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National Agricultural Statistics Service  
Research and Development Division  
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# **Cropland Classification Accuracy as a Function of Training Data Accuracy**

**Association of American Geographers, 2010 Annual Meeting**

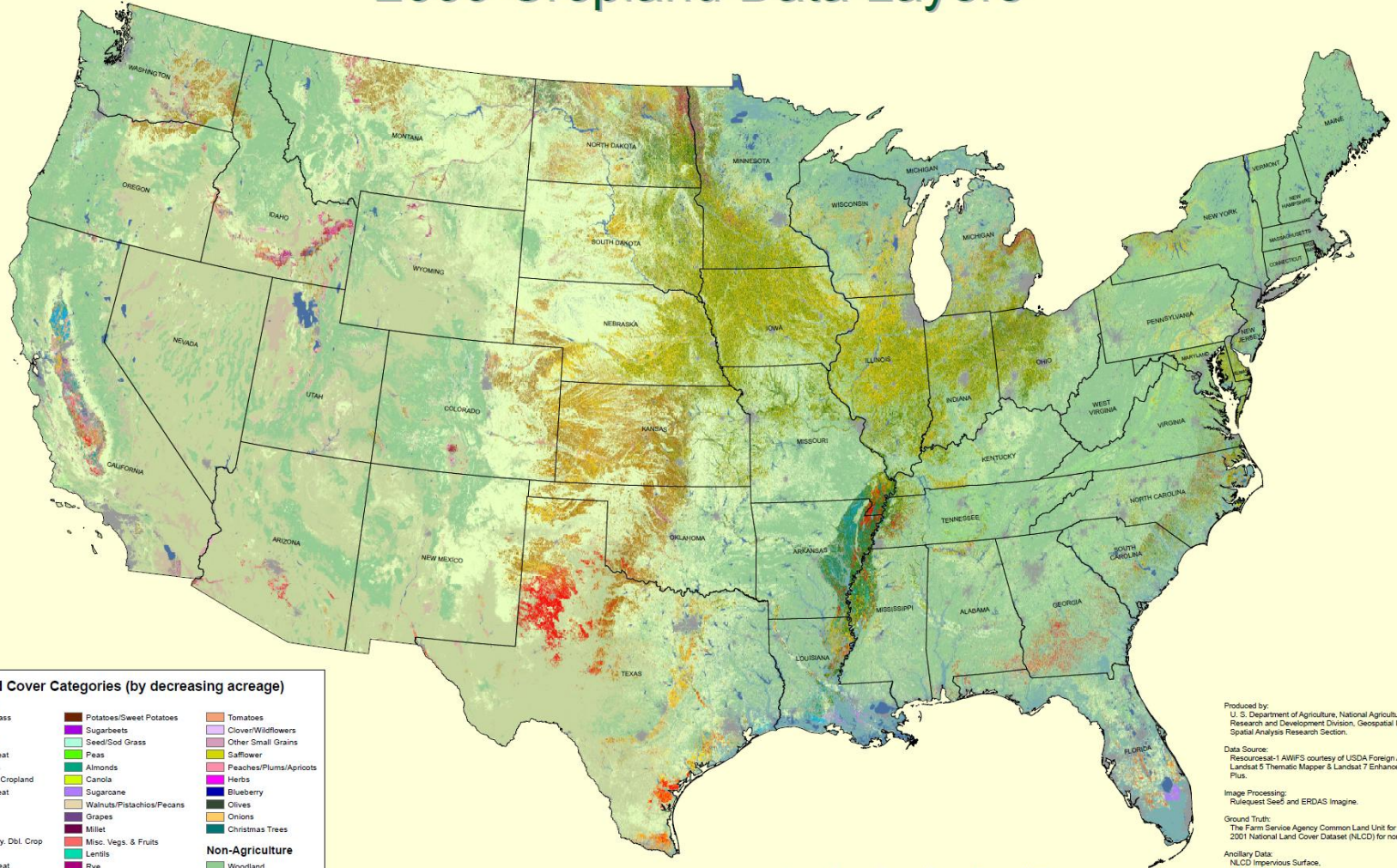


# Study Overview

- Want to understand how potential errors in training data impact decision-tree based land cover classification
  - Especially tailored to mapping efforts within NASS
  - Primarily in regions dominated by common commodity crops
- Hypothesis : Classification accuracy decreases as training data accuracy decreases
  - By how much?
  - Is there a threshold?
  - What's the relationship?
  - Is it linear?
  - Are there scenarios where it improves the outcome?
- Chose 3 states to test these questions
  - Iowa
  - Idaho
  - North Dakota

# Operational land cover mapping within NASS

## 2009 Cropland Data Layers



**Land Cover Categories (by decreasing acreage)**

Agriculture		
Pasture/Grass	Potatoes/Sweet Potatoes	Tomatoes
Corn	Sugarbeets	Clover/Wildflowers
Soybeans	Seed/Grass	Other Small Grains
Winter Wheat	Peas	Safflower
Other Hays	Almonds	Peaches/Plums/Apricots
Fallow/Idle Cropland	Canola	Herbs
Spring Wheat	Sugarcane	Blueberry
Alfalfa	Walnuts/Pistachios/Pecans	Olives
Cotton	Grapes	Onions
Sorghum	Millet	Christmas Trees
W. Wht./Soy. Dbl. Crop	Misc. Veggies & Fruits	
Rice	Lentils	
Durum Wheat	Rye	<b>Non-Agriculture</b>
Barley	Other Tree Nuts/Fruits	Woodland
Sunflowers	Apples/Cherries/Pears	Shrubland
Oats	Flaxseed	Urban/Developed
Dry Beans	Other Crops	Wetlands
Peanuts	Aquaculture	Water
Oranges	Citrus	Barren
		Perennial Ice/Snow

Produced by:  
U. S. Department of Agriculture, National Agricultural Statistics Service,  
Research and Development Division, Geospatial Information Branch,  
Spatial Analysis Research Section.

Data Source:  
Resourcesat-1 AWIFS courtesy of USDA Foreign Agricultural Service,  
Landsat 5 Thematic Mapper & Landsat 7 Enhanced Thematic Mapper  
Plus.

Image Processing:  
Rulequest See5 and ERDAS Imagine.

Ground Truth:  
The Farm Service Agency Common Land Unit for crops classes, and  
2001 National Land Cover Dataset (NLCD) for non-agricultural classes.

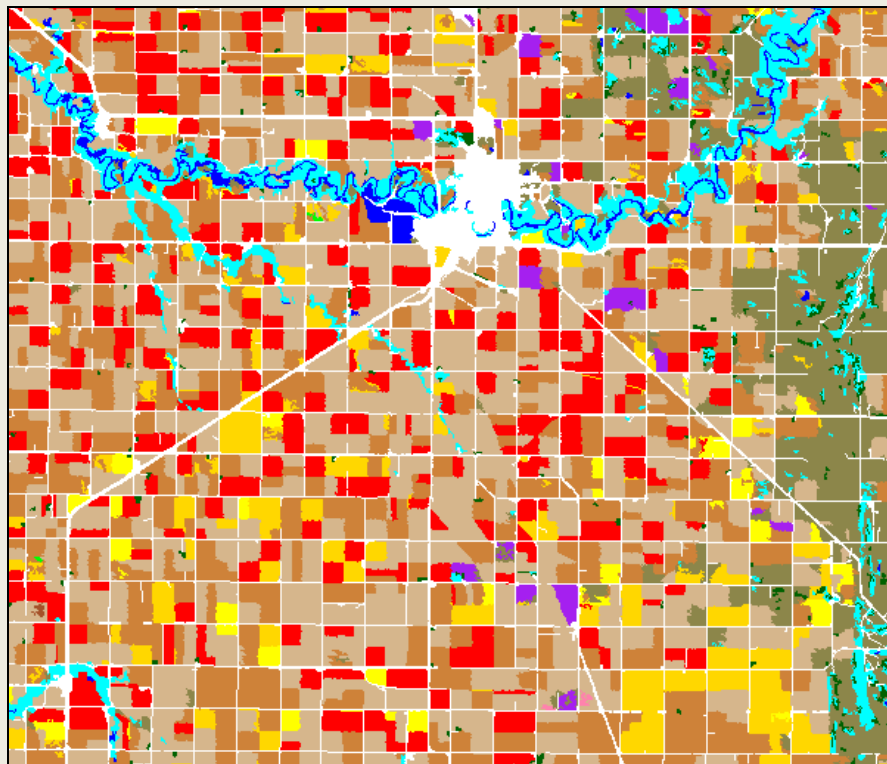
Ancillary Data:  
NLCD Impervious Surface,  
NLCD Forest Canopy,  
National Elevation Dataset, and  
Moderate Resolution Imagery Spectroradiometer (MODIS) 16 day  
Normalized Difference Vegetation Index (NDVI) composite.

Projection:  
Albers Equal Area Conic Projection, NAD 1983 datum.

Map Production:  
ESRI ArcGIS 9.3.1.



# Example Classification Subset



**CDL Classification**

(red = sugar beets, brown = soybeans  
tan = spring wheat, gold = corn,  
yellow = sunflowers)

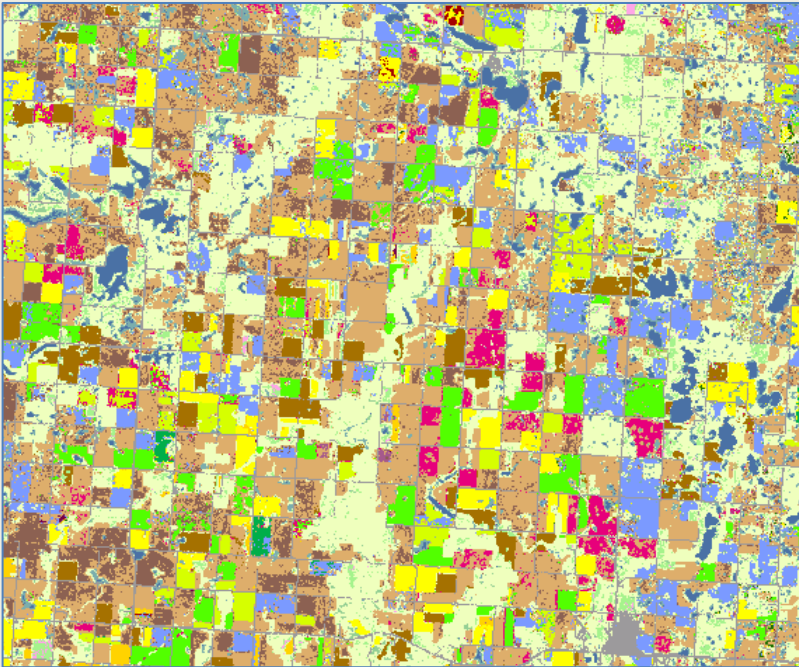


**Resourcesat-1 AWiFS, 6 July 2007**

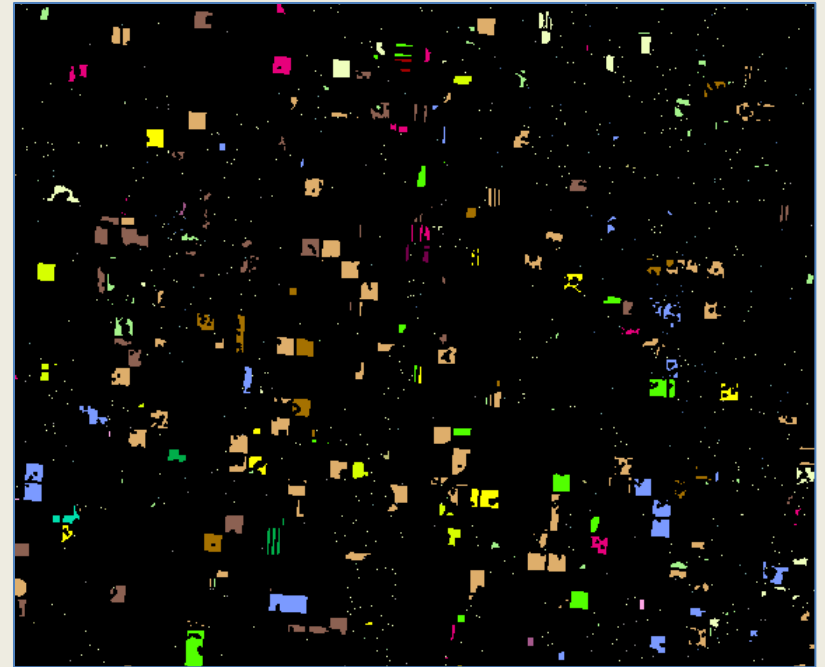
(red =SWIR band,  
green=NIR band,  
blue=red band)

# Accuracy Assessment

Each classification tested against independent set of ground truth data to determine overall and within class accuracies



Example classification subset



Example validation subset

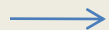
# Degradation methodology

Altered sample files with  
X'th row scrambled

Column with land cover  
category value

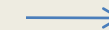
DG	DH	DI	DJ	DK	DL	DM	DN
133	22	3474	8460	434	0	0	1
132	21	3474	8460	436	0	0	5
133	24	4020	9158	435	0	0	1
138	24	2939	9249	440	0	0	1
133	20	3711	9183	439	0	0	1
134	22	3711	9183	439	0	0	1
133	23	3324	9359	437	0	0	1
134	30	4205	9011	440	0	0	1
140	28	3759	9042	438	0	0	1
134	25	3770	9252	436	0	0	1
135	24	2864	9244	439	0	0	1
132	23	3324	9359	437	0	0	1
134	24	3759	9042	437	0	0	1
134	23	4205	9011	438	0	0	1
134	24	3433	9104	435	0	0	1
134	23	3372	8987	439	0	0	1
133	23	3372	8987	440	0	0	1
134	22	3695	8875	438	0	0	181

Every  
row

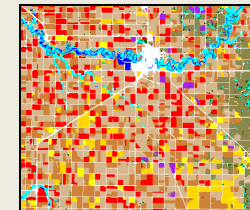


DG	DH	DI	DJ	DK	DL	DM	DN
133	22	3474	8460	434	0	0	1
132	21	3474	8460	436	0	0	5
133	24	4020	9158	435	0	0	181
138	24	2939	9249	440	0	0	1
133	20	3711	9183	439	0	0	5
134	22	3711	9183	439	0	0	1
133	23	3324	9359	437	0	0	1
134	30	4205	9011	440	0	0	1
140	28	3759	9042	438	0	0	1
134	25	3770	9252	436	0	0	5
135	24	2864	9244	439	0	0	1
132	23	3324	9359	437	0	0	1
134	24	3759	9042	437	0	0	1
134	23	4205	9011	438	0	0	1
134	24	3433	9104	435	0	0	1
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133	23	3372	8987	440	0	0	5
134	22	3695	8875	438	0	0	181

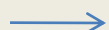
Run classifier



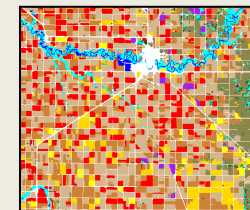
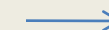
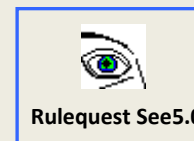
Output land cover map



Every  
other  
row

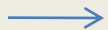


DG	DH	DI	DJ	DK	DL	DM	DN
133	22	3474	8460	434	0	0	1
132	21	3474	8460	436	0	0	5
133	24	4020	9158	435	0	0	1
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133	20	3711	9183	439	0	0	1
134	22	3711	9183	439	0	0	1
133	23	3324	9359	437	0	0	1
134	30	4205	9011	440	0	0	1
140	28	3759	9042	438	0	0	1
134	25	3770	9252	436	0	0	5
135	24	2864	9244	439	0	0	1
132	23	3324	9359	437	0	0	1
134	24	3759	9042	437	0	0	1
134	23	4205	9011	438	0	0	1
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134	23	3372	8987	439	0	0	1
133	23	3372	8987	440	0	0	1
134	22	3695	8875	438	0	0	181

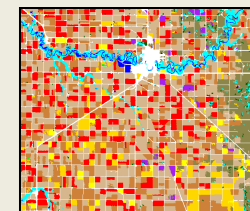
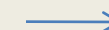


Original sample file  
with no known errors  
(dozens of columns,  
hundreds of thousands  
of rows in reality)

Every  
third  
row



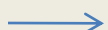
DG	DH	DI	DJ	DK	DL	DM	DN
133	22	3474	8460	434	0	0	1
132	21	3474	8460	436	0	0	1
133	24	4020	9158	435	0	0	181
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133	20	3711	9183	439	0	0	1
134	22	3711	9183	439	0	0	1
133	23	3324	9359	437	0	0	1
134	30	4205	9011	440	0	0	1
140	28	3759	9042	438	0	0	1
134	25	3770	9252	436	0	0	1
135	24	2864	9244	439	0	0	1
132	23	3324	9359	437	0	0	5
134	24	3759	9042	437	0	0	1
134	23	4205	9011	438	0	0	1
134	24	3433	9104	435	0	0	1
134	23	3372	8987	439	0	0	1
133	23	3372	8987	440	0	0	1
134	22	3695	8875	438	0	0	181



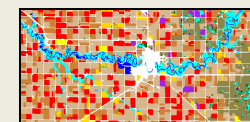
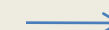
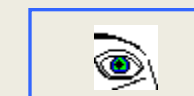
Etc.



Every  
forth  
row













DG	DH	DI	DJ	DK	DL	DM	DN
133	22	3474	8460	434	0	0	1
132	21	3474	8460	436	0	0	5
133	24	4020	9158	435	0	0	1
138	24	2939	9249	440	0	0	1
133	20	3711	9183	439	0	0	1
134	22	3711	9183	439	0	0	1
133	23	3324	9359	437	0	0	1
134	30	4205	9011	440	0	0	1
140	28	3759	9042	438	0	0	1









# 2009 Iowa Cropland Data Layer

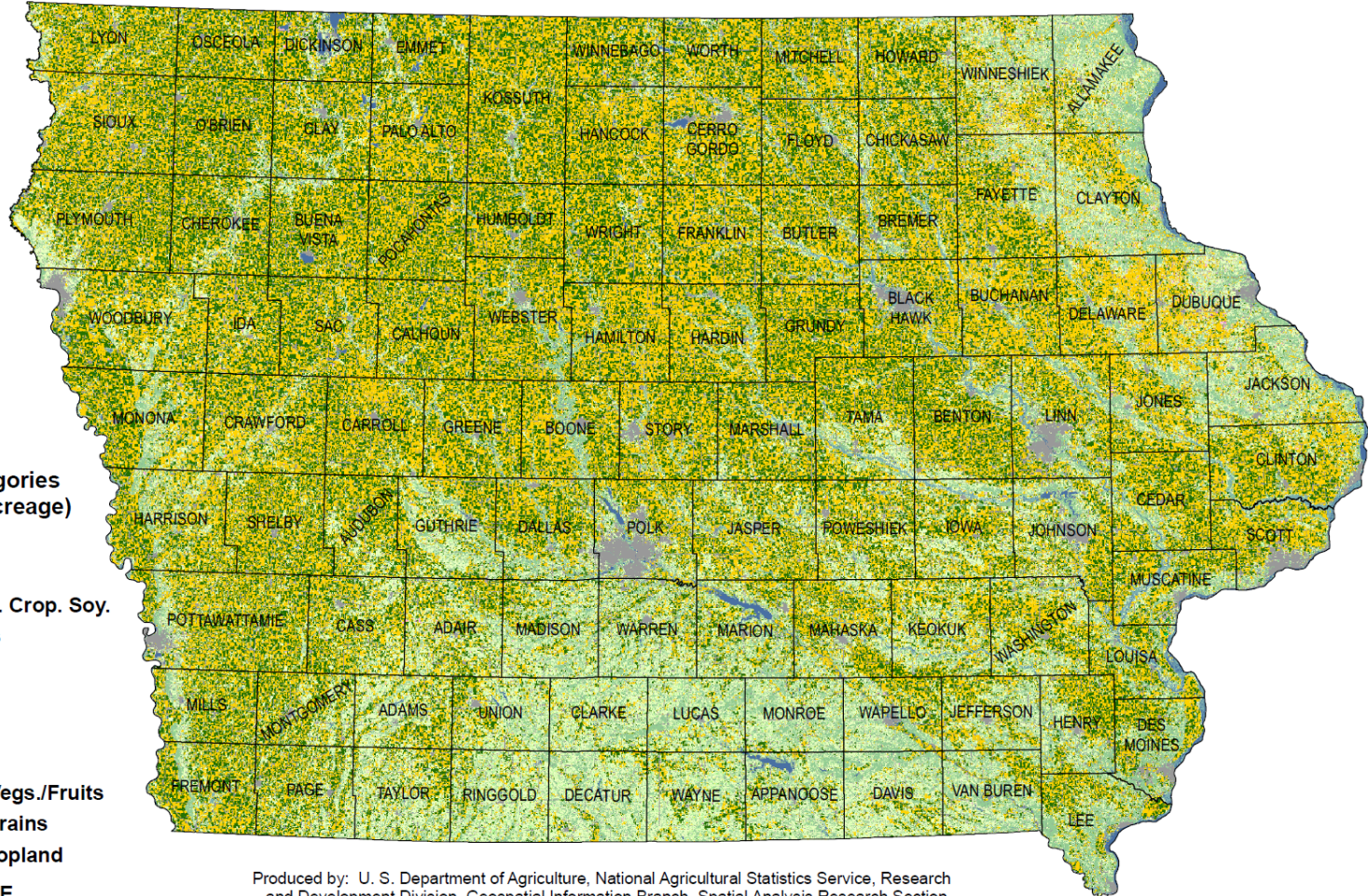
## Land Cover Categories (by decreasing acreage)

### AGRICULTURE

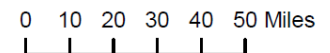
-  Corn
-  Soybeans/Dbi. Crop. Soy.
-  Pasture/Grass
-  Other Hays
-  Alfalfa
-  Oats
-  Winter Wheat
-  Other Crops/Veg./Fruits
-  Other Small Grains
-  Fallow/Idle Cropland

### NON-AGRICULTURE

-  Urban/Developed
-  Woodland
-  Wetlands
-  Water
-  Barren
-  Shrubland

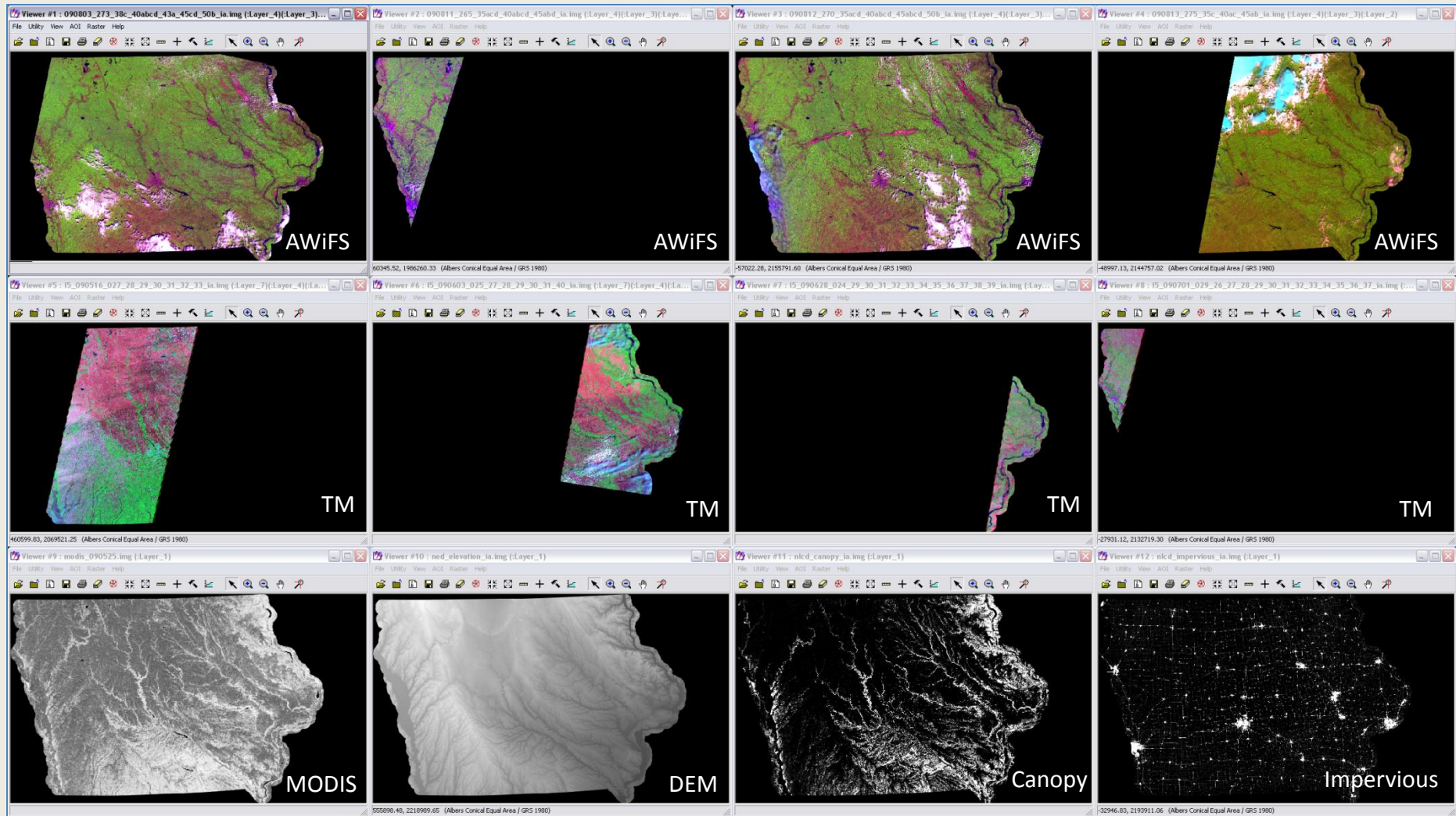


Produced by: U. S. Department of Agriculture, National Agricultural Statistics Service, Research and Development Division, Geospatial Information Branch, Spatial Analysis Research Section.  
 Data Source: Resourcesat-1 AWiFS courtesy of USDA Foreign Agricultural Service, Landsat 5 Thematic Mapper.  
 Image Processing: Rulequest See5 and ERDAS Imagine.  
 Ground Truth: The Farm Service Agency Common Land Unit for crops classes, and 2001 National Land Cover Dataset (NLCD) for non-agricultural classes.  
 Ancillary Data: NLCD Impervious Surface, NLCD Forest Canopy, National Elevation Dataset, and Moderate Resolution Imagery Spectroradiometer (MODIS) 16 day Normalized Difference Vegetation Index (NDVI) composite.  
 Projection: UTM zone 15, WGS84 datum.  
 Map Production: ESRI ArcGIS 9.3.



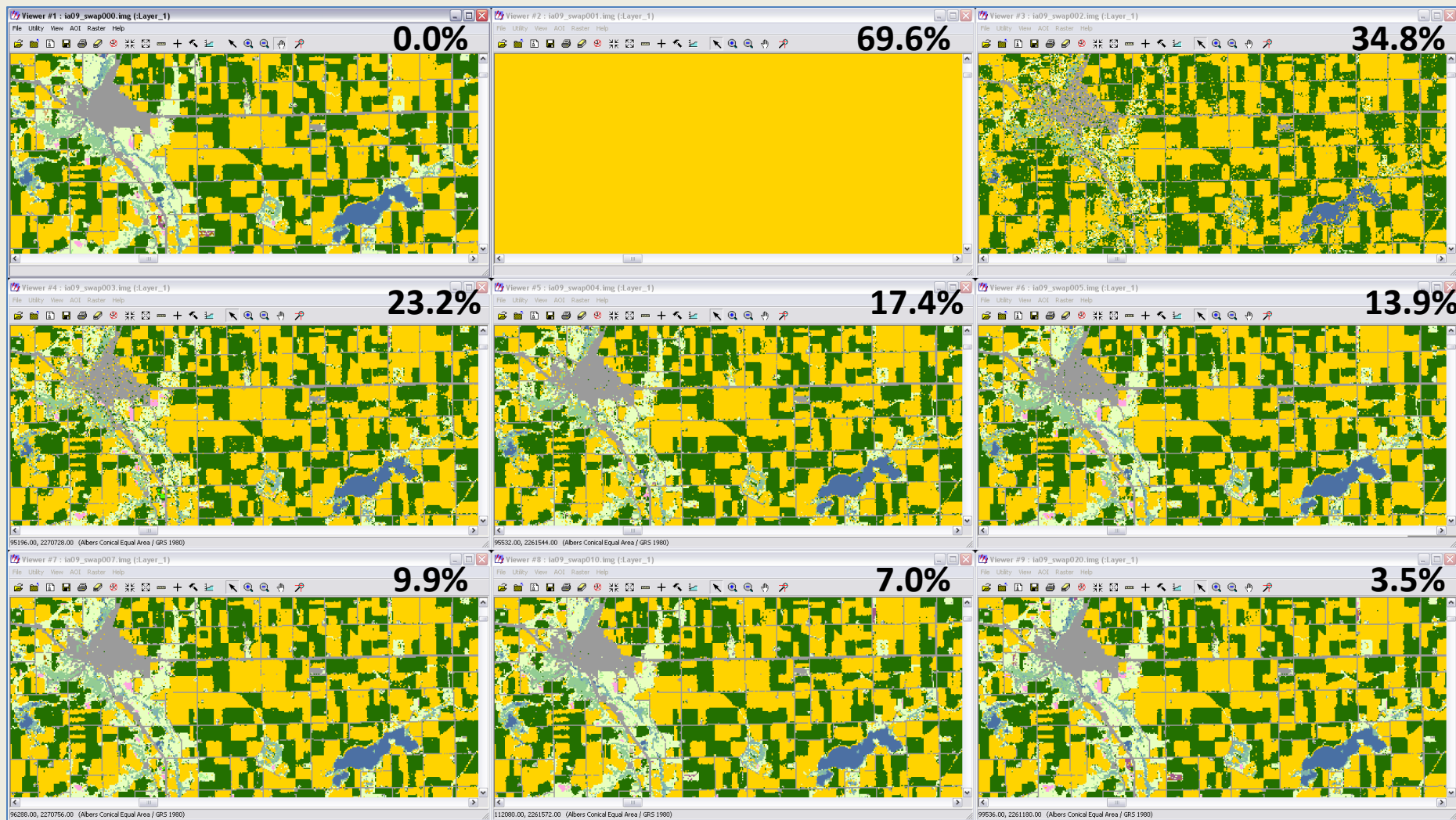


# Iowa '09 CDL input layer examples



Scenes of data actually used: 10 AWiFS, 10 TM, 2 MODIS NDVI, DEM, Canopy, and Impervious  
(dates ranged from 1 April '09 – 8 August '09)

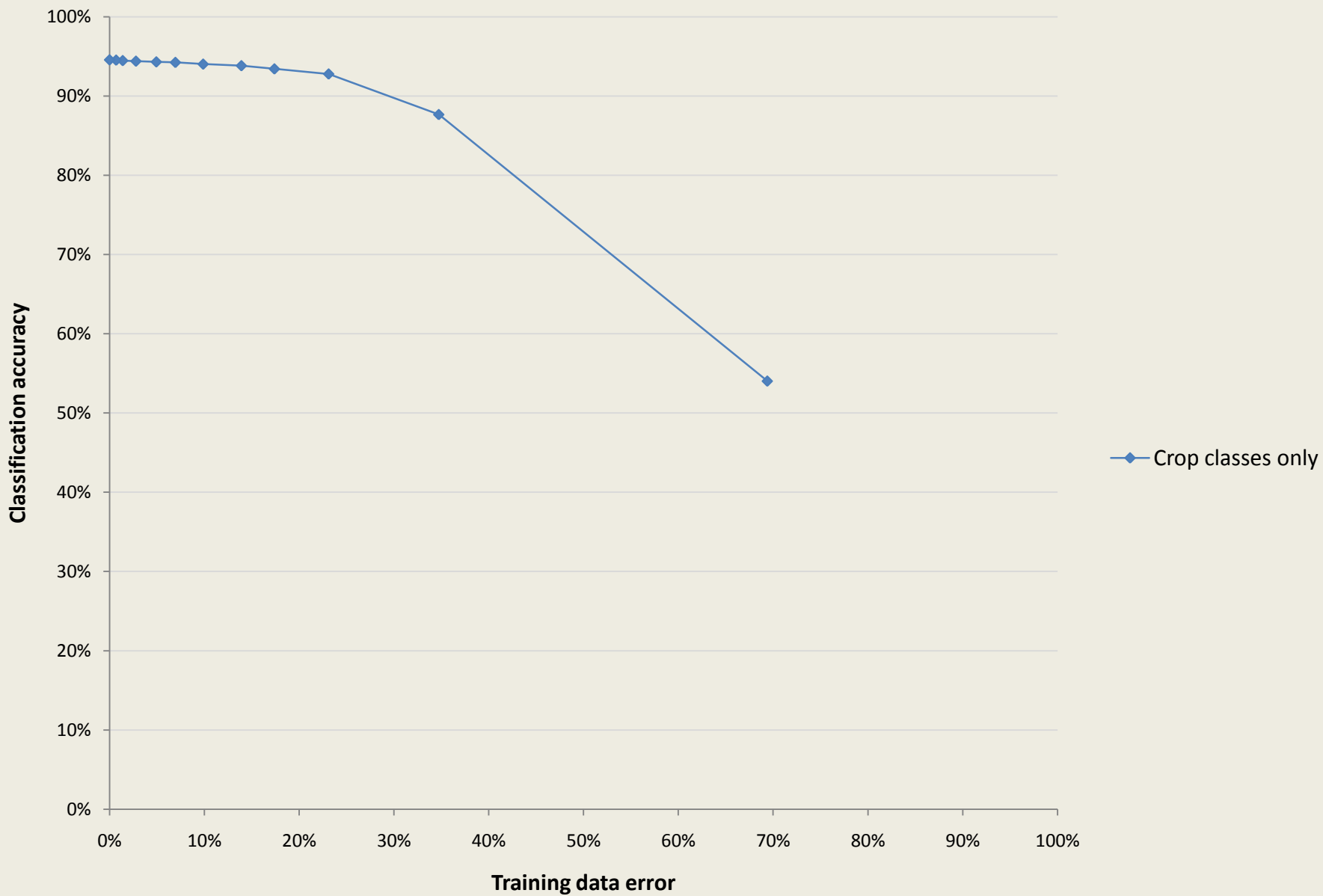
# Iowa classifications with training data error %



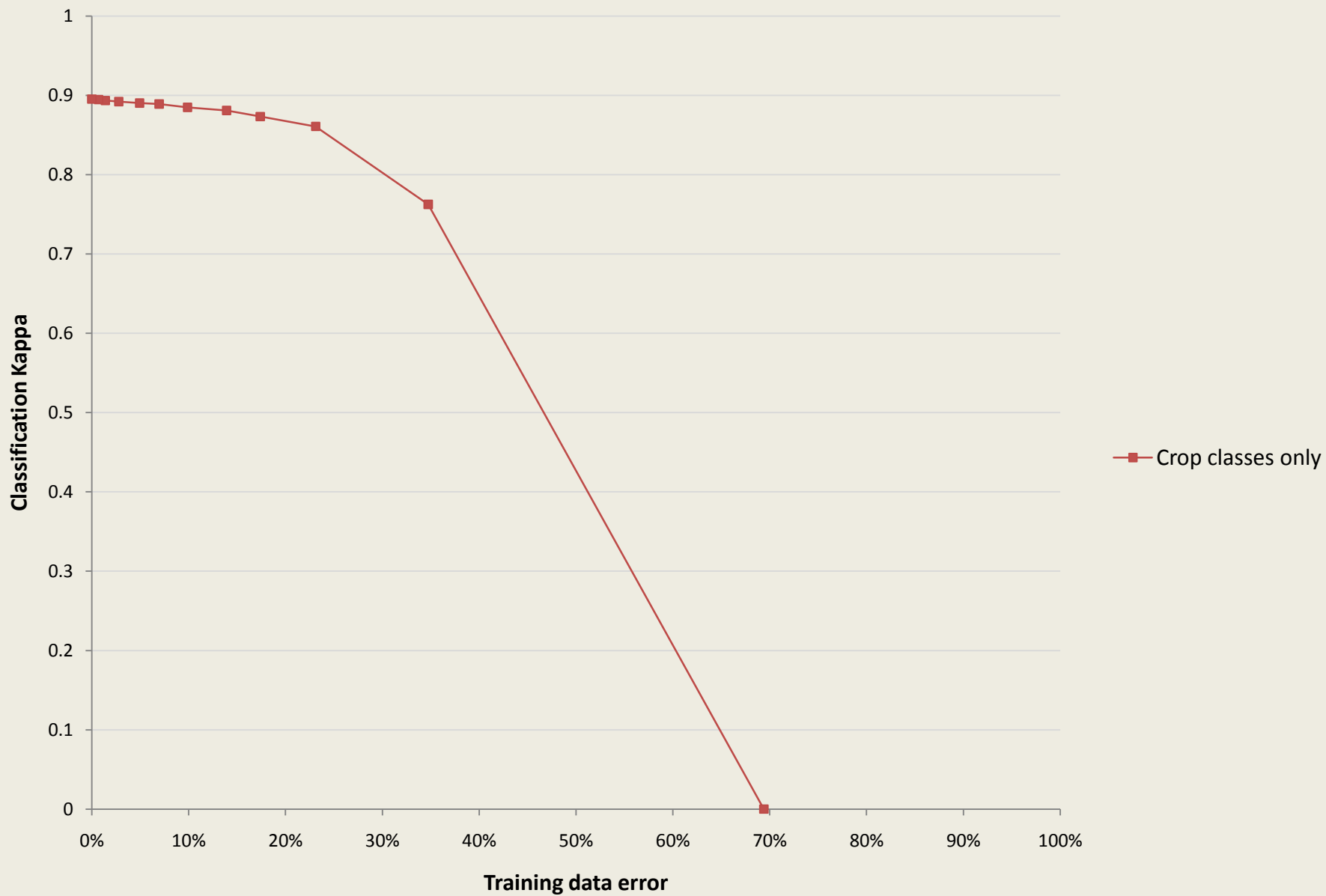
gold = corn, dark green = soybeans

Total scene has 46,474,682 pixels, 755,116 (1.6%) chosen for training

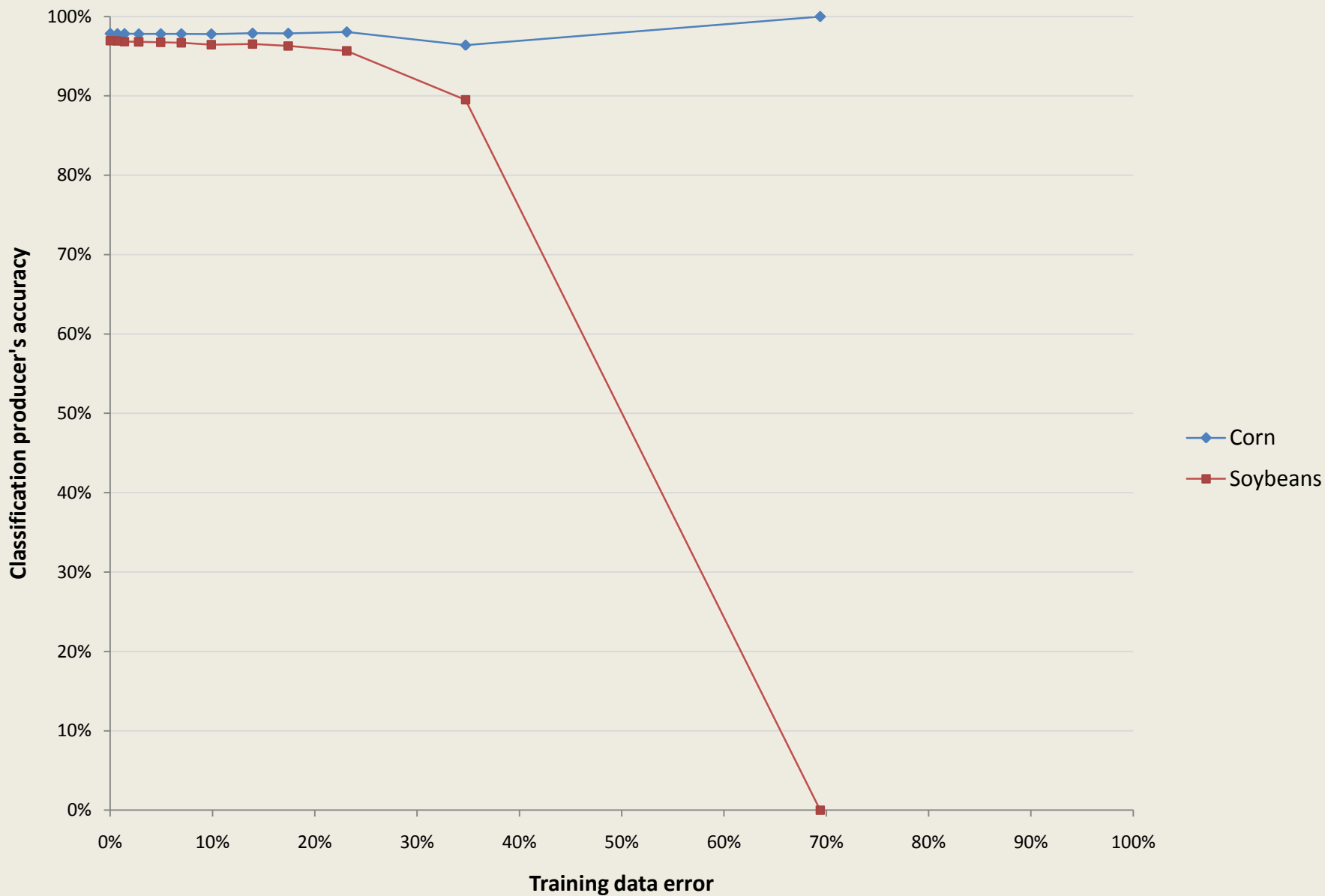
## Iowa '09 CDL, Classification accuracy v. training data error



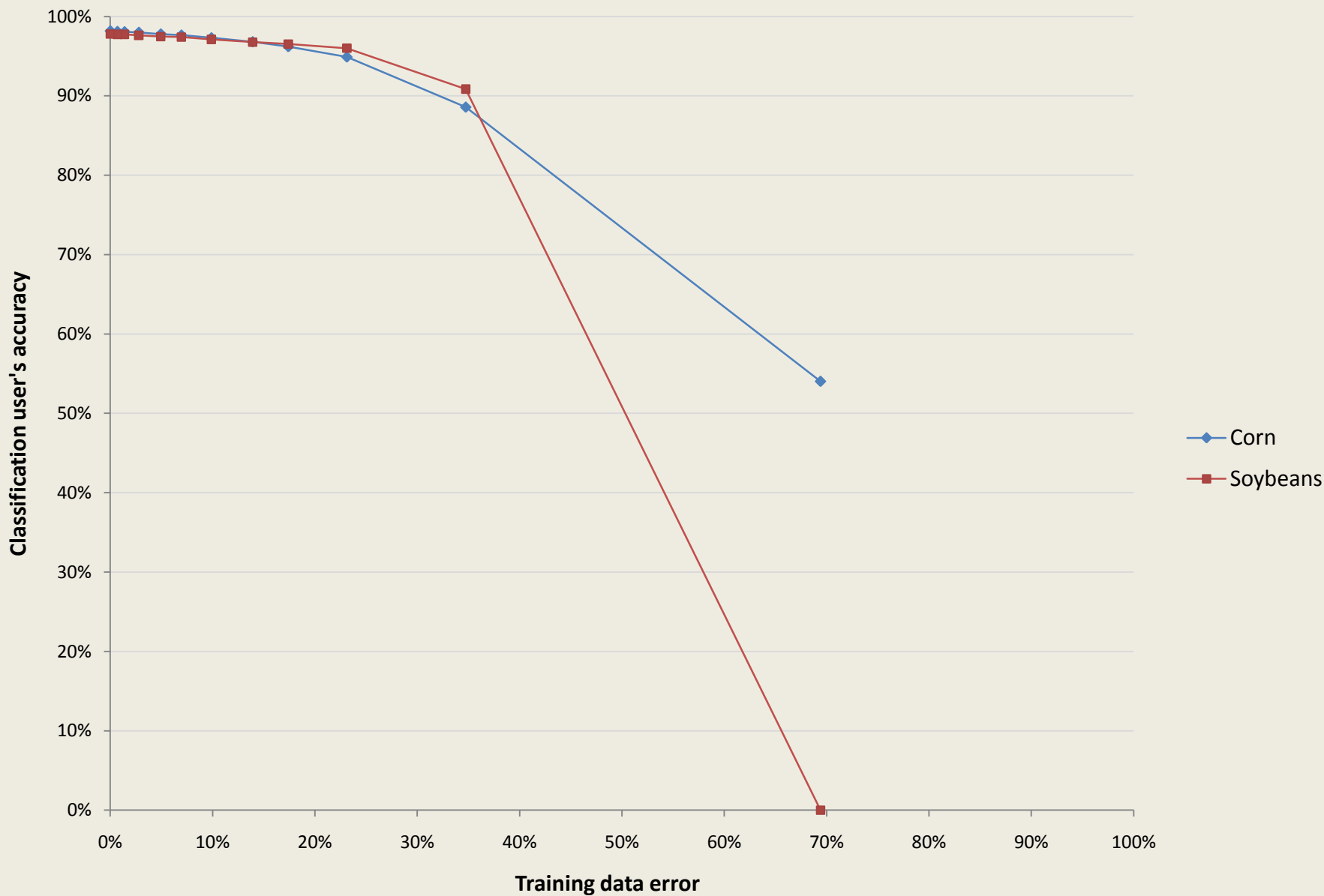
# Iowa '09 CDL, Classification Kappa v. training data error



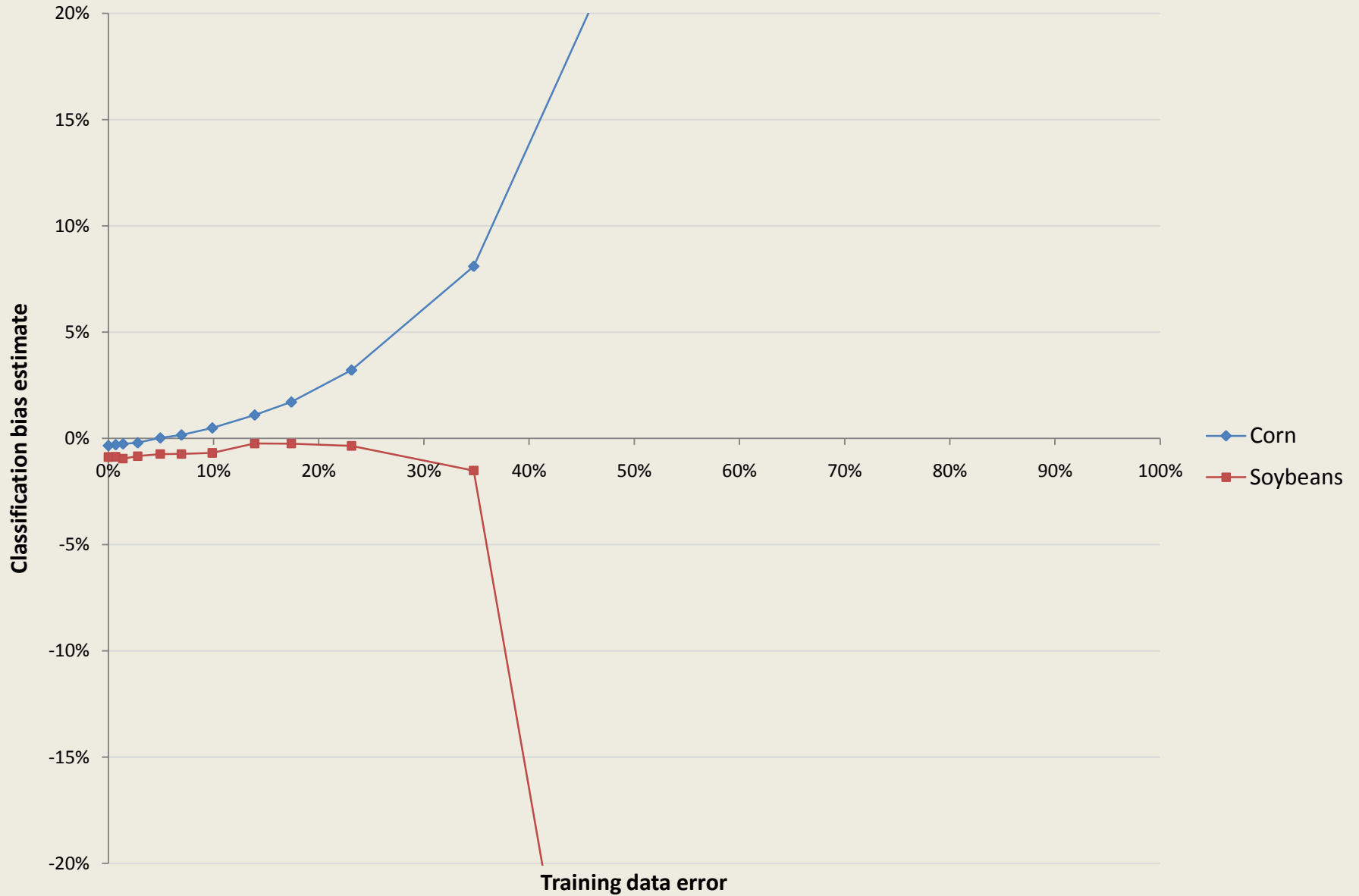
# Iowa '09 CDL, Classification producer's accuracy v. training data error



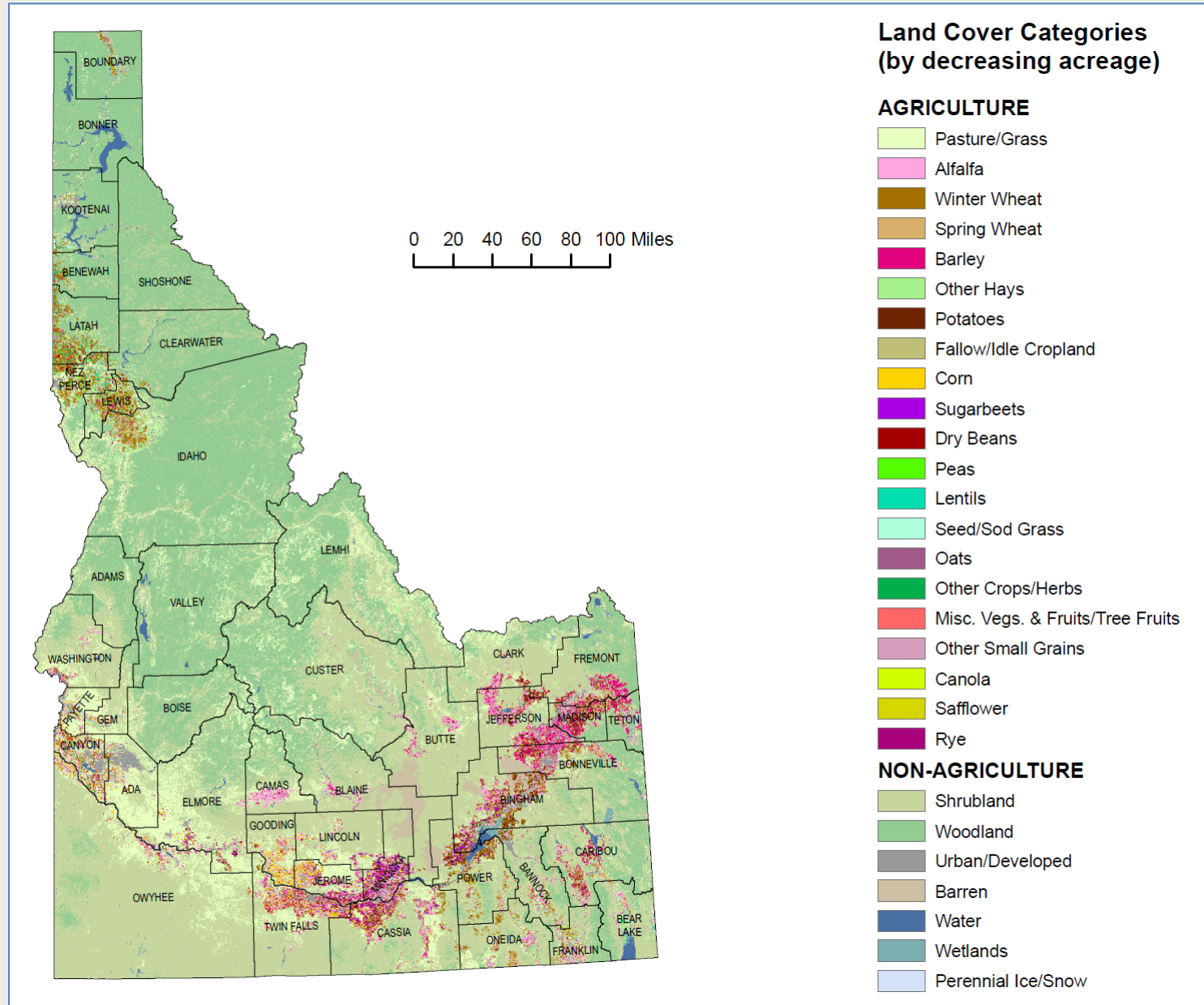
# Iowa '09 CDL, Classification user's accuracy v. training data error



# Iowa '09 CDL, Classification bias v. training data error

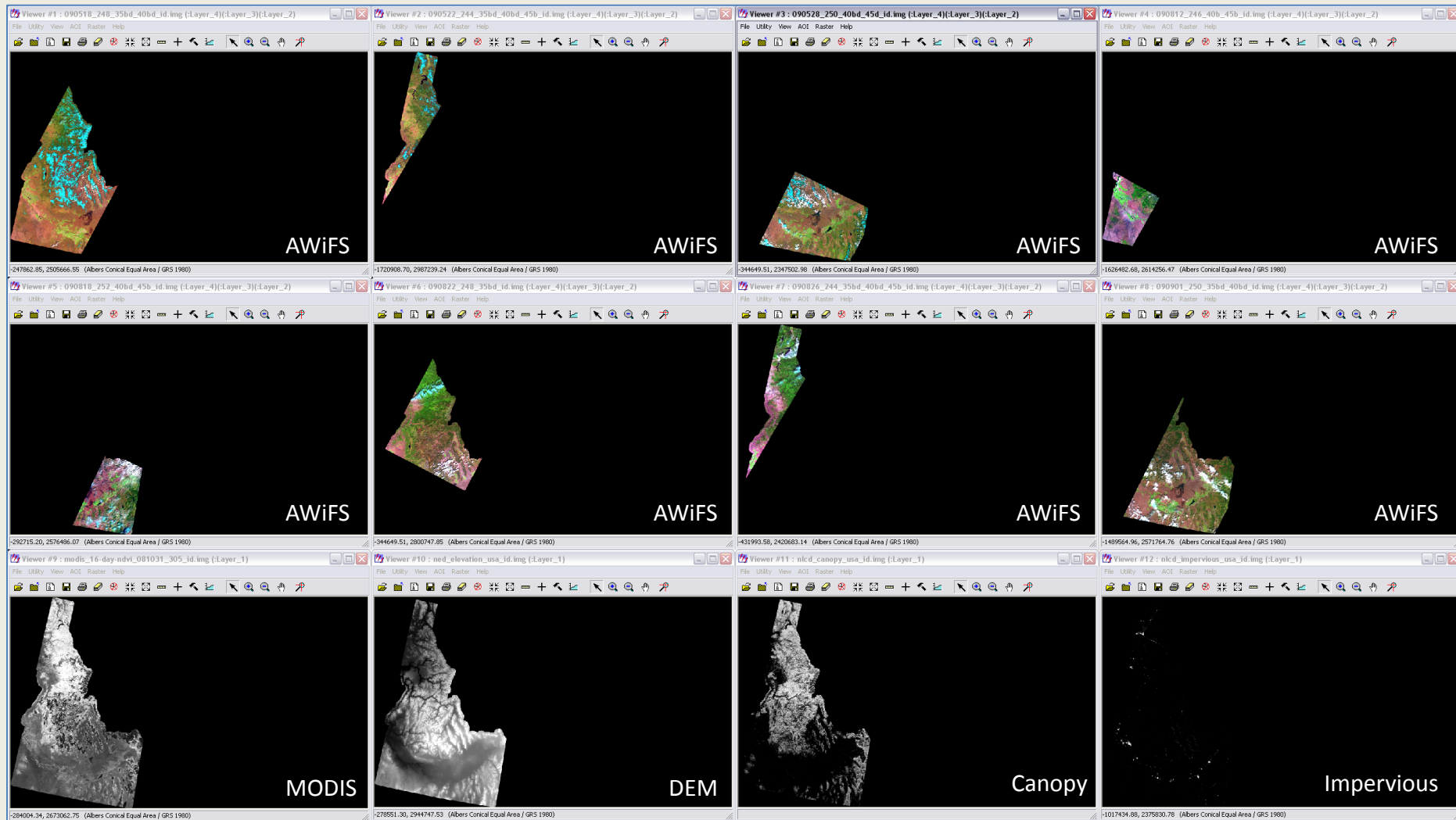


# 2009 Idaho Cropland Data Layer





# Idaho '09 CDL input layer examples



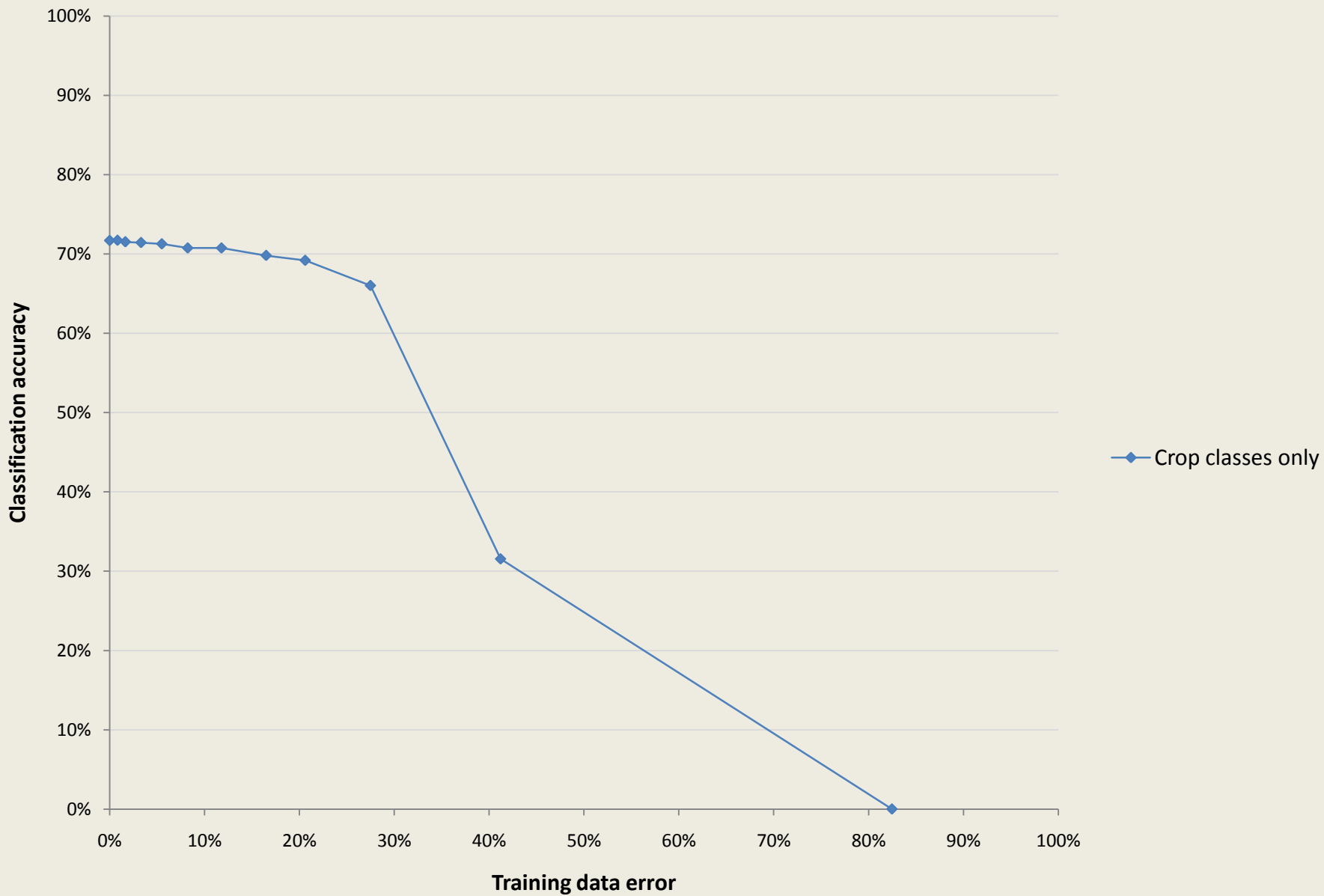
Scenes of data actually used: 15 AWiFS, 7 MODIS NDVI, DEM, Canopy, and Impervious  
(dates ranged from 29 September '08 – 1 September '09)

# Idaho classifications with training data error %



Total scene has 69,018,509 pixels, 891,793 (1.3%) chosen for training

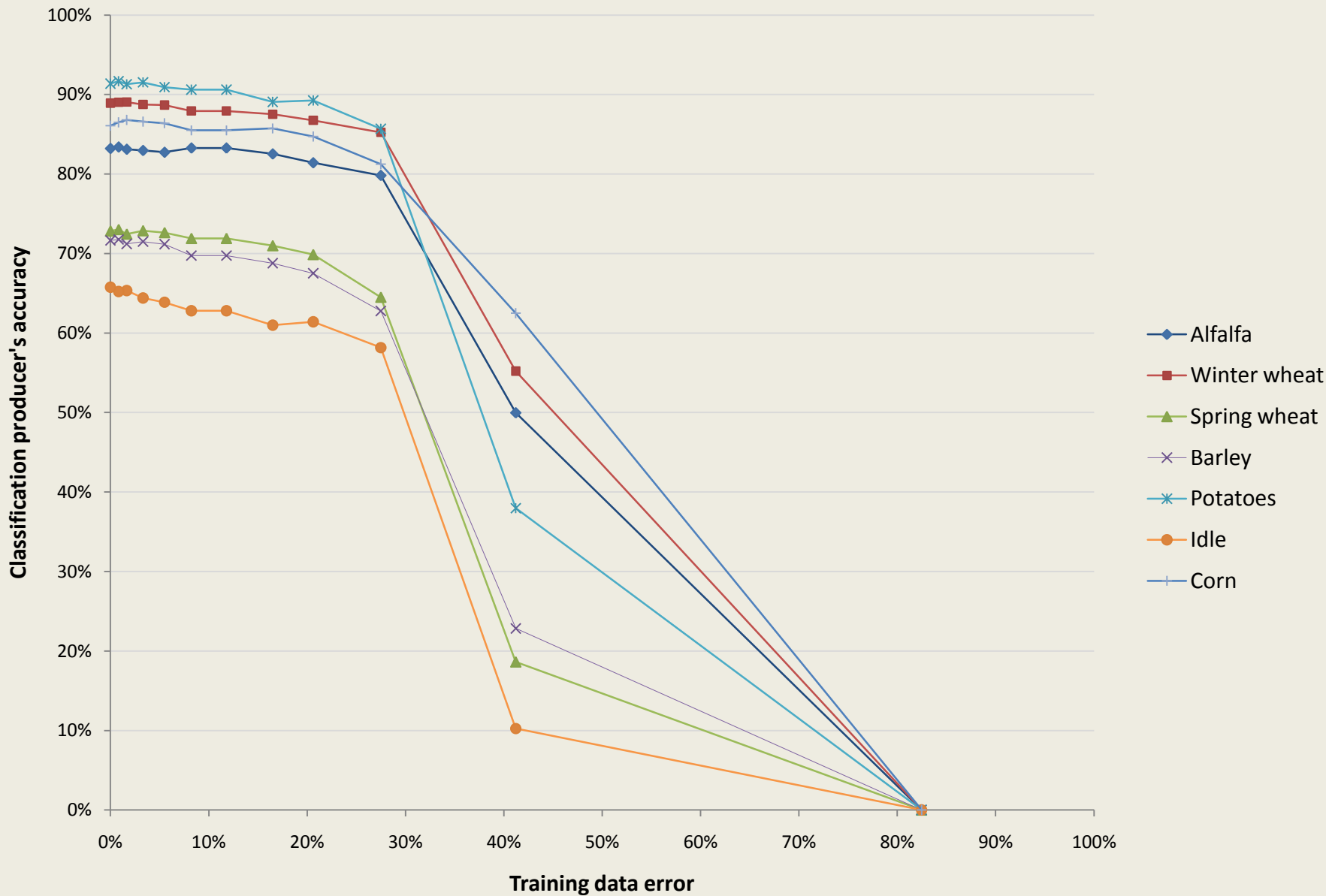
# Idaho '09 CDL, Classification accuracy v. training data error



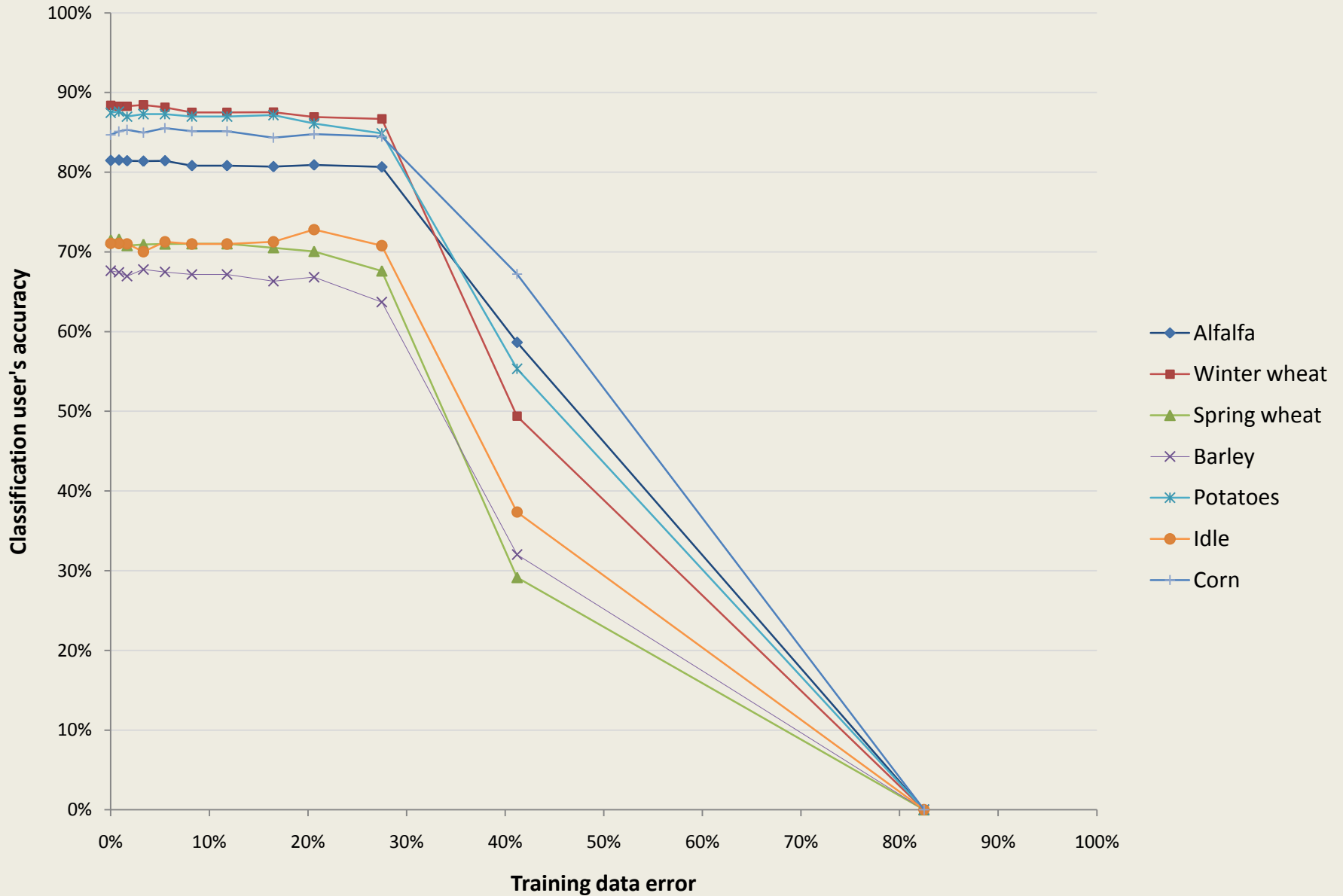
# Idaho '09 CDL, Classification Kappa v. training data error



# Idaho '09 CDL, Classification producer's accuracy v. training data error



# Idaho '09 CDL, Classification user's accuracy v. training data error



# Idaho '09 CDL, Classification bias v. training data error



# 2009 North Dakota Cropland Data Layer

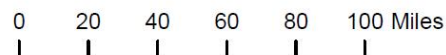
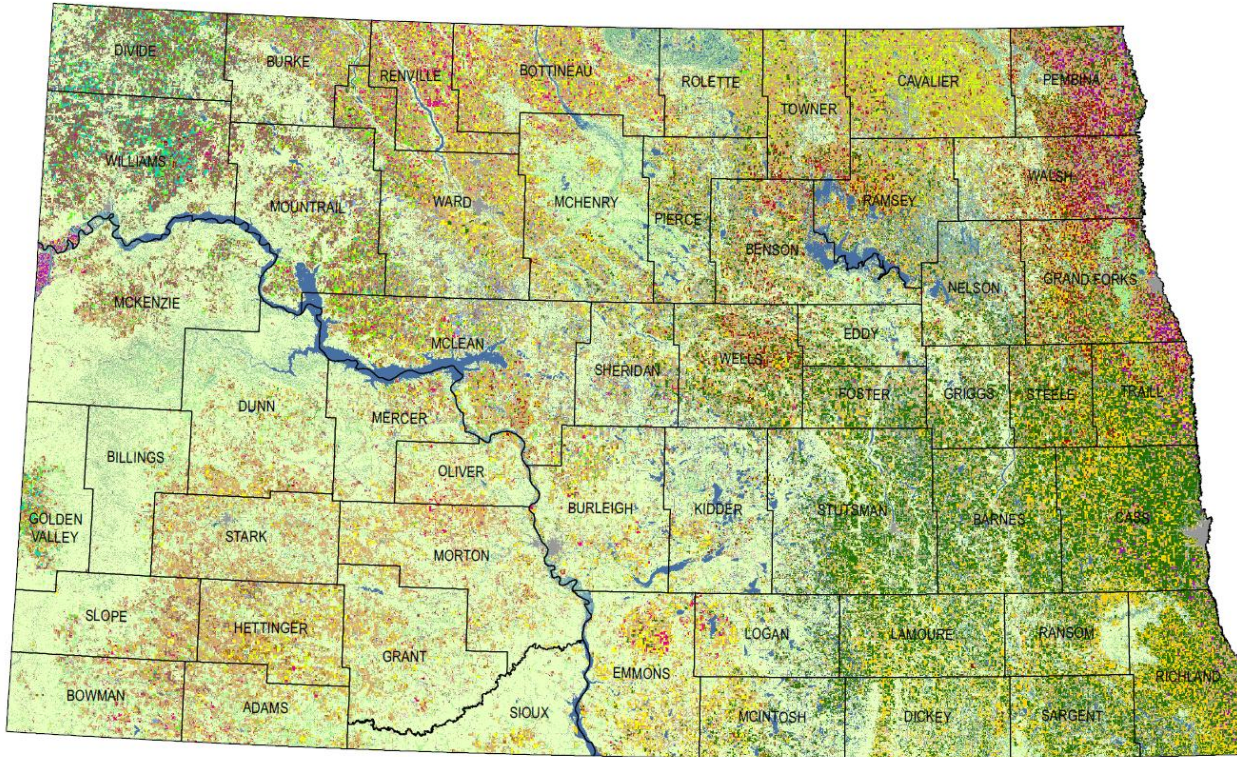
## Land Cover Categories (by decreasing acreage)

### AGRICULTURE

- Pasture/Grass
- Spring Wheat
- Soybeans
- Other Hays
- Corn
- Durum Wheat
- Canola
- Sunflowers
- Dry Beans
- Barley
- Winter Wheat
- Peas
- Alfalfa
- Fallow/Idle Cropland
- Flaxseed
- Sugarbeets
- Lentils
- Oats
- Potatoes
- Other Crops/Vegs./Fruits
- Millet
- Safflower
- Sorghum
- Rye
- Seed/Sod Grass

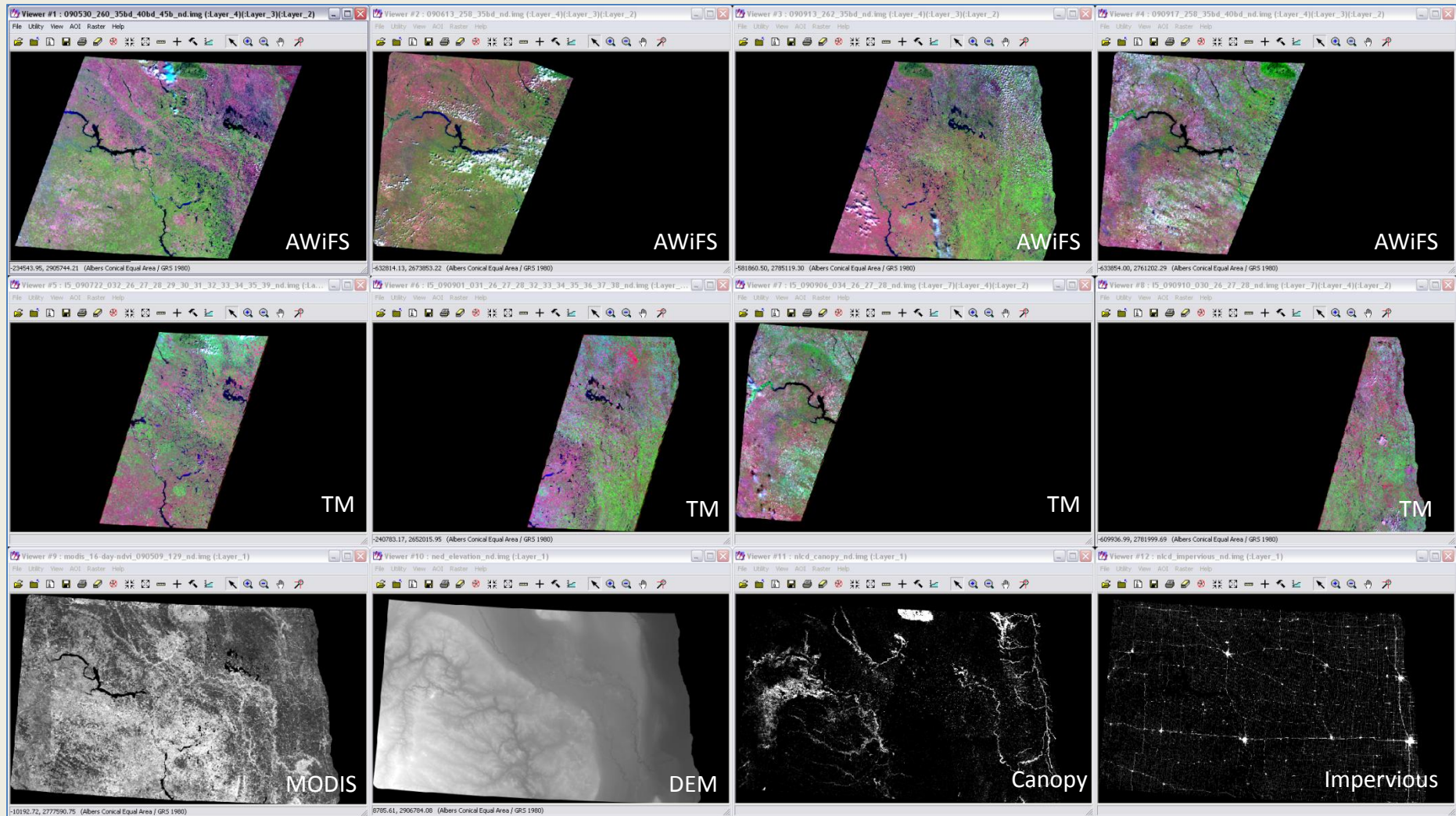
### NON-AGRICULTURE

- Urban/Developed
- Wetlands
- Water
- Woodland
- Shrubland
- Barren



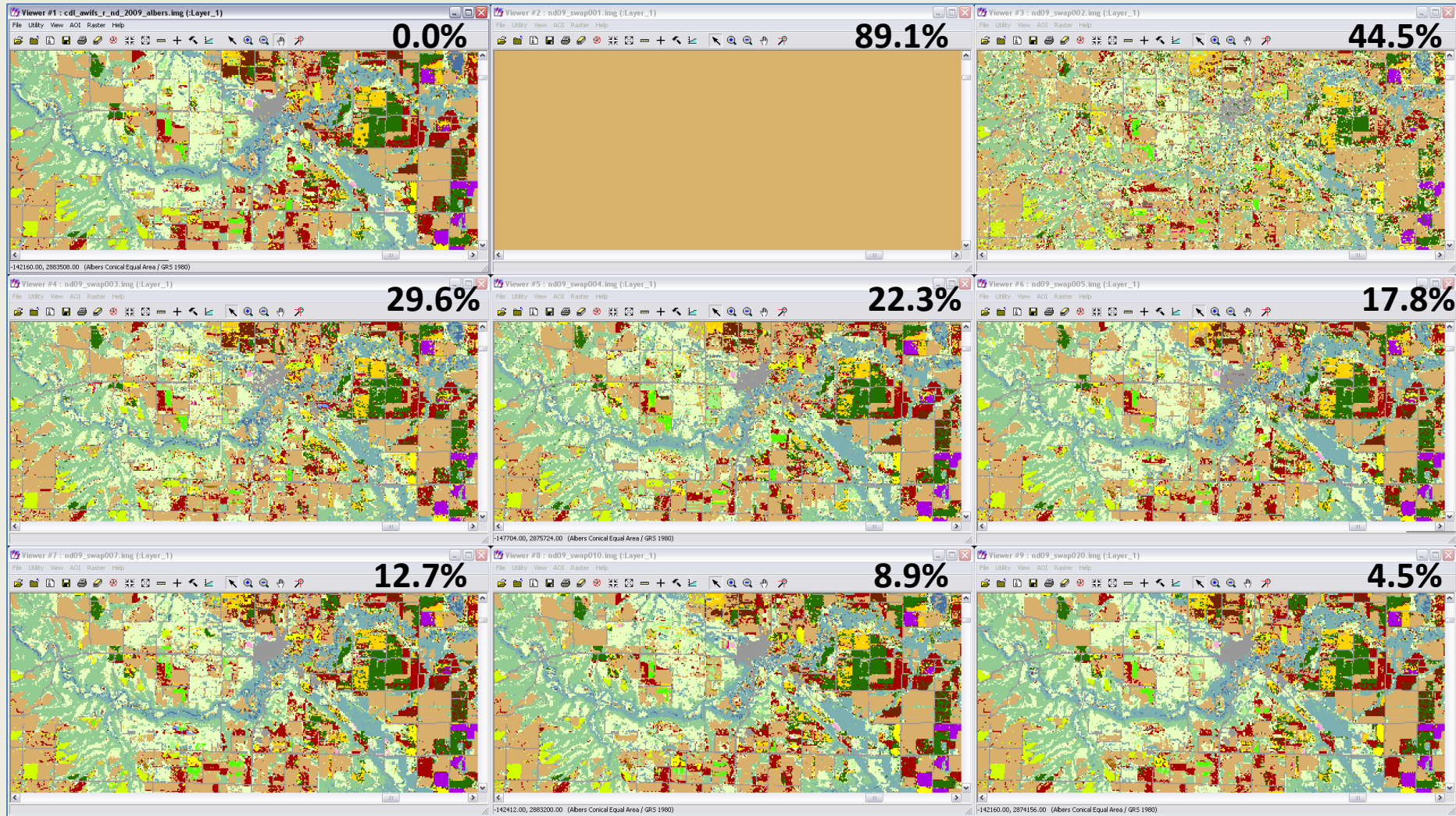


# North Dakota '09 CDL input layer examples



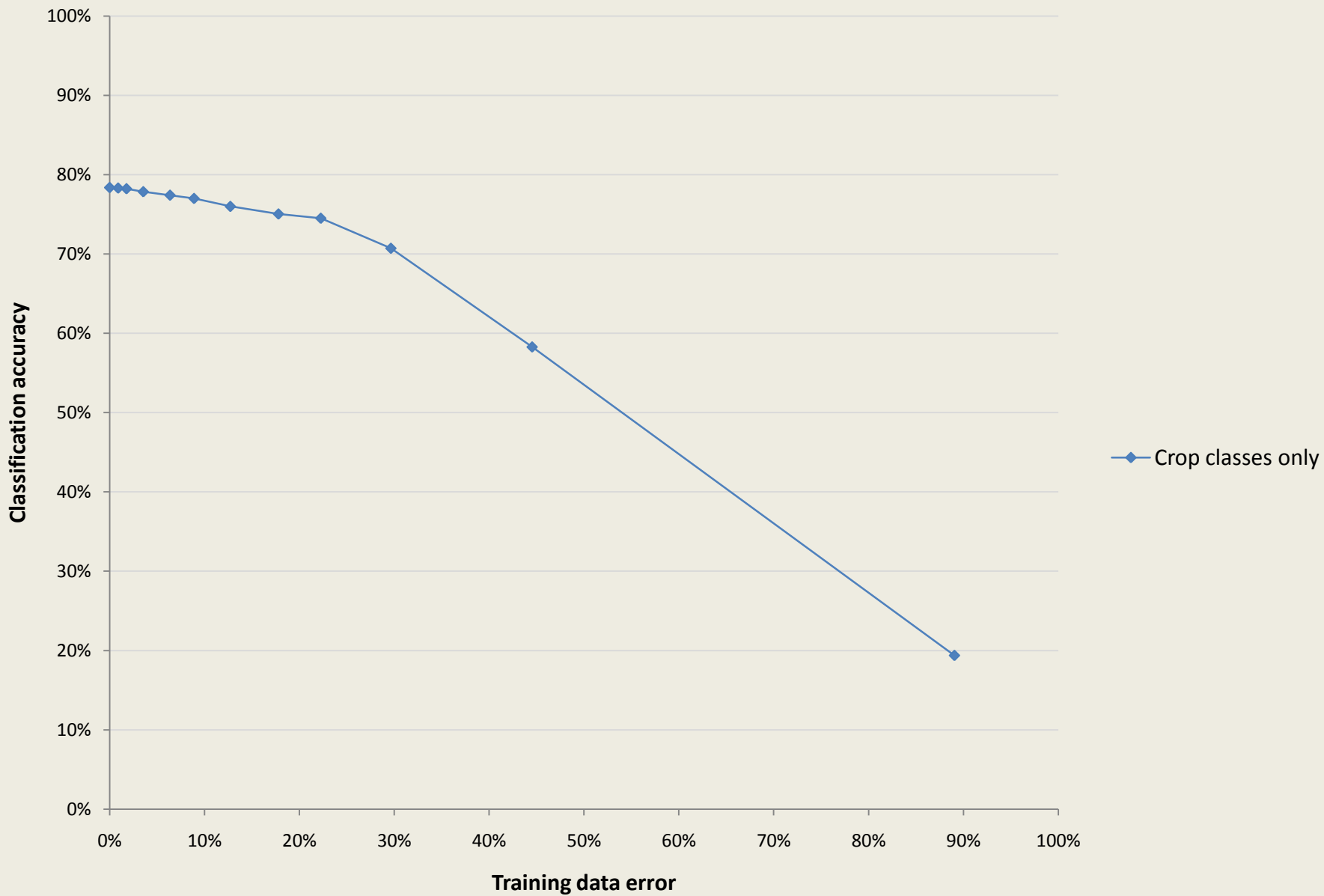
Scenes of data actually used: 14 AWiFS, 13 TM, 1 MODIS NDVI, DEM, Canopy, and Impervious  
(dates ranged from 6 May '09 – 17 September '09)

# North Dakota classifications with training data error %

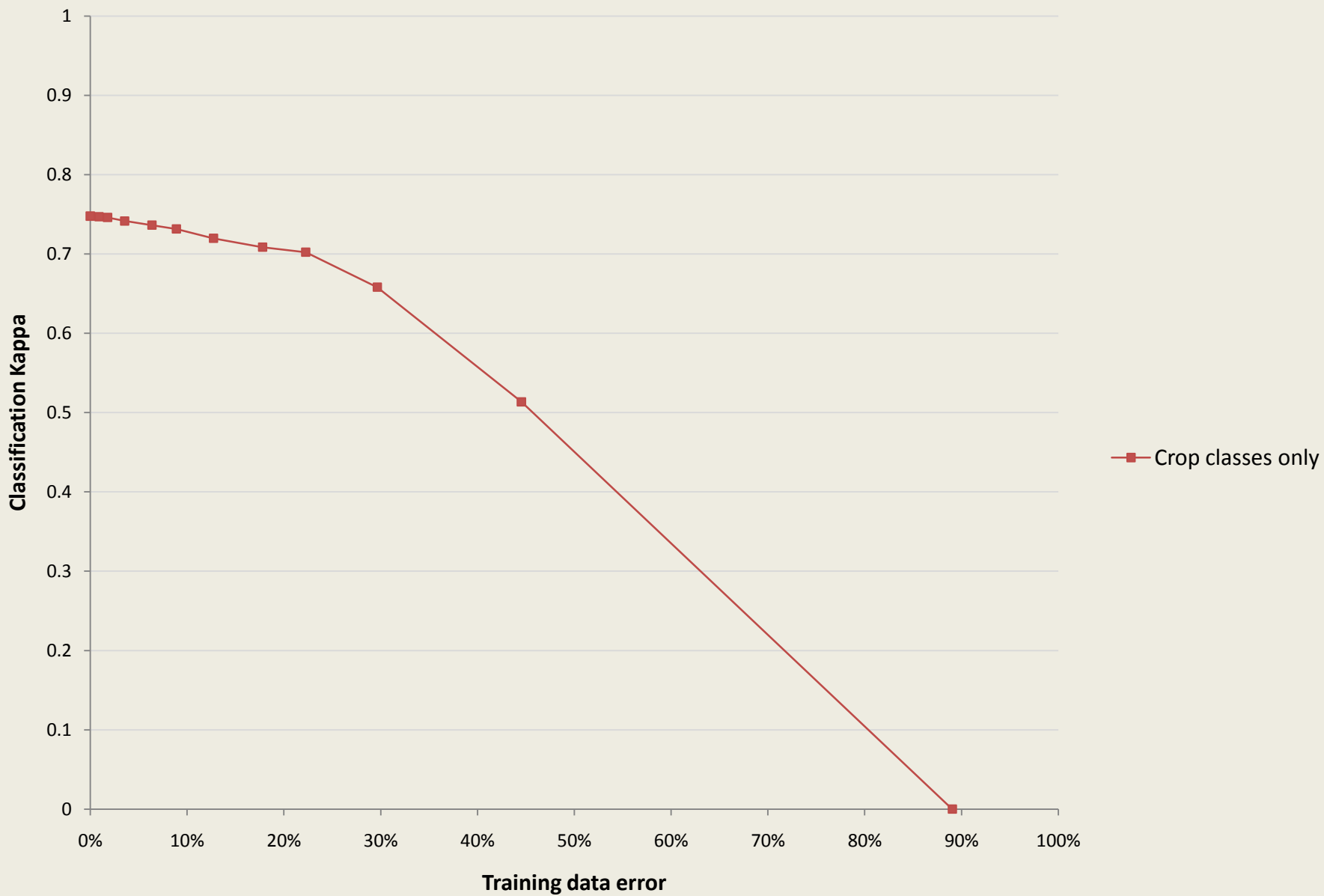


Total scene has 58,388,946 pixels, 737,633 (1.3%) chosen for training

# North Dakota '09 CDL, Classification accuracy v. training data error



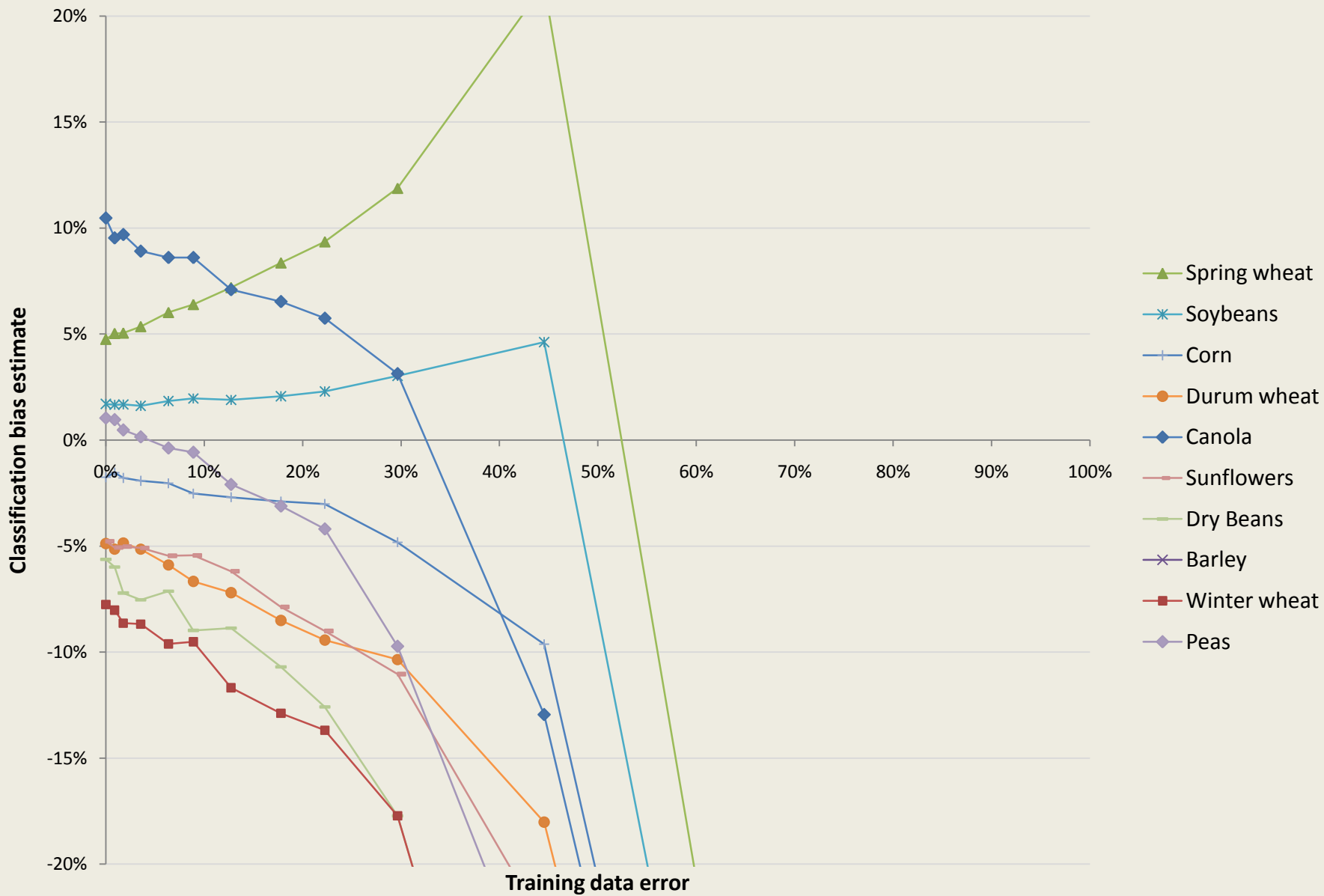
# North Dakota '09 CDL, Classification Kappa v. training data error



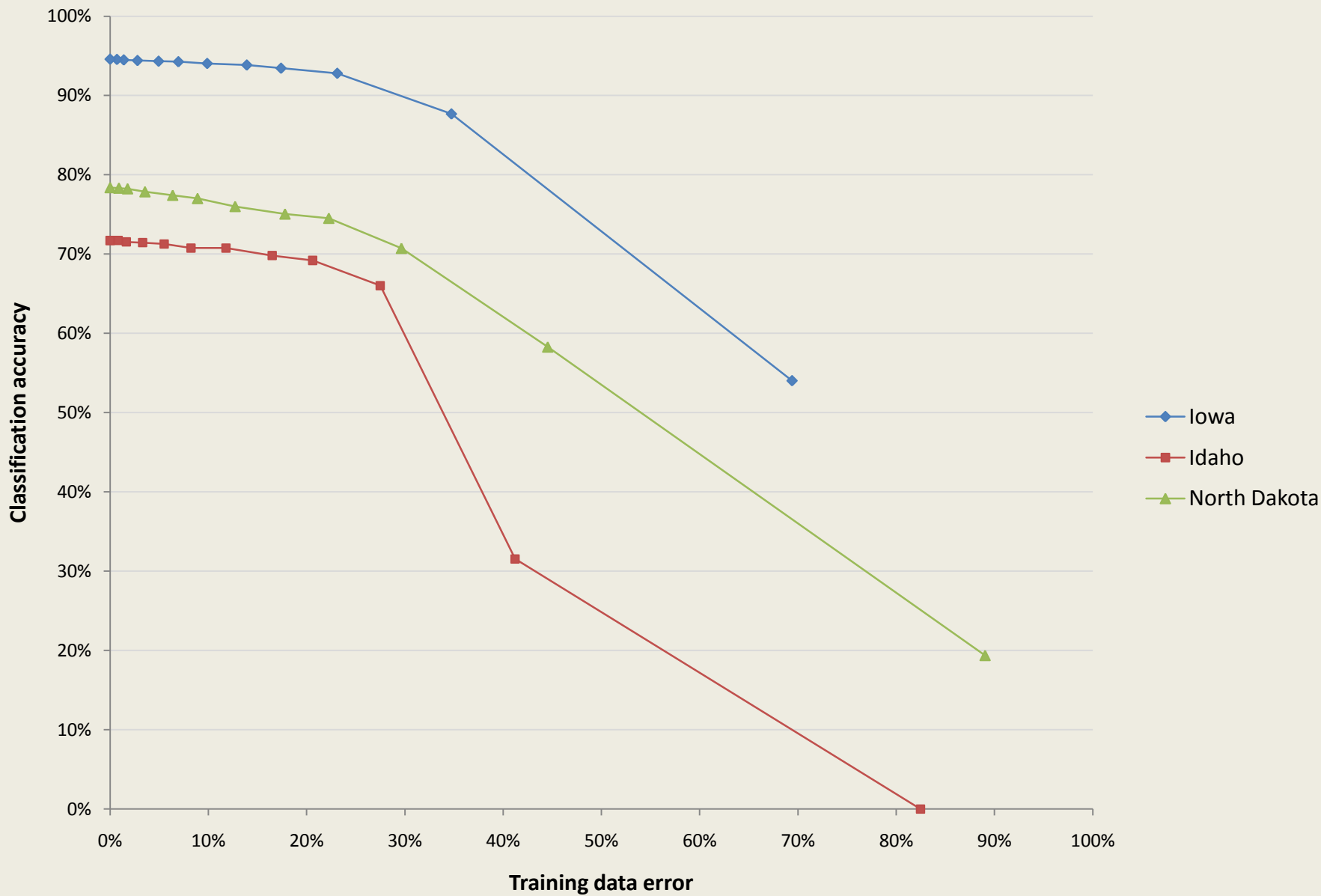




# North Dakota '09 CDL, Classification bias v. training data error

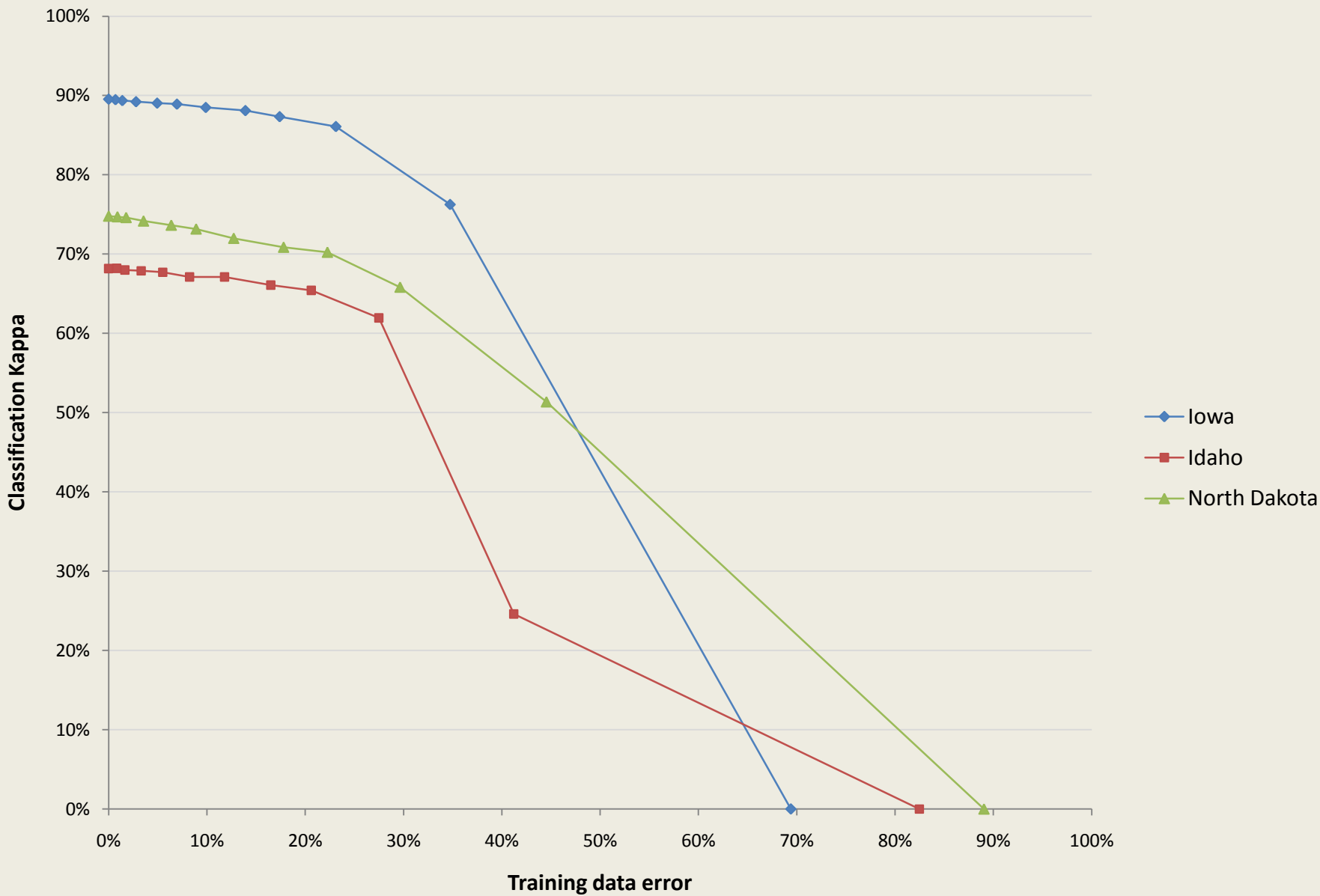


# '09 CDL, Cropland classification accuracy v. training data error

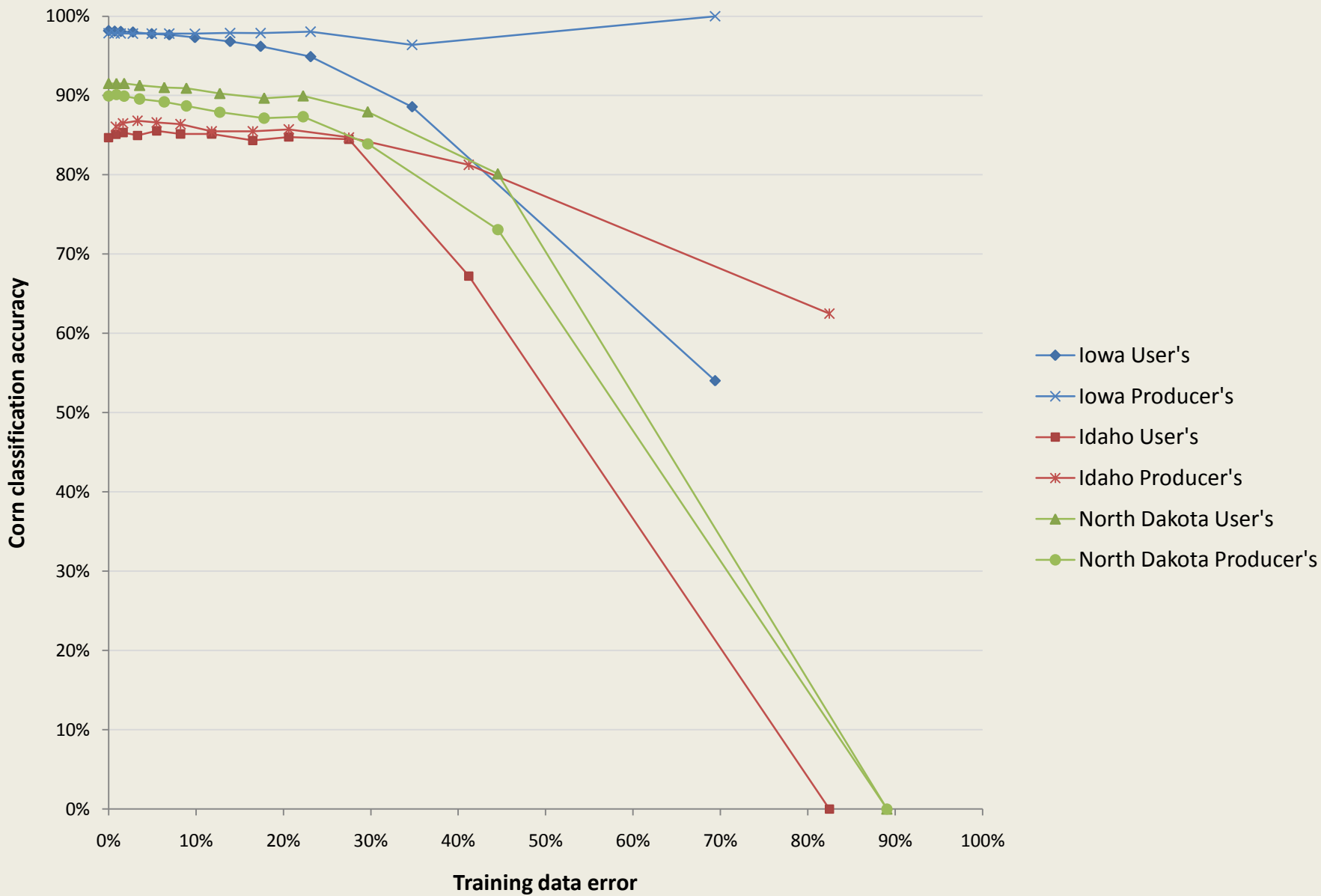




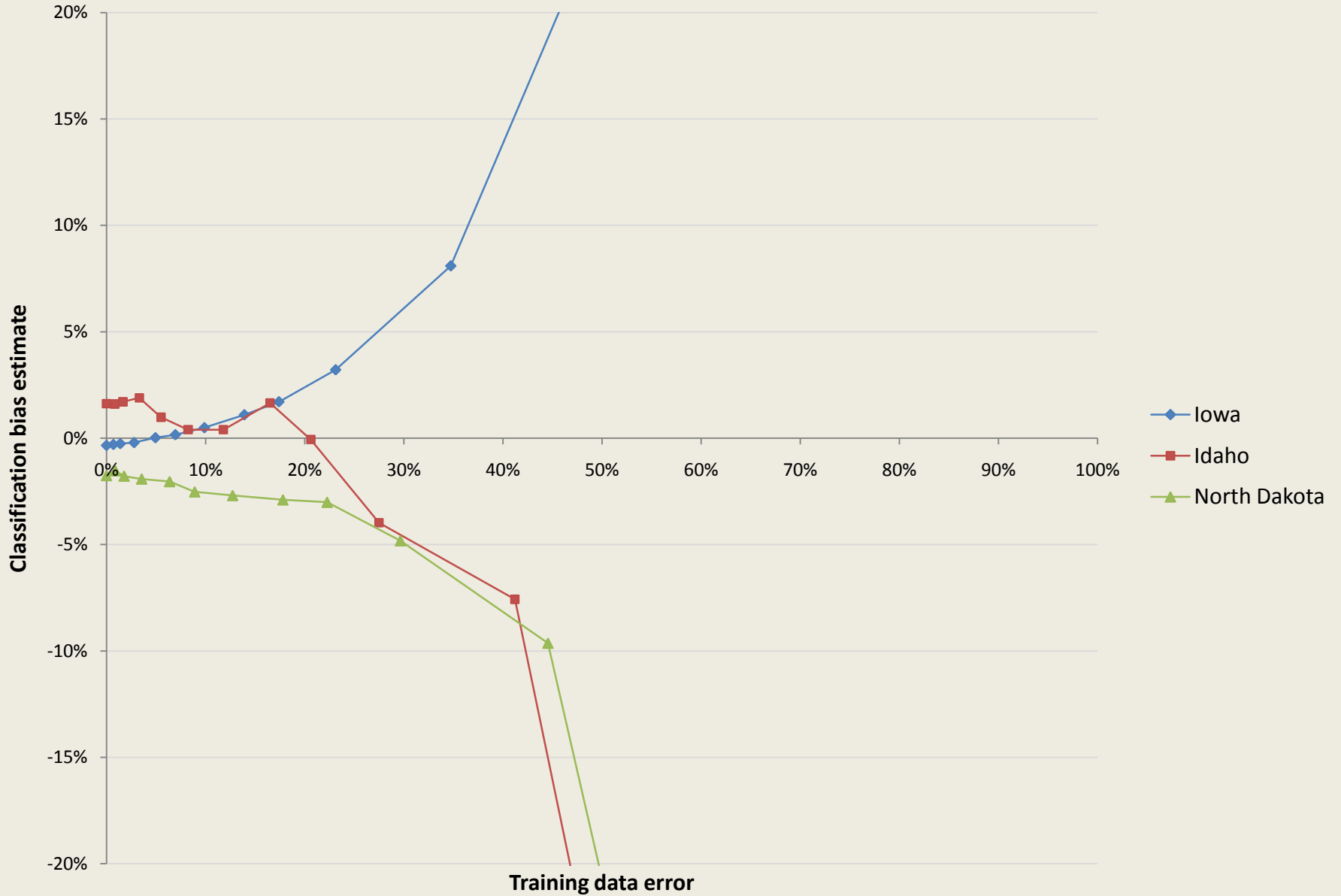
# '09 CDL, Cropland classification Kappa v. training data error



# '09 CDL, Corn classification accuracy v. training data error



# '09 CDL, Corn classification bias v. training data error



# Conclusions

- Degradation of training data.....
  - degrades the classification.
  - has relatively modest impacts on the classification until more than roughly 25% of training data is in error (then it falls rapidly, and thus is not linear).
  - hurts the classification more when lots of classes are present.
  - never improves a classification.
  - impacts differently the areal bias of categories within the classification.

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[www.nass.usda.gov](http://www.nass.usda.gov)