MODIS-based Modeling of Corn and Soybean Yields in the US

American Geophysical Union Fall Meeting - San Francisco - 13 December 2013



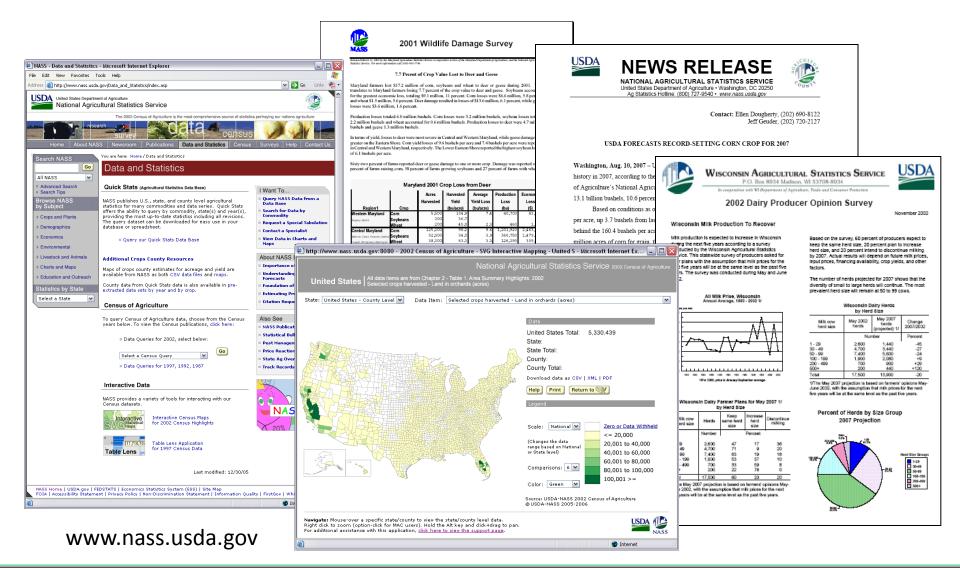


David M. Johnson Geographer United States Department of Agriculture National Agricultural Statistics Service



National Agricultural Statistics Service (NASS)

Provider of timely, accurate, and useful statistics in service to U.S. agriculture

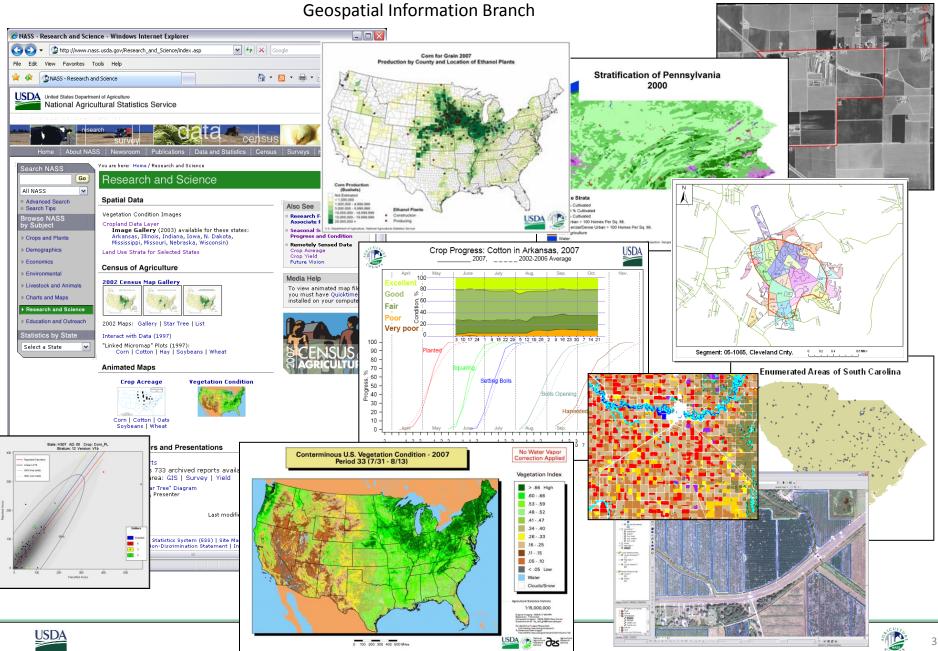




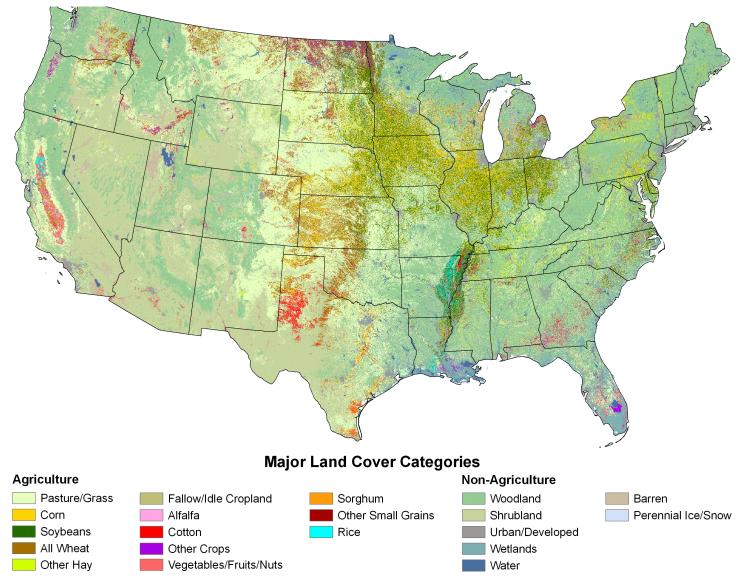
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NASS Research and Development Division



Annually derived Cropland Data Layer (CDL)

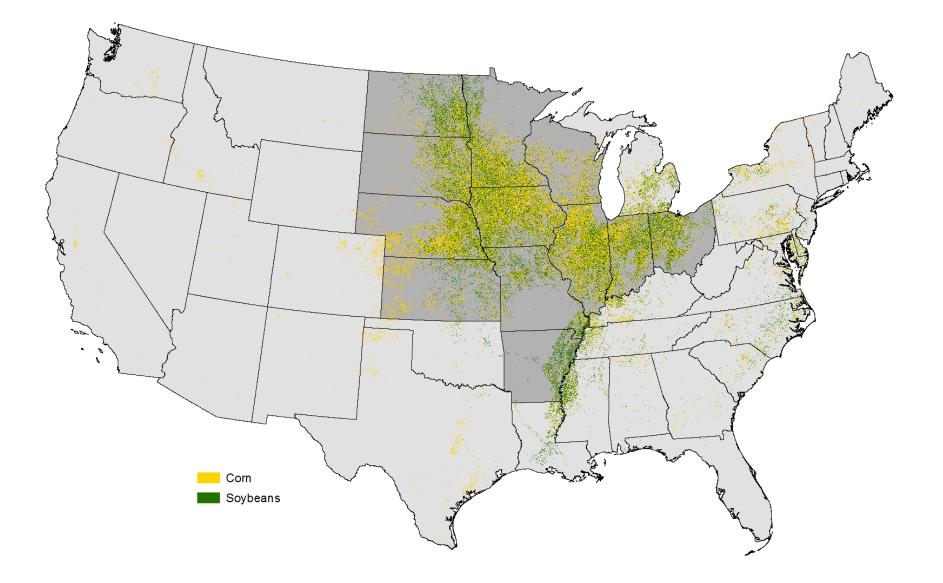


Freely available over Internet via "Cropscape"

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Region with the bulk of corn and soybean production

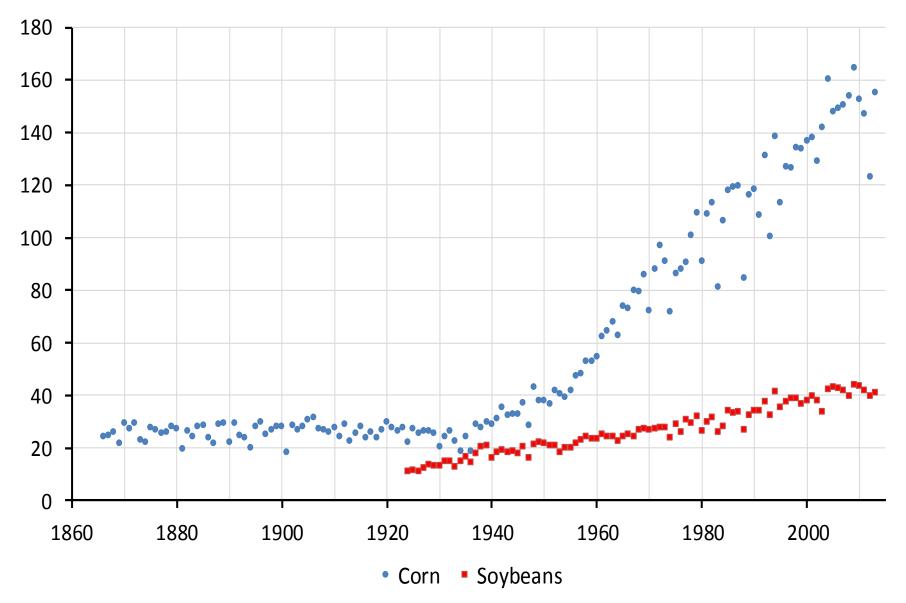








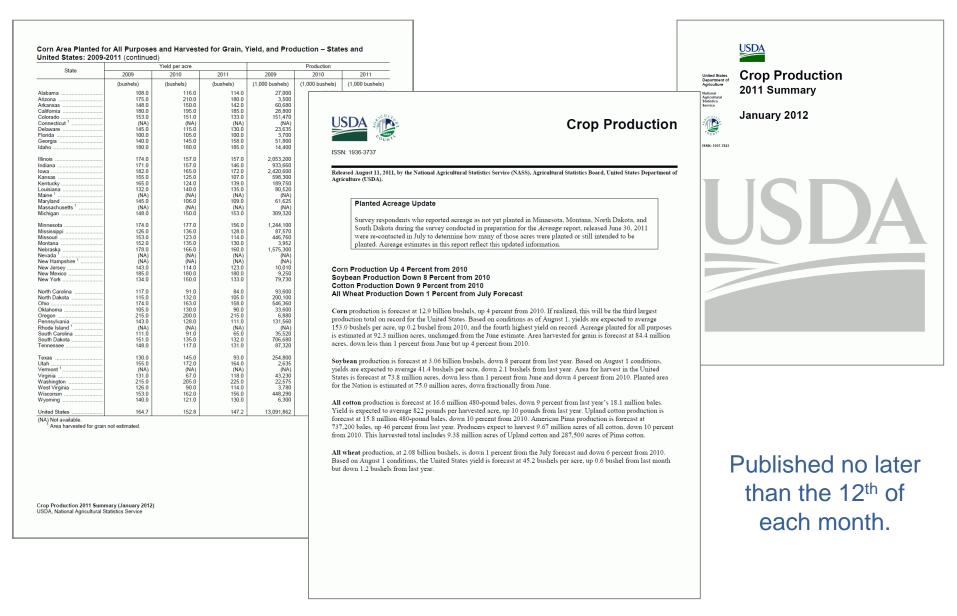
United States Yield (bushels/acre)







NASS Crop Production reports





Yields results primarily derived from two surveys

Agricultural Yield

- Farmer reported survey data of expected crop yields.
- Data obtained throughout the growing season.
- Conducted in all states except Alaska and Hawaii.
- Sample size in the 1000s per state.
- Farm operator contacts are selected from the March Crops/Stocks survey (small grains) and the June Crops/Stocks survey (late season crops and tobacco).
- Primarily telephone based.

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Objective yield

- Corn, Cotton, Soybeans, Wheat, Potatoes.
- Only done in states where the commodities are primarily found.
- Samples selected from areas found in June Area Survey ("Acreage").
- Performed at 100s of sample sites per state.
- Biophysical plant/seed measurements obtained.
- Each plot revisited a few times per season.

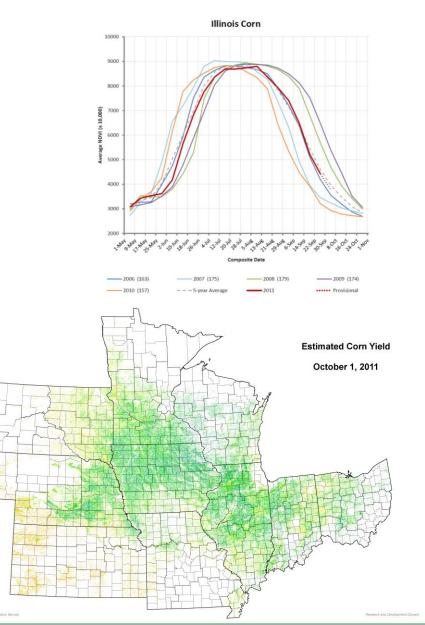




Remote Sensing for Yield Estimation

A third method...

- Premise
 - There is a known relationship between crop
 - Biomass, vigor, "greenness", NDVI
 - and
 - Crop yield
 - Also temperature and rainfall too.
- Utilize MODIS data to obtain biomass and temperature variables
- Utilize Nexrad Ground radar to estimate precipitation
- Produce for national, state, ASD, and county
 - Corn and soybeans only
 - "Speculative" region only
 - i.e. Corn Belt
- Be independent of other methods



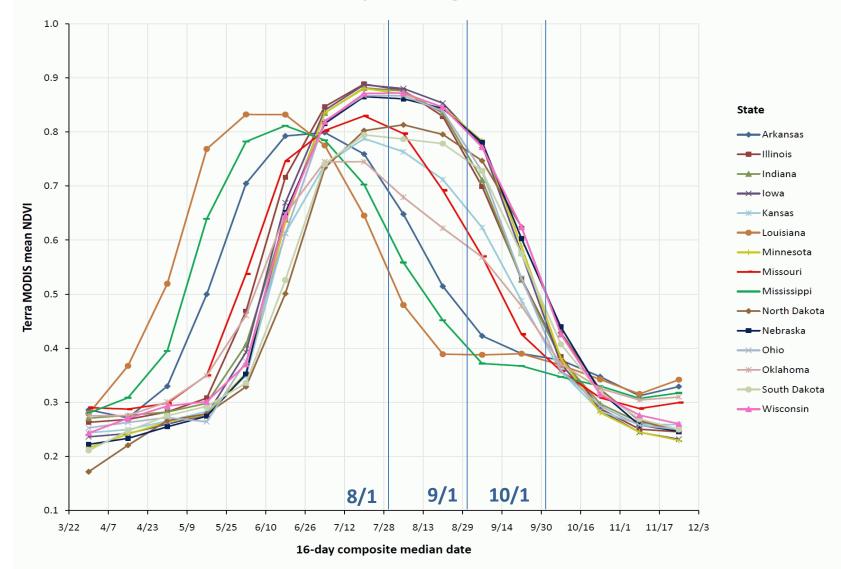






Phenology with Crop Production report timing

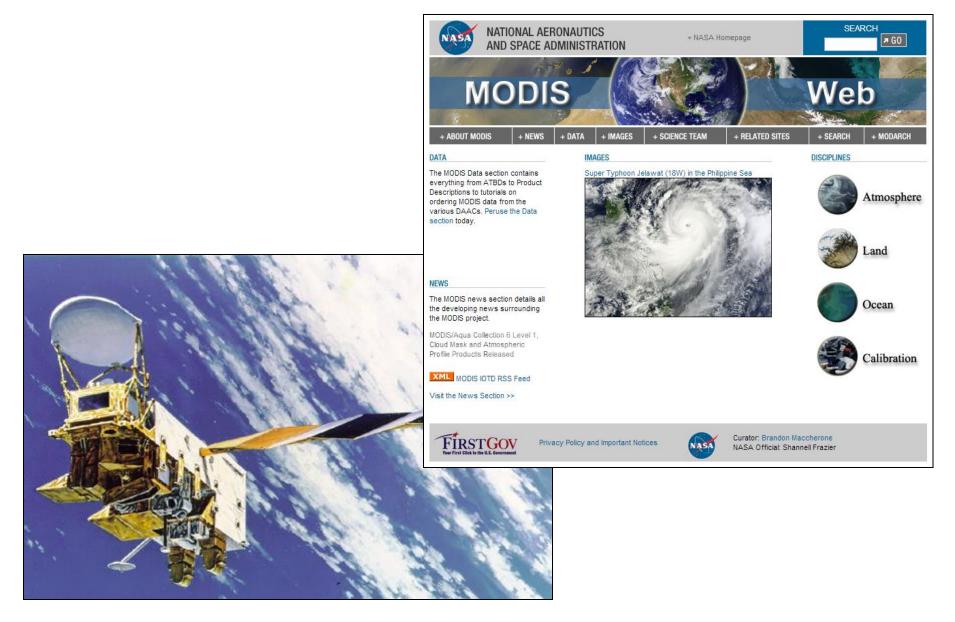
Corn 5-year average 2006-2010





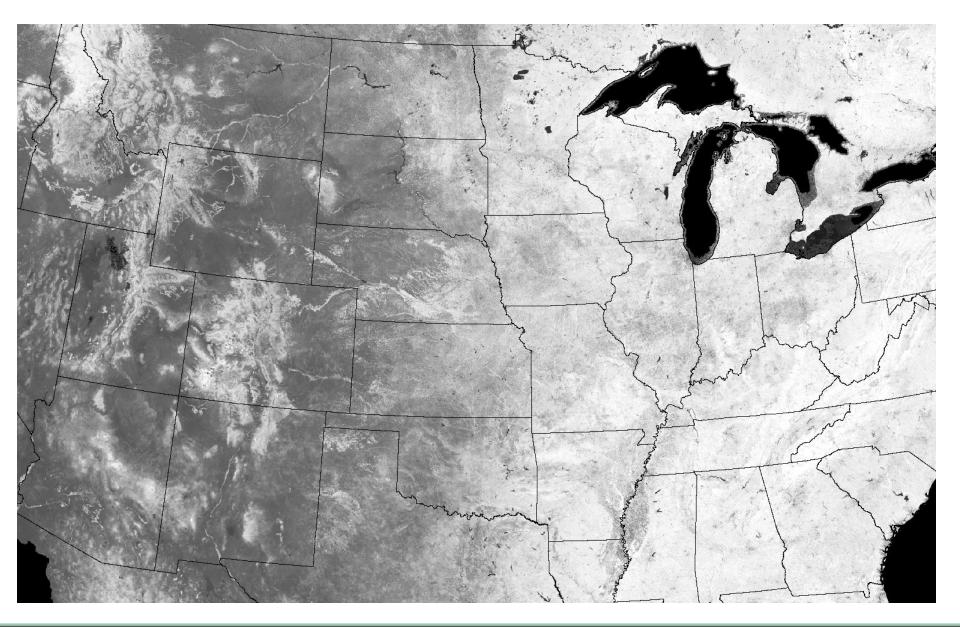


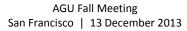
Moderate Resolution Imaging Spectroradiometer (MODIS)





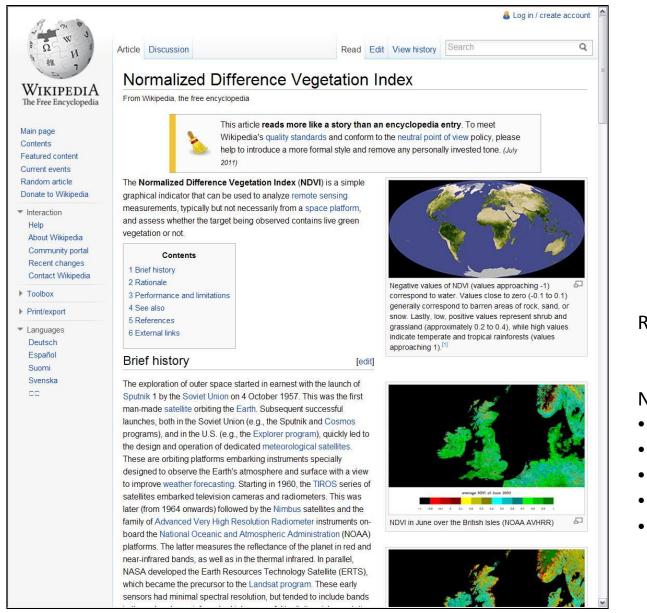
MODIS NDVI data example







Calculation from surface reflectance and use of NDVI



 $NDVI = \frac{(NIR - VIS)}{(NIR + VIS)}$

NIR = near-infrared VIS = visible

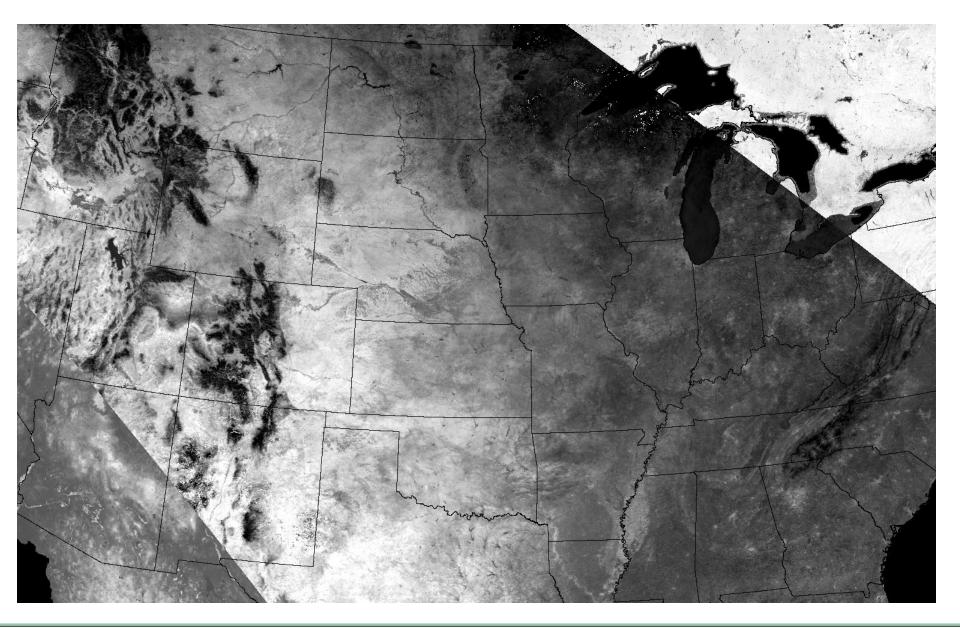
Ranges from -1.0 to 1.0

NDVI is a related to

- Plant health
- Cholophyll content
- "Greenness"
- Biomass
- Vegetation vigor



MODIS LST data example





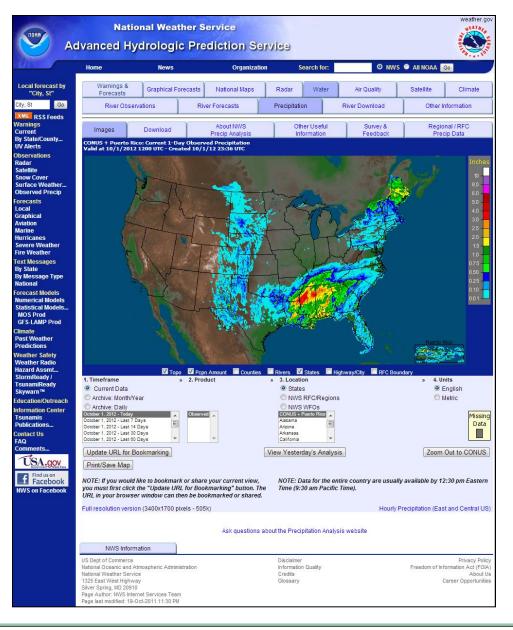


Nexrad-based Precipitation estimates

- GIS-ready product
 - ESRI Shapefile format
- Generated daily
 Little latency
- ~4km grid

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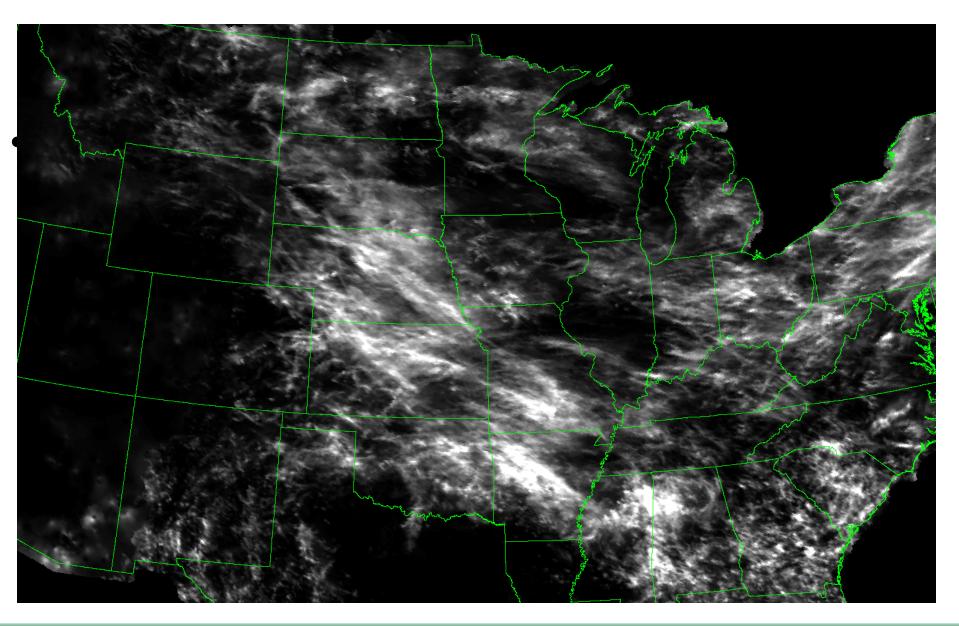
• 2005 - current







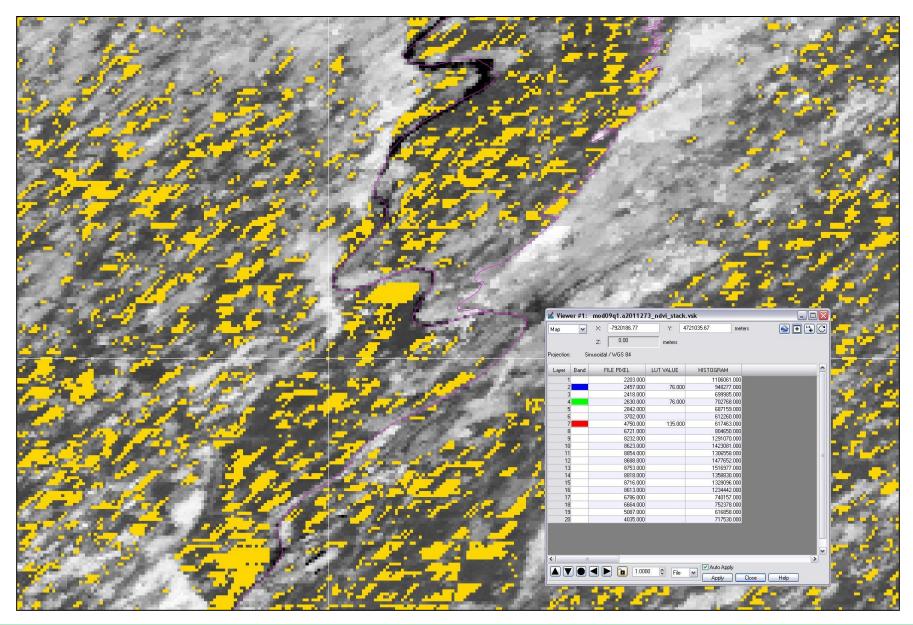
NexRad Rainfall Data example







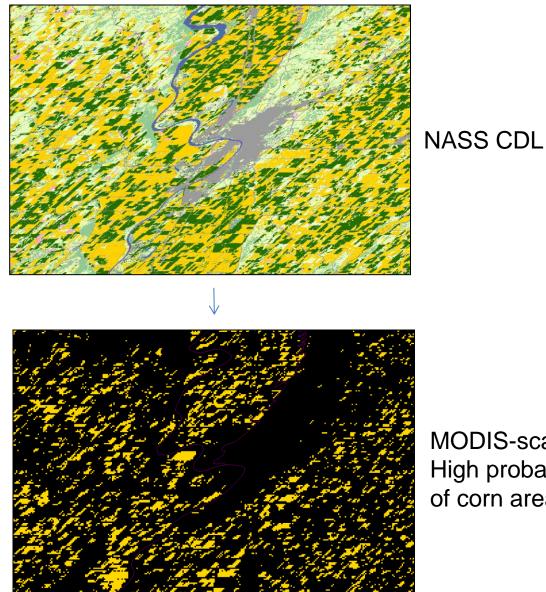
Intersecting corn "mask" with MODIS data



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Establishing the pixels that are only corn



MODIS-scaled High probability sample of corn areas





County-level modeling with "composite" modeling

- Historical NASS county-level yields as dependent variable - 2006-2011
- Analysis over "Speculative" corn and soybean region
- Four timely possible predictor (independent) variables
 - NDVI (Normalized Difference Vegetation Index)
 - derived from Terra satellite MODIS surface reflectance imagery
 - LST (Land Surface Temperature) from day and night
 - derived from Aqua satellite MODIS thermal imagery
 - Precipitation
 - derived from NOAA/NWS Nexrad composite
- Utilizing 8-day composited mosaic products for each
 - Mid-February through late September
- Modeling/mining using Rulequest Cubist software
 - Regression tree based





County-level database developed

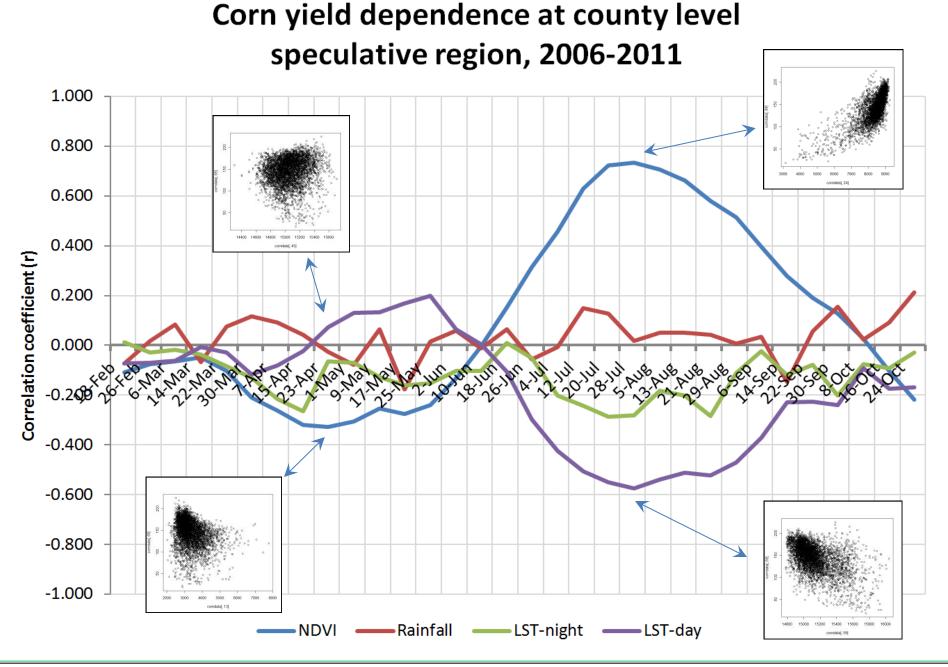
- Potential predictor variables (independent)
 - State (All major production Corn Belt states)
 - County (for each that had a published estimate, ~1000 of them)
 - Year (2006 2011)
 - 32 for each ranging every 8 days from February 18 October 30
 - NDVI
 - Daytime LST (1:30 PM)
 - Nighttime LST (1:30 AM)
 - NWS Precipitation estimates
 - Thus 132 in total
- Forecast variable (dependent)

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- NASS published county level yield
- Sample size to evaluate ~5000 records

															_					
	A	в	С	D	E	F	G	н	1	J	к	L	м	N		L	BM	BN	BO	BP
1	20	119	2009	2263.17	2276.58	2303.59	2348.65	2434.37	2359.19	2266.35	2288.19	2367.71	2512.1	2629.59	28	98.5	15049.9	15100.7	14962.4	22
2	20	69	2009	2288.49	2311.38	2342.2	2339.95	2407.06	2383.9	2340.03	2488.27	2703.84	2684.03	2654.48	28	911.1	14966.6	15021.6	14898	21
3	20	119	2007	2276.42	2251.23	2264.43	2352.61	2469.31	2547.4	2619.07	2684.33	2815.85	2966.01	3141.06	- 33	29.4	15280.7	15183.5	15095.1	21
4	20	\$1	2009	2218.23	2247.03	2304.9	2340.96	2421.18	2349.36	2270.35	2346.69	2446.1	2476.15	2472.75	21	60.4	15013.3	15065.8	14938.8	21
5	31	137	2009	2167.21	2154.48	2139.61	2156.15	2212.33	2278.63	2384.85	2470.95	2573.56	2613.23	2606.06	27	66.6	14578	14519.3	14533.3	21
6	20	175	2007	2232.36	2255.85	2303.21	2438.8	2705.49	2732.7	2735.31	2733.36	2854.66	3053.56	3107.51	32	455	15331.1	15236.9	15109.8	20
7	17	187	2008	1766.18	2556.21	2600.69	2643.94	2644.2	2590.61	2594.14	2706.57	2852.78	3125.75	3459.15	- 34	40.3	14845.2	14543	14629.6	20
8	19	93	2009	1981.28	2104.24	2312.87	2173.23	2118.37	2141.06	2191.56	2315.99	2460.3	2716.56	2887.91	30	16.4	14537.5	14458.7	14470.1	20
9	31	185	2009	2116.76	2196.12	2298.1	2295.07	2297.18	2361.89	2379.85	2411.81	2493.01	2683.23	2960.81	-31	 1678	14508.9	14480.1	14463.9	20
0	17	203	2008	2163.13	2259.22	2355.42	2618.68	2614.77	2551.36	2627.17	2764.98	2829.49	2942.49	3344.19	33	76.6	14873.8	14508.4	14620.5	20
11	17	109	2008	2527.91	2730.11	2685.04	2709.46	2770.32	2661.12	2665.43	2793.58	2972.49	3209.61	3650.63	35	68.4	14885.8	14568.7	14658.5	20
12	31	\$1	2009	2142.37	2188.07	2278.81	2282.22	2292.83	2342.79	2365.32	2406.9	2496	2665.16	2881.11	- 3	76.1	14501.9	14447.6	14462.2	20
13	17	203	2007	964.94	2274.4	2601.68	2709.84	2777.16	2584.11	2568.92	2783.05	3024.46	3052.39	3121.62	3.	14.6	15111.4	14740.3	14567.9	20
14	19	165	2009	2077.56	2064.12	2053.83	2155.59	2357.15	2276.43	2252.18	2341.4	2441.82	2643.78	3136.29	-31	33.3	14608.8	14509.7	14527.9	20
15	20	81	2007	2350.12	2276.67	2247.89	2383.91	2661.31	2660.21	2643.89	2631.8	2883.8	3317.63	3184.67	31	56.2	15275	15216.8	15107.6	20
16	17	175	2008	1352.35	1906.66	2479.59	2545.58	2607.91	2503.92	2507.45	2621.58	2781.29	2988.94	3283.28	3	63.8	14841.5	14532.1	14606	20
17	31	99	2009	2141.86	2118.68	2121.94	2139.96	2196.7	2261.97	2360.63	2444.19	2524.94	2535.12	2579.6	26	55.4	14547.6	14507.1	14520.7	20
8	19	35	2009	2131.72	2203.73	2379.83	2237.11	2157.33	2182.48	2230.26	2375.26	2546.31	2704.29	2947.32	31	62.7	14564.6	14480.9	14480.3	20
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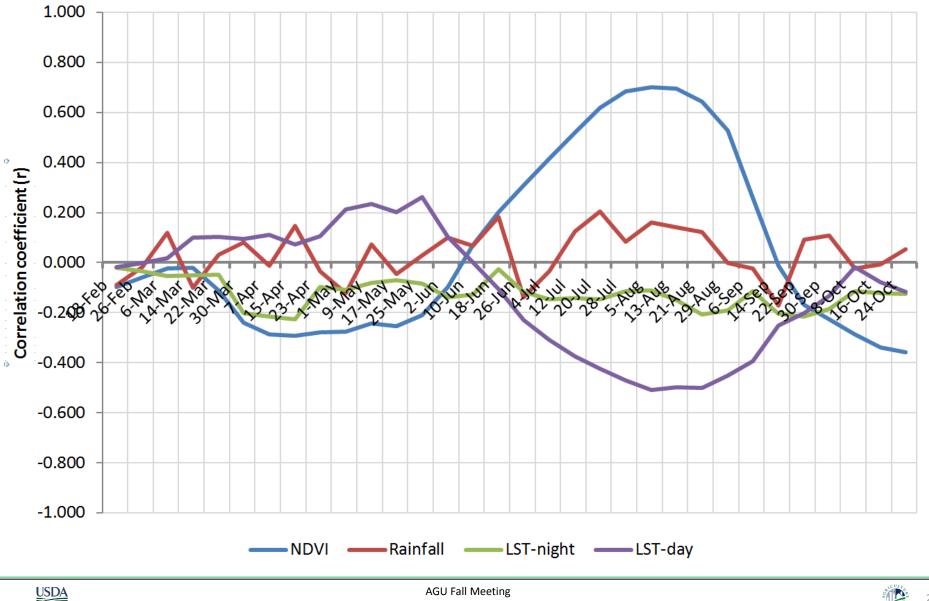








Soybean yield dependence at county level speculative region, 2006-2011



"Voodoo Modeling"

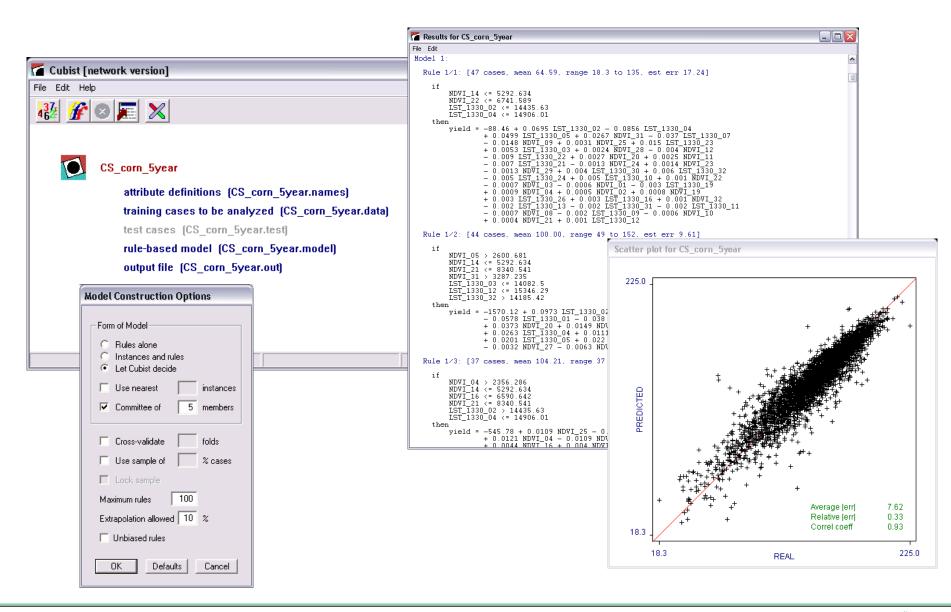
- **Rulequest Cubist**
 - Learning tool to predict continuous rather that discrete outcomes
 - Allow for "composite" predictions using both
 - Instance-based
 - "Nearest neighbor"
 - Predicts the target value of a new case by finding the **n** most similar cases in the training data, and averaging their target values.
 - Model-based, via decision trees and piecewise linear regression
 - Divide and conquer strategy
 - Recursive splitting of training data to minimize intra-subset variation
 - Thus, for composite of instances and models:
 - Cubist finds the n training cases that are "nearest" (most similar) to the case in question. Then, rather than averaging their target values directly, Cubist first adjusts these values using the rule-based model.
 - Also, does "Committee" models
 - made up of several rule-based models. Each member of the ٠ committee predicts the target value for a case and the members' predictions are averaged to give a final prediction

Model Construction Options									
Form of Model									
 Rules alone Instances and rules Let Cubist decide 									
🔲 Use nearest		instances							
Committee of	5	members							
Cross-validate		folds							
Use sample of		% cases							
🗖 Lock sample									
Maximum rules	100								
Extrapolation allowed	10	%							
🔲 Unbiased rules									
OK Defau	ilts	Cancel							





Rulequest Cubist







Example county-level prediction output

Case	Given	Predicted			Case	Given	Predicted
No	Value	Value			No	Value	Value
1	1701.0		+- 24.78		1	501.0	45.30 +- 7.60
2	1703.0	138.24	+- 24.78		2	503.0	32.20 +- 7.60
3	1705.0	64.58	+- 24.78	e .	3	505.0	4.31 +- 9.50
4	1707.0	129,20	+- 24.78	6	4	507.0	4.31 +- 9.50
5	1709.0		+- 24.78		5	509.0	4.31 +- 9.50
6	17011.0		+- 24.78	Ľ.	6	5011.0	4.31 +- 9.50
3 4 5 6 7 8 9	17013.0		+- 24.78	[7	5013.0	
				Ľ.,			4.31 +- 9.50
8	17015.0		+- 24.78	ľ –	8	5015.0	4.31 +- 9.50
	17017.0		+- 24.78		9	5017.0	36.80 +- 7.60
10	17019.0		+- 24.78		10	5019.0	10.88 +- 7.60
• 11	17021.0	122.79	+- 24.78		11	5021.0	39.92 +- 7.60
12	17023.0	79.02	+- 24.78	6	12	5023.0	4.31 +- 9.50
13	17025.0	45.30	+- 24.78	÷	13	5025.0	4.31 +- 9.50
14	17027.0	59.63	+- 24.78	e .	14	5027.0	4.31 +- 9.50
15	17029.0	94.61	+- 24.78		15	5029.0	19.68 +- 7.60
16	17031.0		+- 24.78		16	5031.0	34.14 +- 7.60
17	17033.0		+- 24.78		17	5033.0	28.26 +- 7.60
18	17035.0		+- 24.78		18	5035.0	43.82 +- 7.60
19	17035.0				19		
			+- 24.78	L.	20	5037.0	39.48 +- 7.60
20	17039.0		+- 24.78	ľ.		5039.0	4.31 +- 9.50
21	17041.0		+- 24.78		21	5041.0	39.78 +- 7.60
22	17043.0		+- 24.78		22	5043.0	38.52 +- 7.60
23	17045.0		+- 24.78		23	5045.0	11.48 +- 7.60
24	17047.0		+- 24.78		24	5047.0	4.31 +- 9.50
25	17049.0	66.14	+- 24.78	÷	25	5049.0	4.31 +- 9.50
26	17051.0	62.86	+- 24.78	6	26	5051.0	4.31 +- 9.50
27	17053.0	101.90	+- 24.78	e .	27	5053.0	4.31 +- 9.50
28	17055.0	54.85	+- 24.78		28	5055.0	33.53 +- 7.60
29	17057.0		+- 24.78		29	5057.0	28.21 +- 7.60
30	17059.0		+- 24.78		30	5059.0	12.47 +- 7.60
31	17061.0		+- 24.78	L.	31	5061.0	4.31 +- 9.50
32	17063.0		+- 24.78	Ľ	32	5063.0	31.07 +- 7.60
33	17065.0		+- 24.78	L.	33	5065.0	4.31 +- 9.50
• 34				ſ.	34		
	17067.0		+- 24.78			5067.0	34.07 +- 7.60
35	17069.0		+- 24.78	L	35	5069.0	41.76 +- 7.60
36	17071.0		+- 24.78	l*	36	5071.0	4.31 +- 9.50
37	17073.0		+- 24.78		37	5073.0	36.72 +- 7.60
38	17075.0		+- 24.78		38	5075.0	34.29 +- 7.60
39	17077.0	63.20	+- 24.78		39	5077.0	41.19 +- 7.60
40	17079.0	70.46	+- 24.78		40	5079.0	42.98 +- 7.60
41	17081.0	46.99	+- 24.78		41	5081.0	26.32 +- 7.60
42	17083.0	94.22	+- 24.78		42	5083.0	10.81 +- 7.60
43	17085.0	123.96	+- 24.78		43	5085.0	34.80 +- 7.60
44	17087.0		+- 24.78	*	44	5087.0	4.31 +- 9.50
45	17089.0		+- 24.78	4	45	5089.0	4.31 +- 9.50
46	17091.0		+- 24.78	ſ.	46	5091.0	22.51 +- 7.60
40	17091.0		+- 24.78	L.	40	5093.0	43.11 +- 7.60
				ſ			
48	17095.0		+- 24.78	L	48	5095.0	36.57 +- 7.60
49	17097.0		+- 24.78	Ľ	49	5097.0	4.31 +- 9.50
50	17099.0	128.53	+- 24.78	f	50	5099.0	4.31 +- 9.50

Corn

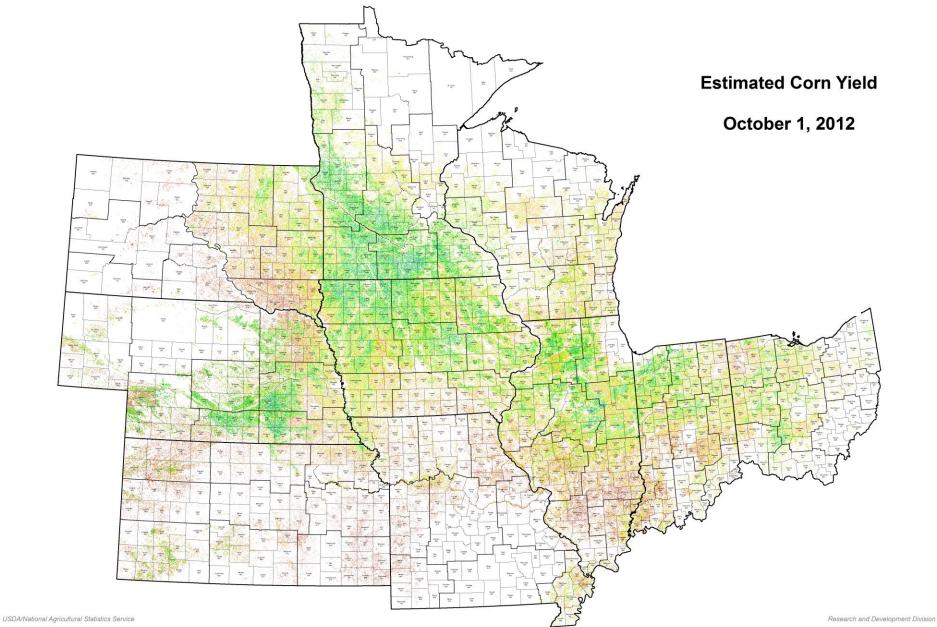
Soybeans

Weight by a 3-year average of harvested acres to derive ASD, state, and region estimates



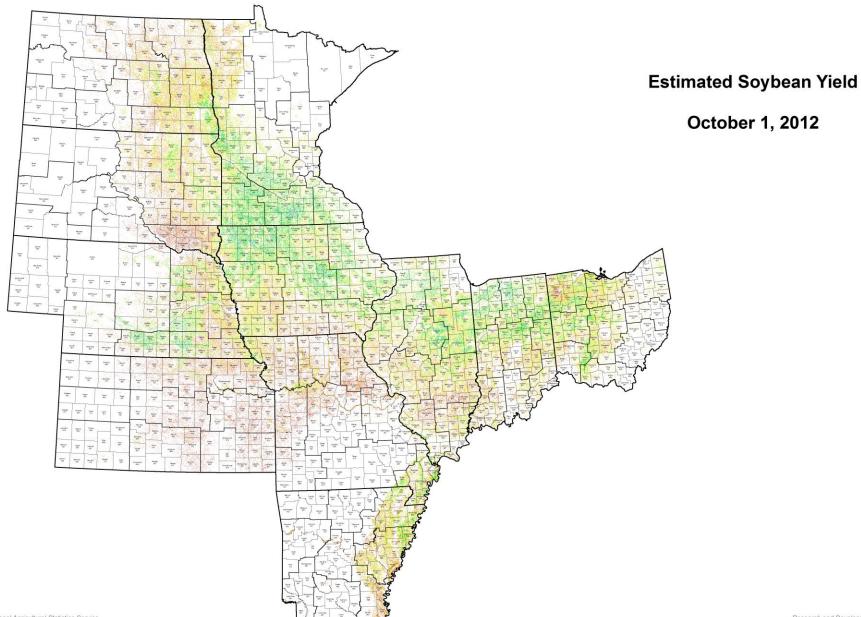


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USDA/National Agricultural Statistics Service

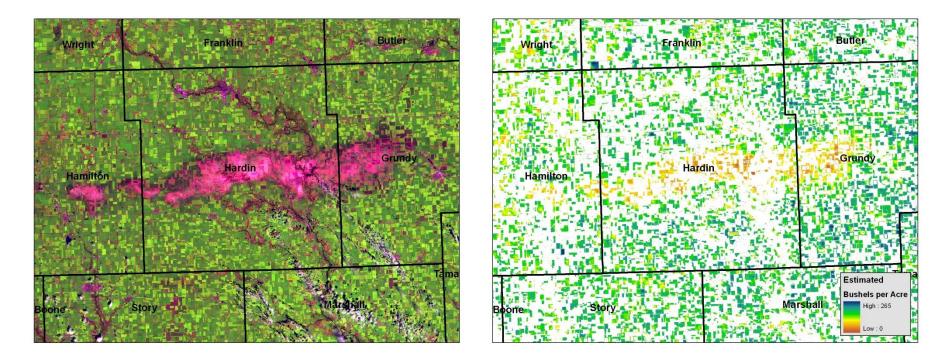
Research and Development Division





Reality check

Scene of a large hailstorm



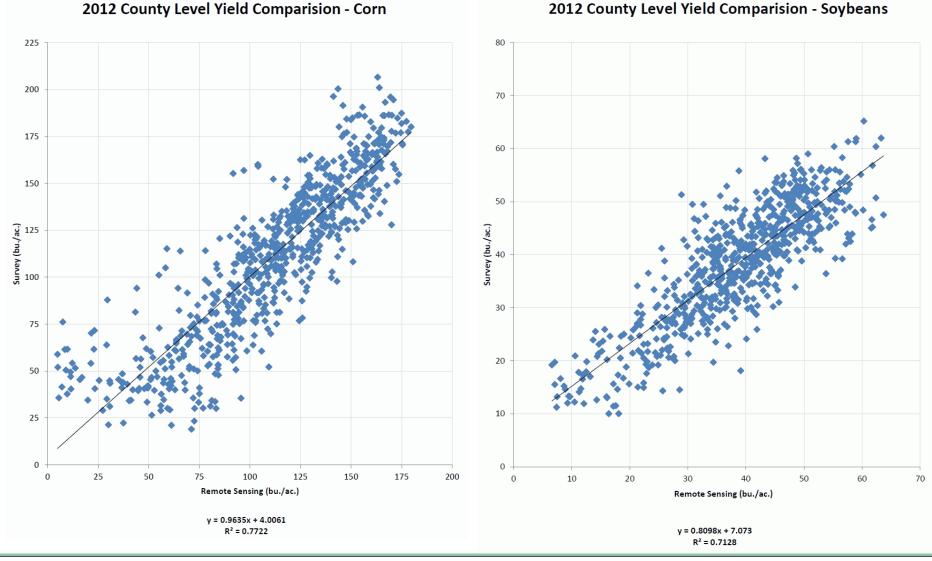




2012 Remote sensing vs NASS yield

State level average error

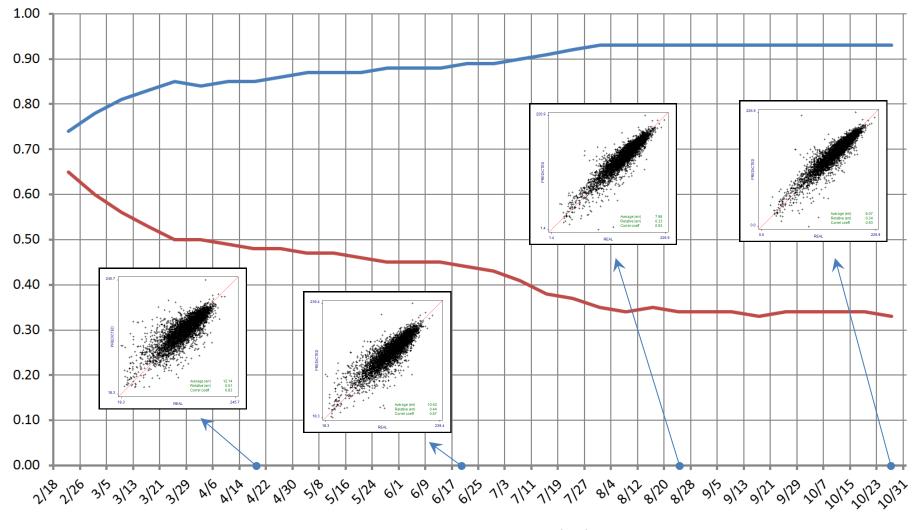
corn = 5.8 bu./ac. soybeans = 3.1 bu./ac.



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Corn yield regression-tree model performance v. data timing county level, speculative region, 2006-2011

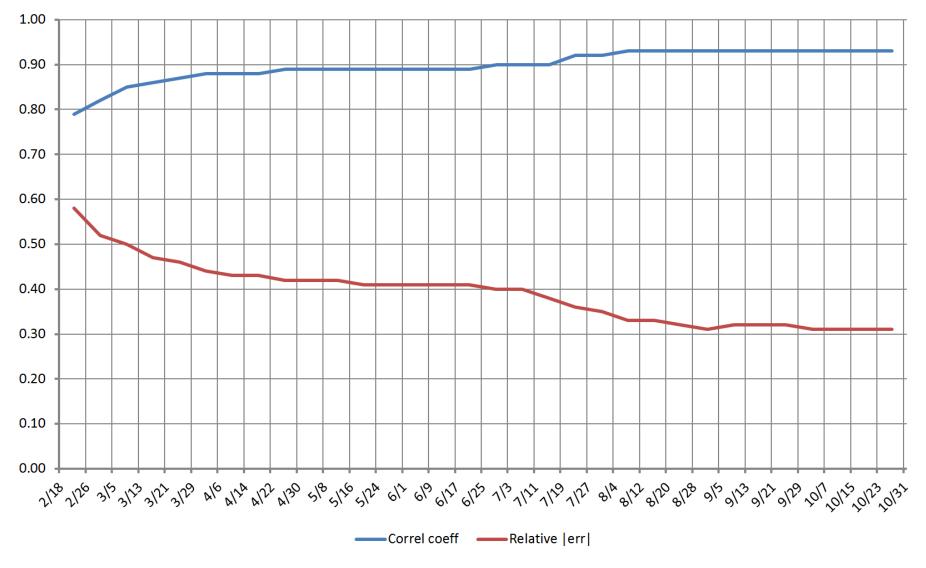


Correl coeff ——Relative |err

The **relative error magnitude** is the ratio of the average error magnitude to the error magnitude that would result from always predicting the mean value; for useful models, this should be less than 1!

The **correlation coefficient** measures the agreement between the cases' actual values of the target attribute and those values predicted by the model.

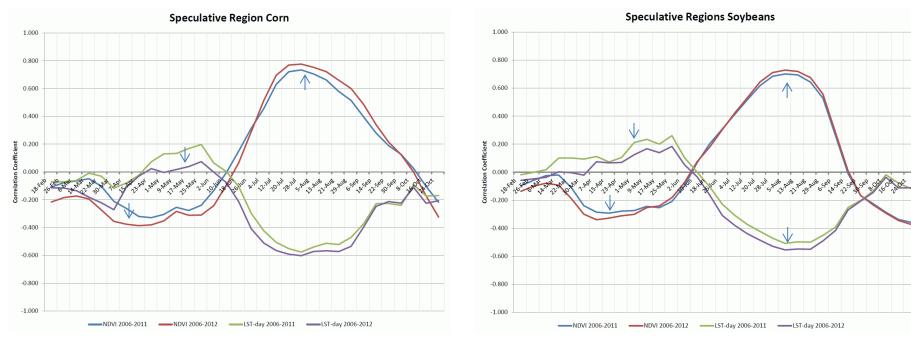
Soybean yield regression-tree model performance v. data timing county level, speculative region, 2006-2011





Models improvements for 2013

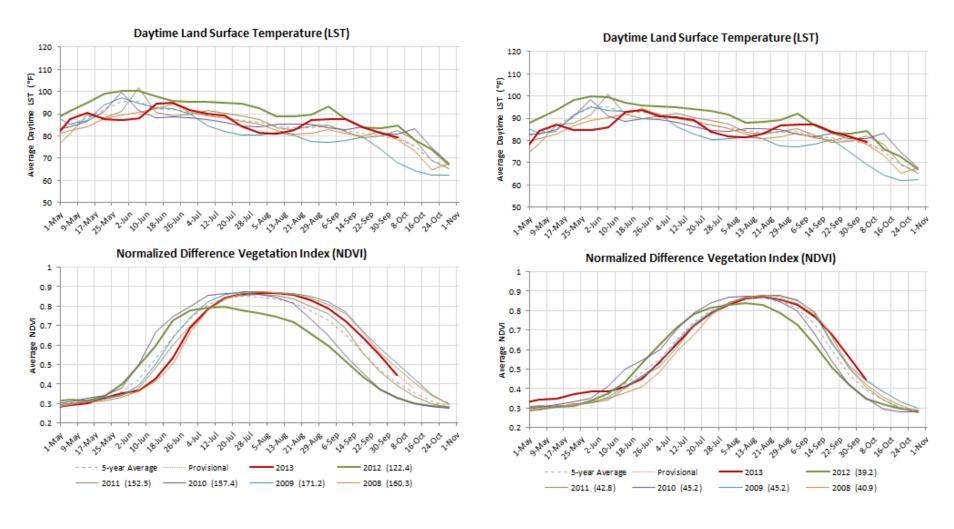
•	Corn	2012	2013
	 relative err 	0.33	0.30
	 correl coeff 	0.93	0.95
•	Soybeans		
	 relative err 	0.31	0.30
	 correl coeff 	0.93	0.94



- Absolute error unchanged
 - ~8.0 bu/ac for corn, ~2.5 for soybeans



MODIS-derived crop dynamics based on CDL areas



Corn

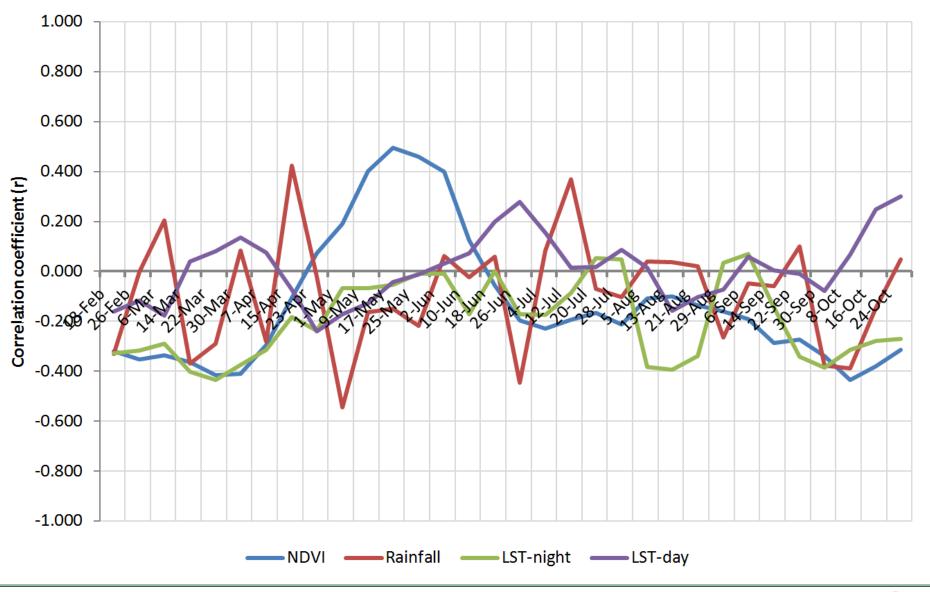


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This winter: Build full understanding all common MODIS derived variables and how they relate to various crops' yields

- Explore fully beyond only corn and soybeans
 - Wheat
 - Rice
 - Potatoes
 - Sorghum
 - Cotton
- Compare the full suite of common MODIS variables
 - NDVI
 - LAI
 - FPAR
 - LST (daytime and nighttime)
 - and more....
- Test Both Terra and Aqua platforms
 - Assess the AM vs PM overpass time
- Look at pixel scale issues
 - 250 m vs. 500 m vs. 1000 m (particularly for NDVI)

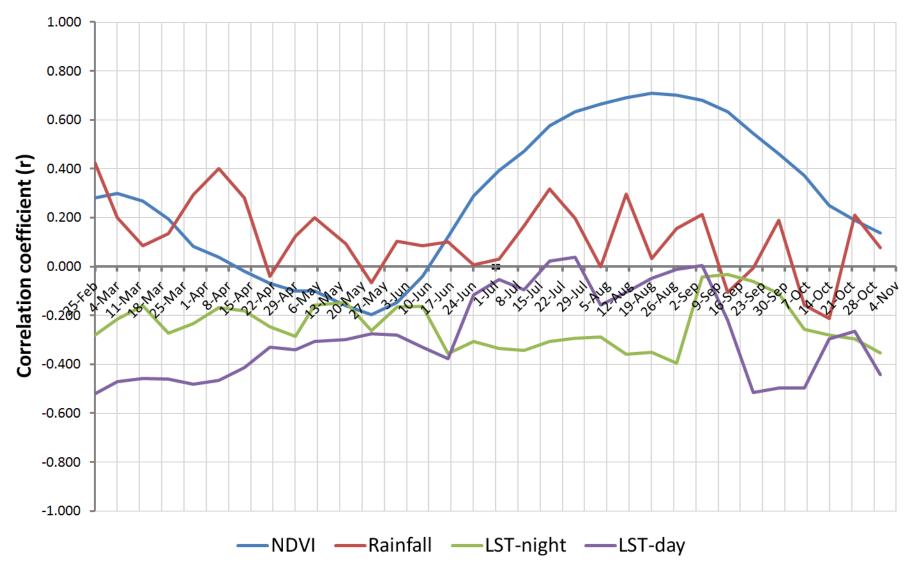
Winter wheat yield dependence at county level Kansas, 2006-2011







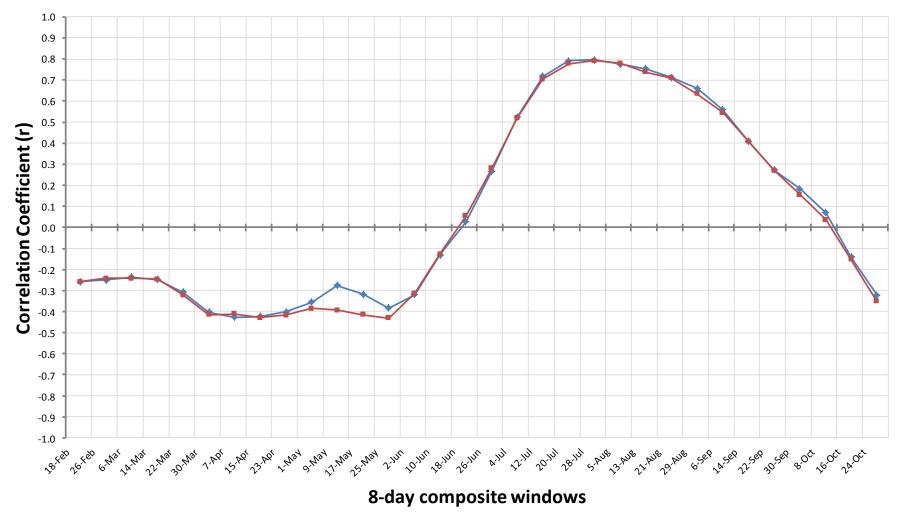
Cotton yield dependence at county level TX & AR, 2005-2011







Corn yield correlations to NDVI from the different MODIS platforms

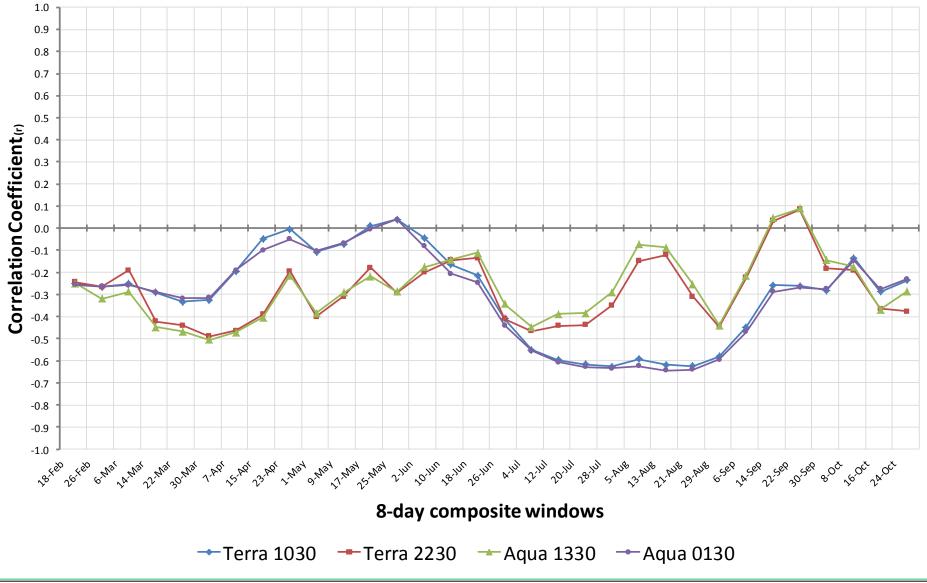


-Terra -Aqua





Corn yield correlations to LST from the different MODIS platforms





In summary

- Corn and soybeans yield predictors
 - NDVI most useful
 - Daytime LST also useful
 - Precipitation not useful
 - Nighttime LST not useful
- Full exploration of other MODIS variables and other crops has begun







Thanks

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www.nass.usda.gov www.nass.usda.gov/Research_and_Science nassgeodata.gmu.edu/CropScape



