



## ROBOTICS FOR SPECIALTY CROPS: PAST, PRESENT AND PROSPECTS

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# Introduction

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- Specialty crops:
  - ▣ Fruits and vegetables, tree nuts, dried fruits, horticulture, and nursery crops (including floriculture).
- In 2007, the value of farm-level production totaled \$50.1 billion, representing more than one-third of the value of U.S. crop production.

# Production challenges

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- A workshop titled “Engineering Solutions for Specialty Crop Challenges,” was held April 24-25, 2007 in Arlington VA.
- Representatives from various specialty crop industries and from several federal agencies planned and organized the meeting.
- Attendees included federal program managers, specialty crop industry producers and representatives, and researchers, educators, and outreach specialists from numerous universities.

# Production challenges

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- Three major industry concerns were identified:
  - Product quality
  - Environmental footprint
  - Labor cost and availability

# Engineering advances needed

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- Attendees anticipated that the following engineering advances and technologies are needed:
  - ▣ improved and readily available sensors to increase knowledge of plant growing conditions and product quality
  - ▣ more efficient application/use of water, nutrients, and chemicals
  - ▣ automated systems that can reduce costs of harvest and cultural practices
  - ▣ better economic models and decision support systems that can improve production and management decisions.

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## Robotics - mechatronics

The marriage of mechanical devices with electrical and computational systems is referred to as 'robotics' or 'mechatronics'.

# Agricultural robots

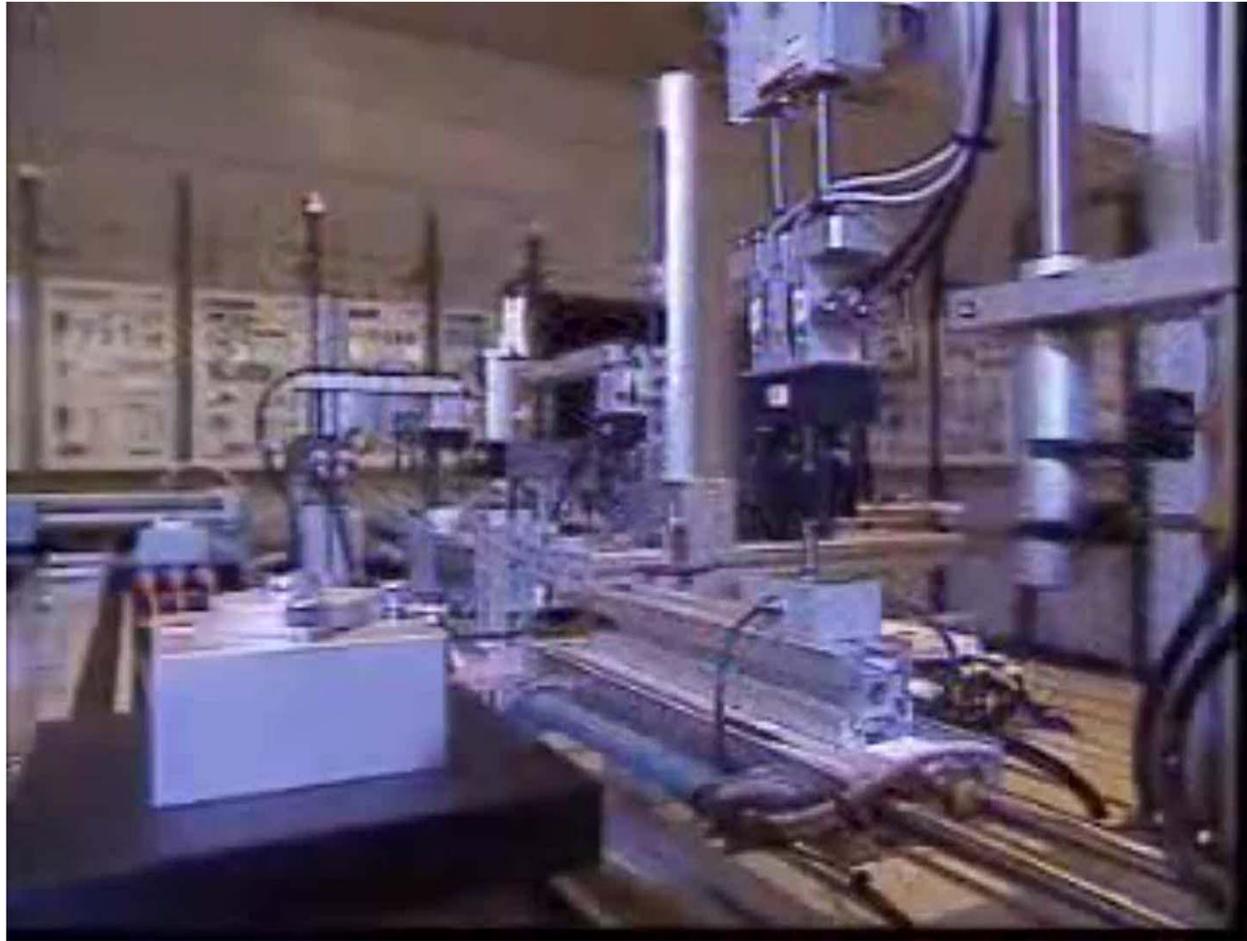
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- An agricultural robot incorporates three basic components\*:
  1. A sensing system to measure important physical and biological properties of the agricultural system
  2. decision-making capabilities for processing information from the sensor system to determine how the agricultural system should be manipulated
  3. actuators to manipulate the agricultural system accordingly.

\*Harrell, R.C., Slaughter, D.C., Adsit, P.D., 1988. Robotics in agriculture. In: Dorf, R.C. (Ed.-in-Chief), International Encyclopedia of Robotics Applications and Automation. John Wiley & Sons, Inc., New York, pp. 1378–1387.

# Industrial robots vs. Ag robots

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# Grain crops 'robotics'

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# Ag robots for specialty crops

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- Large 'lag' with respect to industrial and commodity crop robotics.
- Such systems may offer solutions for
  - ▣ improved and readily available sensors to increase knowledge of plant growing conditions and product quality (SENSING)
  - ▣ more efficient applications/use of water, nutrients, and chemicals (APPLICATION)
  - ▣ automated systems that can reduce costs of harvest and cultural practices (MANIPULATION).



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## SENSING

A new era in remote sensing of crops with unmanned robots.  
Pablo J. Zarco-Tejada, J. A. J. Berni, L. Suárez, and E. Fereres  
17 December 2008, SPIE Newsroom. DOI: 10.1117/2.1200812.1438

# Sensing for specialty crops

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- NEED: periodic and on-demand sensing of plant growing conditions and product quality
  - ▣ Data for plant, soil, environment and fruits.
- Major applications
  - ▣ Disease detection
  - ▣ Yield estimation
  - ▣ Water stress detection
  - ▣ Fruit ripeness estimation.

# Sensing for specialty crops

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- Two major approaches
  - ▣ Sensor networks
    - Periodic and on demand data
    - Large number of simple, low-cost, static sensors
    - Issues of energy, calibration, ruggedness, communication
  - ▣ Robotic scouting
    - On demand data
    - Mobile units carry complex, expensive instruments
    - Issues of autonomous and safe operation.

# Robotic scouting: UAVs

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## □ Aerial imaging

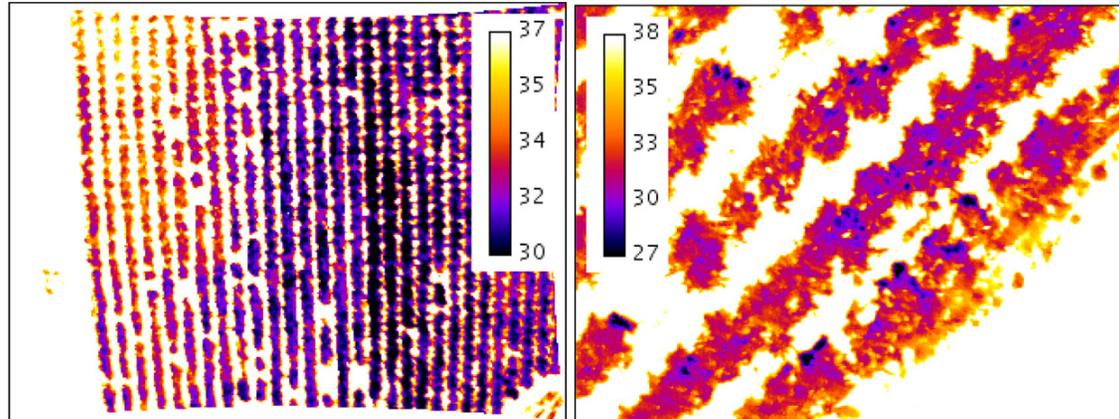


- UAV cost: from ~ \$7k to \$100k.
- Flight height ~ 100 – 2000 ft.
- Resolution ~ 1-10 inch/pixel.

# Robotic scouting: UAVs

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- Data on demand (photos, NDVI, thermal, multi-spectral).
- Example:



Thermal images (expressed in °C) acquired with a helicopter UAV over a peach orchard showing water-stress spatial variability (left), and within-crown thermal differences as a function of stress (right)

- Camera cost: from \$100 to \$15k or more.

# Robotic scouting: UAVs

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## □ Issues

### □ Cost

□ Automatic take-off & landing still not robust

□ Trained operator required

□ In USA, the FAA has not allowed operation yet

□ Legal issues (safety, privacy)

□ Product integration

■ Data interpretation

■ Decision making.



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APPLICATION

# Precise applications for specialty crops

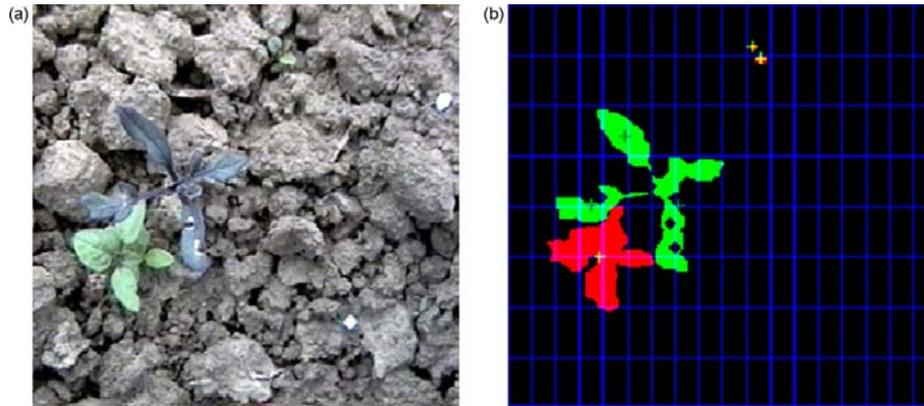
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- NEED: more efficient application/use of water, nutrients, and chemicals
- Focus on pesticide spraying
- Two major approaches:
  - ▣ Vision based
  - ▣ Seed map based

# Vision based spraying

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- Weeds are detected using computer vision



- Micro-spraying is performed on weeds; precision up to 0.5 inches;
- Issues: the lack of robust weed sensing technology is the main limitation.

# Vision based spraying

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- Precision spraying can be performed by
  1. A robotic implement on a tractor
  2. A special-purpose mobile robot
- A recent start-up is focusing on (2) and uses vision and machine learning; venture capital has been invested.

# Seed map based spraying

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- Location of plant seeds is recorded with sub-inch accuracy (seed map)
- Inter-plant space is sprayed based on seed map
- Issues:
  - ▣ High equipment cost for RTK GPS
  - ▣ Localization robustness
  - ▣ An “open-loop” approach
  - ▣ Only at research level (UC Davis, Danish Inst. Of Agricultural Science)

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MANIPULATION

# Manipulation for specialty crops

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- NEEDS: Automated
  - Mechanical weeding
  - Pruning
  - Thinning
  - Harvesting
  - Nursery transplanting
- Robotic harvesting is the “Holy Grail” of agricultural robotics.

# Robotic harvesting

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- Technical feasibility has been shown for apples, citrus, strawberries, and other crops.



- Closed-loop approach: visually detect fruit, reach it and grasp using visual feedback.

# Robotic harvesting

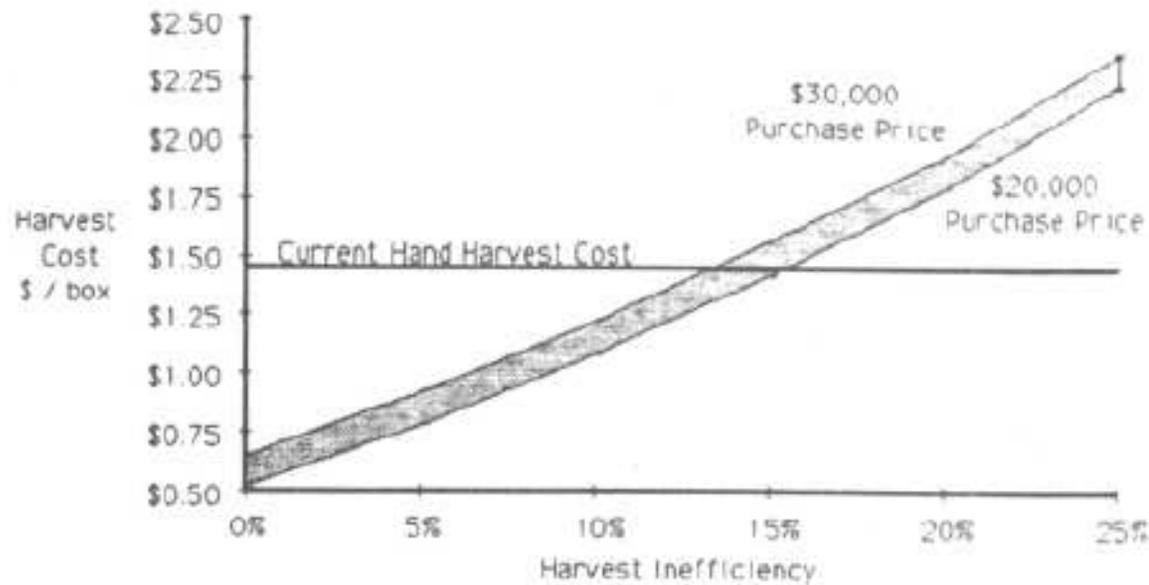
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- Technical issues:
  - Un-harvested fruits (harvest inefficiency)
    - Not detected; not reached; dropped - bruised
  - Picking throughput (fruits/sec)
    - $\sim 5$  sec/fruit; human is  $\sim 1$  sec/fruit
- Financial issues
  - Robot picking cost (human is  $\sim 1$  c/fruit)
  - Seasonal utilization
  - Segmented market

# Robotic harvesting

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## □ Economic study for citrus (1989)



## □ Lack of updated studies.

# Robotic fruit harvesting prospects are positive

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- Labor issue is becoming critical
  - ▣ Increased cost; harvest loss risk
- Cost of robotic manipulators is falling
  - ▣ From \$100k to \$10k or less
  - ▣ Massively parallel -> high throughput
- Cost of computing is low and power is tremendous
- Robotics theory has matured

# Robotic fruit harvesting prospects are positive

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- Renewed funding interest
  - Washington Apple Commission plans major effort
  - Venture capitalists are looking into agriculture

- BlueRiver



- Harvest Automation



# Why are we still at the prototype level?

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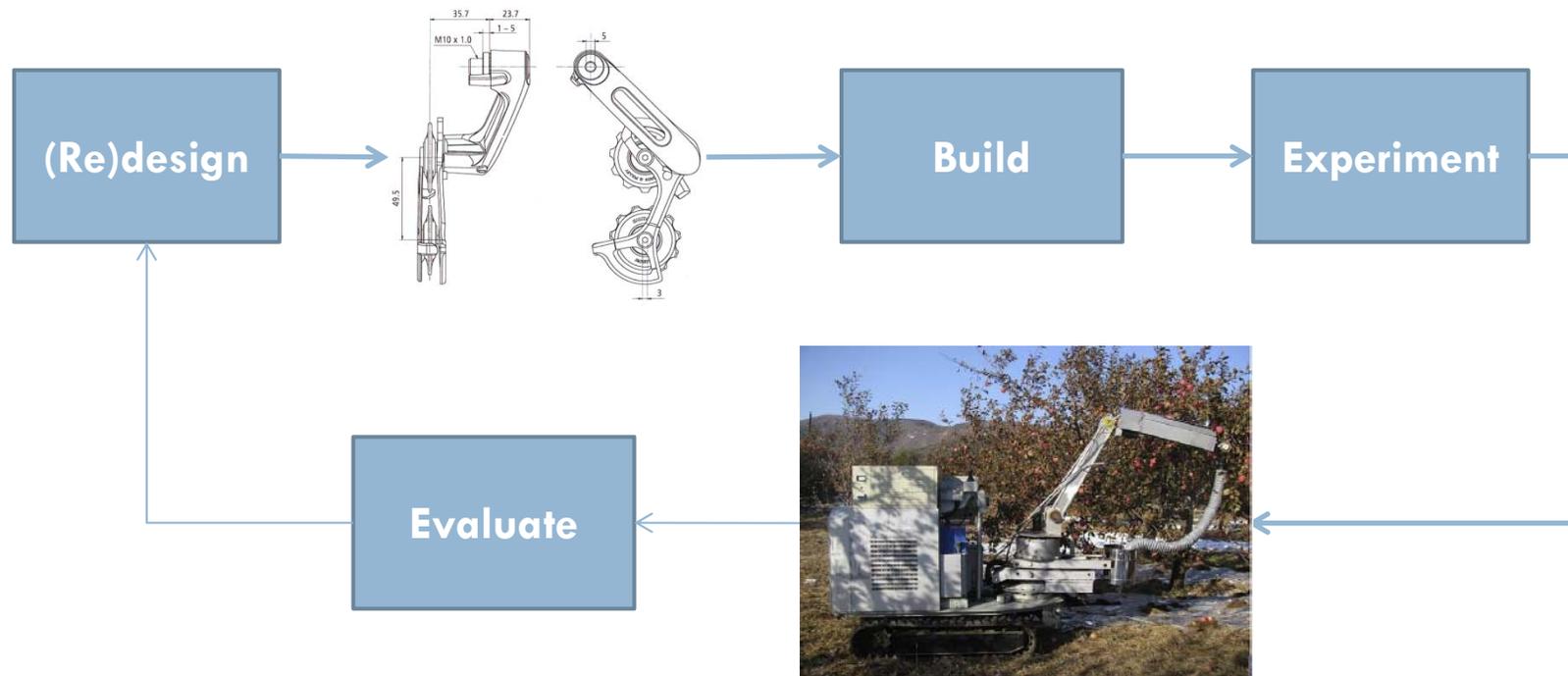
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## Major obstacle

The cost of designing and testing robotic fruit harvesters is very high.

# Current development cycle

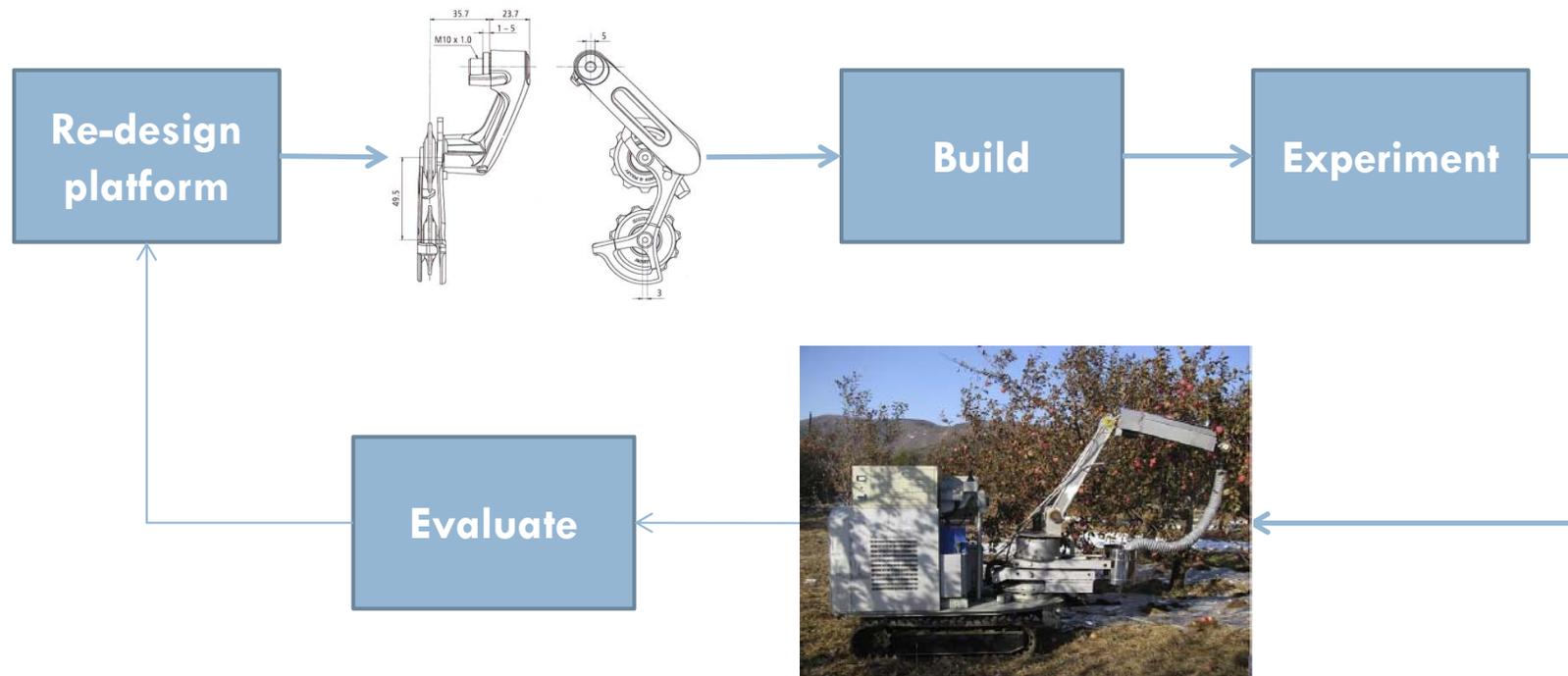
- Design, build, evaluate...



# Limitations

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- Development cycle : “Re-design, build, evaluate”
  - cycle relies on actual harvesting tests
  - costly & slow ( $\sim 1$  cycle/year).



# More limitations

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- Experimental evaluations are not readily transferable to different

**Machines**



**Training systems & orchard layouts**



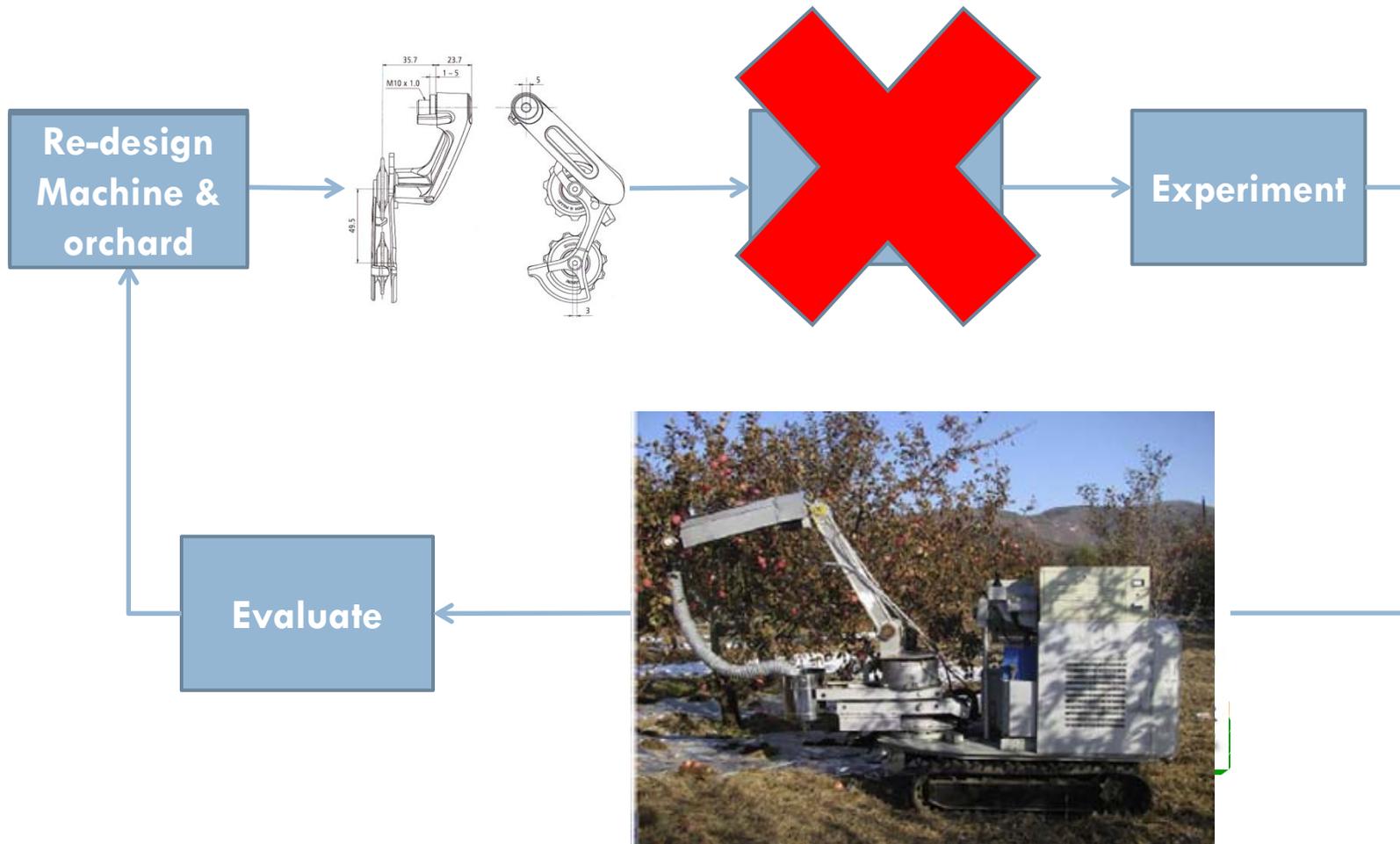
# Productivity of mechanized systems

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- Depends on the interactions of several factors:
  - Tree canopy shape and fruit distribution
  - Orchard layout
  - Machine design: geometry, size, structure, kinematics, dynamics, mobility
  - Fruit transport and logistics.

# 'Digital harvesting'

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# 'Digital harvesting' simulator

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Tree training system & orchard layout



3D fruit distributions



Machine kinematics



# Funding sources for 2012

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## Canning Peach Mechanization Research Fund



## California Pear Advisory Board



# Measuring fruit locations

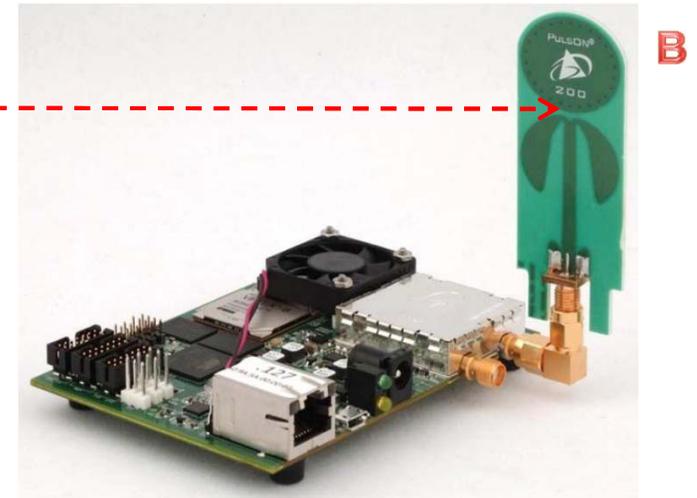
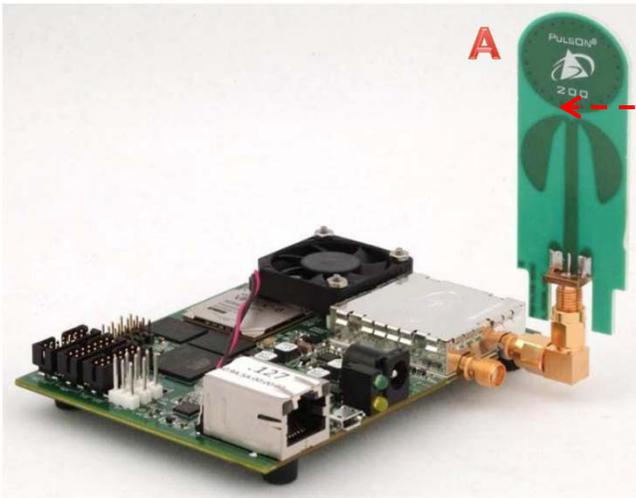
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- Existing approaches: very few reported
  - String & plumb bob (1966)
  - Manipulator & inverse kinematics (1991)
  - Surveying with theodolite (1994)
  - 3D vision (2000)
- Measurement rates  $< 1$  fruit/minute.

# Main idea: track picker's hand position

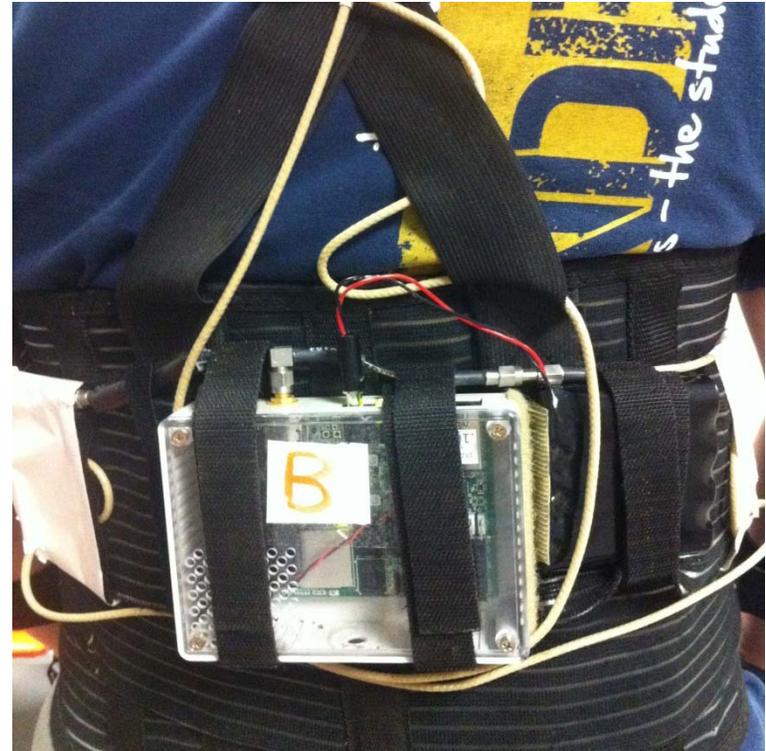
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- Use ranging devices & trilateration



# Wearable devices

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# Trailer with radio beacons

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# Methodology



# Ruddick Ranch (8/21/2012)

## 15 Bartlett pear trees

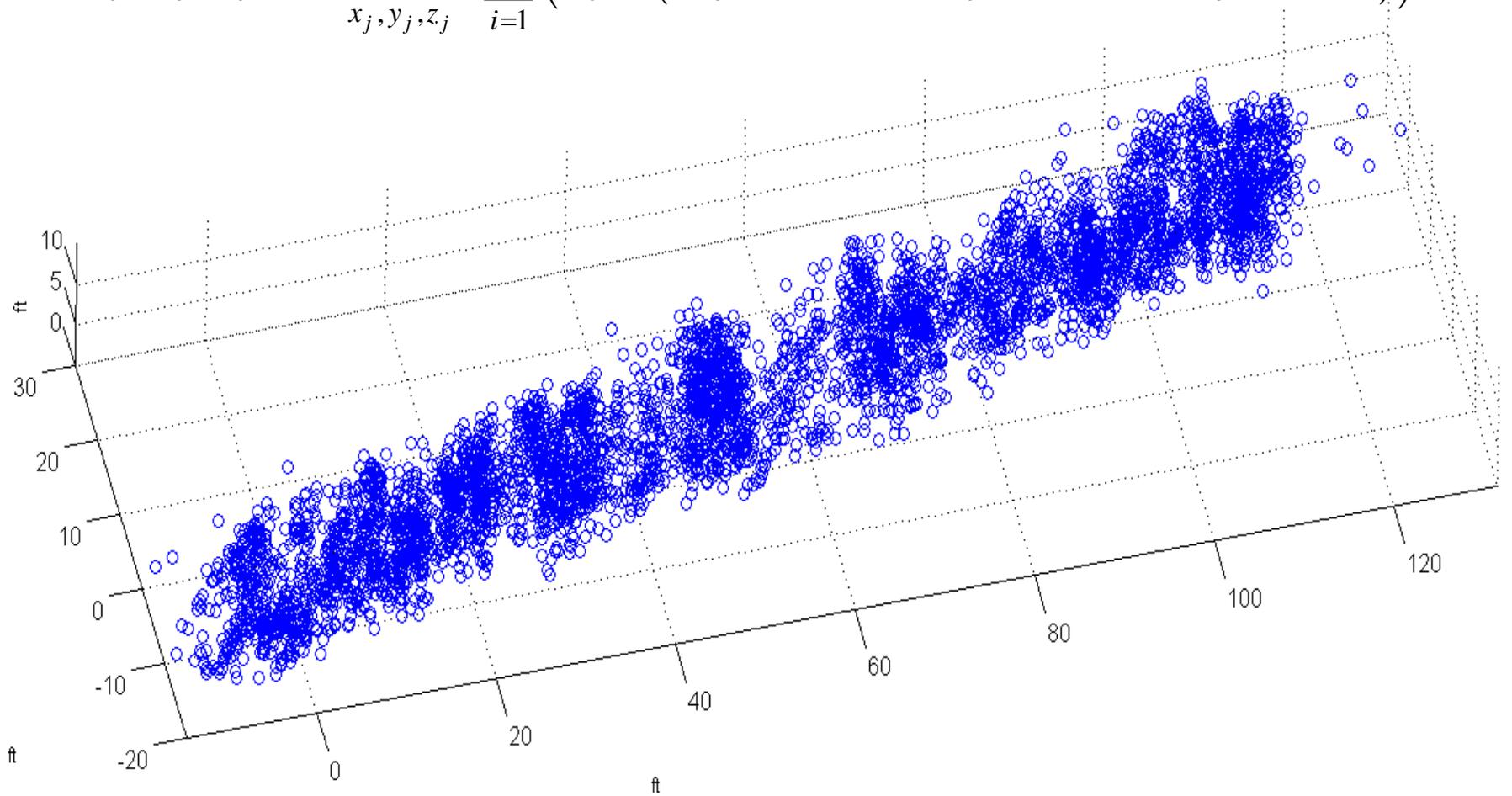
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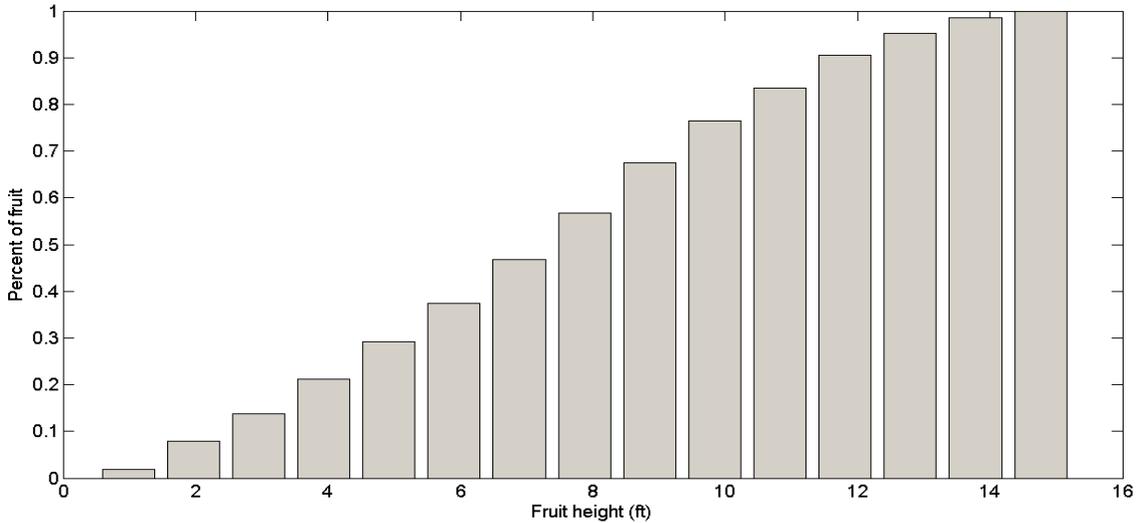
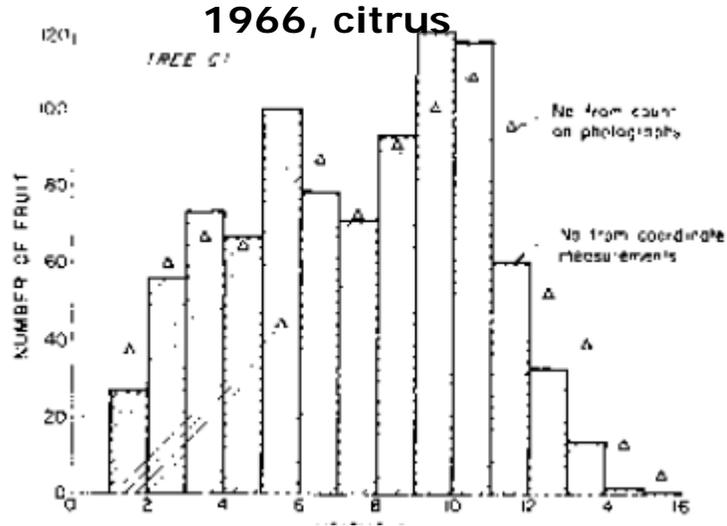
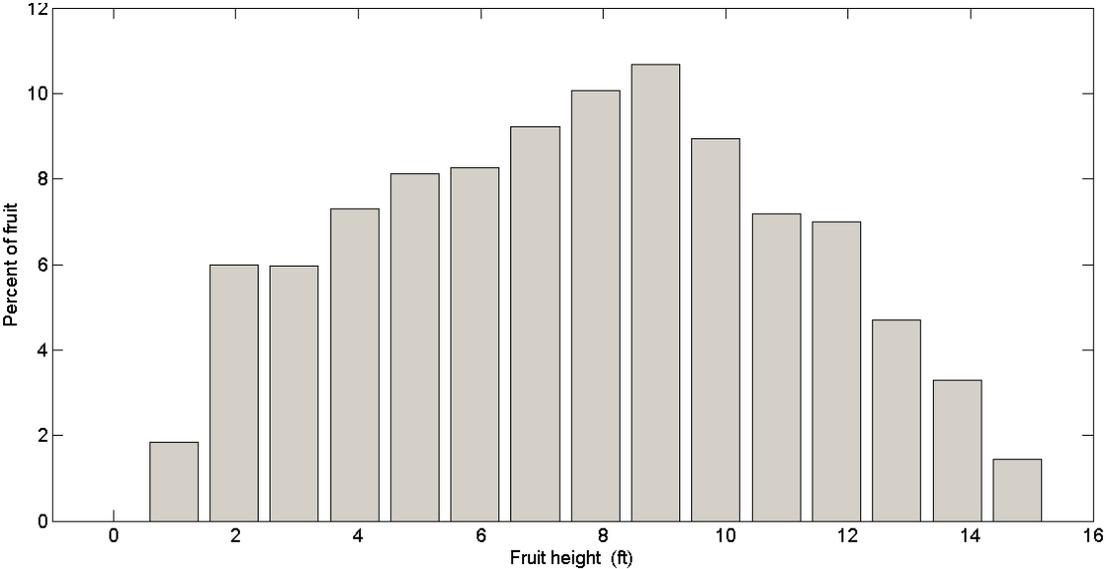
# Fruit locations in the canopies

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$$(x_j^*, y_j^*, z_j^*) = \arg \min_{x_j, y_j, z_j} \sum_{i=1}^4 \left( \hat{r}_{ij}^2 - \left( (x_j - bx_i)^2 + (y_j - by_i)^2 + (z_j - bz_i)^2 \right) \right)$$



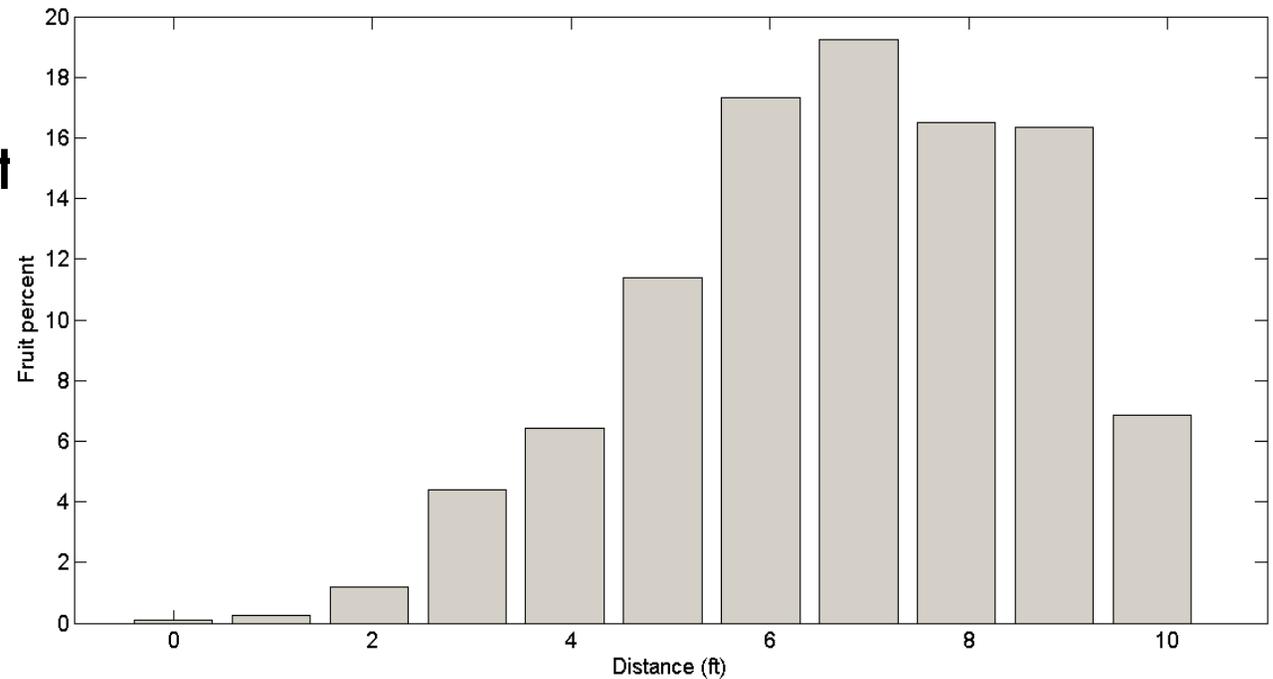
# Fruit height distributions



# Fruit horizontal distances from row centers

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- $E(d) = 6.74$  ft,  
and  $\sigma = 1.88$  ft



# 2013: To be continued...

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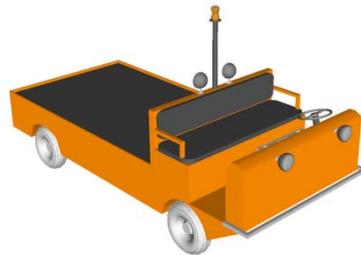
Tree training system & orchard layout



3D fruit distributions



Machine kinematics



## Design tool



Worker kinematics



# THANK YOU!

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