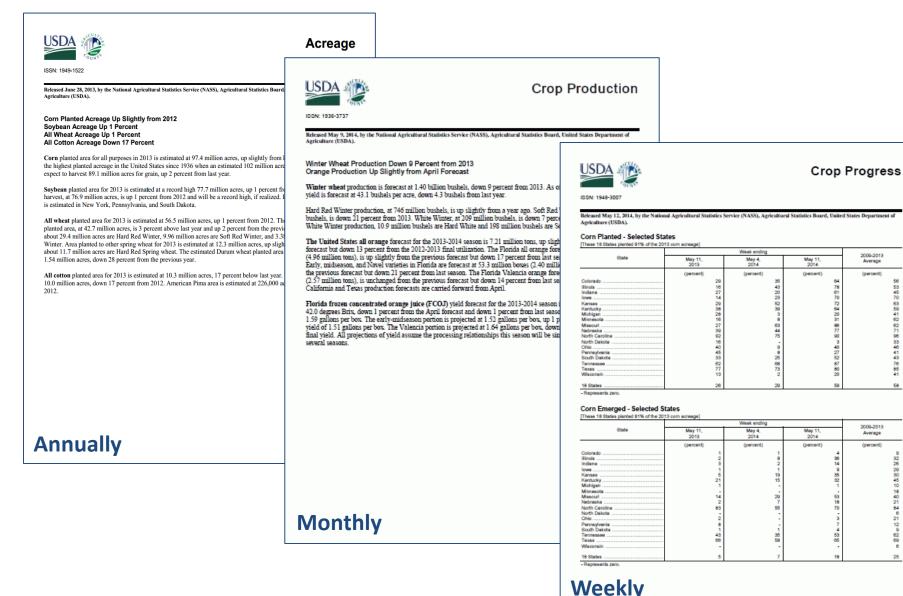
Satellite Remote Sensing Imagery Quality and Timeliness: Considerations for Use in Regional Estimation of Crop Production



**United States Department of Agriculture** National Agricultural Statistics Service David M. Johnson, Geographer BLS Data Quality Workshop, December 1, 2017 dave.johnson@nass.usda.gov

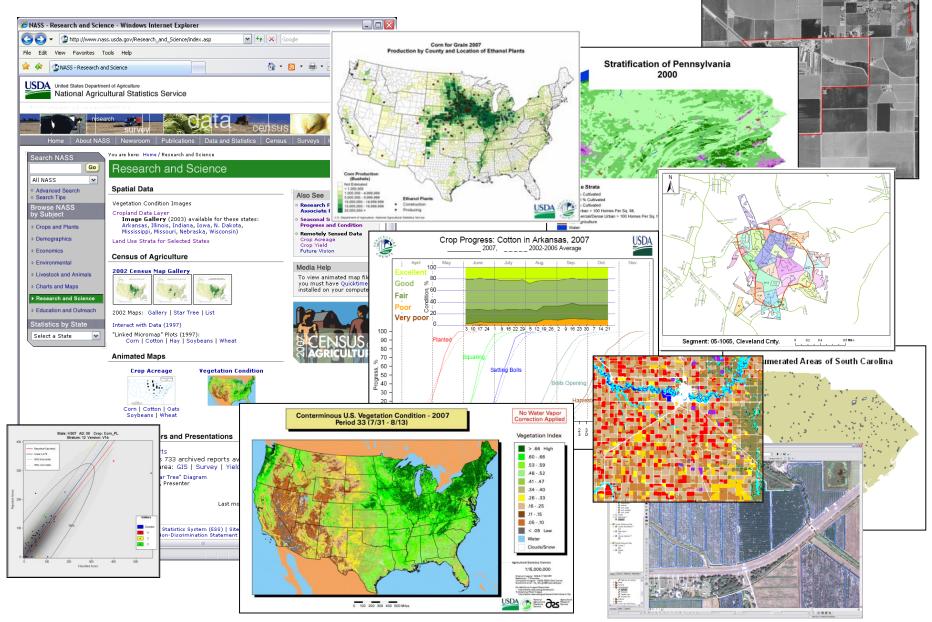


#### Most popular crop reports from NASS

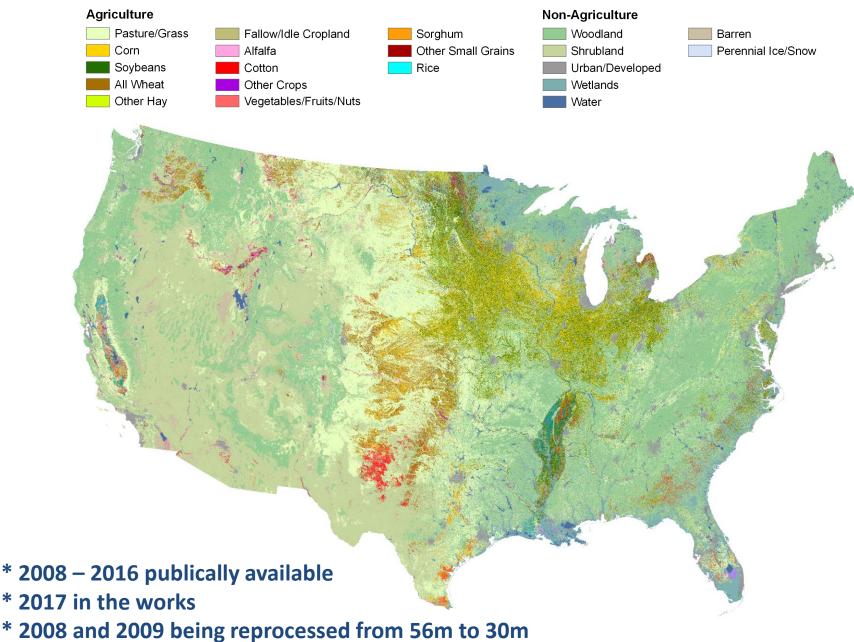


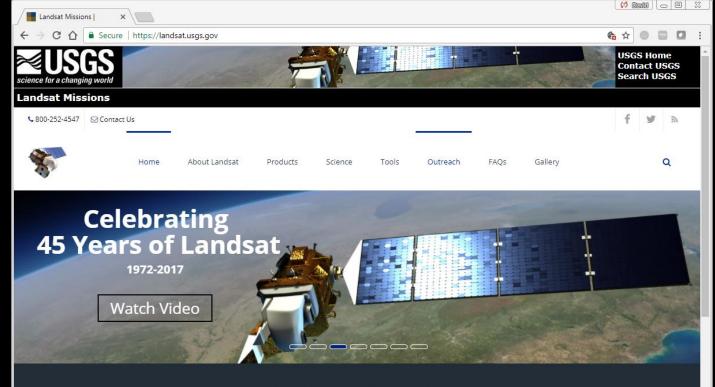
#### **NASS Research and Development Division**

#### **Spatial Analysis Research Section**



## Land cover mapping - Cropland Data Layer (CDL)





*Landsat* represents the world's longest continuously acquired collection of spacebased moderate-resolution land remote sensing data.

Download Data Now!

#### Landsat Headlines



November 29, 2017 -Delivery changes to Band 4 Solar/Sensor Zenith/Azimuth Angle Bands

A recent software release to the Earth Resources Observation and Science (EROS) Center Science Processing Architecture (ESPA) on-demand interface changes the delivery to users of the Band 4 Solar/Sensor zenith/azimuth angle bands.



November 21, 2017 -Landsat Analysis Ready Data for Alaska and Hawaii Available

USGS Landsat Analysis Ready Data (ARD) for Alaska and Hawaii are now available for download from EarthExplorer. This completes the release of U.S. Landsat ARD for all 50

states. (Read More)



November 8, 2017 – New Video Introduces Landsat Analysis Ready Data

A new video introducing Landsat Analysis Ready Data (ARD) has been added to the Landsat ARD webpage, as well as the Landsat Media Library. (Read More)



November 1, 2017 -Upcoming Infrastructure Maintenance

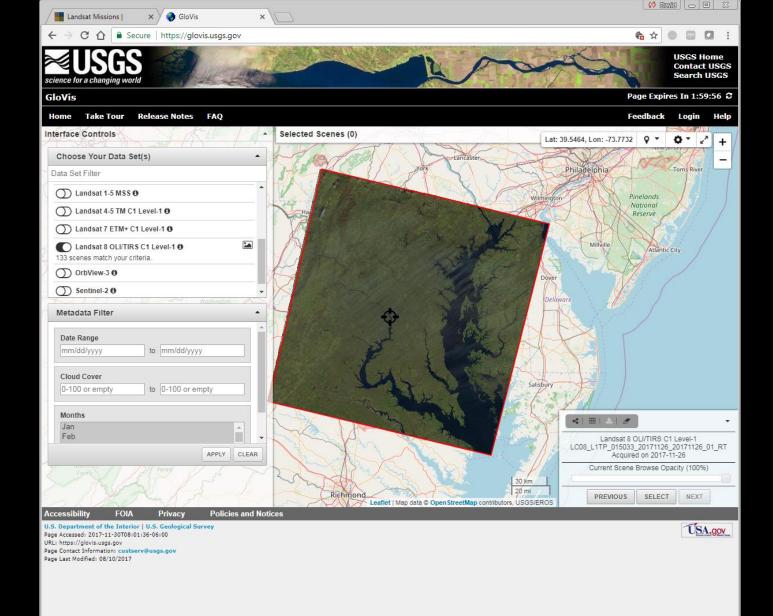
On Tuesday, November 7, 2017, the USGS EROS Center in Sioux Falls, South Dakota will temporarily halt Landsat data processing at 11:00 am CST, and all data distribution from EarthExplorer, GloVis, the LandsatLook Viewer, and ESPA at 3:00 pm CST due to planned required infrastructure maintenance.

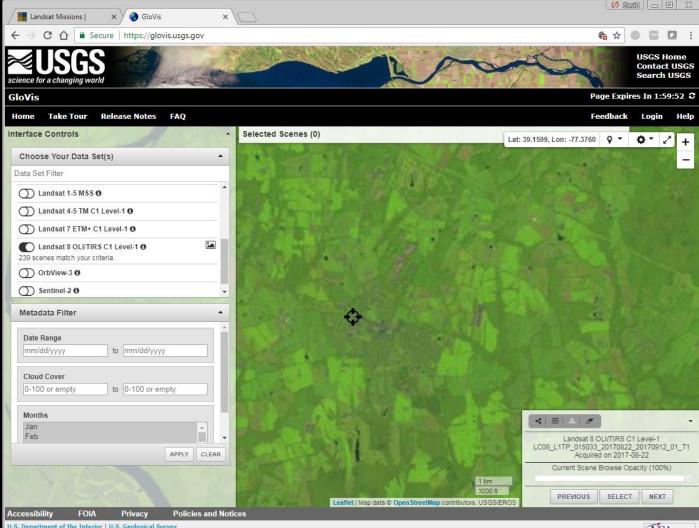




October 30, 2017 - Landsat Analysis Ready Data Available USGS Landsat Analysis Ready Data (ARD) for the conterminous United States are now available for

download from EarthExplorer. (Read More)





U.S. Department of the Interior | U.S. Geological Survey Page Accessed: 2017-11-30T08:01:36-06:00

URL: https://glovis.usgs.gov

Page Contact Information: custserv@usgs.gov

Page Last Modified: 08/10/2017

USA.gov

#### 2017: June 16 – 22

#### Agricultural areas

Landsat 8

**Sentinel 2a** 



**DMC** Deimos

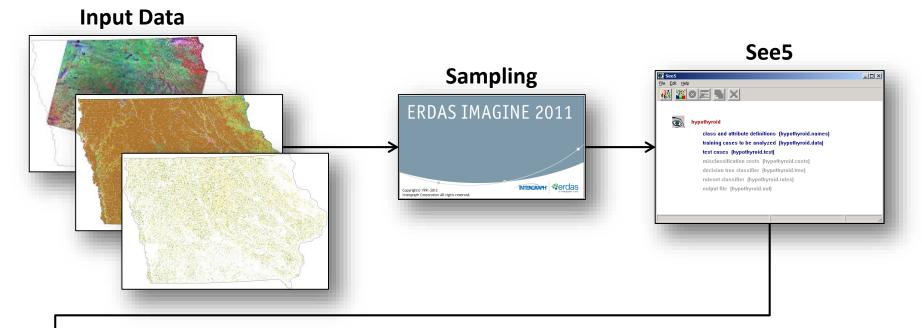
DMC UK2

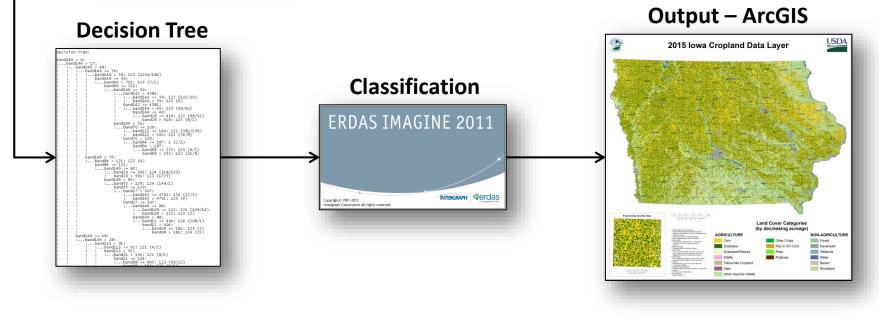
Resourcesat-2 LISS3

### Available satellite imagery 2017: June 16 – 22



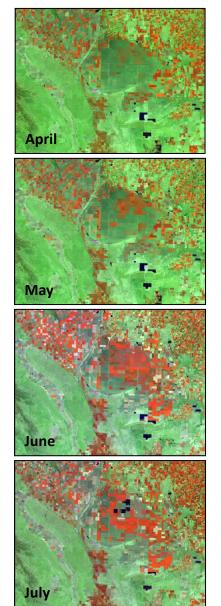
### **CDL Processing Flow**

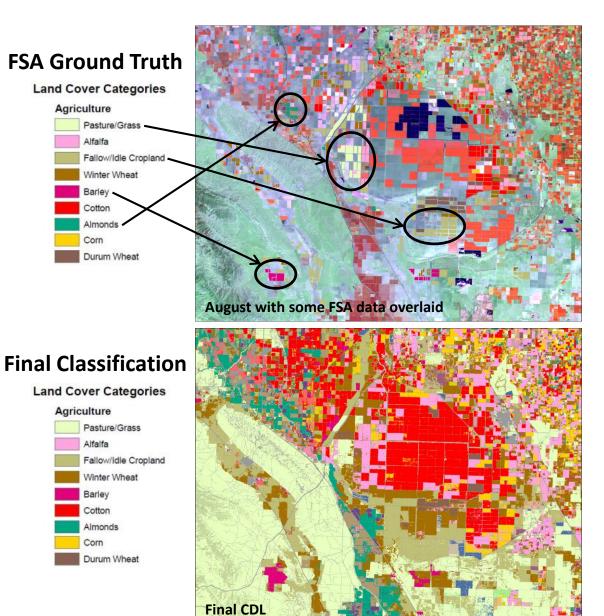




### Classification

#### **False Color IR Imagery**





#### **Accuracy Assessments**

USDA, National Agricultural Statistics Service, 2014 Colorado Cropland Data Layer STATEWIDE AGRICULTURAL ACCURACY REPORT

Crop-specific covers o	only *Correct	Accuracy	Error	Kappa
OVERALL ACCURACY**	2,630,488	85.5%	14.5%	0.812

Cover	Attribute	*Correct	Producer's	Omission		User's	Commission	Cond'l
Туре	Code	Pixels	Accuracy	Error	Kappa	Accuracy	Error	Kappa
Corn	1	419737	90.76%	9.24%	0.895	90.22%	9.78%	0.889
Sorghum	4	83214	62.32%	37.68%	0.611	64.72%	35.28%	0.635
Soybeans	5	1058	43.25%	56.75%	0.432	72.47%	27.53%	0.724
Sunflower	6	5760	39.64%	60.36%	0.395	70.61%	29.39%	0.705
Barley	21	7176	71.52%	28.48%	0.715	81.00%	19.00%	0.810
Winter Wheat	24	1100020	93.26%	6.74%	0.905	94.21%	5.79%	0.918
Millet	29	75109	67.86%	32.14%	0.671	76.85%	23.15%	0.762
Alfalfa	36	196153	89.75%	10.25%	0.891	85.60%	14.40%	0.848
Other Hay/Non Alfalfa	37	84626	63.33%	36.67%	0.624	85.92%	14.08%	0.854
Sugarbeets	41	4679	63.13%	36.87%	0.631	90.28%	9.72%	0.903
Dry Beans	42	9406	62.72%	37.28%	0.626	69.54%	30.46%	0.694
Potatoes	43	6104	89.74%	10.26%	0.897	93.79%	6.21%	0.938
Fallow/Idle Cropland	61	625989	88.08%	11.92%	0.855	89.23%	10.77%	0.869

#### **Classified area vs June enumerated**

	ν manγ acres are inside this blue tract b	D - CROPS AND LAN oundary drawn on the photo (ma	p)?	4 17 -
	v Iwould like to ask about each field ins FIELD NUMBER Total acresin field	ide this blue tract boundary and i 01 02 628 828 •	ts use during 2000. 03 04	28
<u>3.</u> 4.	Crop or land use. [Specify] Occupied farmstead or dwelling Weste, unoccupied dwellings, buildings and structures, roads, ditches, etc.	843 		
a 🛃 🕹	Woodand Padure Permanent (not in coprotation) Cropland (used only for padure	856 856	842 842 84	42 56
8.	Ide gopjand - Ide all during 2000 Two cropsplartied in this teld or two uses of the s cop. [Specify second crop or u A	ise] 844	No DYes DNo DYes No 844 844 844 84	
<b></b> 11.	Acresite to be planted Acresitingsted and to be intigeted [/f double cropp include acreage of each crop ingated] Winter Wheat (include acvert crop)	540 540	620 620 62 540 540 54	
17.	(include cover crop) For grain or seed Ryce (include cover crop) (Exclude rycgrass) For grain or seed	541 541 547 547 548 548	• • •	41 47 48
		1		
ALC:		V.		
Cover	X Classified	Y Enumerated	<b>.</b> .	
Туре	Acres	Acres		A. 70
Corn	637.0	640.0		
Soybean	1.0	0.0		
Wheat	0.0	0.0		
Alfalfa	0.0	0.0	8	<b>•</b>
				10
Non-Ag.	2.0	0.0		

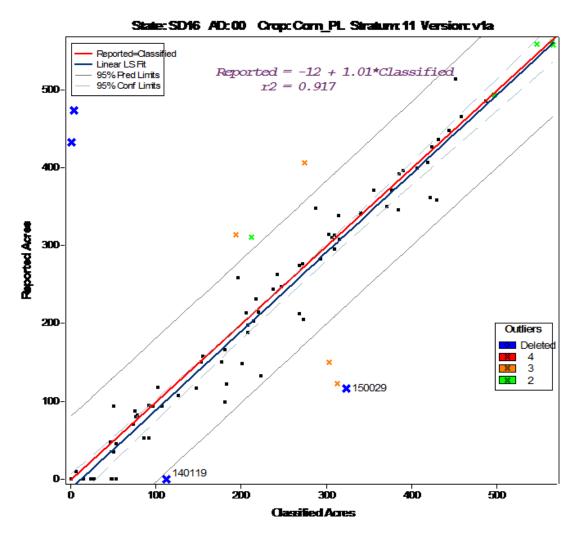
#### At three sites

							Cover	X Classified	Y Enumerated			<b>.</b>	
							Туре	Acres	Acres				-
							Corn	176.0	190.0		l	a d <sup>al</sup>	
ł,						<b>1</b>	Soybean	302.0	290.0	-	$\geq$		
			<u></u>				Wheat	0.0	0.0		l es		i  1
							Alfalfa	3.5	0.0	7			-
-										a∦ !			
							Non-Ag.	158.5	160.0				
		X	Y										j.
4	Cover Type	Classified	Enumerated						ÇA 🖸			12	
5		Acres	Acres		A Presi								2
5	Corn	34.0 177.0	21.0 155.0			<b>.</b>							. 2017 1 1 1
	Soybean Wheat	4.5	0.0	and and a						<u>- </u>		. <u>La "</u>	É i
1 1	Alfalfa	2.5	0.0					X	Y				4
				s PE			Cover	^ Classified	T Enumerated	26		S.C.	
L)							Туре	Acres	Acres				
	Non-Ag.	422.0	464.0				Corn	637.0	640.0				
2							Soybean	1.0	0.0				
e e		-11 a. Triba				$\mathbf{i}$	Wheat	0.0	0.0				
							Alfalfa	0.0	0.0	8			
												4	
į							Non-Ag.	2.0	0.0	1		2 P. 🖕	

### 10 sites and so forth....



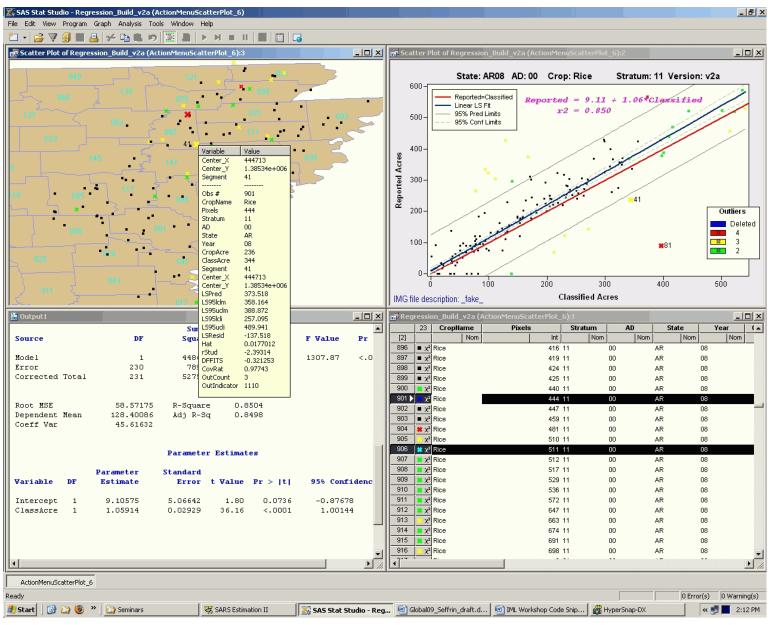
#### **Acreage Regression Estimation**



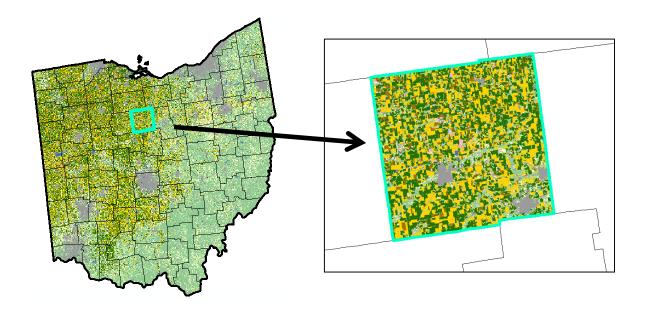
IMG file description: \_oct\_subset\_30m\_

We don't just "pixel count" from CDL to estimate acreage

#### **SAS-based Regression Estimate system**



#### à la Bob Seffrin



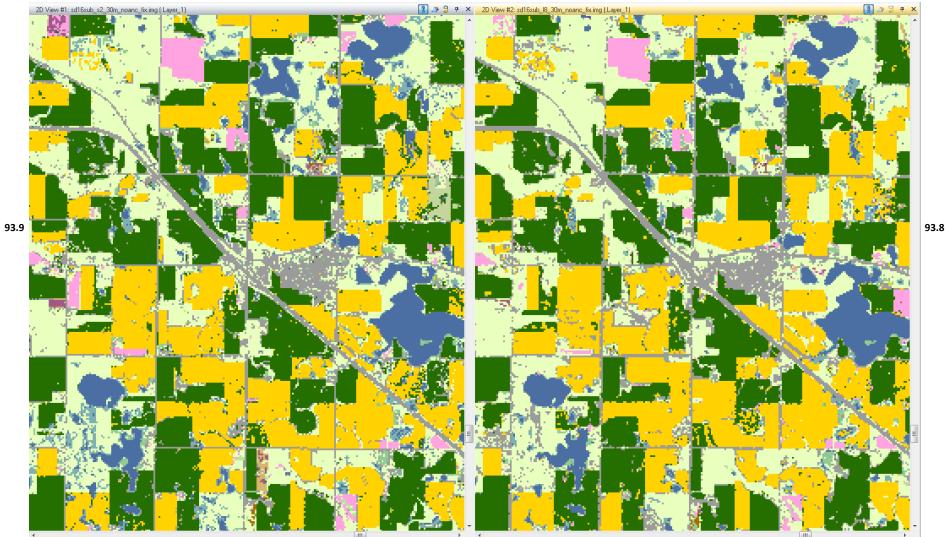
### **County Estimates**

- Use Battese-Fuller estimator with nested design
- Apply state-strata level regression parameters
- Adjust intercept based on segments in county
- Ag Statistics Districts Est = Sum of County Estimates

#### **Classification comparison #1**

#### 30m Sentinel-2a

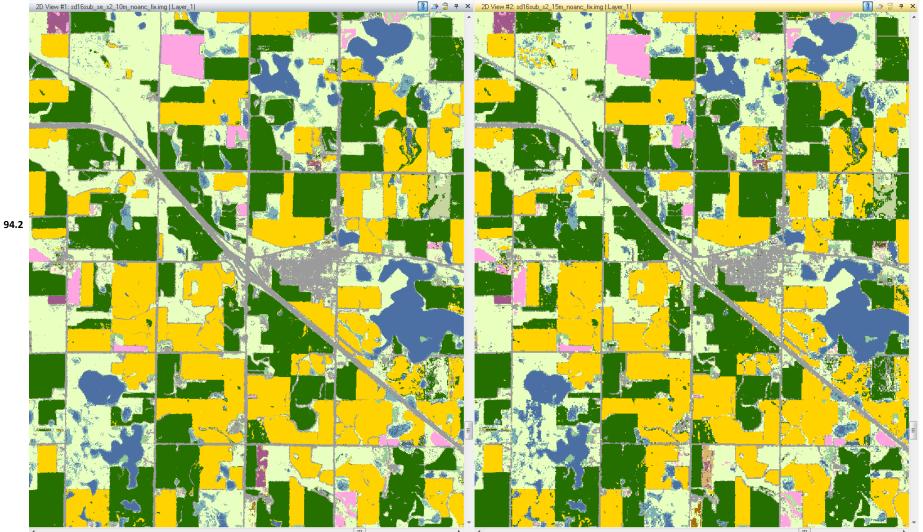
30m Landsat 8



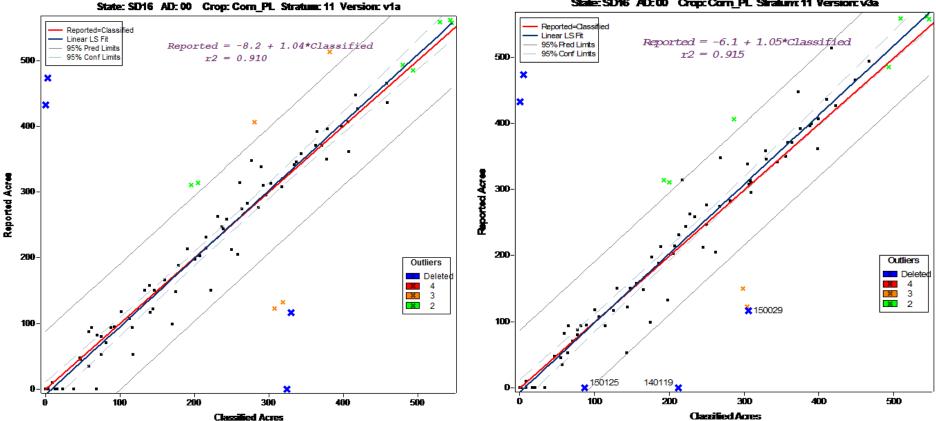
#### **Classification comparison #2**

#### 10m Sentinel-2a

#### 15m Sentinel-2a



#### 60m vs 15m regression analysis - corn

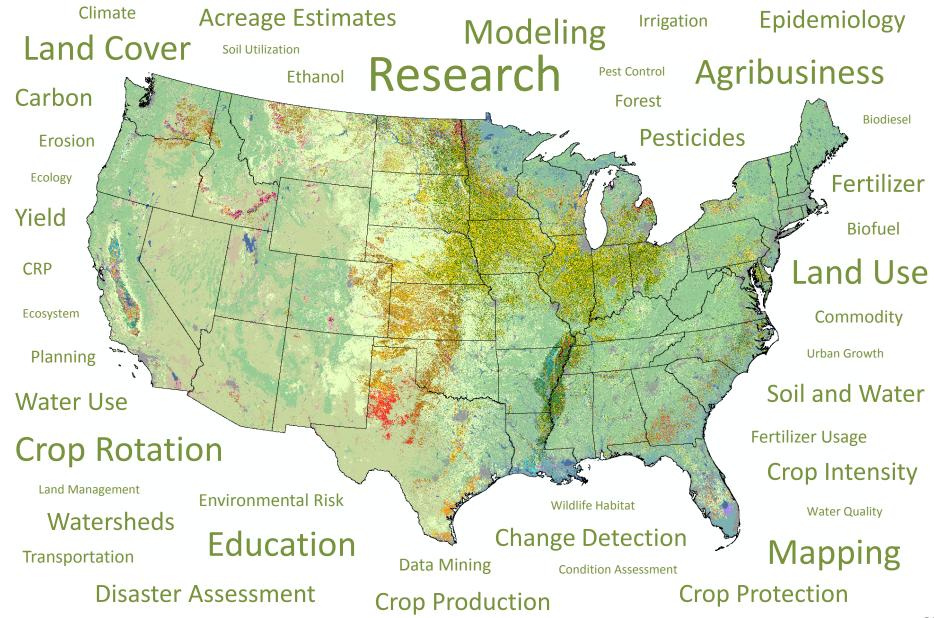


State: SD16 AD:00 Crop: Corn PL Stratum: 11 Version: v3a

IMG file description: 60m fix

VIG file description: \_sub\_s2\_15m\_noan

### **CDL Applications**







A weather system that brought rain to Tehran and Iran's Caspian Sea coastline and Alborz mountains helped raise dust inland in late November, 2017...

Continue Reading

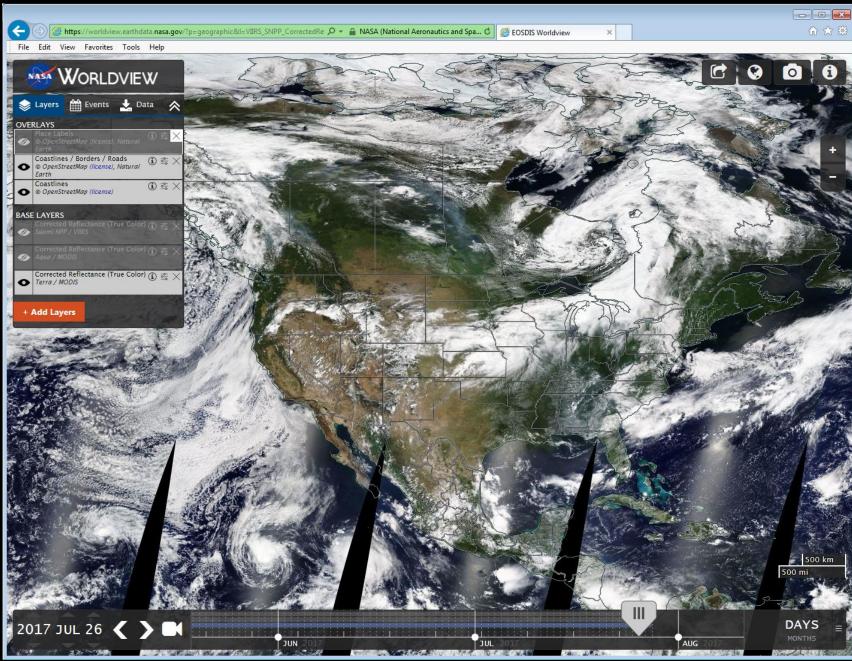


#### Disciplinary Teams

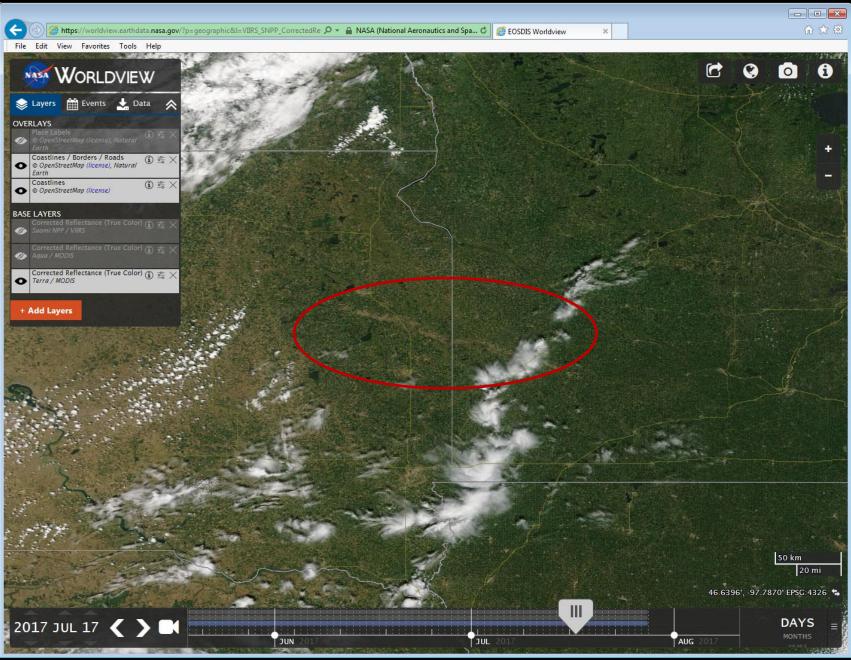


Learn More About Today's Image

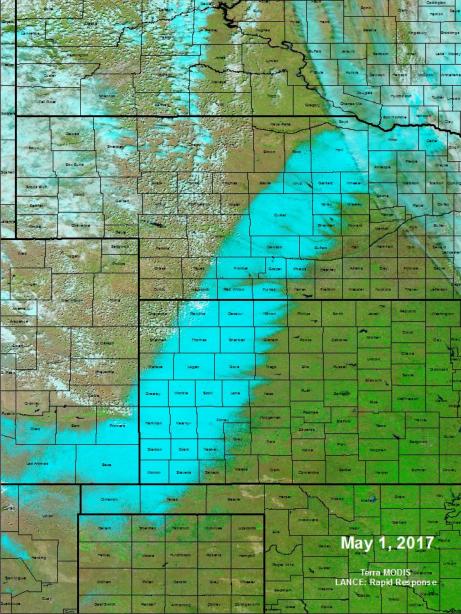
### https://worldview.earthdata.nasa.gov/

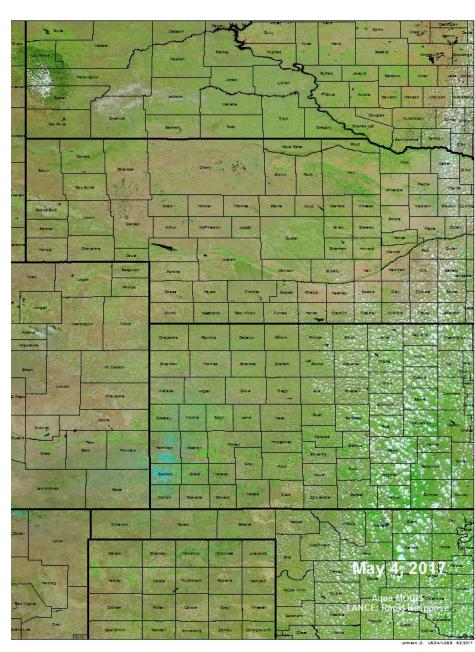


#### Hail example



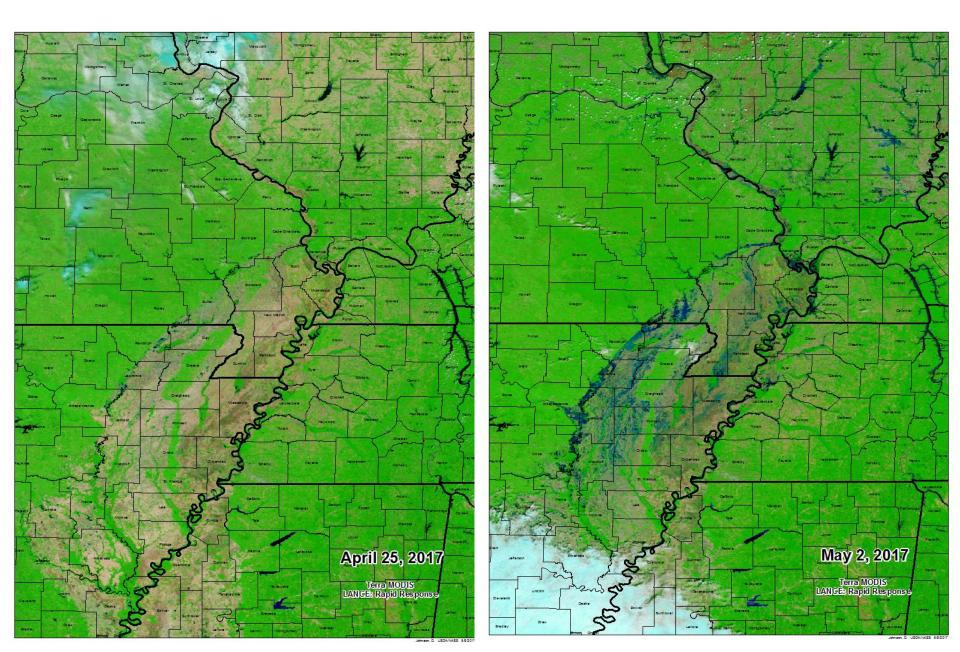
### **MODIS Imagery - Snow event**



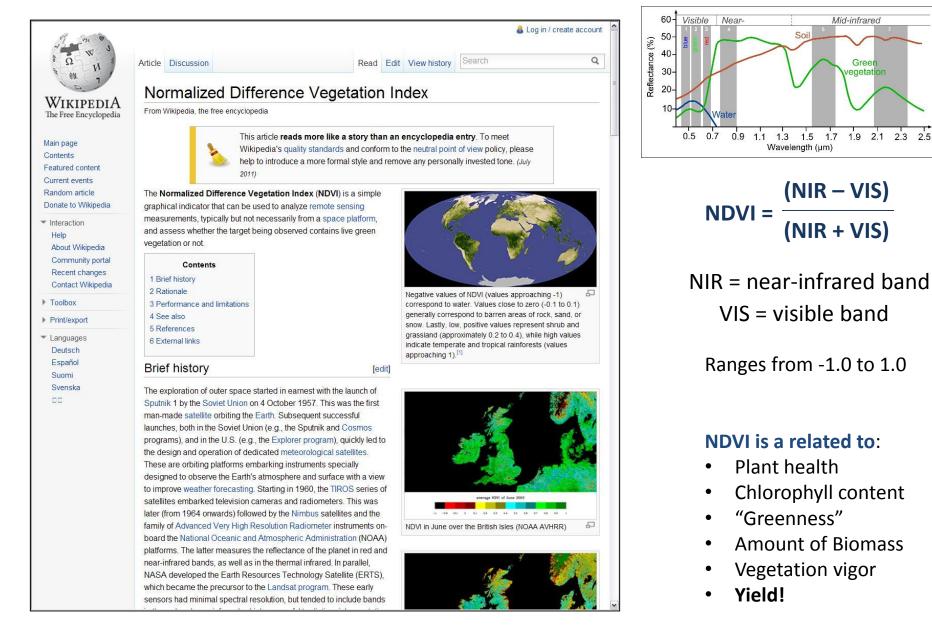


Johnson, D. USDA/NASS 5/2/2017

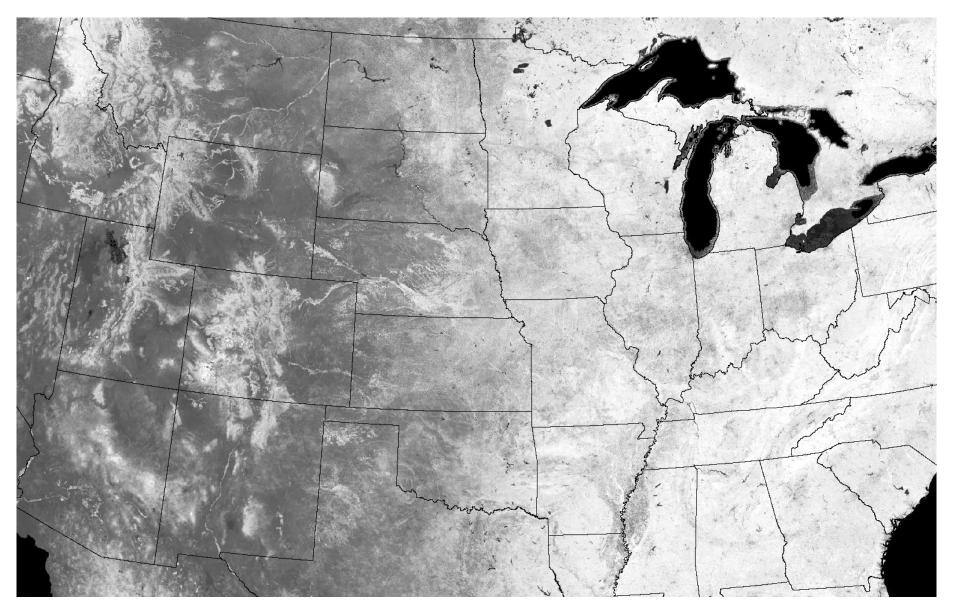
### **MODIS Imagery - Flood event**



### **Calculation and use of NDVI**

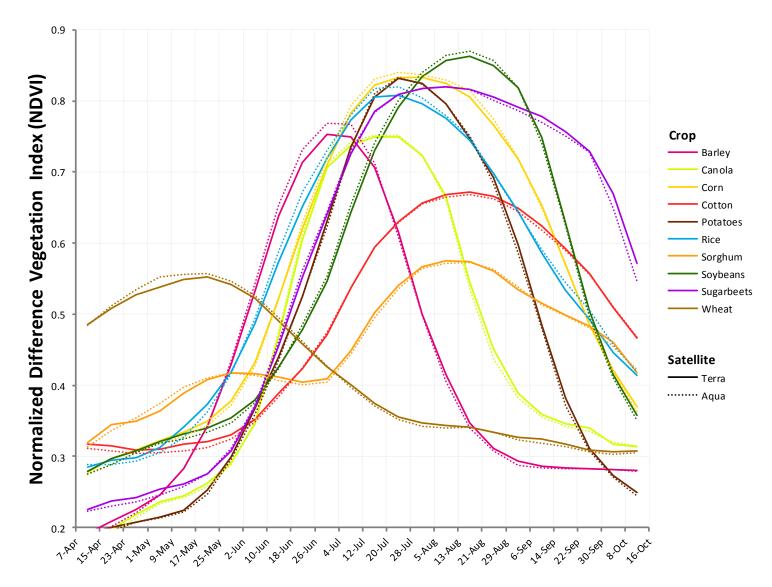


### **MODIS NDVI 8-day composite imagery example**



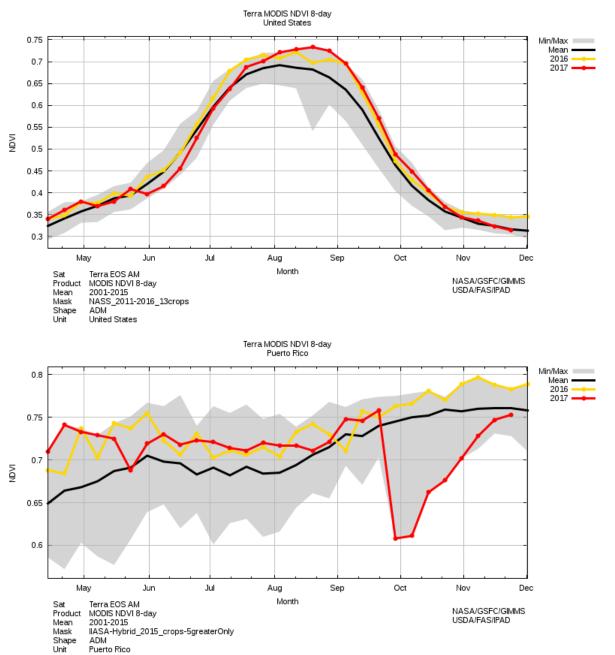
Lighter shades, greater NDVI

#### **Average NDVI phenologies over United States**



#### Signals isolated using crop specific masks

#### **Real-time tracking of NDVI**



### NASS recent efforts on remote sensing of crop yields

#### Premise

- Positive relationship between crop yield and biomass – plant vigor - "greenness" - NDVI
- Negative relationship between crop yield and land surface temperature

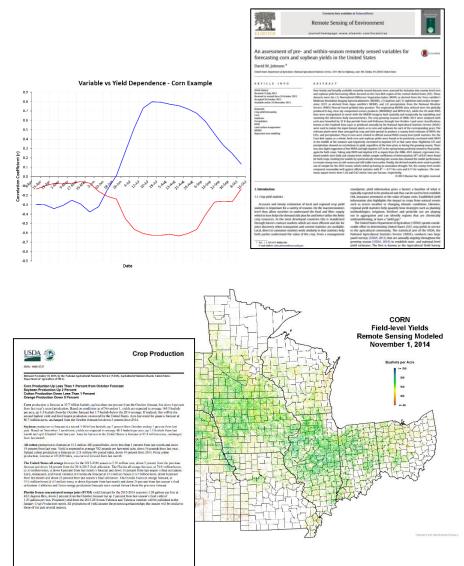
#### Utilize time-series MODIS satellite data to obtain biomass and temperature estimates throughout the growing season

- Use Cropland Data Layer to isolate known crop areas
- Then use them in an empirically-based prediction model based on historical imagery and NASS county-level yield statistics

# Run model at National, State, and County levels

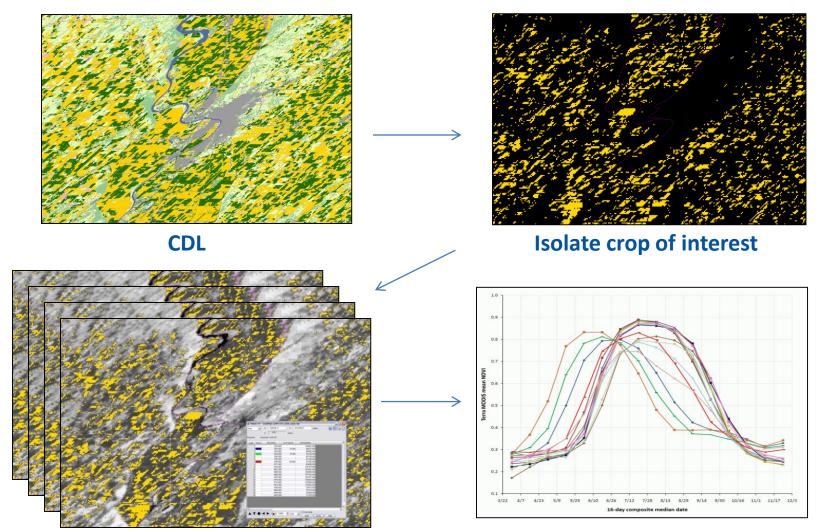
- Integrating over season approach
- Using decision trees (Rulequest Cubist)
- Corn and Soybeans operational currently

# Perform within crop season at monthly intervals



#### Must be timely, accurate, and useful

# Intersecting of crop "mask" with time-series of MODIS data

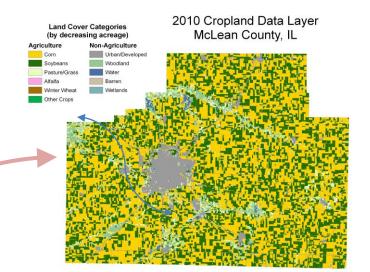


Intersect crop mask with MODIS time series and then spatially average those pixels

### **County-level time-series database has been built** 2006 -> present

Every eight day "window" through the growing season

- Observed average value of NDVI
- Observed average value of LST

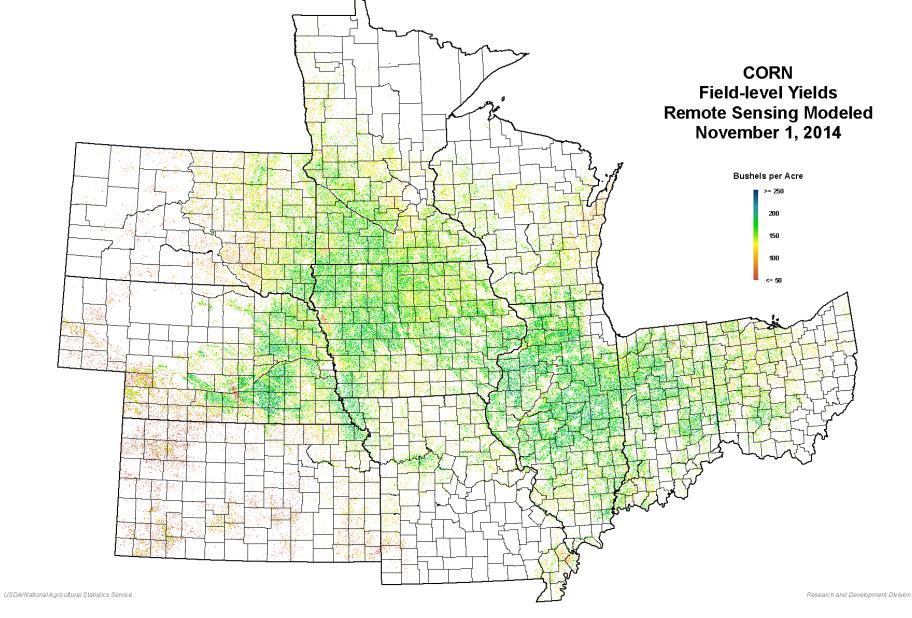


#### For every county we also know

NASS published yield

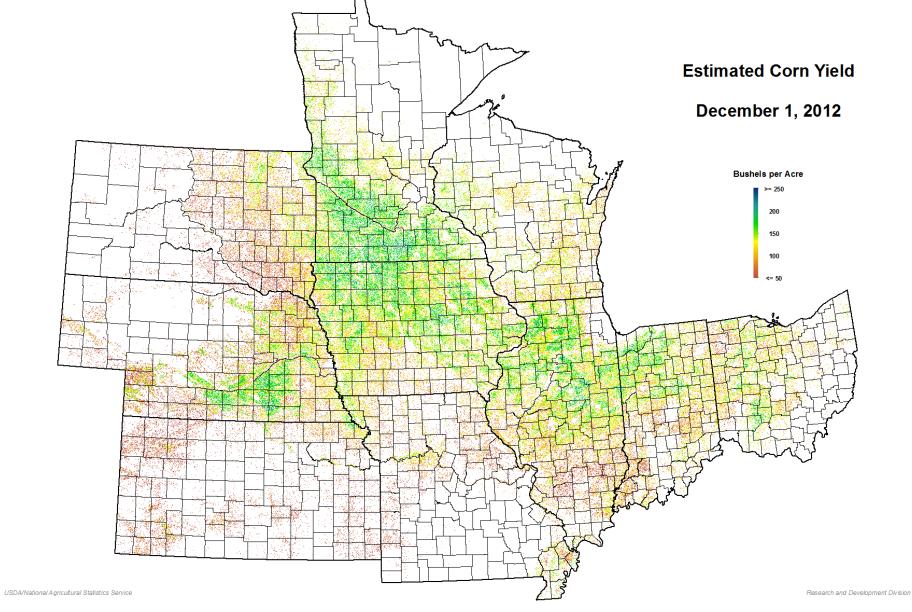
state	county	year	yield	NDVI	DLST	23-A <mark>p</mark> r	1-May	9-May	17-May	25-May	2-Jun	10-Jun	18-Jun	26-Jun	4-Jul	12-Jul	20-Ju
17	93	2010	168.6	$\sim$		2980.6 <mark>7</mark> 3	3264.547	3862.318	3866.409	3990.153	5893.765	7558,422	8108.906	8559.559	8826.817	8930.85	8935.61
17	95	2010	159.9	$\sim$	~~~~~	2992.195	2977.046	3203.401	3377.778	3848.078	5771.909	7450.655	8012.679	8310.476	8537.081	8645.006	8702.09
17	97	2010	?	$\sim$	$\sim$	3337 <mark>.</mark> 864	3836.186	5403.508	4459.949	4311.525	5635.508	7575.966	7776.288	8082.441	8575.966	8689.22	8524.83
17	99	2010	162.9	$\sim$		2844.2	2971.099	3332.757	3486.144	3937.887	6032.626	8059.657	8344.285	8565.456	8784.721	8903.283	8939.18
17	101	2010	156.9	$\sim$	$\sim\sim\sim$	3874.741	3988.952	4349.537	4930.498	5452.197	6205.825	7687.592	8022.794	8396.337	8589.467	8770.097	8827.56
17	103	2010	173.6	$\sim$		2782.144	2805.154	3018.966	3094.823	3512,443	5563.73	7306.284	8102.134	8554.694	8918.413	8949.941	8862.48
17	105	2010	166.6	$\sim$	1 m	2816.087	3007.861	3371.65	3608.025	3988.44	6100.426	8236.687	8503.526	8751.071	8872.207	8945.958	8805.10
17	107	2010	155.7	$\sim$	han	2706.578	2846.325	3249.914	3622.797	4402.778	6661.937	8254.186	8426.524	8515.872	8656.284	8755.741	8782.62
17	109	2010	141.8	$\sim$	$\wedge$	3104.659	3240.878	3479.558	3558.215	4056.188	6151.458	7440.174	7913.743	8068.891	8437.704	8697.558	8783.25
17	111	2010	171	$\sim$	~~	3079.982	3283.63	3659.725	3626.941	3909.787	5483.311	7163.395	8116.372	8578.756	8844.542	8908.066	8886.72
17	113	2010	169.5		$\wedge \dots \wedge$	2727.7	2899.42	3316.302	3573.485	4053.523	6484.888	8304.741	8585.89	8846.408	8960.349	8953.985	8860.08
17	115	2010	168.5	$\sim$	1	2791.229	2943.968	3442.016	3862.389	4880.537	7264.473	8749.776	8793.049	8769.201	8787.036	8828.984	8797.02
17	117	2010	146.9	$\sim$	$\sim\sim\sim$	3213.265	3342.063	3617,414	4030.15	4799.472	6474.308	8021.675	8511.256	8646.314	8885.395	8920.521	9010.37
17	119	2010	166.3	$\sim$	/~~~v	3282.816	3405.388	3712.914	4087.371	4860.512	6581.844	8129.837	8410.098	8587.161	8826.583	8879.477	8854.28
17	121	2010	149.7	$\sim$	$\sim\sim\sim$	3524.353	3534.901	3931.625	4260.018	4856.34	6251.207	7828.069	8242.445	8450.967	8601.465	8853.656	8952.45
17	123	2010	163.3	$\frown$	~~ <u> </u>	2748.186	2881.69	3261.708	3448.857	3851.546	5647.397	7541.326	8205.306	8567.037	8799.09	8891.405	สจิศิ4.80

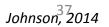
#### Map Output



#### Normal year

#### **Map Output**

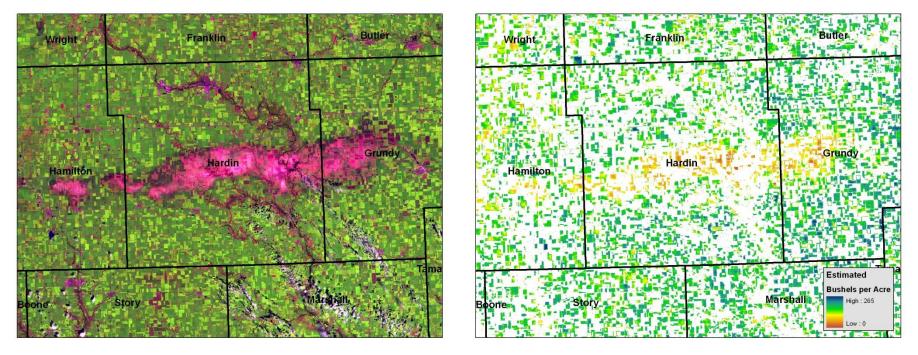




#### **Drought year**

# Localized example of yield map variability

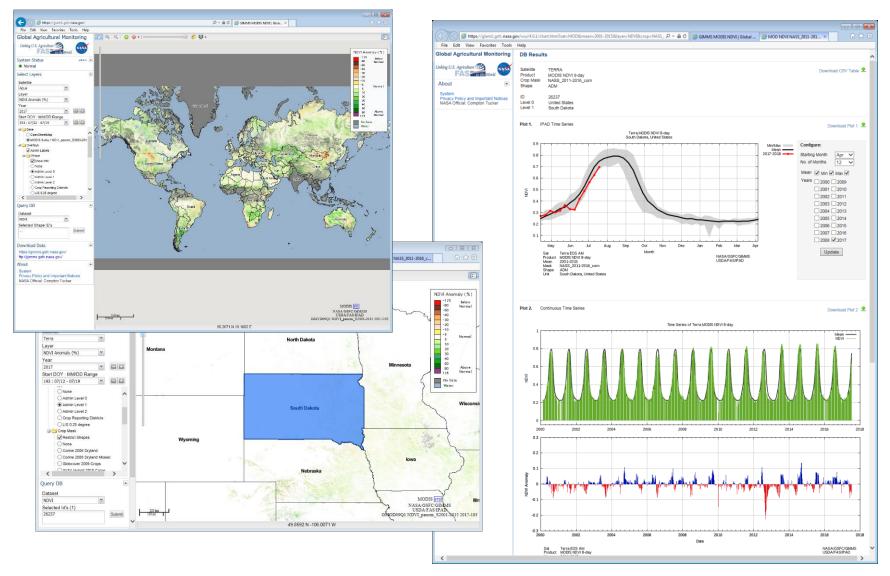
#### Scene of a large hailstorm



Landsat image

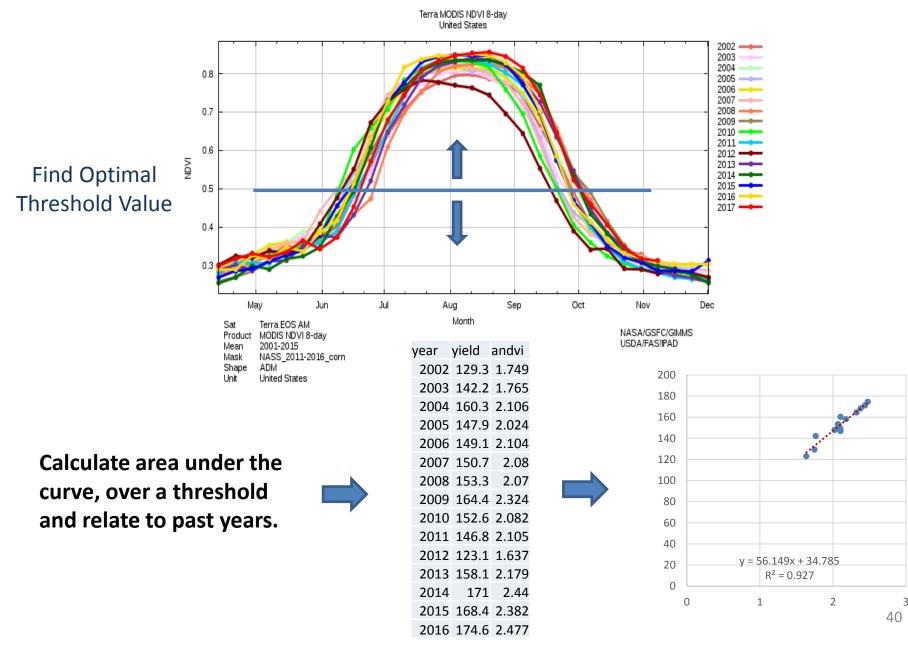
Modeled yields from MODIS

#### USDA Foreign Agricultural Service/NASA GLAM https://glam1.gsfc.nasa.gov/

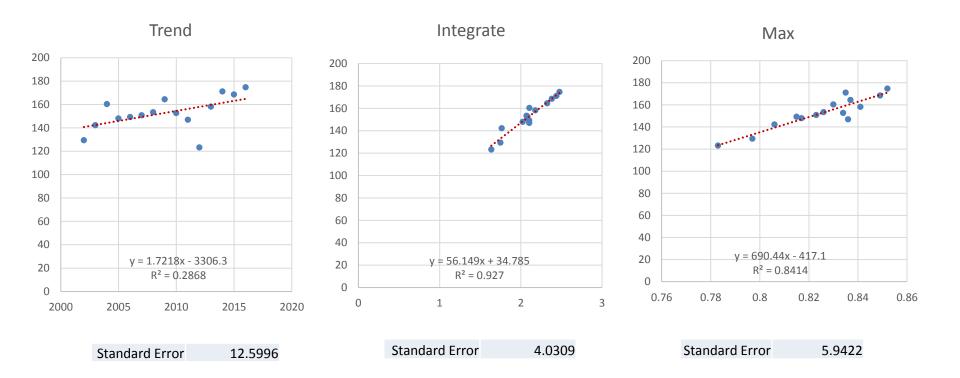


Highly already customized tool for time series analysis and display <sup>39</sup>

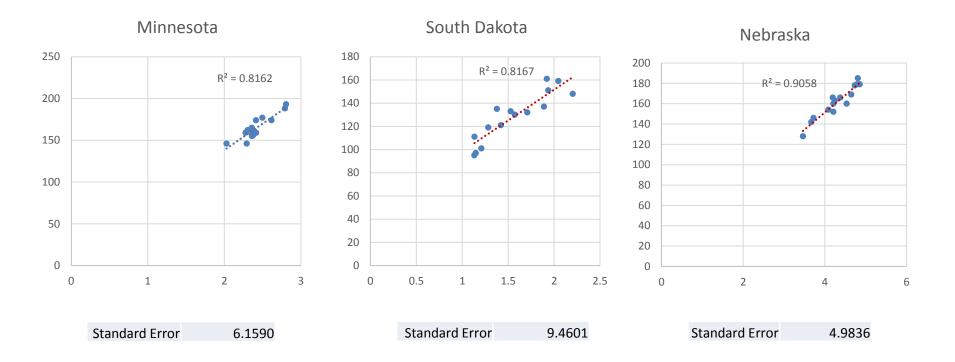
## Also, shifting to a simpler model construction



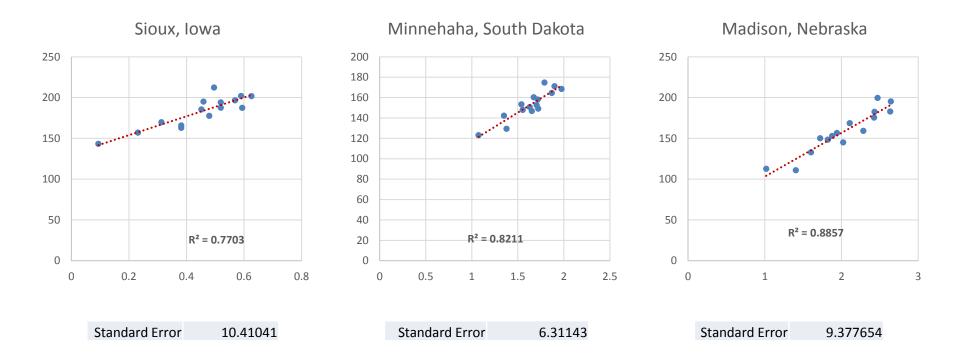
## USA national-level simplistic corn yield model



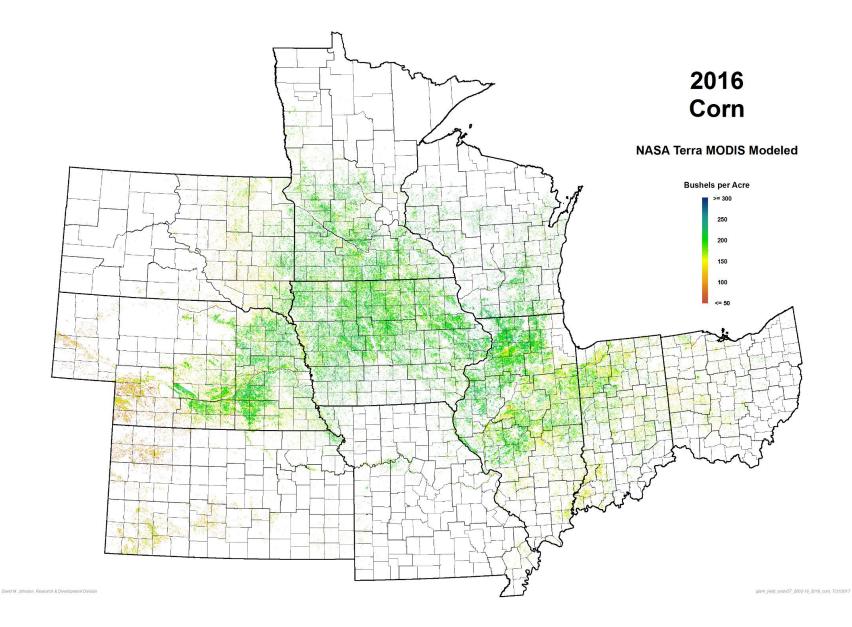
### **State-level simplistic yield modeling**



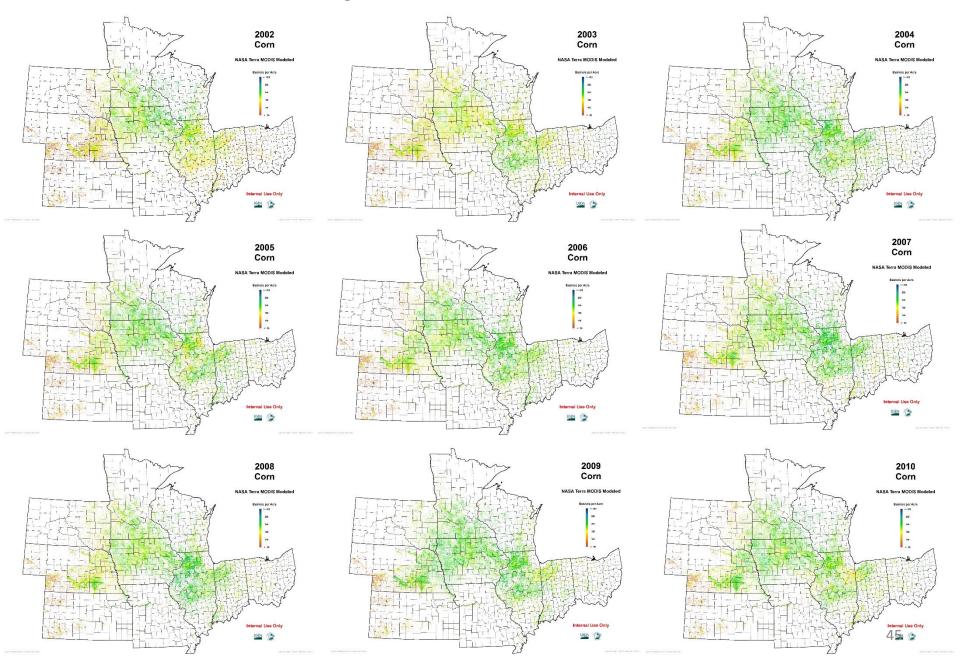
## **County-level simplistic yield modeling**



### Map output still possible



#### And easy to create time series...



# Summary of Remote Sensing for Crop Production Estimation

#### • Strengths

- Good areal coverage
- Solid temporal coverage
- Many free data sources
- Better sensors on the way
- Little data latency
- Fine spatial detail
- Simple statistical models seem to be as good as complicated ones
- Cheap computing and analytics has been a boon

#### Weaknesses

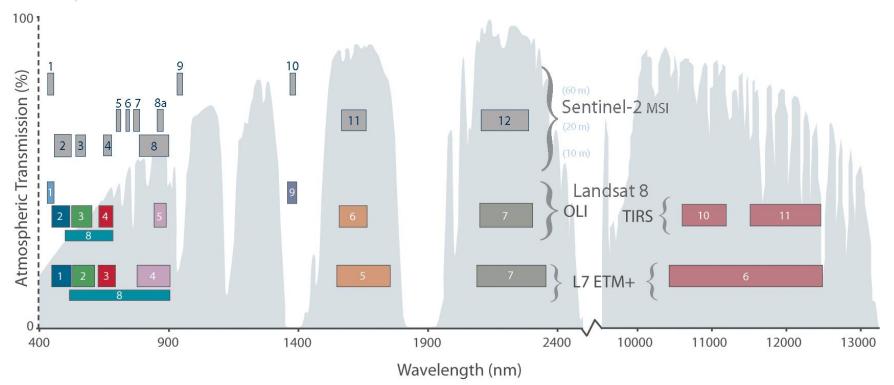
- Computationally intensive
- Integrative skill set required
- Calibration of datasets always ongoing
- Measurement uncertainties difficult to quantify
- A variety of noise sources are present
- No long-term history
- In situ validation lacking
- Past utility was oversold



David M. Johnson, Geographer BLS Data Quality Workshop, December 1, 2017 dave.johnson@nass.usda.gov



## Sentinel-2 vs Landsat 7 & 8 spectral bands



Comparison of Landsat 7 and 8 bands with Sentinel-2

# Acreage estimate for a crop in stratum h

$$\hat{y}_h = N_h [\bar{y}_h + b_h (\bar{X}_h - \bar{x}_h)]$$

 $N_h = Number \ of \ frame \ units(segments \ in \ frame)$ 

- $\bar{y}_h = sample$  mean per segment of reported acres of crop cover
- $b_h = Slope \ of \ the \ regression \ of \ acres \ in \ segment \ on \ pixel(acres)$
- $\overline{X}_h = population mean pixels(acres)in segment$
- $\bar{x}_h = sample mean pixels(acres) in segment$

### Estimate of county total for a crop, stratum

$$\hat{T}_{(BF)hc.} = N_{hc} \Big[ \hat{\beta}_{0h} + \hat{\beta}_{1h} \overline{x}_{hc} + \delta_{hc} \overline{u}_{hc.} \Big]$$

$$\overline{u}_{hc.} = \overline{y}_{hc.} - \hat{\beta}_{0h} - \hat{\beta}_{1h} \overline{x}_{hc.} \qquad \text{(residual)}$$

1) if  $\sigma^2_{\text{within}} = 0$ , use  $\delta = 1$ , 2) if  $\sigma^2_{\text{between}} = 0$ , use  $\delta = 0$ , 3) if < 2 segments use  $\delta = 0$ , 4) if  $\sigma^2_{\text{within}} = 1.0$ , use  $\delta = 0$ , 5) otherwise use  $\delta = \gamma$