

Agricultural Classification of Multi-Temporal MODIS Imagery in Northwest Argentina Using Kansas Crop Phenologies

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RESEARCH QUESTIONS

Can I...

- ▶ develop a phenological classification toolset?
- ▶ extract crop signatures from Kansas data?
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OUTLINE

1. Background
2. Study Areas
3. Data and Methods
4. Results and Discussion
5. Conclusion

BACKGROUND

DEFORESTATION IN ARGENTINA

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- ▶ 1998 to 2002: 940,000 ha deforested
- ▶ *Ley de Bosques* passed in 2007
 - ▶ Classified red, yellow, and green areas

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DEFORESTATION IN ARGENTINA

Table: Deforestation in Argentina, 2006 to 2011

Time Period	Hectares Deforested
2006 to <i>Ley de Bosques</i> (2007)	573 296
Ley de Bosques to OTBN (2009)	473 001
OTBN to 2011	459 108
Total	1 505 405

DEFORESTATION IN ARGENTINA

- ▶ Deforestation has remained extremely high
- ▶ The effect of the the *Ley de Bosques* has been questioned

SOY AND ITS EFFECTS

- ▶ Argentina's soybean cultivation has continually increased
 - ▶ 5 million ha in 1993 to 19 million ha in 2011

SOY AND ITS EFFECTS

- ▶ Soy production highly mechanized
- ▶ Over 99 percent of Argentine soy is genetically modified
 - ▶ Resistance to glyphosate = heavy pesticide use
- ▶ Capital requirements cut out small producers

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- ▶ **Capital requirements cut out small producers**

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- ▶ Deforestation research has neglected to analyze specific crop cover

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Questions

- ▶ What if two crops have similar VI values on a single date?
- ▶ How does one determine a crop's VI values?

VEGETATION INDICIES

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VEGETATION INDICIES

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What if two crops have similar VI values on a single date?

Answer

Use imagery from multiple dates.

TIME SERIES IMAGES

NASA's Moderate Resolution Imaging Spectroradiometer (MODIS) Sensor

- ▶ Terra and Aqua satellites
- ▶ Each images the Earth once per day
- ▶ Composite 16-day NDVI imagery at 250-meter resolution

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TIME SERIES IMAGES

Time Series Image (TSI)

- ▶ Each band is a 16-day VI composite
- ▶ Bands are sequential composites
- ▶ Contains enough bands to cover an entire growing season

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TIME SERIES IMAGES

Key Points

- ▶ A TSI pixel shows VI values over time
- ▶ Each crop's phenology exhibits a unique temporal signature

CROP TEMPORAL SIGNATURES

Graphics/wardlowCropSignatures.png

(From Wardlow and Egbert 2005)

PHENOLOGICAL CLASSIFICATION

Question

How does one determine a crop's VI values?

PHENOLOGICAL CLASSIFICATION

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How does one determine a crop's VI values?

Answer

Existing approaches require training sites.

PHENOLOGICAL CLASSIFICATION

Problem

What if you don't have training sites?

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Sounds a lot like hyperspectral remote sensing...

TIME SERIES IMAGES

Idea

Could we use a hyperspectral-like method to fit known crop signatures to unknown pixels?

TIME SERIES IMAGES

Graphics/transformations.pdf

Two-Step Filter (TSF) method from sakamoto2010a-two-step

- ▶ Two steps: (1) wavelet smoothing and (2) curve fitting
- ▶ Curve fitting can fit reference signature to unknown pixels

TSF Equation 1

$$RMSE = \left[\frac{1}{365/s} \sum_{x = j(0), j(1) \dots}^n (f(x) - g(x))^2 \right]^{\frac{1}{2}}$$

where

- ▶ n is the number of dates in the TSI
- ▶ $f(x)$ is the temporal signature for a given pixel in a dataset
- ▶ x is the DOY, as defined by $j(y)$

TSF Equation 2

$$g(x) = yscale \times h(xscale \times (x + tshift))$$

where

- ▶ *yscale* and *xscale* are coefficients controlling the vertical and horizontal scaling of a reference signature $h(x)$
- ▶ *tshift* is a constant representing the horizontal shift, in days, of $h(x)$
- ▶ x is the DOY

TSF METHOD

Graphics/transformations.pdf

TSF Equation 1

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Minimizing Equation 1 with appropriate constraints on *yscale*, *xscale*, and *tshift* will find the fit of a reference signature to a pixel.

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Answer

The TSF equations allow the classification of unknown pixels using a library of crop signatures.

STUDY AREAS

KANSAS STUDY AREA

- ▶ 2012 Kansas top crops:
 - ▶ Winter wheat
 - ▶ Corn
 - ▶ Soy
- ▶ Ground truth:
USDA Cropland Data
Layer

Graphics/KSstudysite.pdf

Graphics/argentinaOverview_landscape.pdf

Table: Deforestation in Pellegrini, 2001 to 2011

Period		Hectares Cleared	Percent of Land
2001	to2005	5.968	0.9
2006	to2011	75.249	10.8

DATA AND METHODS

RESULTS AND DISCUSSION

CONCLUSION