion of the state in 2019 (5 project phases)
- Project specifications called for integration of digital image analysis and traditional photo-interpretation
  - Human Photo Interpretation (PI)
  - Object Based Image Analysis
  - Seamless integration of LiDAR, RADAR, SSURGO, field reference data.



#### NWI UPDATE MAPPING PROCESS



#### MINNESOTA NWI UPDATE PHASE II PROJECT



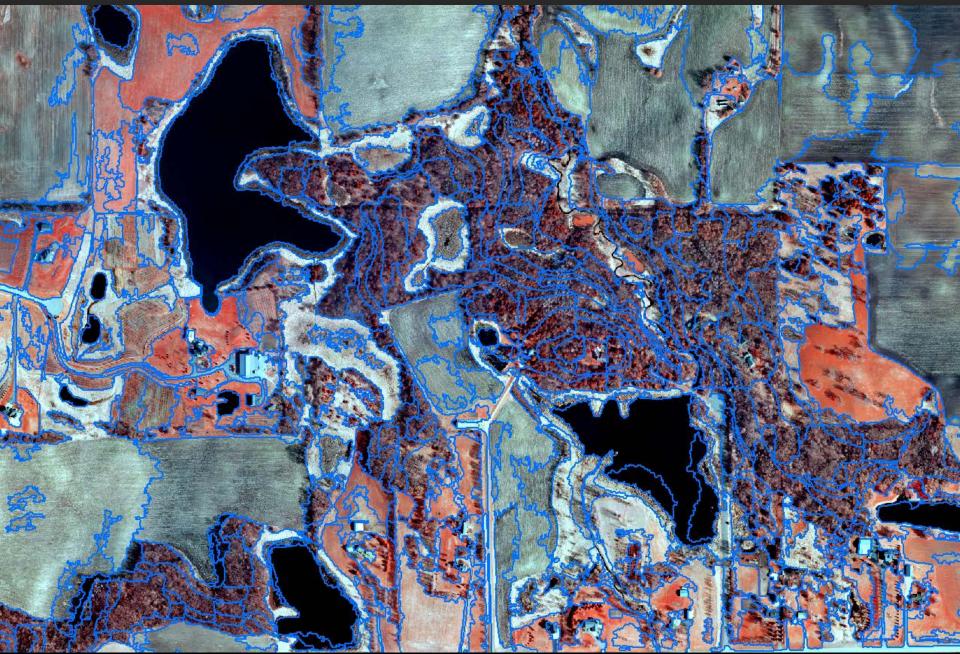
Project Location Twin Cities Area 13 Counties 7149 sq. mi.

Base Imagery (Spring Leaf-off) 0.3 to 0.5m pixel resolution 4 band (R,G,B, CIR) 548 Total Images (1-2 gb ea.) 1.2 TB of Imagery & related data

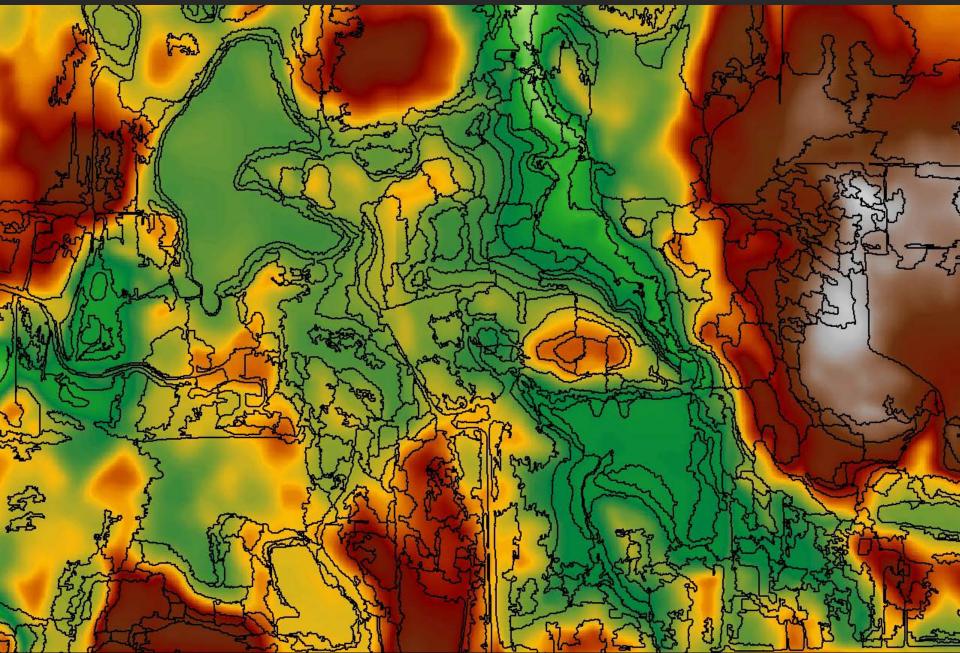
### FALSE COLOR IR SPRING IMAGERY ~1:2000 SCALE



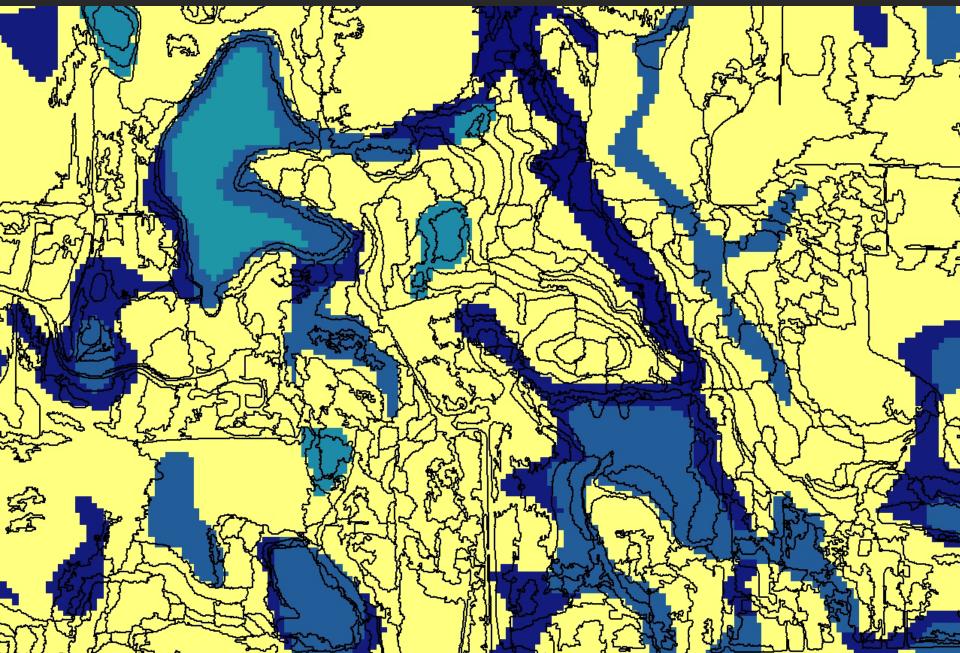
#### SPRING IMAGERY AT 1:5000 SCALE



#### LIDAR DERIVED DEM

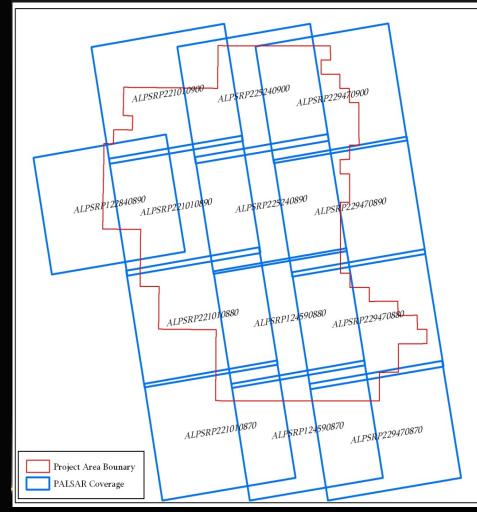


#### SSURGO HYDRIC SOILS



### ALOS PALSAR SATELLITE RADAR

- PALSAR archival data provided by the Alaska Satellite Facility
- PALSAR processing done by Ducks Unlimited Inc.
  - Speckle reduction
  - Unsupervised clustering
  - Manual identification of clusters associated with wooded wetlands (trees + surface water = double bounce)
  - These clusters also were also associated with parking lots (cars + pavement = double bounce)
  - OBIA context information is used to constrain application of PALSAR to forested areas



#### PALSAR WOODED WETLANDS

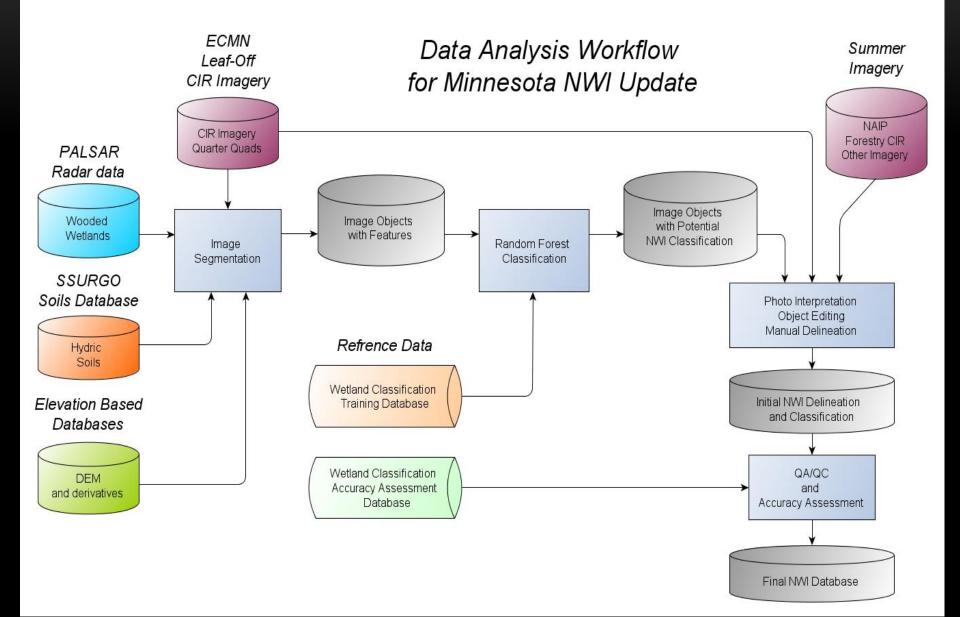


#### FIELD DATA COLLECTION

- Field data collection for classification reference
  - 1,346 polygons in total were selected for field verification
  - Used to create photo guide for training PI team
  - Used to train RandomForest algorithm
- Field data collection for accuracy assessment
  - Independent data collection by the University of Minnesota
  - Production team never sees this dataset



#### NWI UPDATE MAPPING PROCESS



#### SEGMENTATIONS DELIVERED TO INTERPRETERS

Wooded Wetlands identified with PALSAR

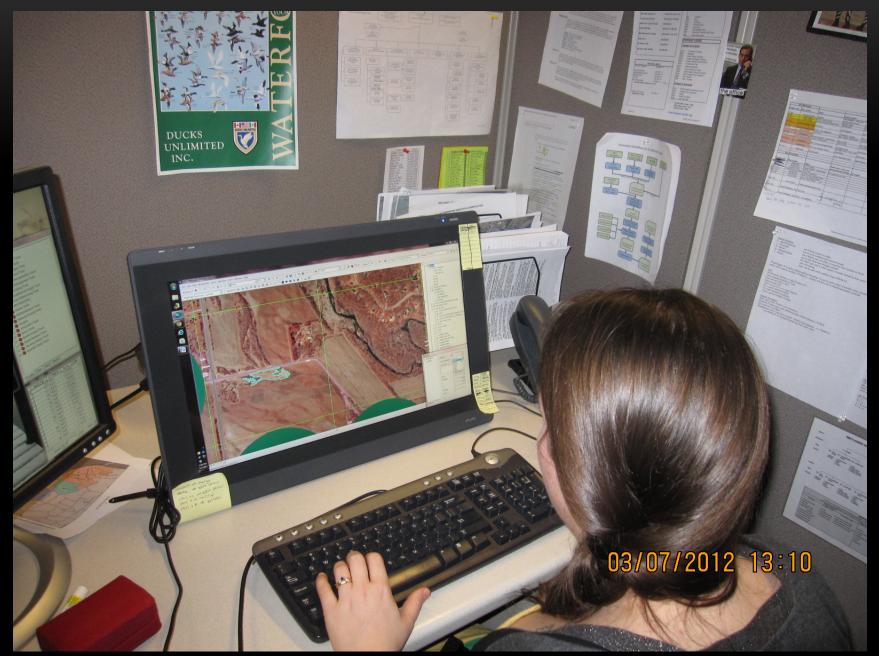
Stream bed wetlands boundary derived from hydrologic flow modeling and LiDAR DEM

Topographic contour lines in forested areas derived from LiDAR DEM

> L1/L2 (6' depth) boundary derived from in-situ sonar bathymetry

Forest / non-forest boundary based on spectral (NDVI) information and image object texture

#### PHOTO INTERPRETATION USING SEGMENTATION



#### SELECTING VS DIGITIZING

#### Extracted feature boundary





Hand drawn boundary

#### Time to complete: 6 seconds

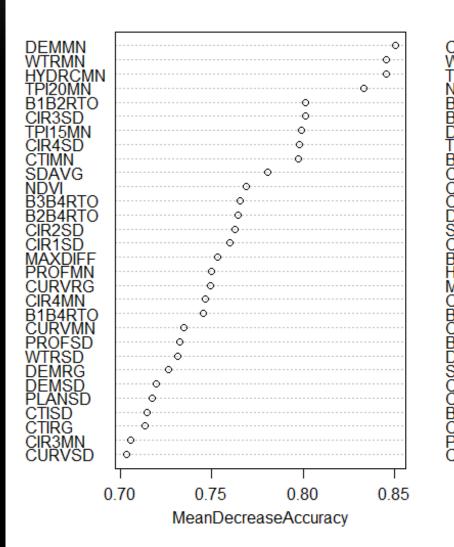
#### Time to complete: 71 seconds

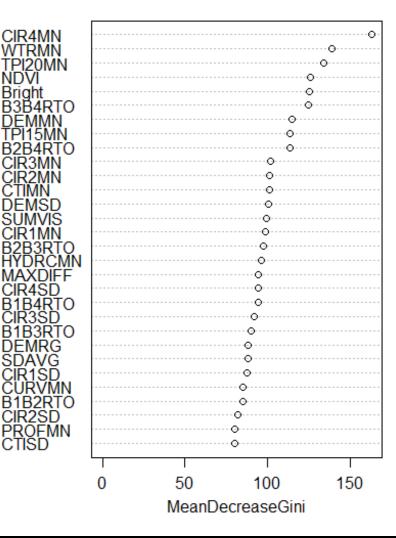
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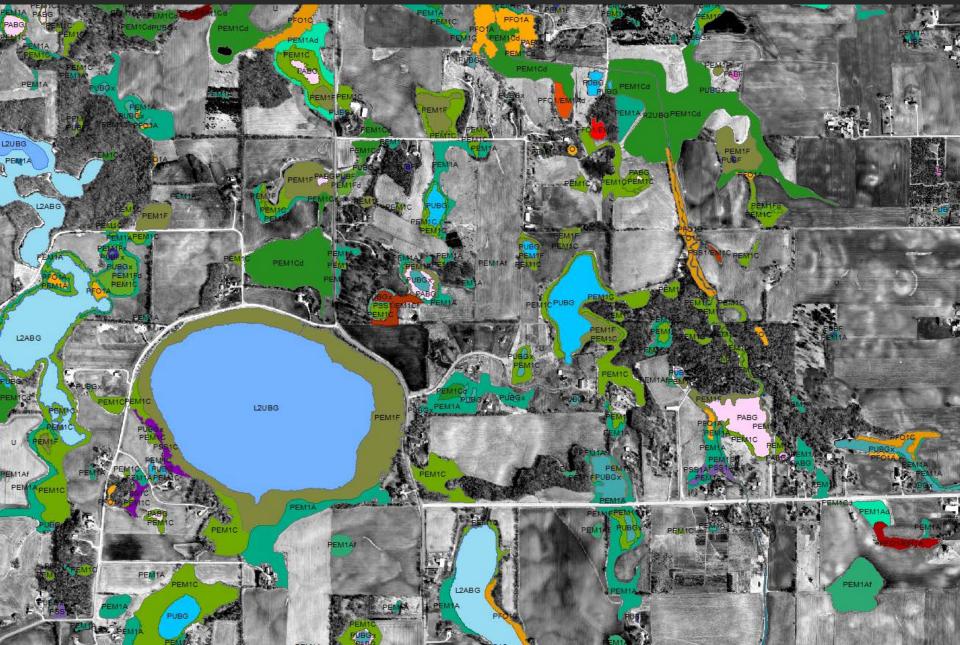
#### RANDOM FOREST CLASSIFICATION

#### **RF.RFWR**





#### NATIONAL WETLAND INVENTORY



#### SUCCESSFUL OBIA / HUMAN INTEGRATION:

- Rule #1 Automated processes do not replace people
  - When difficult decisions need to be made, humans have the advantage in intelligence, judgment, prior experience, subject matter expertise, and in our ability to communicate with each other. A skilled person will beat any affordable program-based alternative
- Rule #2 Algorithms: not as smart as people, but they don't get bored or tired
  - See Rule #1; When you need to make lots & lots of simple decisions, it is good to engage the services of a computer
- Rule #3 Humans are not ideally suited for mindless repetitive tasks
  - People get bored when they spend hundreds of hours on a simple task that doesn't require them to exercise their intelligence, judgment, subject matter expertise or the ability to communicate with each other
- Rule #4 Human time is expensive and computer time is cheap