







# Crop Yield Monitoring at Field Scale with Satellite Data

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#### Introduction



- Evaluation was carried out within NASA's "Private Sector Small Constellation Satellite Data Product Pilot" program
  - https://earthdata.nasa.gov/esds/small-satellite-data-buy-program/csdap-pilotevaluation

Evaluation of very-high spatial resolution (VHR) satellites (<5 m)</li>

- Objective:
  - Evaluation of VHR satellite data (Planet, Worldview-3) for agricultural monitoring tasks

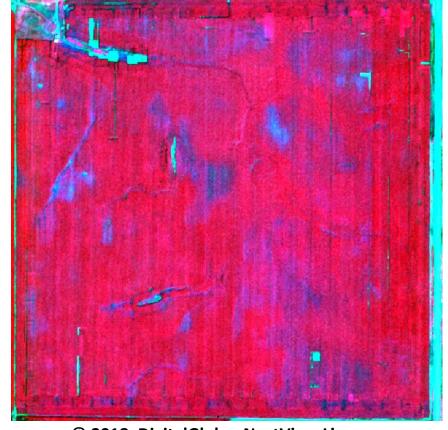


# Objectives of the VHR study for Ag



 To evaluate VHR data (1-3 m) on capturing a field-level yield variability

- Importance:
  - Insurance industry
  - Fine scale crop production mapping
  - Crop yield gap analysis
  - Guiding sample collection
    - e.g. yield cuts



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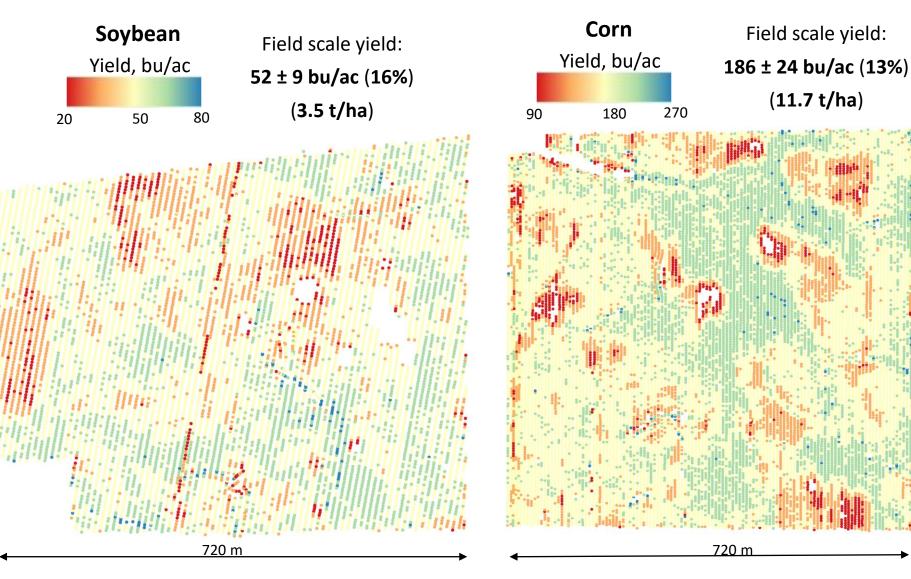


# Ground data: crop yields at field scale





Field scale yields for corn and soybean (Hamilton County, IA, USA). Provided by Iowa State University



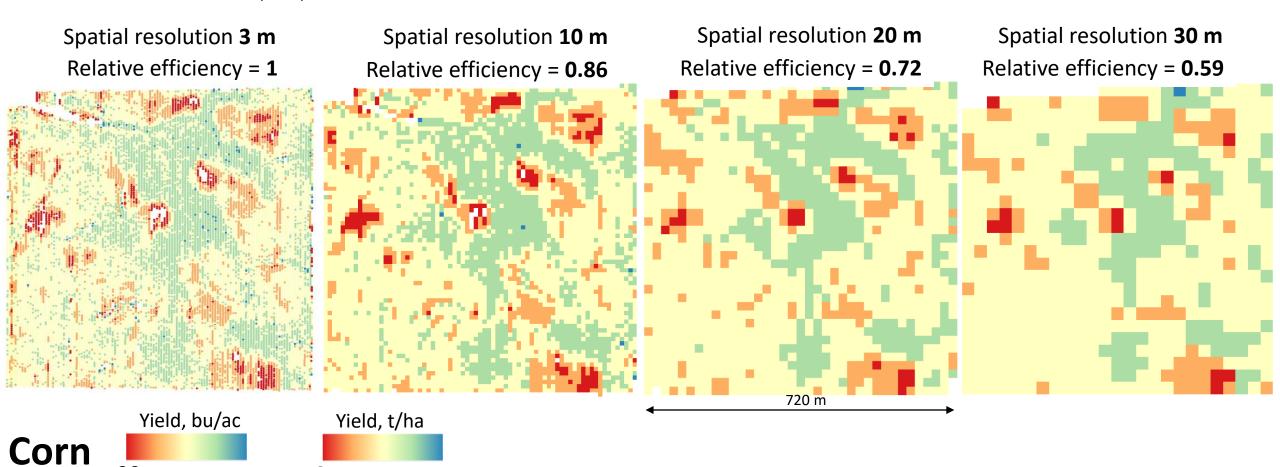


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## Effect of spatial resolution on capturing yield variability



- Simulating yields at various spatial resolutions
  - Relative efficiency = Variance Yield(@Xm) / Variance Yield(@3m)
     where X is 10, 20, 30

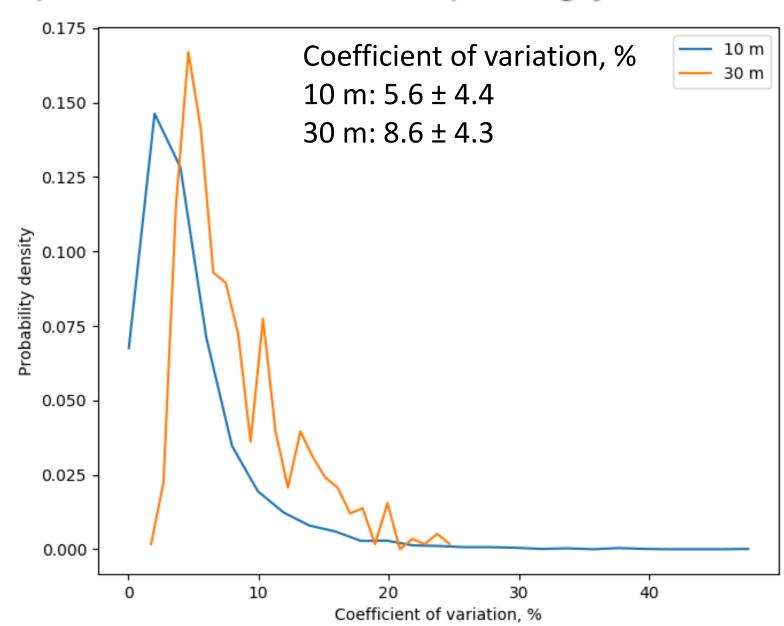




#### Effect of spatial resolution on capturing yield variability



Yield variation inside the pixel





#### Remote sensing data



- Maxar: WorldView-3 (WV-3)
  - 2 images
  - 8 spectral bands
- Planet: PlanetScope
  - 270 unique scenes, ~84,000 km²
  - Near-daily coverage
  - 4 spectral bands
- Harmonized Landsat Sentinel-2 (HLS)
- Atmospheric correction
  - LaSRC: Land Surface Reflectance Code adaptation from Landsat 8 and Sentinel-2 (Vermo et al. 2016)



WV-3, 21 July 2018, 1.2 m, Red-Green-Blue

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Planet/PlanetScope, 30 August 2018, 3 m, Red-Green-Blue



### Results: WV-3 16

WV-3 Corn 21 July 2018, DOY=202

NDVI:  $\frac{\rho_{NIR} - \rho_{RED}}{}$ 

 $\rho_{NIR} + \rho_{RED}$ 

WDRVI:  $\frac{0.1\rho_{NIR}-\rho_{RED}}{2.1}$ 

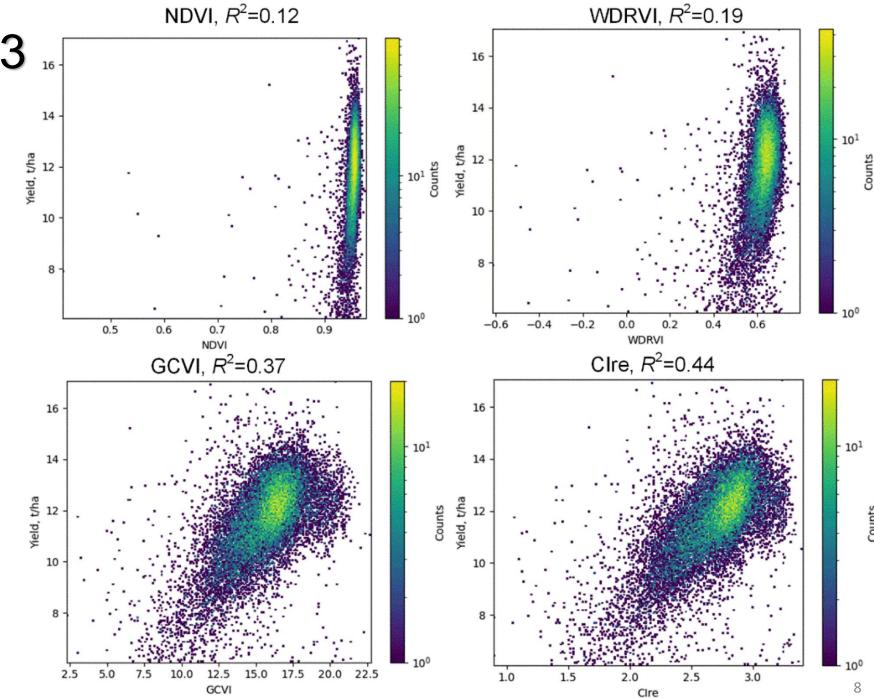
 $0.1\rho_{NIR} + \rho_{RED}$ 

GCVI:  $\frac{\rho_{NIR}}{2}$  - 3

 $\rho_{GREEN}$ 

Cire:  $\frac{\rho_{NIR}}{\rho_{NIR}}$  – 1

 $\overline{
ho_{RED-EDGE}}$ 



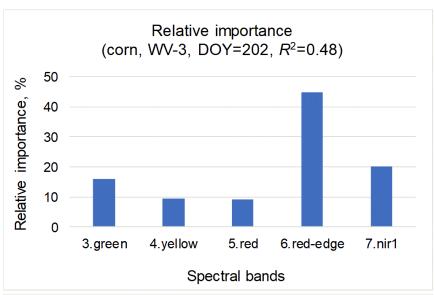


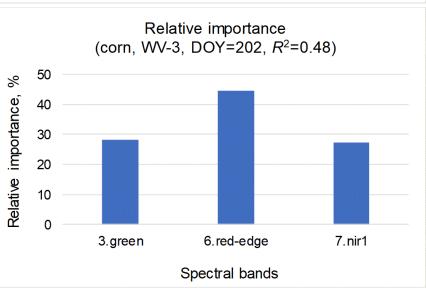
#### Results: WV-3

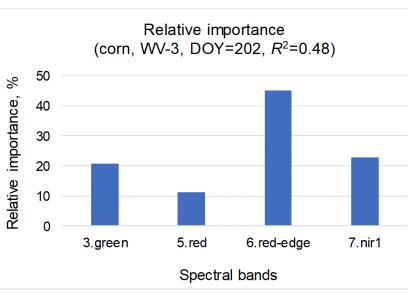


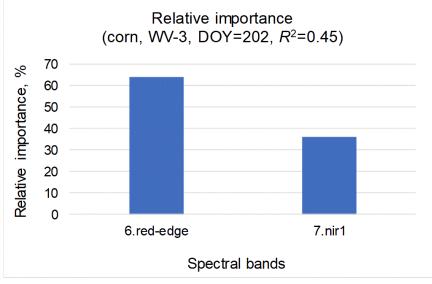
WV-3 Corn 21 July 2018, DOY=202

The red-edge (0.726  $\mu$ m), green (0.552  $\mu$ m), and NIR (0.831  $\mu$ m) bands were the most important in explaining yield variance at field scale











#### Results: PlanetScope

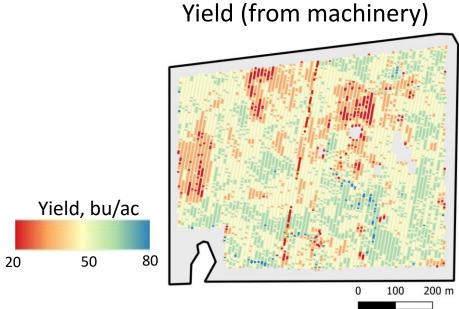


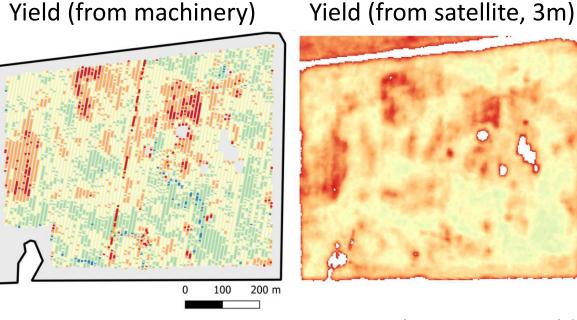
#### Planet, soybean, R<sup>2</sup> over time

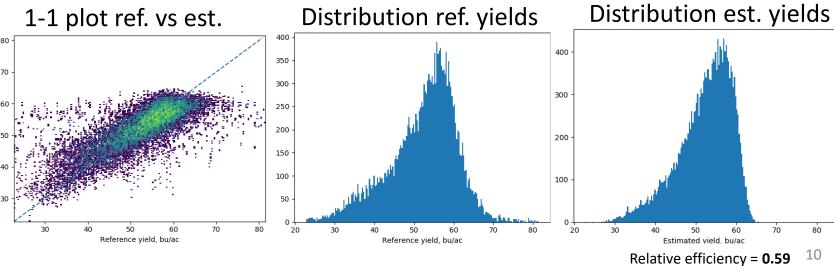
Maximum R<sup>2</sup>=0.59 for DOY=242 (30 August)



Year and DOY Relative importance PlanetLab spectral bands





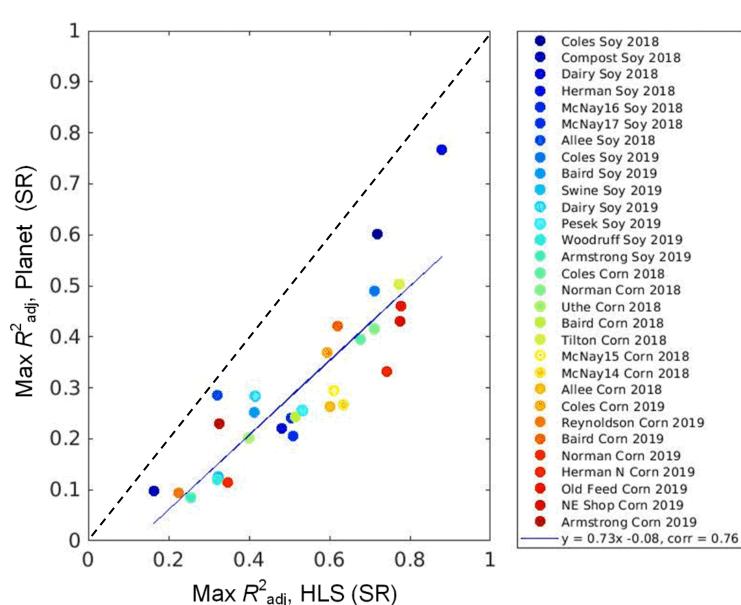




#### Results: PlanetScope vs HLS



- Planet (3 m)
  - R<sup>2</sup>: 0.09 to 0.77 (0.30 ± 0.16)
- HLS (30 m)
  - R<sup>2</sup>: 0.21 to 0.88 (0.56 ± 0.19)

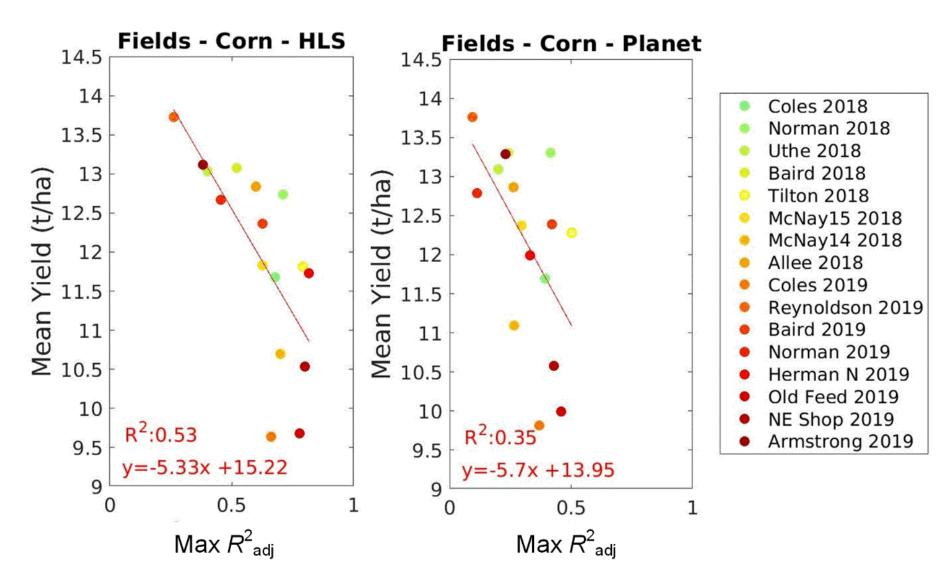




#### Results: PlanetScope vs HLS



 Dependence of the coefficient of determination R<sup>2</sup> for the best HLSand Planetbased models on field yield for corn





#### Conclusions



- Capturing in-field yield variability depends on the spatial resolution of remote sensing data:
  - e.g., 10-m data explain 86% and 30-m ~60% compared to 3-m
- Most important spectral bands explaining corn/soybean yield variability
  - green/yellow, red edge and NIR
- The high temporal frequency of Planet allows identification of the optimal date for yield assessment
  - 4 PlanetScope's spectral bands at 3 m explained from 10% to 75% of in-field corn/soybean yield variability



Assessing within-Field Corn and Soybean Yield Variability from WorldView-3, Planet, Sentinel-2, and Landsat 8 Satellite Imagery

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Crop yield estimation at different growing stages using a synergy of SAR and optical remote sensing data

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