

# Automatic Imputation for an Area Survey

Tara Murphy<sup>1</sup>, Arthur Rosales<sup>1</sup>, Luca Sartore<sup>1,2</sup>, Denise Abreu<sup>1</sup>

<sup>1</sup>National Agricultural Statistics Service, <sup>2</sup>National Institute of Statistical Sciences



**United States Department of Agriculture**  
National Agricultural Statistics Service



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# Background: June Area Survey (JAS)

- United States Department of Agriculture (USDA) National Agricultural Statistics Service's (NASS) largest annual survey
- Provides key indications for many agricultural aspects, including:
  - Planted acreage for most row crops and small grains
- Measures the incompleteness of the NASS List Frame



# Background: June Area Survey (JAS)



- Area-frame based
- Segments of land sampled and remain in survey for five years
- Sampled segments divided into tracts representing unique land operating arrangements

# Unique Nonresponse Challenges for JAS Tracts

- Data collection based mostly on in-person and telephone interviews
- Extensive screening activities are needed to identify in-scope land tracts, especially for new segments



# Unique Nonresponse Challenges for JAS Tracts

- Land-use arrangements may change during the five-year sample period
- **Digital records of tract boundaries** have historically never been created, making it difficult to link external, ancillary data to those tracts



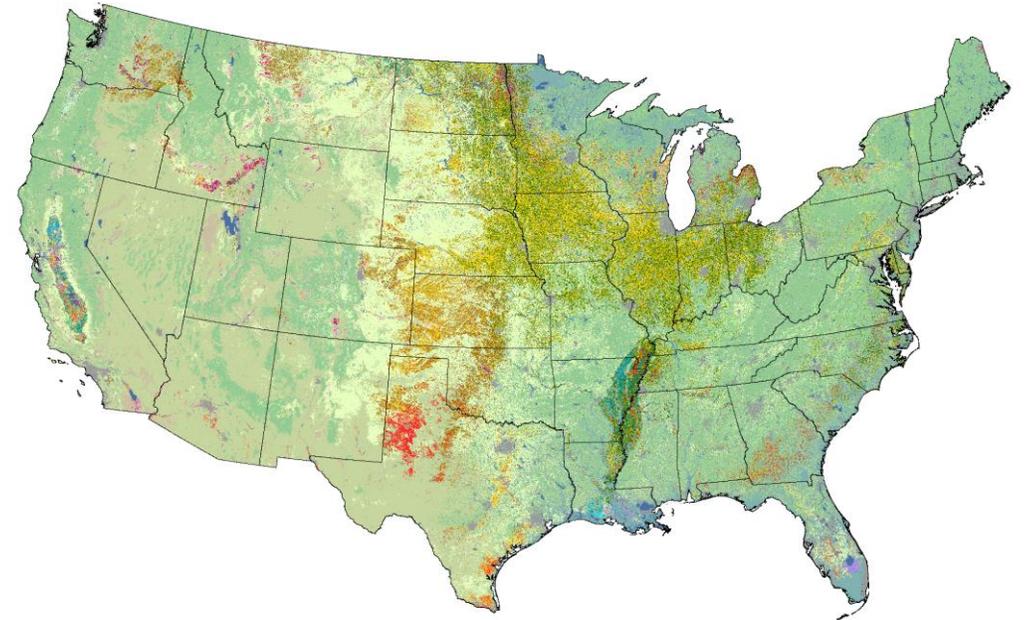
# Availability of New Data

- Beginning 2021, digitization of all in-sample tract boundaries performed
- Resulting in **geospatially-referenced** record of tracts
- Allowing them to be linked to other data for estimation and imputation



# Auxiliary Data: Cropland Data Layer (CDL)

- Crop-specific land cover classification product created by USDA NASS
  - Raster product at 30-meter resolution
  - Available for the conterminous U.S. annually since 2008
  - Pixel-level crop data for over 100 crop categories
  - **Only available at the end of the year**



# Auxiliary Data: Farm Service Agency (FSA)

- FSA-578 Form
  - Available for all land associated with a USDA program in a calendar year
  - Provides crop information (what producers are growing and where)
- FSA Common Land Units (CLUs)
  - Smallest unit of land that has a permanent contiguous boundary, common land cover and land management, common owner, & common producer
  - **CLUs linked to corresponding FSA-578 data**



[https://www.agridatainc.com/Home/Products/Mapping%20Features/Land%20Resource%20Intelligence/FSA%20Field%20Boundaries%20\(CLU\)](https://www.agridatainc.com/Home/Products/Mapping%20Features/Land%20Resource%20Intelligence/FSA%20Field%20Boundaries%20(CLU))

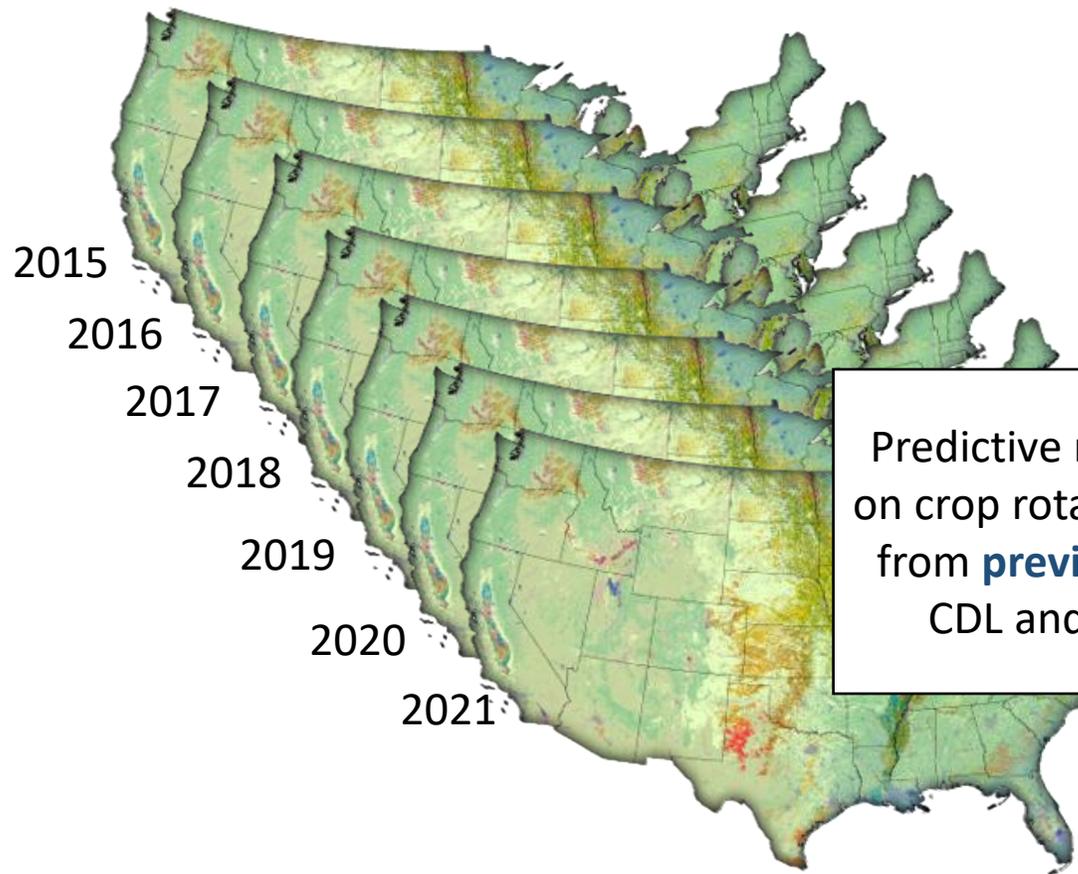


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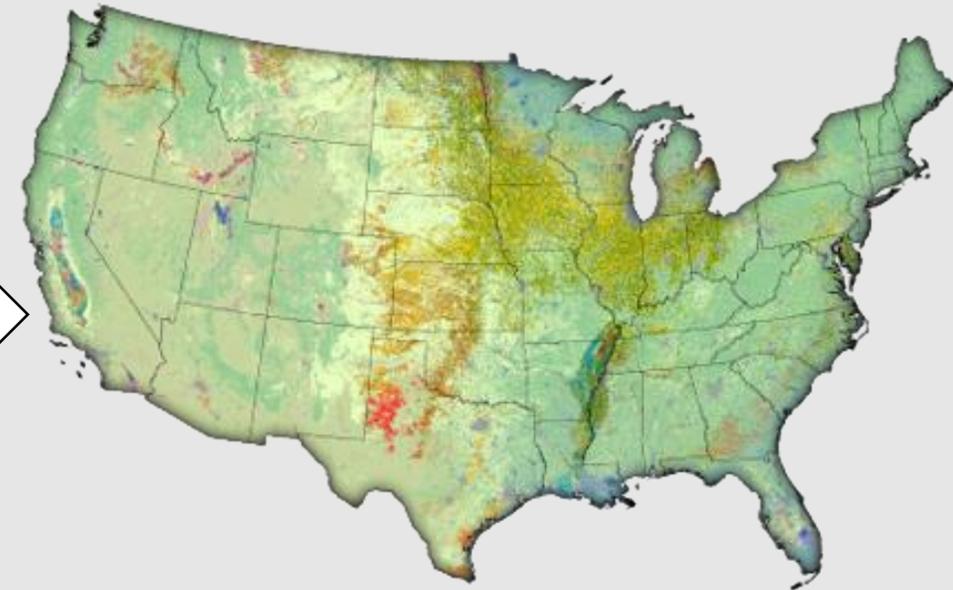


# New Auxiliary Data: Predictive CDL (PCDL)

- Offers probabilistic land-cover classification
- Allow for an in-season estimate of land-cover



Predictive model based on crop rotation patterns from **previous years** of CDL and FSA data



2022 PCDL

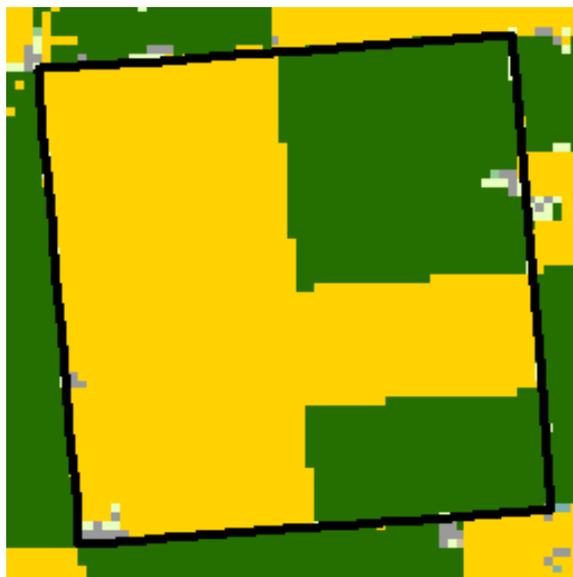
# Entropy Layer of PCDL

- Designed to provide a sense of confidence in the PCDL for the area of interest
- Low entropy = High predictability
- High entropy = Low predictability



# Entropy Layer of PCDL

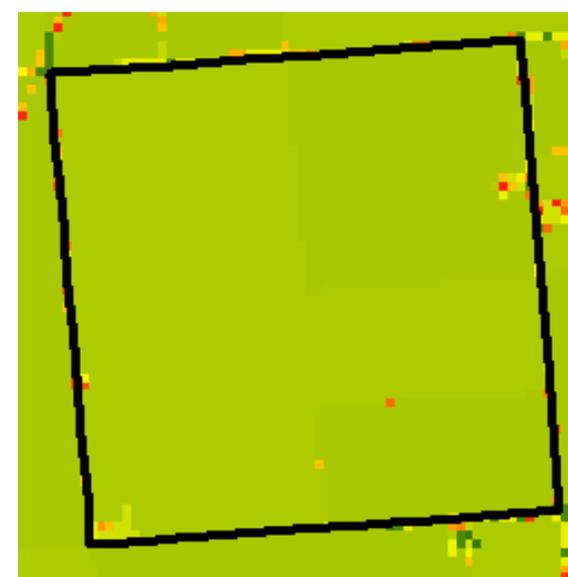
 Corn  
 Soybeans



FSA CLUs linked to 578  
data & Official CDL  
"Truth"



Predictive CDL

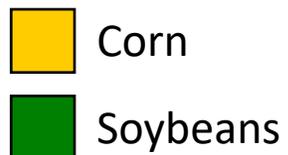


Entropy Layer  
Low Entropy

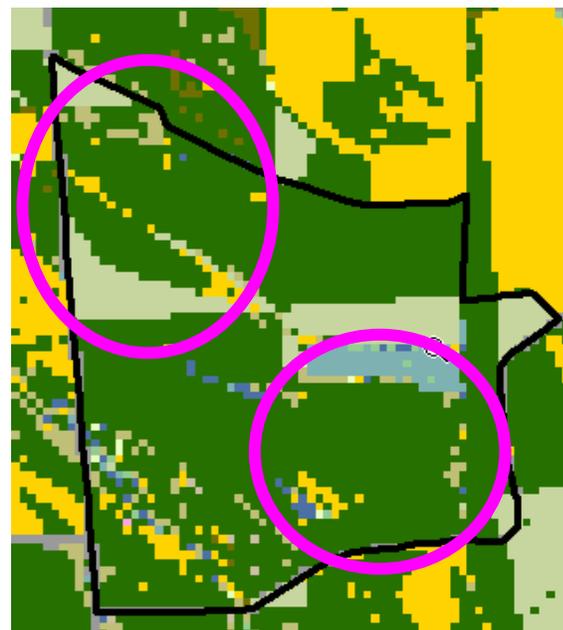
**Entropy**  
 High  
Low



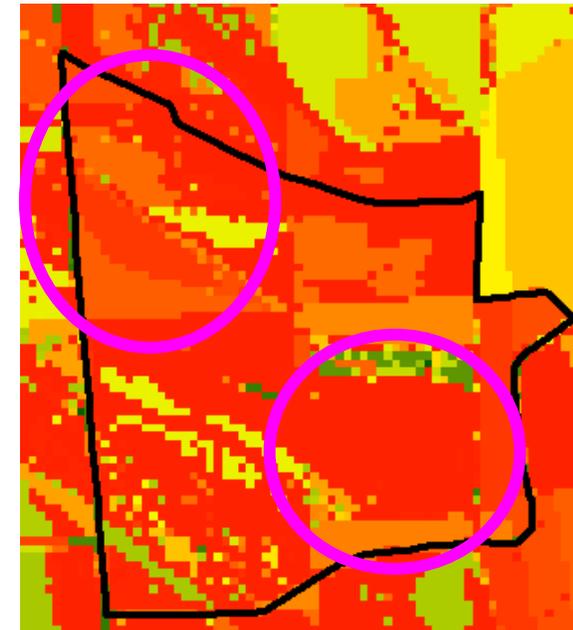
# Entropy Layer of PCDL



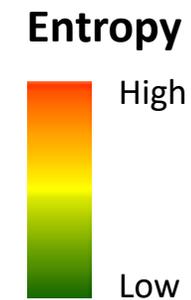
FSA CLUs linked to 578  
data & Official CDL  
"Truth"



Predictive CDL



Entropy Layer  
High Entropy



# Research Question

- Can **automatic imputation** be performed at the tract-level by incorporating digitized tracts, the PCDL and the Entropy Layer?



# Data Preparation

- 2019 & 2021 JAS survey data utilized
- PCDL and FSA data summarized within digitized tract boundaries and linked to respective JAS survey data
- Data was subset to “low hanging fruit” records:
  - Number of PCDL crops  $< 2$
  - Mean entropy of tract  $< 0.1$  (Sartore, et al., 2022)
  - Digitized tract acres between 10 and 1000 acres



# Imputation Model

- Cubist model
  - Used to predict FSA crop acreage at the crop level, based on PCDL, entropy, and other covariates (e.g., lat/long, state, sampling stratum)
  - Implemented using *caret cubist* packages in R software
- Model fit on 2019 data
- Predictions made on 2021 data



# Results: Corn

- Color coded by JAS response:

- 0 = manually estimated
- 1 = reported

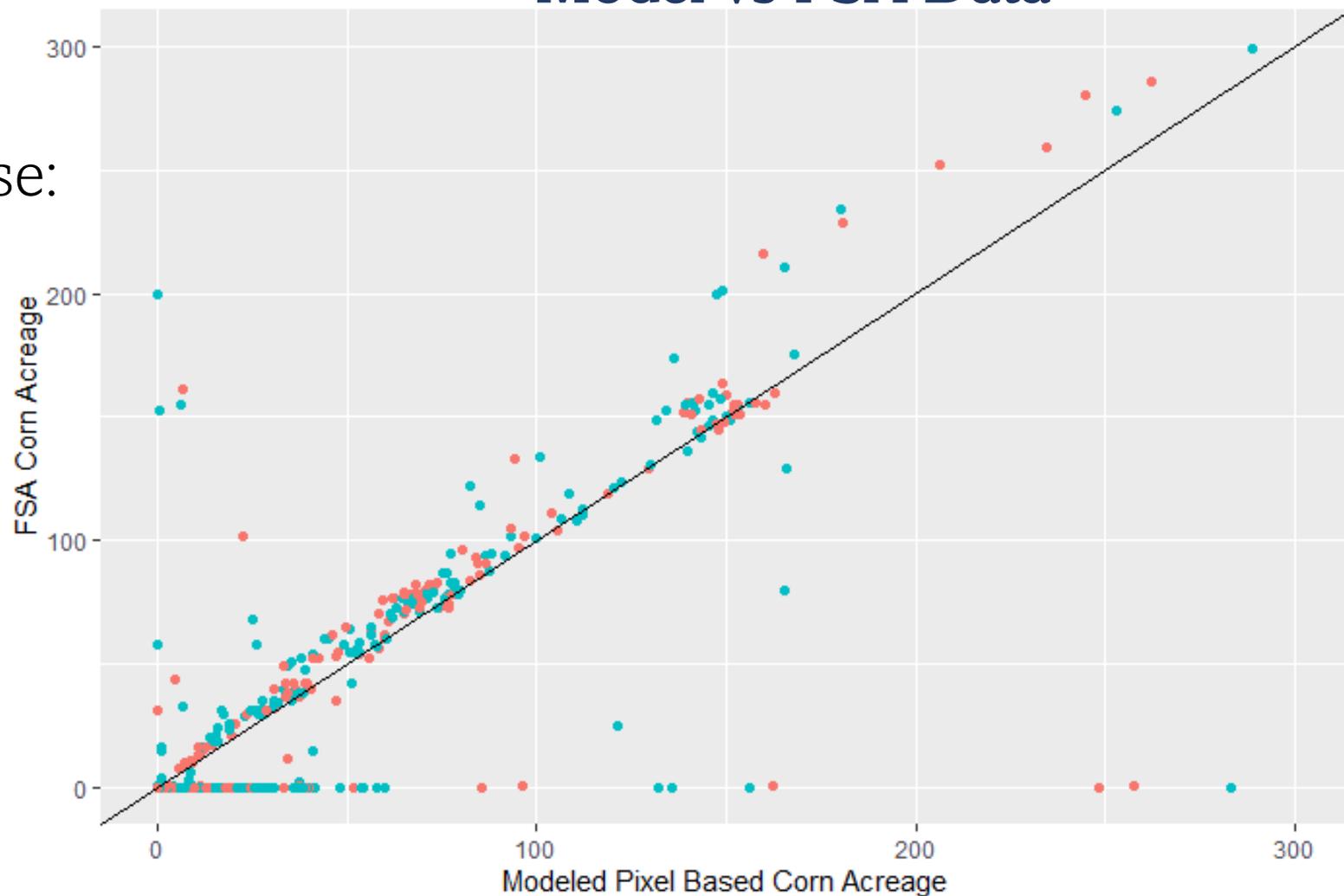
- $R^2 = 0.781$

- MAE = 4.783 acres

- Important model variable:

- PCDL Corn

## Model vs FSA Data



# Results: Corn

- Color coded by JAS response:

- 0 = manually estimated
- 1 = reported

- $R^2 = 0.807$

- MAE = 2.797 acres

## PCDL vs FSA Data



# Results: Soybeans

- Color coded by JAS response:

- 0 = manually estimated
- 1 = reported

- $R^2 = 0.86$

- MAE = 2.82 acres

- Important model variables:

- PCDL combined soy, PCDL soy, digitized tract acres, state, entropy mean, entropy median, latitude, longitude

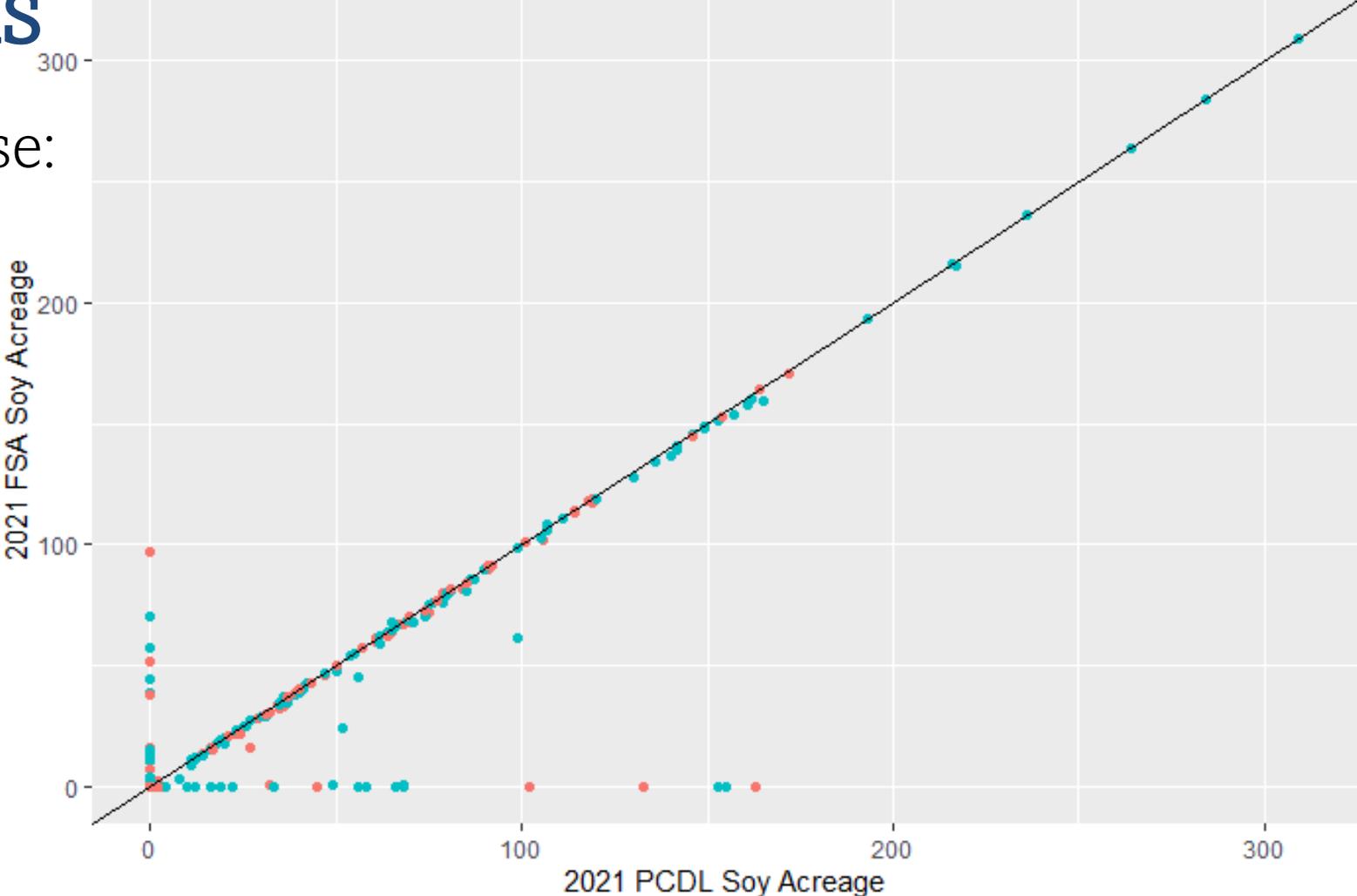
## Model vs FSA Data



# PCDL vs FSA Data

# Results: Soybeans

- Color coded by JAS response:
  - 0 = manually estimated
  - 1 = reported
- $R^2 = 0.884$
- MAE = 1.336 acres



# Discussion

- PCDL outperforms imputation model for “low hanging fruit”
  - Automatic imputation can easily be performed
- However, “low hanging fruit” represents a small portion of records



# Future Research

- Expand beyond “low hanging fruit” records
- Find optimal level of entropy where PCDL is useful for the purposes of imputing JAS tract nonresponse
- Improve imputation model by incorporating additional auxiliary data
  - Economic data
  - Environmental data



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# Thank you!

Tara.Murphy@usda.gov



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