



# Simulation of adaptive site-specific irrigation control performance with spatially variable rainfall

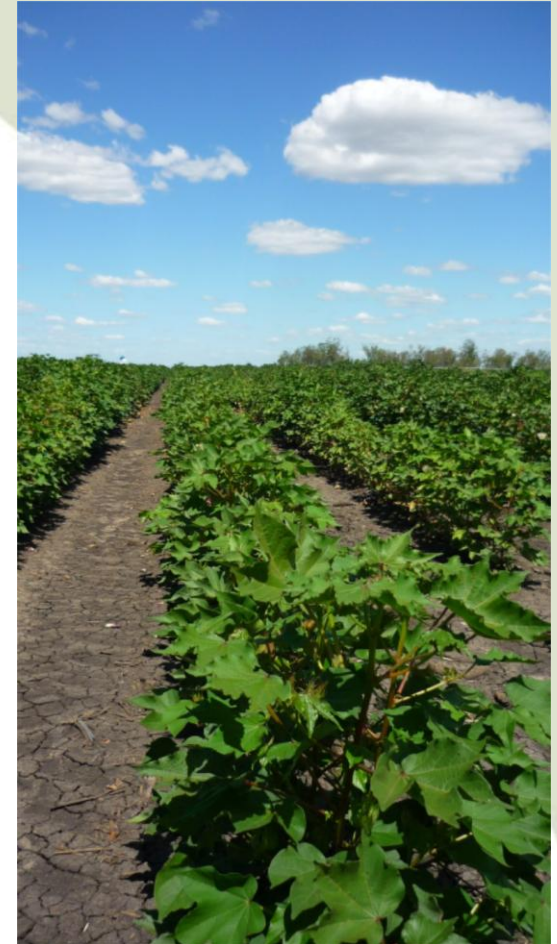
Alison McCarthy, Nigel Hancock and Steven Raine

*mccarthy@usq.edu.au*

# Overview



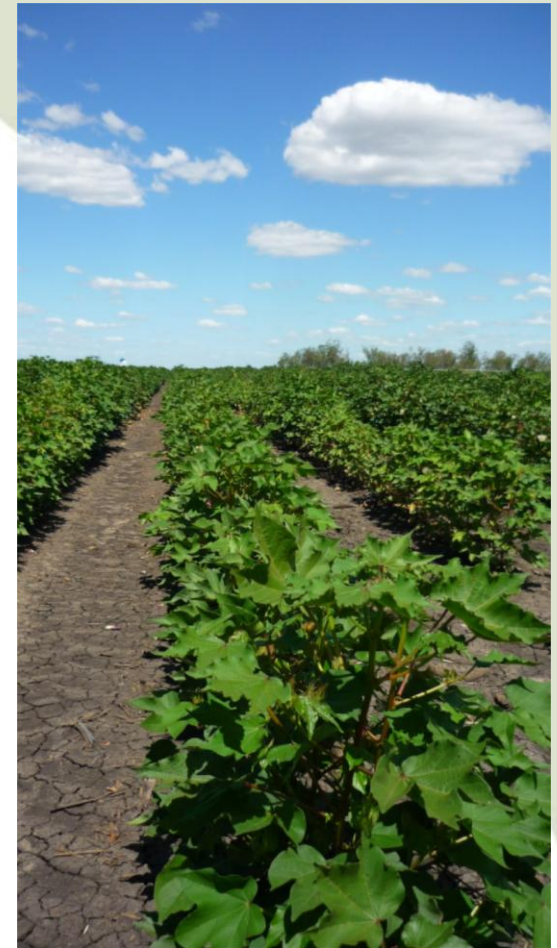
- Background – irrigation management and control
- ‘VARlwise’ software for simulation of irrigation control



# Overview



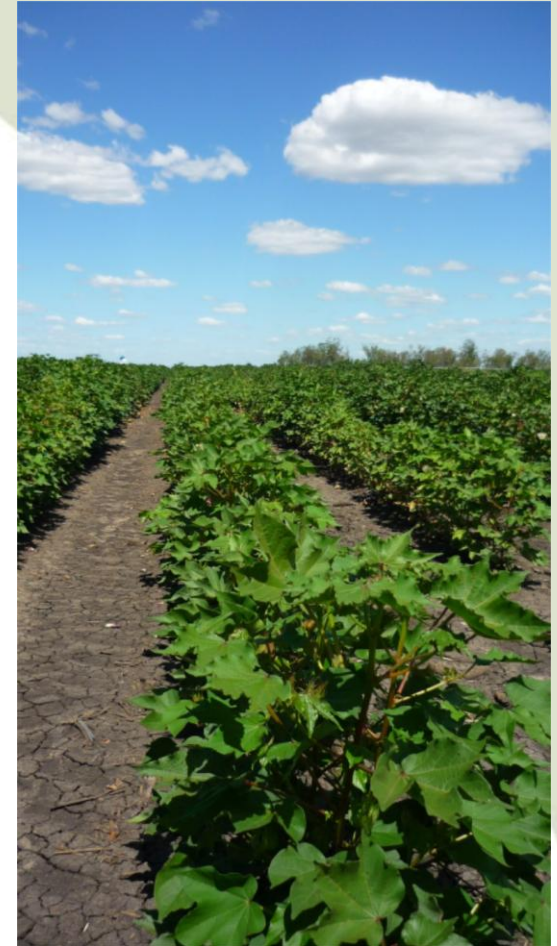
- Background – irrigation management and control
- ‘VARlwise’ software for simulation of irrigation control
- Spatially variable rainfall



# Overview



- Background – irrigation management and control
- ‘VARlwise’ software for simulation of irrigation control
- Spatially variable rainfall
- Simulation of adaptive control strategies
- Conclusion





# Background – 1. irrigation management



1. Improves water use efficiency
2. Can use plant, soil and/or weather data
3. Automatic implementation enables real-time control



# Background – 1. irrigation management



1. Improves water use efficiency
2. Can use plant, soil and/or weather data
3. Automatic implementation enables real-time control

*But:*

4. Affected by spatial and temporal variability



# Background – 2. ‘adaptive’ control



# Background – 2. ‘adaptive’ control



- Accounts for *temporal variability* in the field
- Accounts for *spatial variability* in the field



# Background – 2. ‘adaptive’ control



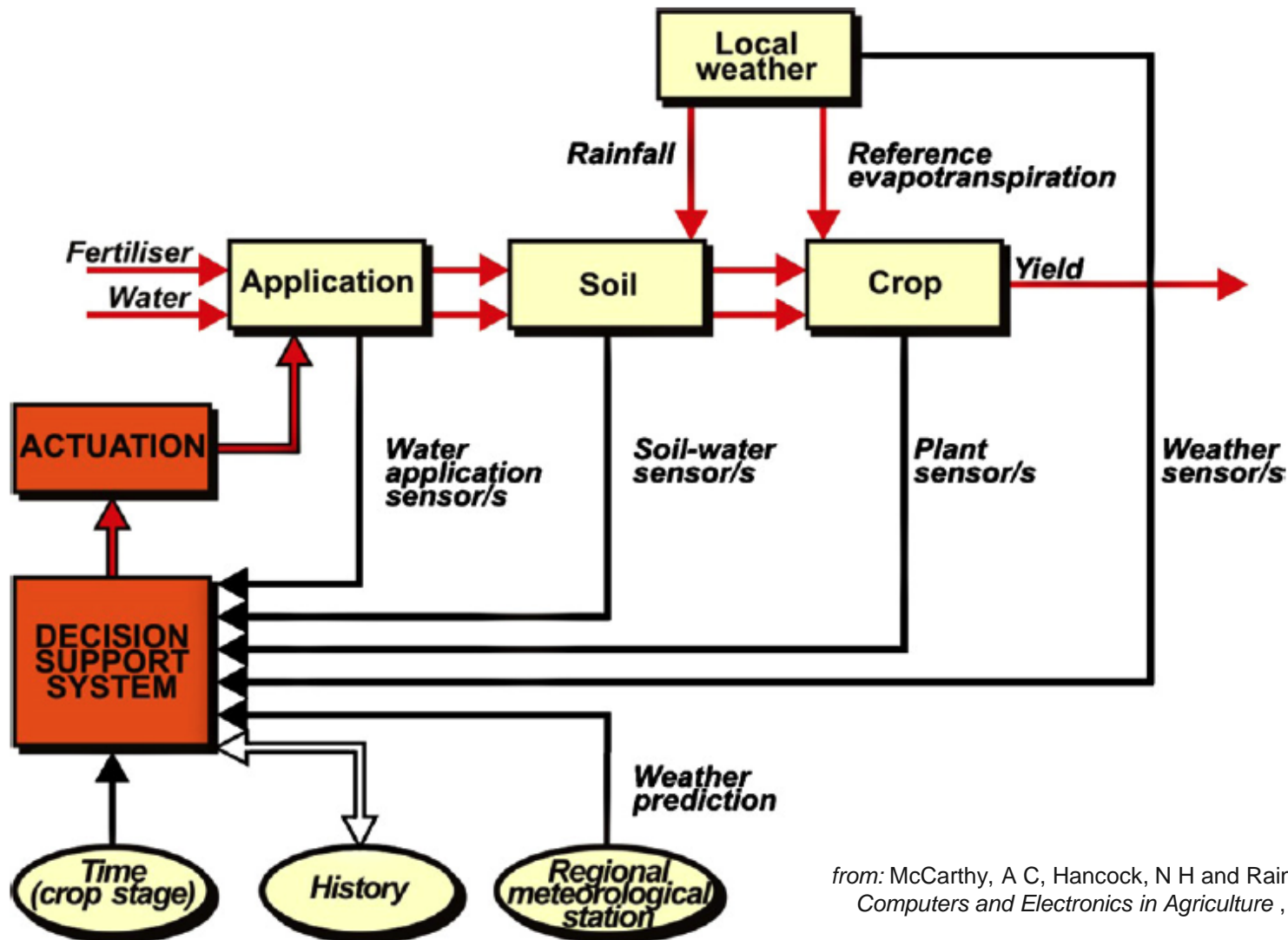
- Accounts for *temporal variability* in the field
- Accounts for *spatial variability* in the field
- Adjusts applicator hardware to improve irrigation timing/volume

# Background – 2. ‘adaptive’ control



- Accounts for ***temporal variability*** in the field
- Accounts for ***spatial variability*** in the field
- Adjusts applicator hardware to improve irrigation timing/volume
- Can improve crop water use efficiency while maintaining or improving crop yield

# Control simulation framework 'VARlwise'

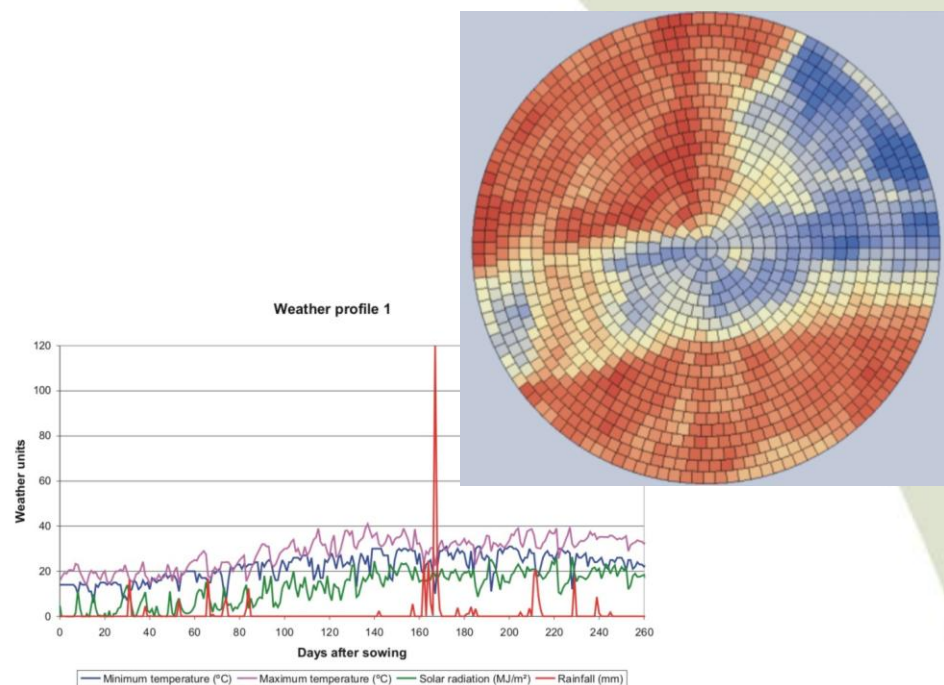


from: McCarthy, A C, Hancock, N H and Raine, S R (2010)  
*Computers and Electronics in Agriculture*, 70(1) 117-128

# Control simulation framework 'VARlwise'



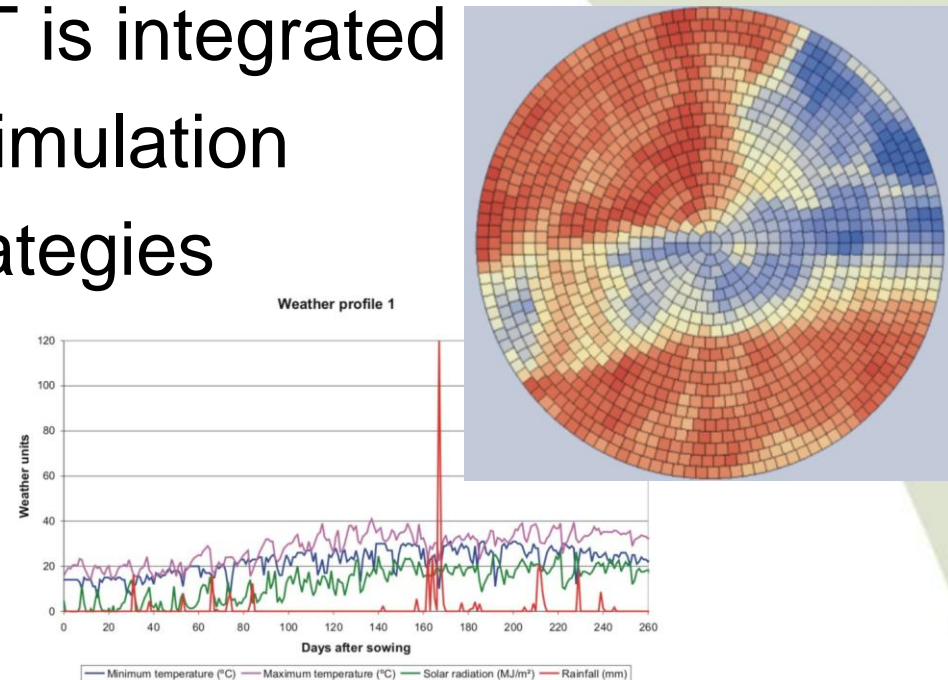
- Input of field scale variations down to one square metre
- Input of data at any temporal scale



# Control simulation framework 'VARlwise'



- Input of field scale variations down to one square metre
- Input of data at any temporal scale
- Cotton model OZCOT is integrated for irrigation control simulation
- Integrates control strategies for CPIMs (and LMIMs)





# Adaptive control strategies:

## 1. *Iterative Learning Control (ILC)*

- Uses the error between the ***measured*** and ***desired*** soil moisture deficit after the previous irrigation,
- . . . to ***adjust*** the irrigation volume of the next irrigation event.

# Adaptive control strategies:

## 1. *Iterative Learning Control (ILC)*

- Uses the error between the *measured* and *desired* soil moisture deficit after the previous irrigation,
- . . . to *adjust* the irrigation volume of the next irrigation event.
- ‘Learns’ from history of prior error signals to make better adjustments.

# Adaptive control strategies:

## 2. *'Model Predictive Control' (MPC)*

- A *\*calibrated\** cotton model simulates and *predicts* the next required irrigation, i.e. volumes and timings
  - according to evolving crop/soil/weather input
  - separately for all cells/zones
  
- *\*Calibration\** is adjusted according to *measured* (or emulated) crop response

# Adaptive control strategies:

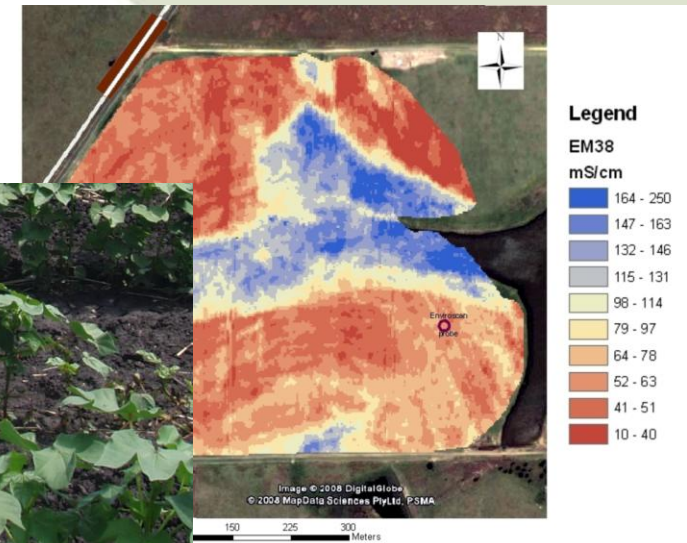
## 2. '*Model Predictive Control*' (MPC)

- A \*calibrated\* cotton model simulates and *predicts* the next required irrigation, i.e. volumes and timings
  - according to evolving crop/soil/weather input
  - separately for all cells/zones
- \*Calibration\* is adjusted according to *measured* (or emulated) crop response
- We might *choose* to implement the MPC irrigation scheme that maximises the final cotton yield (as predicted by the evolving model)

# Evaluating variability in VARlwise

- VARlwise can be used evaluate effect of variability of:

- weather
- soil
- plant



on yield/WUE and control performance



# Spatial variability of rainfall



- Spatial variability of natural rainfall in Queensland summer cropping areas can be substantial



# Spatial variability of rainfall



- Spatial variability of natural rainfall in Queensland summer cropping areas can be substantial
- Typically weather data is only measured for a single point nearby
- Spatial variability of rainfall is often unquantified
- The effect on irrigation optimisation is unknown



# Spatial variability of rainfall

## – Case study methodology

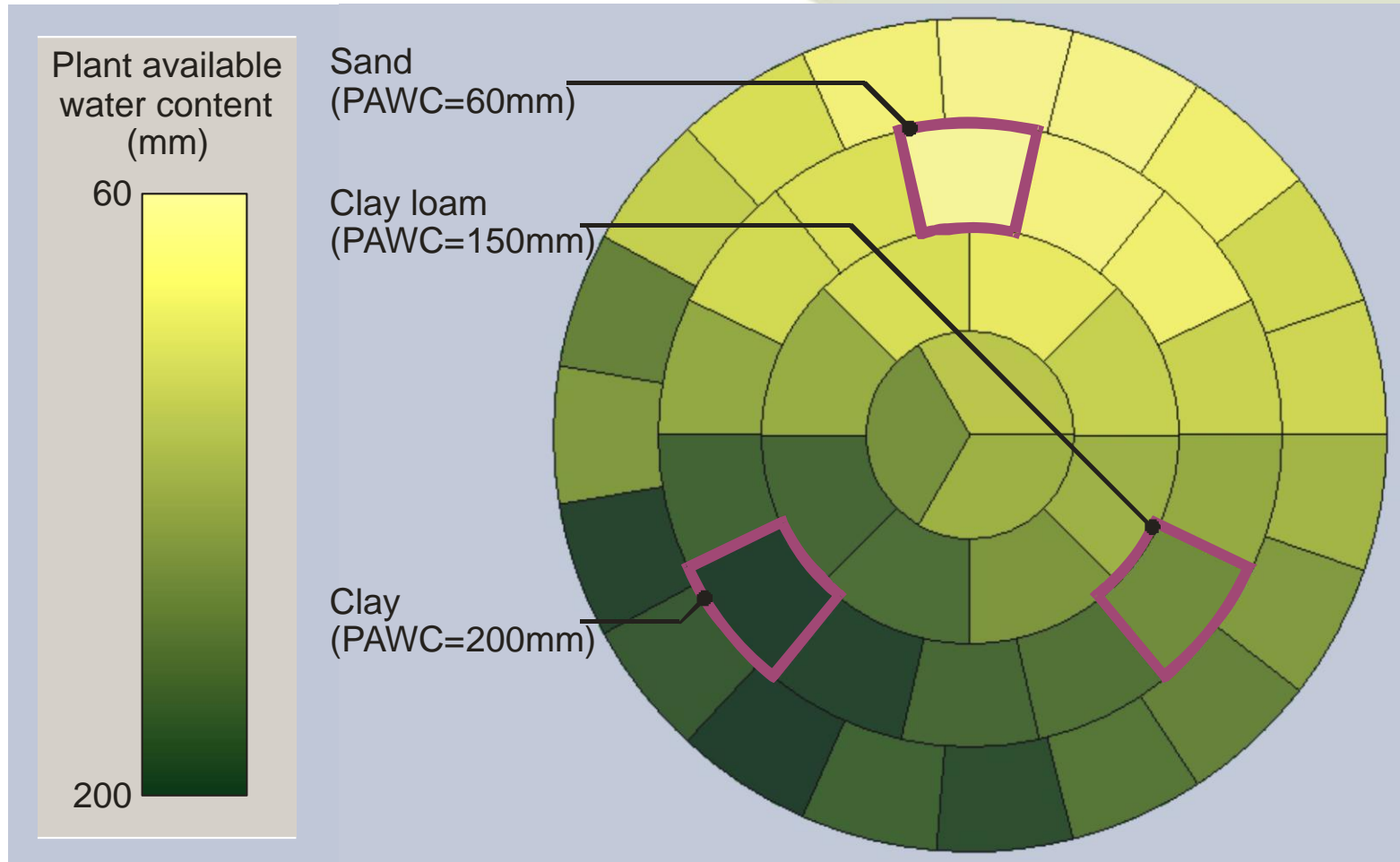
- Rainfall was *spatially varied* by applying a Gaussian distribution of variability to the rainfall measurement in each cell
- Average rainfall values were obtained from SILO weather data
- Two amounts of imposed variability, 20% and 50%, were evaluated
- Ten replicates of each rainfall pattern were simulated

# Spatial variability of rainfall

## – Case study inputs

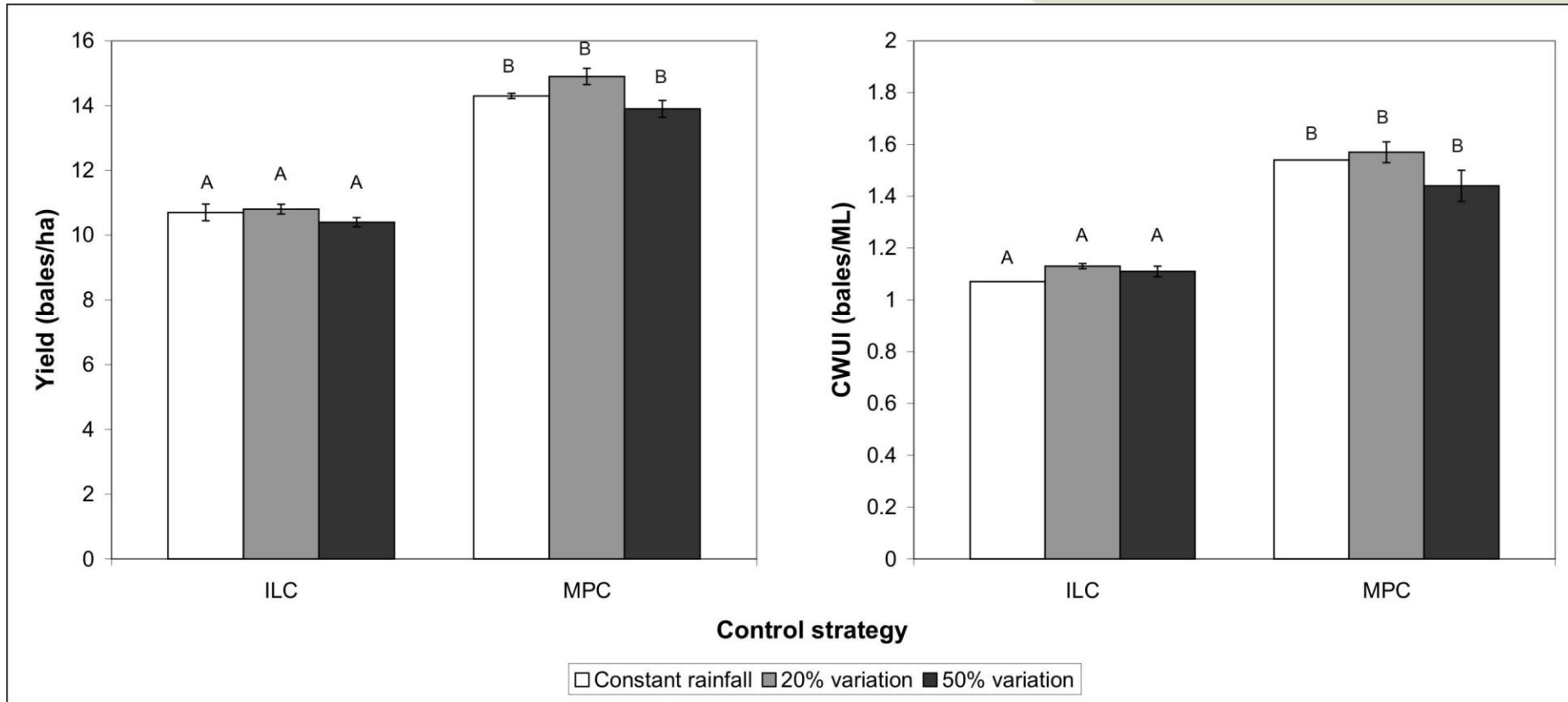
- Cotton was sown on a 400 m diameter centre pivot-irrigated field on 4 October
- The field was automatically divided into 44 cells
- Irrigations occurred until 14 March of the following year
- Initial stored soil moisture = 50 mm

# Case study inputs – soil variability





# Spatial variability of rainfall – Case study results



# Conclusions



- 1. Advanced process control can be successfully applied to irrigation**
- 2. Advanced process control is robust to rainfall variability**  
**i.e. accommodates major rainfall variability with minimal loss of yield**

# Acknowledgements



- Australian Research Council
- Cotton Research and Development Corporation



# Simulation of adaptive site-specific irrigation control performance with spatially variable rainfall

Alison McCarthy, Nigel Hancock and Steven Raine

*mccarthy@usq.edu.au*