Robotic Technologies for Automated High-throughput Plant Phenotyping

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Agenda

- I. Machine Vision Systems
 - Automated Plant Population and Interplant Spacing Sensing
 - Plant 3D Reconstruction and Characterization under Indoor and Outdoor Conditions
- 2. Mobile Robotic Platforms

I. Machine Vision Systems

- I) 2D Color Video and Sequential Images
- 2) Stereo 3D Imaging
- 3) Time-of-Flight (ToF) of Light 3D Imaging

Manual population and spacing data collection

Labor intensive, time consuming, error-prone





Seed germination and survival, yield trials



http://www.firstseedtests.com/data3/images/3stand_counting_crew.jpg



http://www.agweb.com/the-knirks/

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http://farmlandforecast.colvin-co.com



Corn plant identification - some of the challenges



- Intertwined Leaves



- a weed interference, damaged plants



- Double plants, plants in the background



I) 2D color video and sequential imaging

Challenges in lighting conditions, real-time processing







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Real-time mosaicking - a "long" panorama image of a crop row





Examples of real-time crop row reconstruction







Color image processing for stem center identification



Original



Color Transformation



Bayes classification



K-Mean clustering



Robust crop row center-line fitting - excluding small weed plants









Plant Identification and Spacing Measurement

Squenced Frame Window: Sequenced-Frame Length=27

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Software Interface





Conclusions on 2D imaging

- Plus:
 - Reliable mosaicking and identification for corn plants between V2 and V3 (partly V4).
 - Allow error checking and manual recording.
- Delta:
 - Cannot deal with larger plants.
 - Rely on good color sensor and ideal lighting.
 - Cannot work under direct sunlight.



2) Stereo Vision system



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Real-time Stereo-On-a-Chip

Graphical User Interface







Color 🗹

Debug Information

SYS VEISION 1. 11 DES DIgital Stereo mienace	Function Stereo 👻 Get 3D
Auto Exposure	
Can not process on camera	
MMX presence: 7	Conf: 7 曽 Disparity: 16 曽
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Device/Device 1: #VIDERE_DESIGN MDS-STH	
#0055050400083260	
Can get color image	
Can process on chip	
max_x_offset <= 0	
Open device 0	
[User] Proc mode none	Initial Start Stop
[User] Proc mode disparity	

Color



A Complicated Example









Method Overview



Skeletonization and leaf & intersection code lists





Result of identified stem centers





Robotic Phenotyping – Sorghum Biomass Traits



Genetic architecture of sorghum biomass yield-associated traits identified using high-throughput, fieldbased phenotyping technologies

PI: Patrick Schnable Co-PIs: Maria Salas Fernandez, Lie Tang Sponsors: USDA, DOE





Back-View

Stereo camera



Stereoscopic sensing

Robot platform



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Robotic platform - PhenoBot











Video on scouting short plants



Video on scouting tall plants





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3D point-cloud





3D point-cloud with higher image resolution





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Need a clean segmentation at the first place.

Conclusions on stereo imaging

- Plus:
 - Utilization of depth information enables segmenting of individual plants.
 - High resolution and low-cost sensors are available.
- Delta:
 - The depth data generated by stereo head is often noisy and inaccurate.
 - Direct sunlight and variable lighting still impose challenges.
 - High computational cost for more sophisticated algorithms.



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Data Processing

+

Stem matching & image mosaicking







Inter-plant Spacing Sensing

Inter-plant distance measurement



 $S_{d_1} = 6.32$



 $S_{d_2} = 10.87$



 $S_{d_3} = 4070.37$



 $S_{d_1} = 6.32$ $d_1 = 20.09$ cm

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 $S_{d_2} = 10.87$ $d_2 = 20.38$ cm



 $S_{d_3} = 4070.37$ $d_3 = 26.41 \text{ cm}$



Manually measured distance was 20.21 cm.

Cotton plant stem detection



amplitude image











"vesselness" filtered





Cotton plant stem detection







Software Demo





Good news: a new stand analyzer has been developed

- Analyzes crop stands in real-time (> 5 mph) and under any lighting conditions.
- Works for corn plants from V2 V9 (~7-8" and above). Potential to measure and count other crops.
- 3) Measures population, interplant spacing, and estimate stem diameter simultaneously.
- 4) Measure multiple rows (2 8) simultaneously.
- 5) GPS-ready for individual crop stand georeferencing and mapping.
- 6) High corn plant stand counting accuracy (>97%) has achieved in preliminary field test.

Indoor phenotyping using ToF 3D imaging

Objective: 3D vision algorithm for the phenometrics related to plant structure and growth such as the number of leaves, leaf length, leaf locations, plant volume, and plant height



3D holographic reconstruction

-Low image resolution necessitates multiple views









3D holographic reconstruction and characterization





Color image of corn plant





Amplitude image from PMD Nano camera



Distance image from PMD Nano camera



Calibration between 2D and 3D camera

• The Rotation matrix and translation between 2D and 3D camera:

 $R_{3D22D} = R_{L22D}R_{3D2L}$

 $t_{3D22D} = R_{L22D} t_{3D2L} + t_{L22D}$

 $Q_{2D} = R_{3D22D} Q_{3D} + t_{3D22D}$



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Acquiring multiple views





3D Registration

The relationship between 2D camera and target array is achieved

 $Q_{2D} = R_{2D}Q_w + t_{2D}$

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• Convert different 3D point cloud data view to consistent world coordinate system defined by the target array:

$$Q_w = R_{2D}^{-1}(R_{3D22D}Q_{3D} + t_{3D22D} - t_{2D})$$











Physical parameter measurements

- Leaf skeleton estimation
 - Singular Value Decomposition (SVD) regression method

 $\tilde{y} = x \sin(\varphi) + y \cos(\varphi)$

$$z = a\tilde{y}^4 + b\tilde{y}^3 + c\tilde{y}^2 + d\tilde{y} + e$$

- Leaf length
- Leaf width
- Leaf area
- Leaf collar height



Results and Discussion

3D reconstruction result of plant 1







Stem and leaf recognition result





Robotic sampling and plant treatment - sensor and hand coordination







2. Mobile Robotic Platforms - AgRover – A Field Scouting Robotic Vehicle



- Self-leveling on slopes
- Adjustable clearance
- 4-Wheel-Drive
- 4-Wheel-Steering
- Manual wireless
 - Operation
- GPS-based autoguidance





AgRover Prototype



















AgRover-2 Scouting Robot





Thank You!



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