

Detecting signals of local adaptation in widespread species

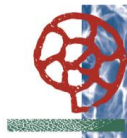


Department of
Environment and Conservation

Our environment, our future



NCCARF
National
Climate Change Adaptation
Research Facility



CENTRE for
**ECOSYSTEM
MANAGEMENT**

EDITH COWAN UNIVERSITY



**University of the
Sunshine Coast**
The best of both worlds

What should we plant?

Current paradigm for revegetation projects:

Use locally sourced seed to:

- maintain current patterns of genetic variation
- avoid problems of:
 - ❖ poorly adapted germplasm
 - ❖ genetic contamination of local populations
 - ❖ outbreeding depression



Image: P Tilyard

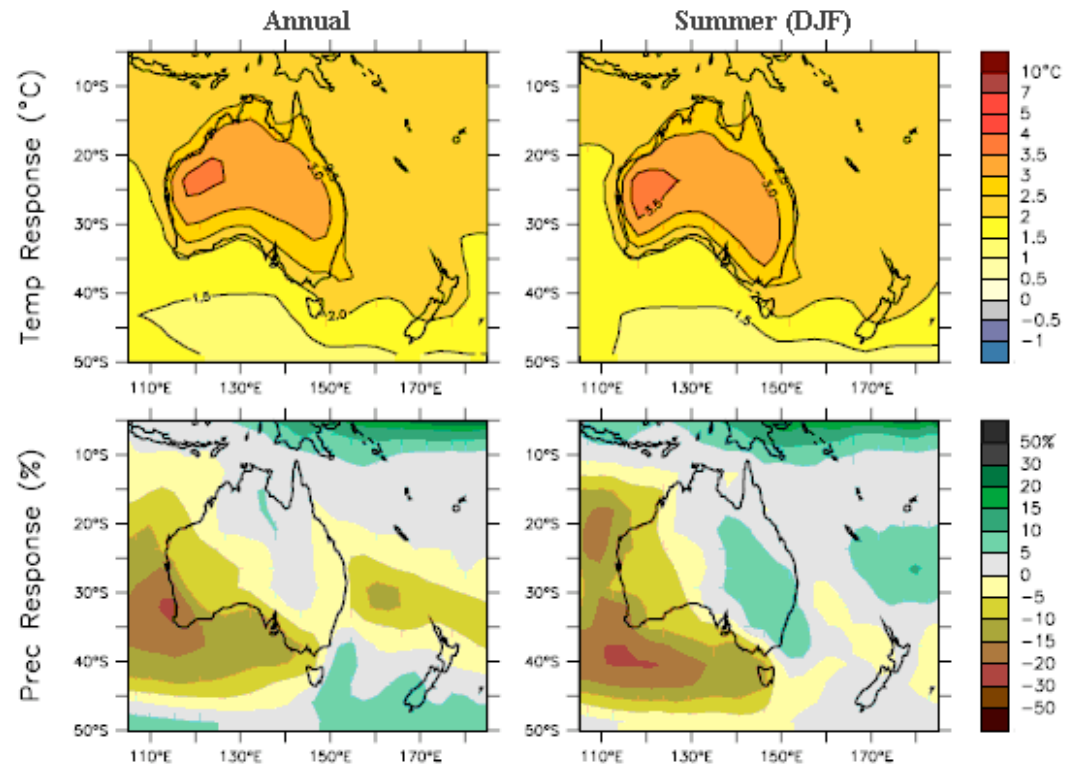


But is this appropriate when environmental conditions are changing?

Models of climate change in Australia predict shifts in geographic and temporal patterns of:

- Rainfall
- Temperature
- Drought
- Flooding
- Fire frequency

Forecast change in Temperature and Precipitation by 2099



Annual mean and summertime (December-January-February) changes for the period 1980-1999 vs. 2080-2099. Image credit: [2007 IPCC report](#).

Widespread foundation species live in diverse environments

Is there adaptive genetic variation across environmental gradients?

If so, can we exploit it to:

- facilitate the persistence of plantings?
- maintain ecosystem functions?



Eucalyptus tricarpa, Victoria

Mean annual rainfall 440-1200 mm
Mean annual temperature 11-17°C



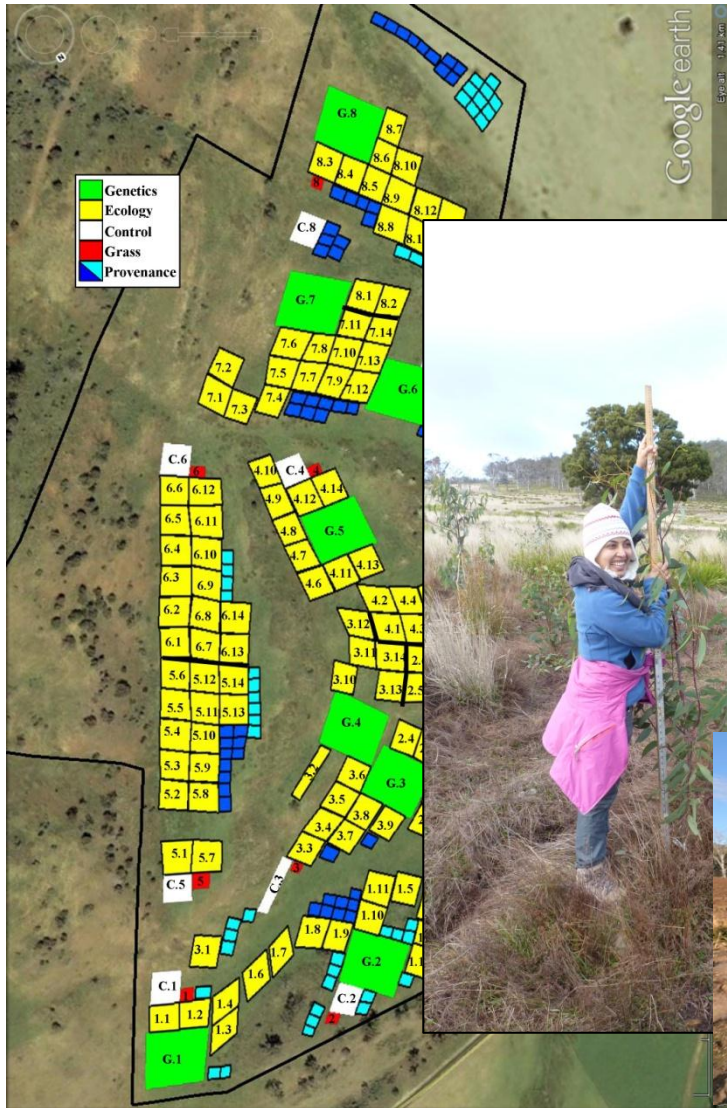
Images: E. Mclean

Detecting adaptive variation

Field trials are best

But ...

- expensive
- time consuming
- long term

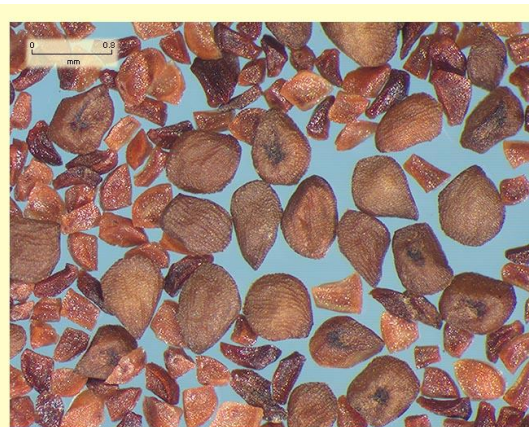


A genomic approach to:

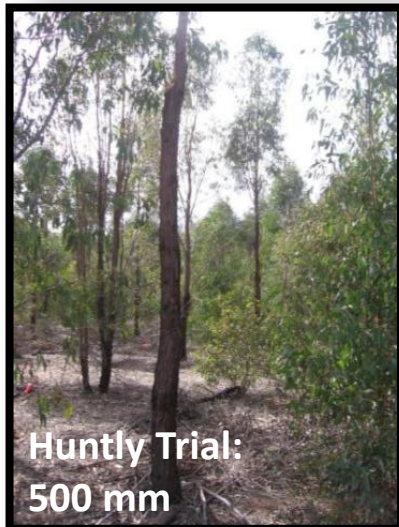
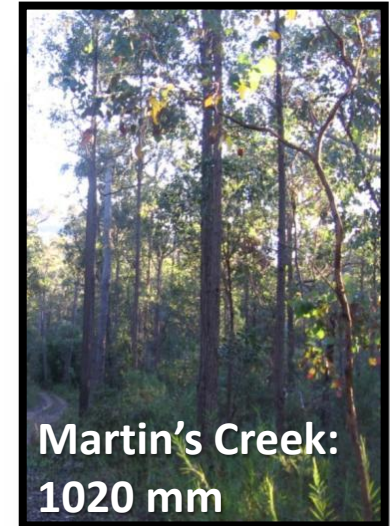
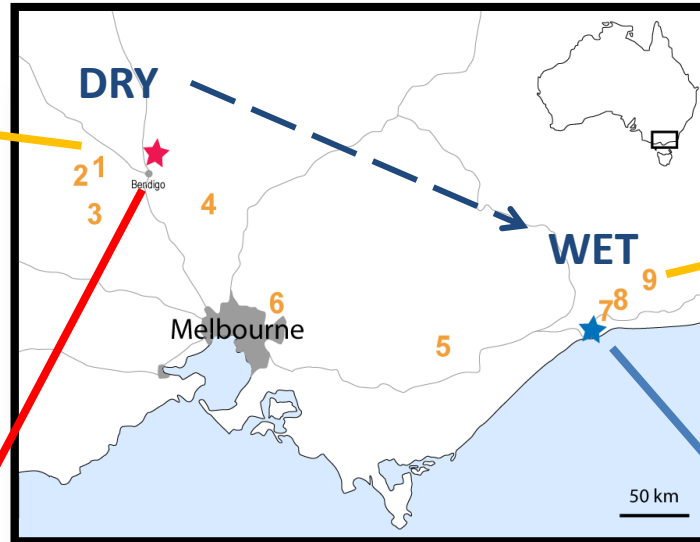
- studying adaptation in restoration species
- developing seed transfer guidelines



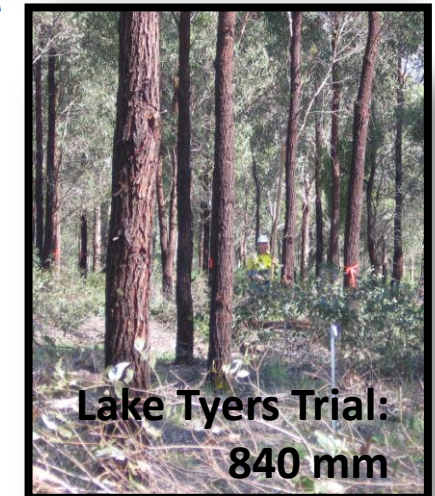
Image: P Tilyard



Eucalyptus tricarpa grows across a rainfall gradient in southeastern Australia



- **Nine provenances**
- **Genomic Analysis:**
30 trees/provenance
DARtseq markers
- **Functional Trait analysis:**
10 trees/provenance in the wild
and in 12 y.o field trials.



Functional trait measurements

Plant, Cell & Environment

PCE

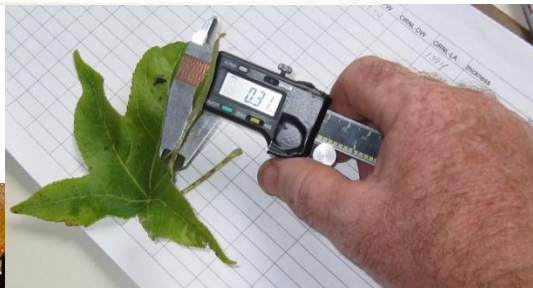
Plant, Cell and Environment (2014) 37, 1440–1451

doi: 10.1111/pce.12251

Original Article

Plasticity of functional traits varies clinally along a rainfall gradient in *Eucalyptus tricarpa*

Elizabeth H. McLean^{1,2}, Suzanne M. Prober², William D. Stock³, Dorothy A. Steane^{4,5}, Brad M. Potts⁴, René E. Vaillancourt⁴ & Margaret Byrne¹



Morphology

Leaf size

Leaf thickness

Leaf density

Specific leaf area

Circumference of main stem

Total cross-sectional area

Tree height

Trait plasticity

Physiology

Cellulose ¹³C

Leaf ¹³C

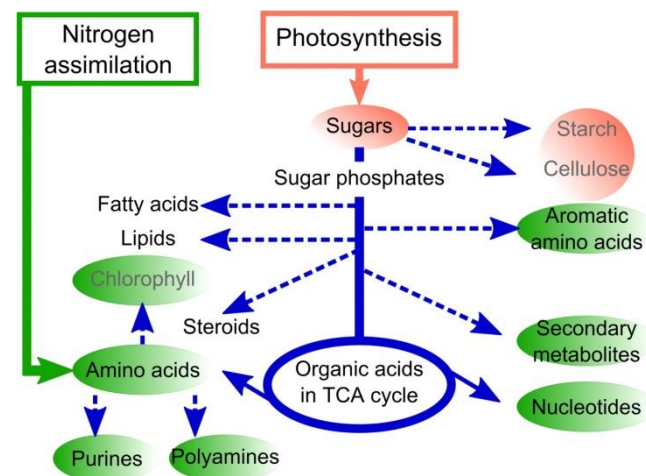
Leaf ¹⁵N

C_{mass}

C/N ratio

Leaf N_{area}

Leaf N_{mass}



Genomic analysis

DArTseq

Genome-wide genotyping using next generation (short-read) sequencing of a set of restriction fragments from whole genome

- (1) Presence/absence data
- (2) Sequence data (70 bp)



Thousands of markers
distributed across
genome



274 individual trees

6,544 DArTseq markers

35 climatic variables -> 15 climatic variables

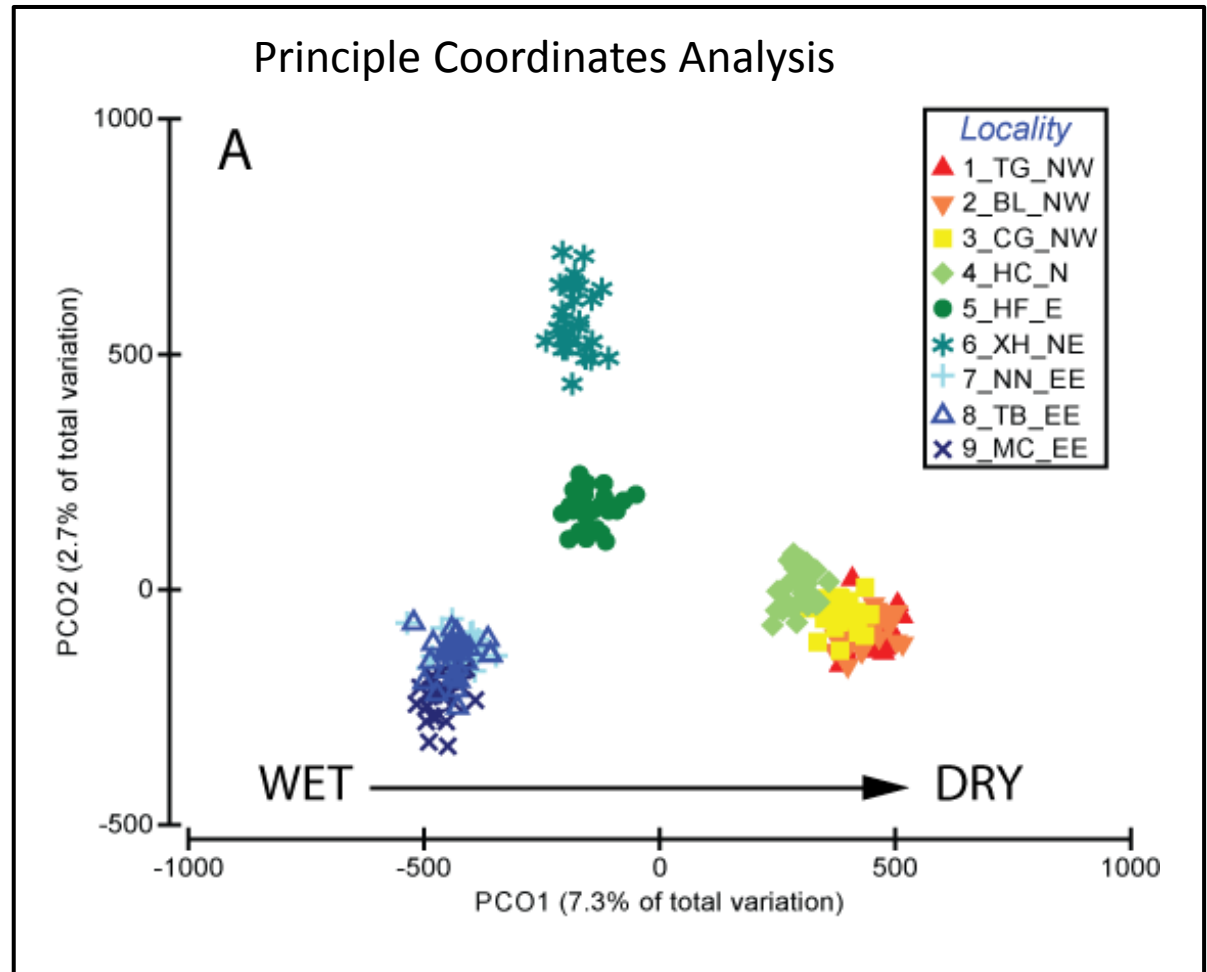
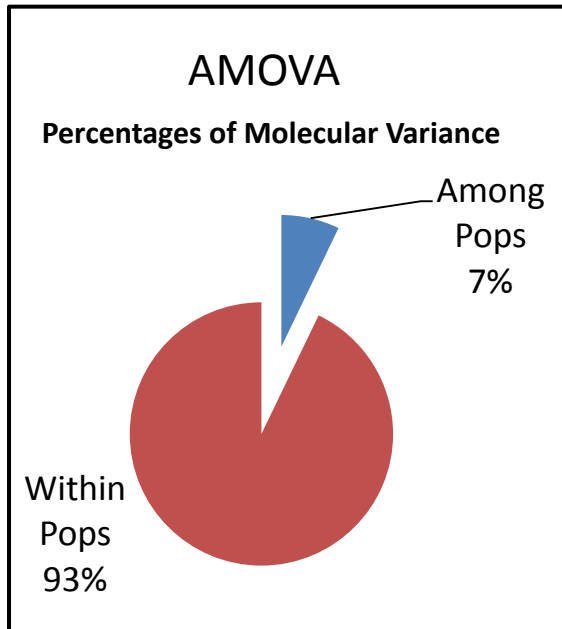
15 functional traits

There is genetic structure across populations

274 trees

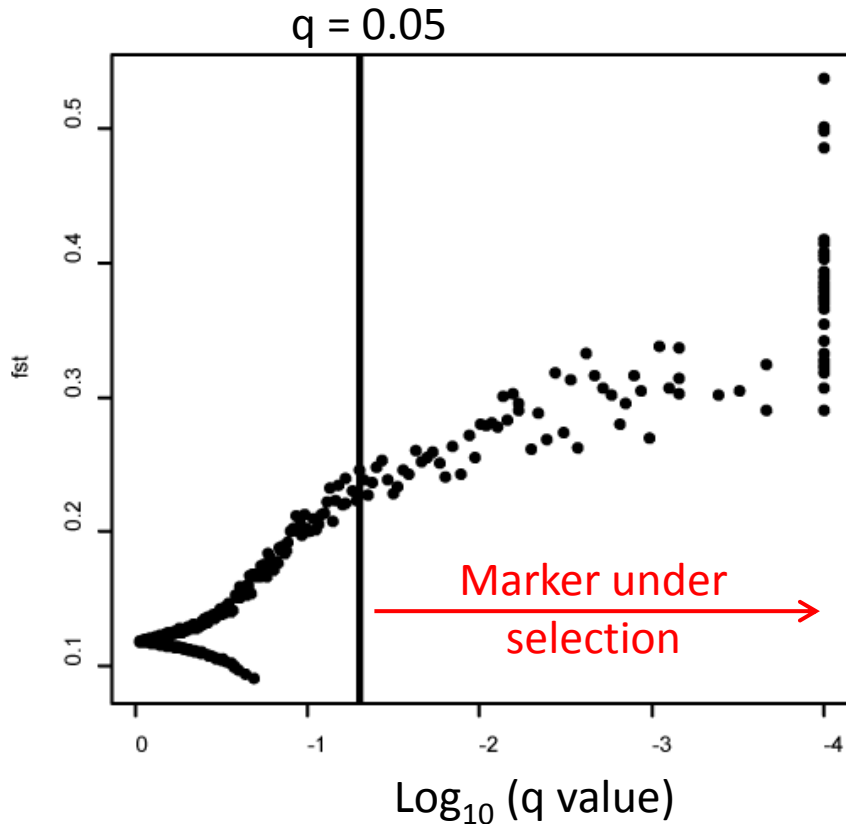
9 populations

6,544 markers



... but how much of it is adaptive?

Fst outlier analysis identifies markers that may be under selection

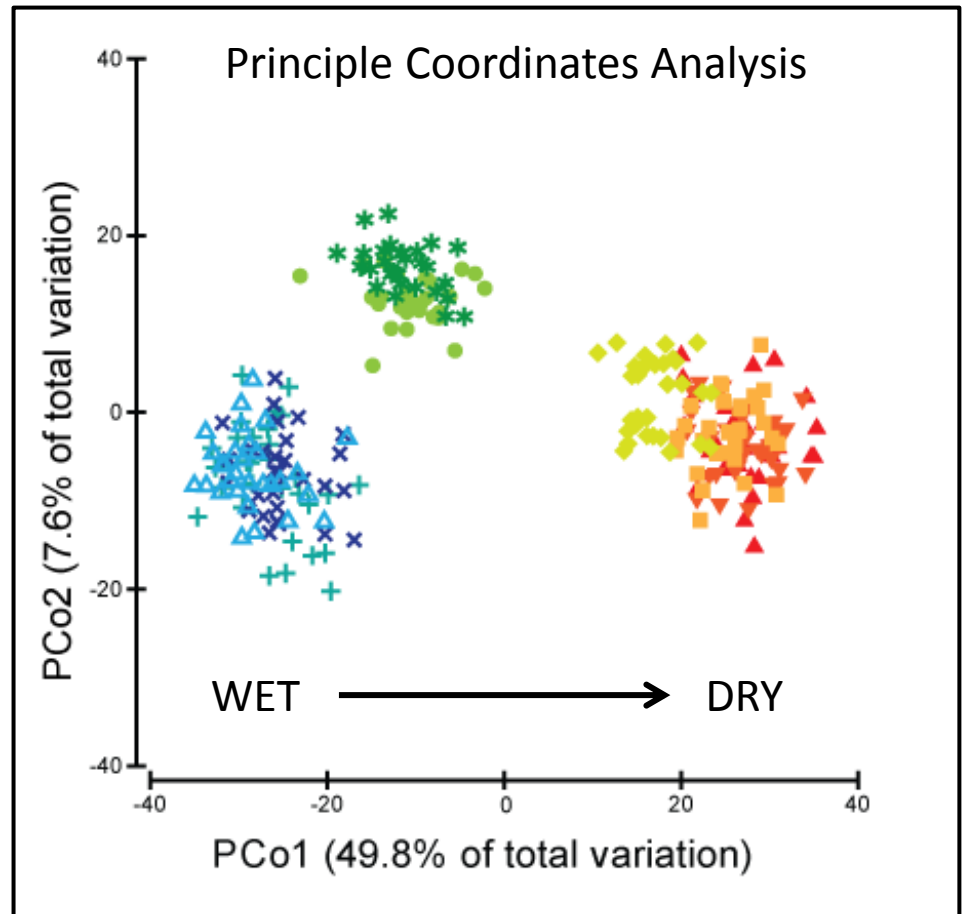
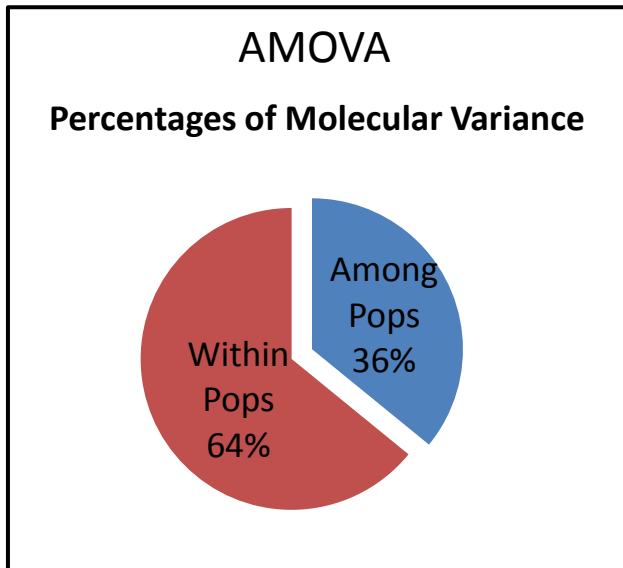


F_{st} – degree of inbreeding within sub-populations relative to the whole population (based on allele frequencies)

Outlier analysis – plots marker F_{st} values against the probability that the allele frequency of a marker differs more among subpopulations than would be expected from chance (drift).

Outliers provide an 'adaptively enriched genetic space'

274 trees
9 populations
94 outlier markers



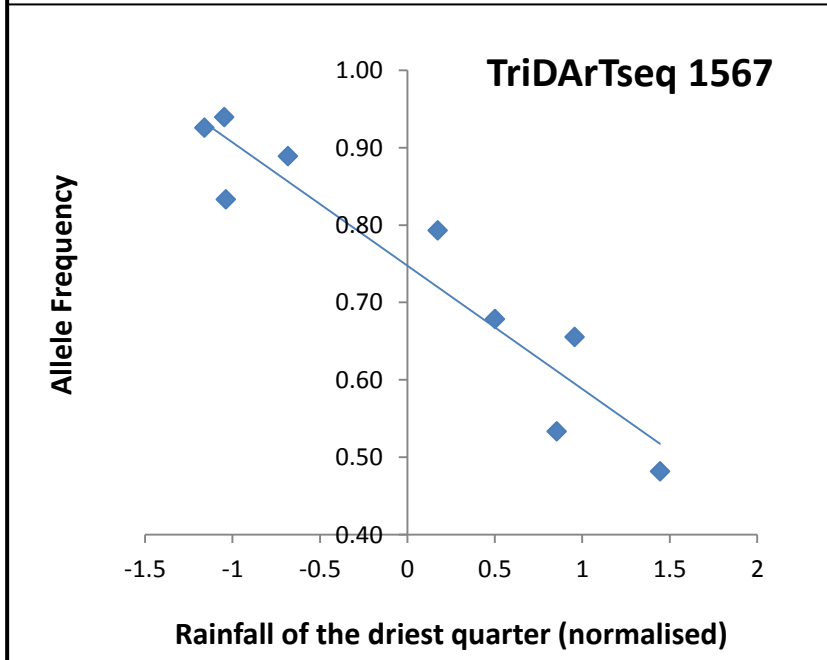
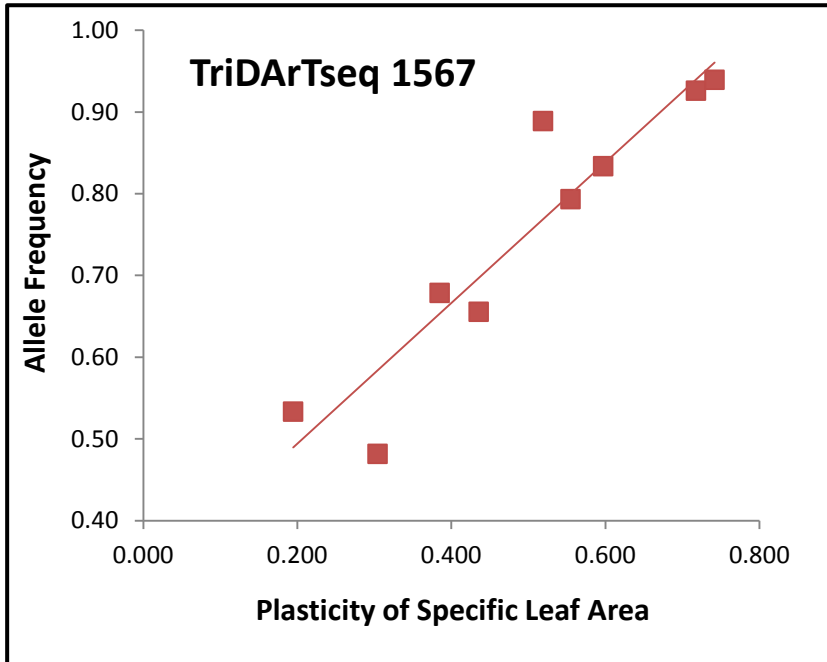
But what evidence is there that the outlier markers are ‘adaptive’?

*Are allele frequencies across populations correlated with
(i) functional trait variation, or
(ii) changes in an environmental variable?*

We did LOTS of linear regressions!

	94 outliers	3,590 neutral
35 Climatic variables	3,290	125,650
15 Soil variables	1,410	53,850
14 Wild population traits	1,316	50,260
28 Common garden traits	2,632	100,520

... correcting for multiple testing using a ‘Dependent False Discovery Rate’ (DFDR) of 5%



All outlier loci were correlated with climate and/or functional traits

100% associated with at least one climate variable (8X*)

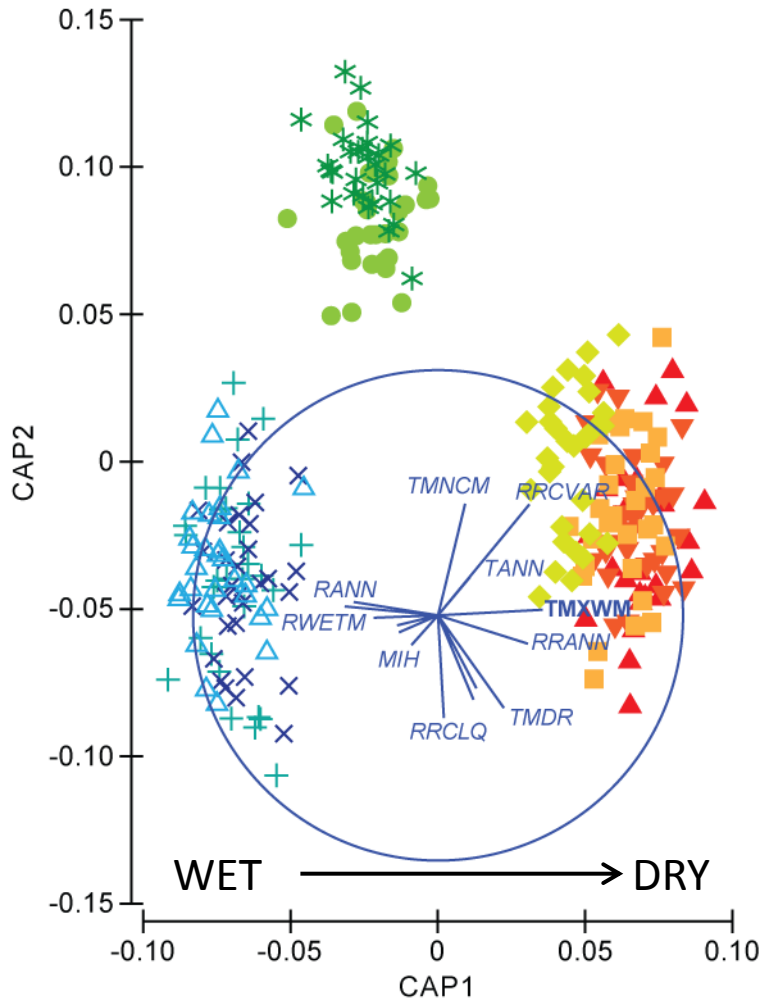
82% associated with at least one functional trait (3X*)

75% associated with both functional traits and climate variables

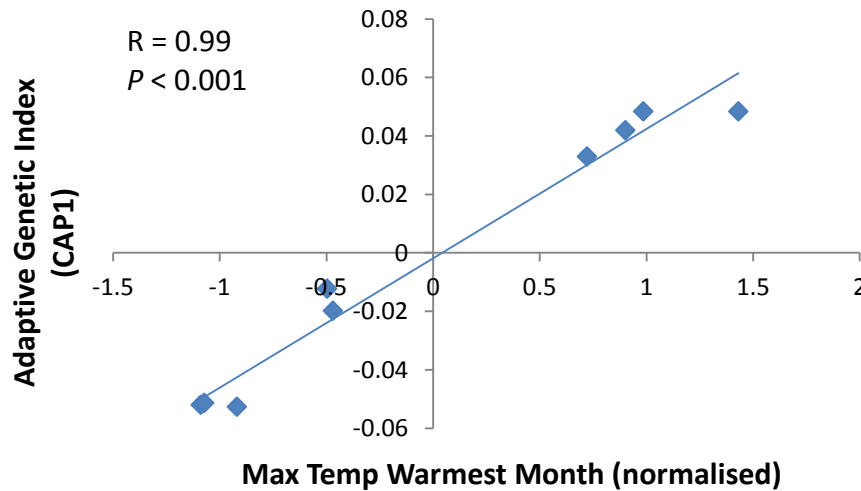
*increase in number of marker-trait associations ($P < 0.001$) relative to neutral markers

Canonical Analysis of Principle Coordinates provides a 'climate-aligned adaptive genetic index'

CAP1 represents the direction of molecular change most closely associated with change in climate.

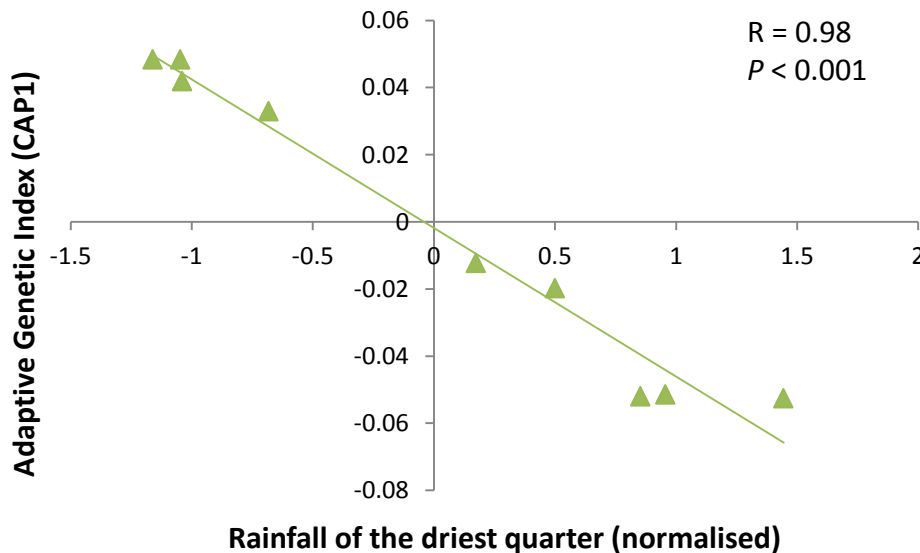


Maximum temperature of the warmest month



Population-level variation in outlier markers, as described by CAP1, is strongly associated with climate variation.

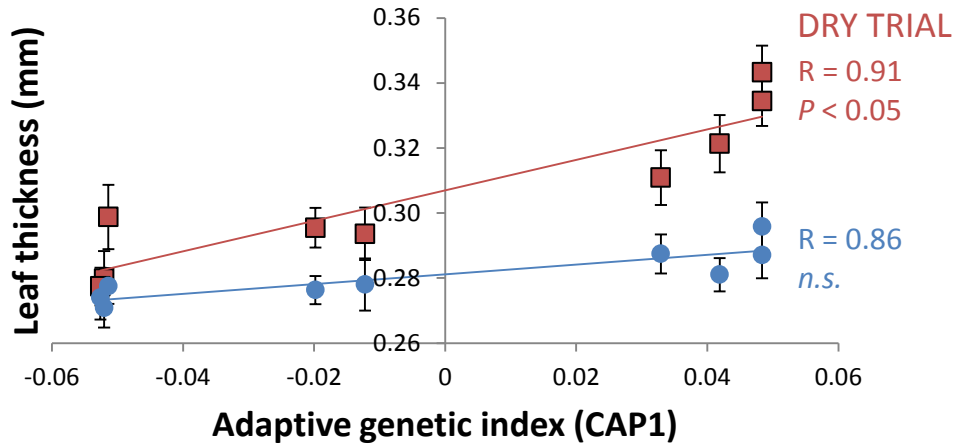
Rainfall of the driest quarter



23/35 climatic variables were significantly ($P < 0.05$) associated with change in CAP1.

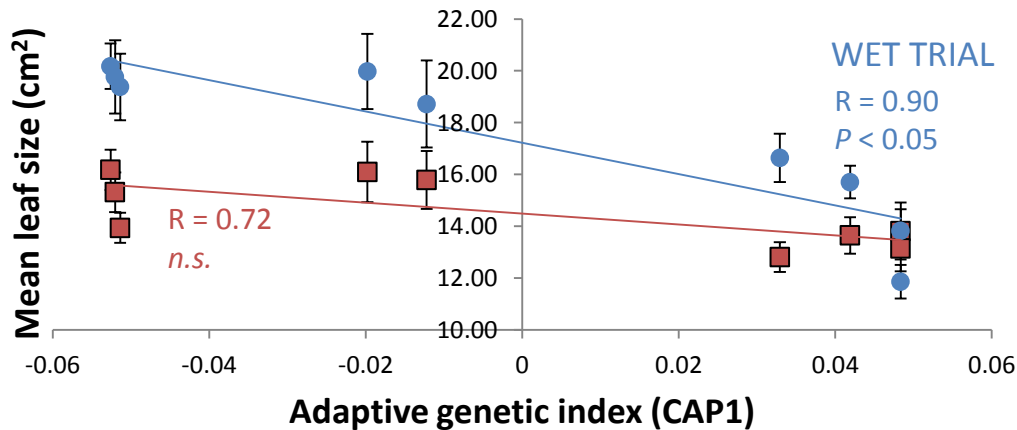
Strongest associations were with factors that contribute to summer aridity.

Leaf thickness



Higher CAP1 = genetically thicker leaves (dry site)

Leaf size



Higher CAP1 = genetically smaller leaves (wet site)

Population-level variation in outlier markers (CAP1) is correlated with quantitative genetic changes in functional traits.

**CAP1 describes molecular genetic change
associated with
adaptation of *E. tricarpa* populations
to increasing aridity.**

CAP1 forms the foundation of a management metric

$$\text{Aridity Index (AI)} = \sum a_i b_i$$

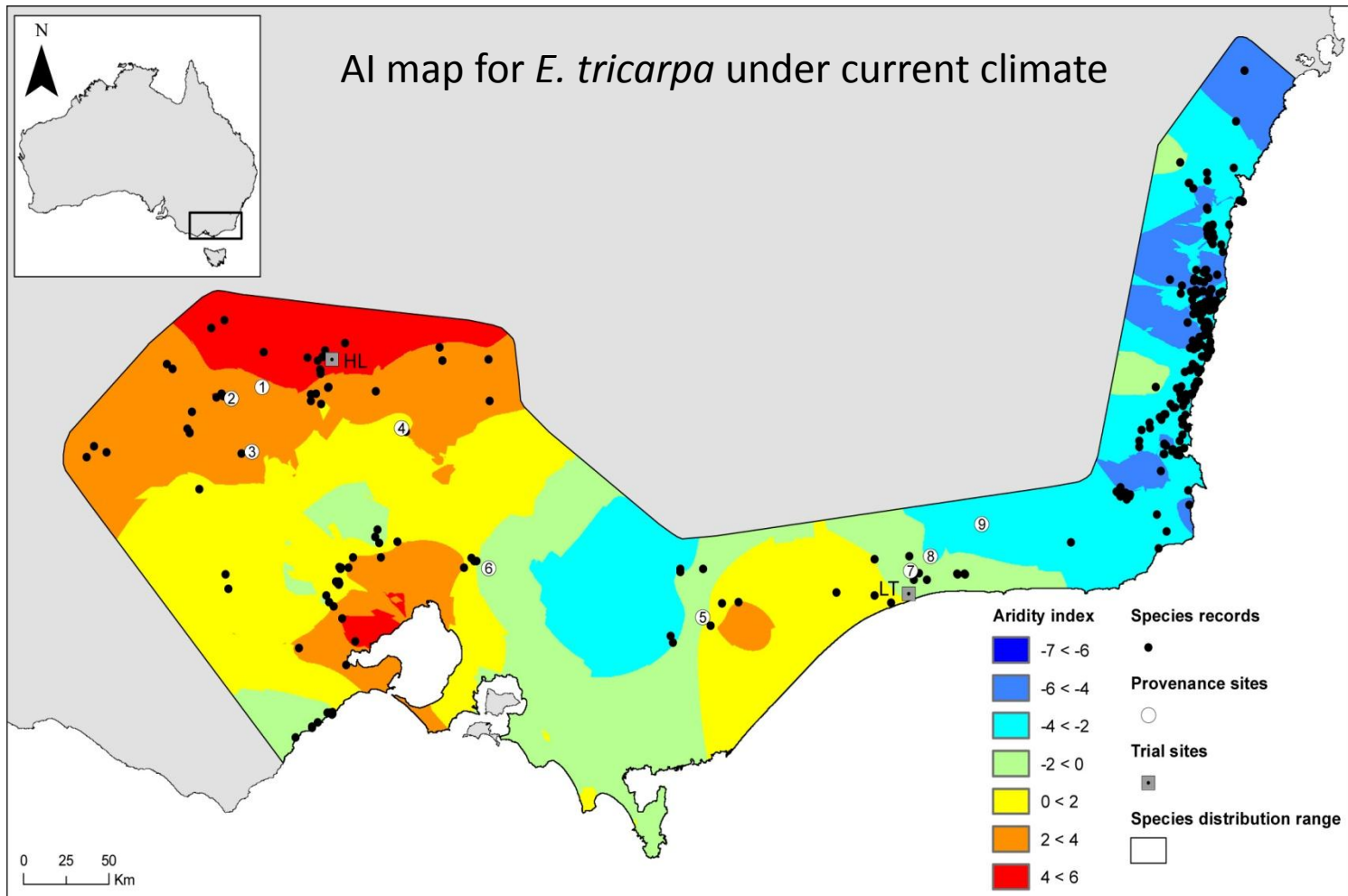
a = normalised climatic variable, x

b = the weighting of the climatic variable on the canonical eigenvector aligned with CAP1

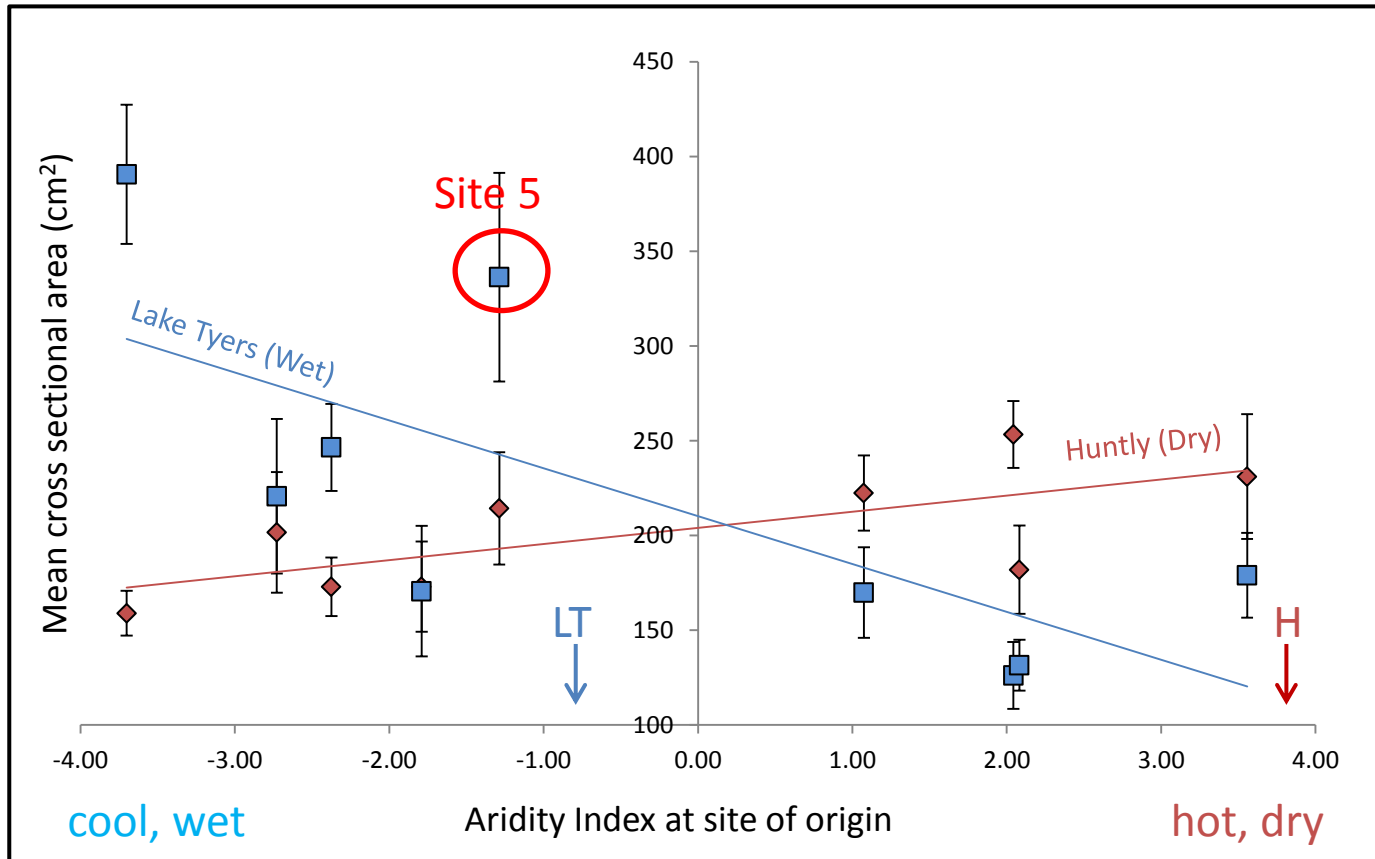
EXAMPLE

$$\begin{aligned} \text{AI}_{\text{TARNAGULLA}} &= (\text{TMXWM}_{\text{TARN}} \times \text{CAP1}_{\text{TMXWM}}) + (\text{RANN}_{\text{TARN}} \times \text{CAP1}_{\text{RANN}}) + (\text{TMNCM}_{\text{TARN}} \times \text{CAP1}_{\text{TMNCM}}) \\ &= (1.431 \times 0.421) + (-1.251 \times -0.343) + (0.068 \times 0.134) \\ &= 2.290 \end{aligned}$$

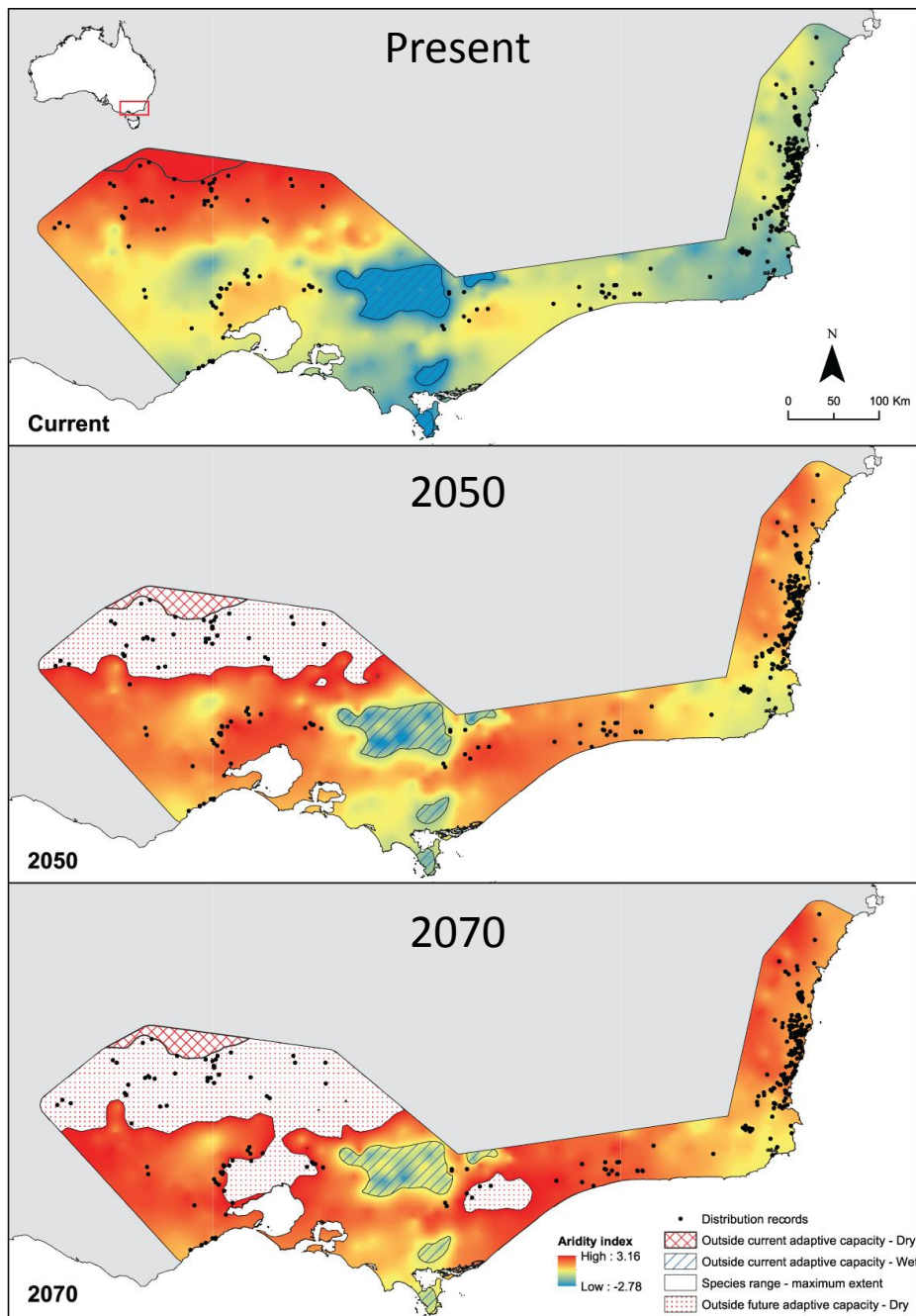
Seed transfer guidelines?



But does it work?



Seed from drier areas (higher AI) grew better at drier trial
Seed from wetter areas (lower AI) grew better in wetter trial



Guidelines for assisted migration?

CSIRO global climate model for 2050 and 2070

The Team



Dr Margaret Byrne

Plant Conservation Geneticist

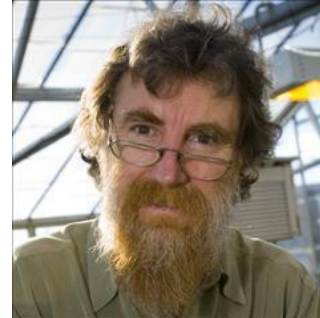
Dept. Conservation and Environment, WA



A/Prof René Vaillancourt

Plant Geneticist

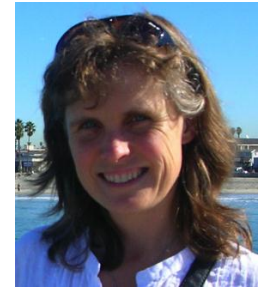
University of Tasmania



Prof Brad Potts

Forest Geneticist

University of Tasmania



Dr Dorothy Steane

Plant Geneticist

University of Tasmania and
University of the Sunshine Coast



Dr Elizabeth McLean

Plant Physiologist

DEC WA/CSIRO



Prof William Stock

Plant Physiologist

Edith Cowan University, WA



Dr Suzanne Prober

Plant Ecologist

CSIRO Ecosystem Science, WA



Peter Harrison

GIS expert

University of Tasmania