

# Re-analyses of historical series of variety trials: lessons from the past and opportunities for the future

Genetic change over time

Breakdown in disease resistance

GEI and climate change

Genomic selection

Ian Mackay

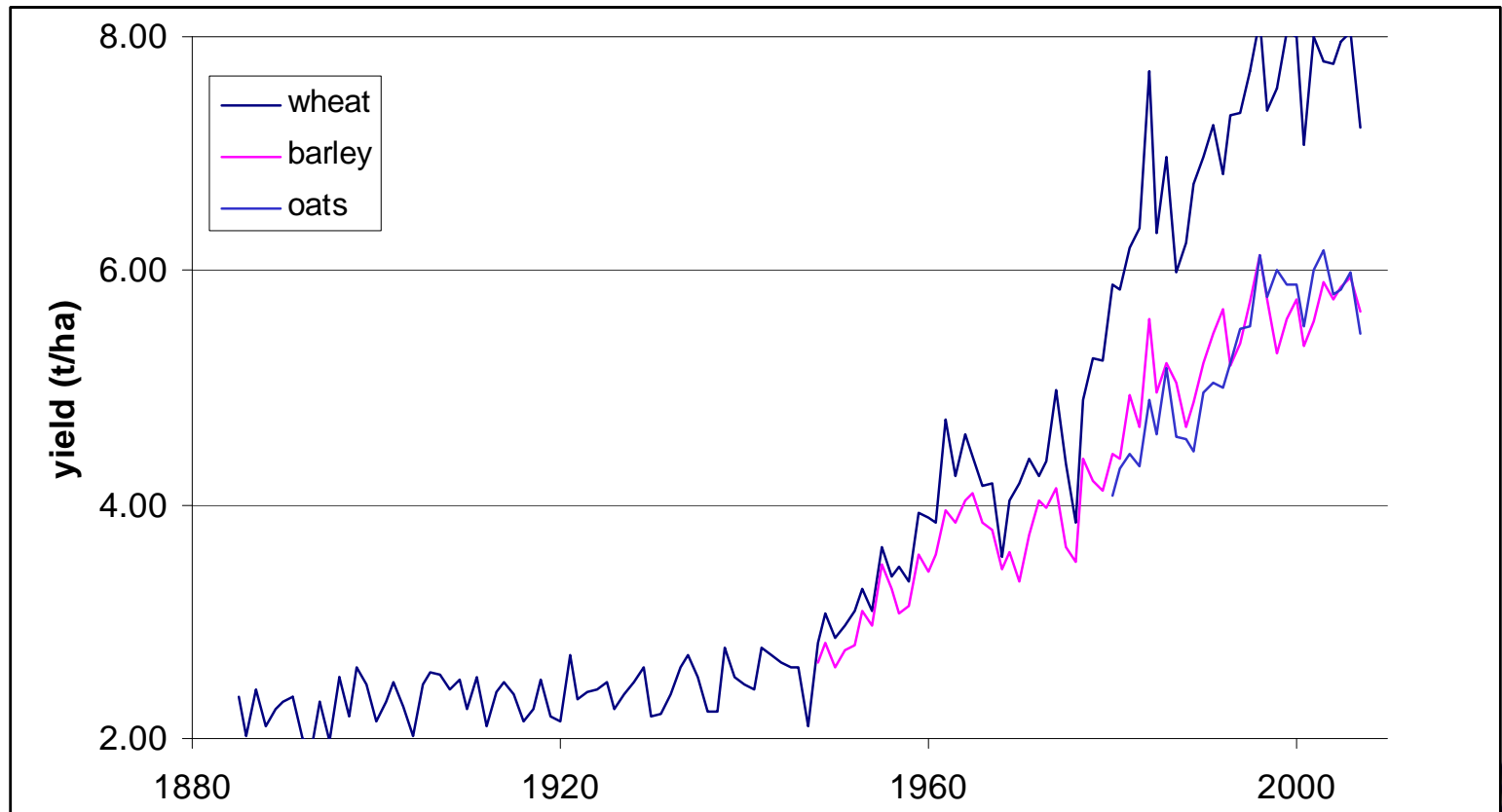
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# DEFRA statistics on national yields

Yields have increased ~3 fold in WW since 1948 as a result of improved varieties and agronomy.



# Re-analyses from 1948-2007

What proportion of yield increase is genetic?

Winter wheat

*Spring barley*

*Winter barley*

*Oil seed rape*

*Forage Maize*

*Sugar beet*



# Data are extensive

|                                 |        |
|---------------------------------|--------|
| Total data points               | 52 909 |
| No. varieties                   | 308    |
| No. years                       | 60     |
| minimum years per variety       | 3      |
| Average life time               | 5.8    |
| Trials                          | 3 590  |
| Trials x varieties missing data | 95%    |

# Statistical analysis kept simple:

Fixed vars + years

Random vars.years + years.trials + vars.trials.years

No data for replicates within trials

No weighting of trials

No chronology (years fitted as a factor not as a covariate)

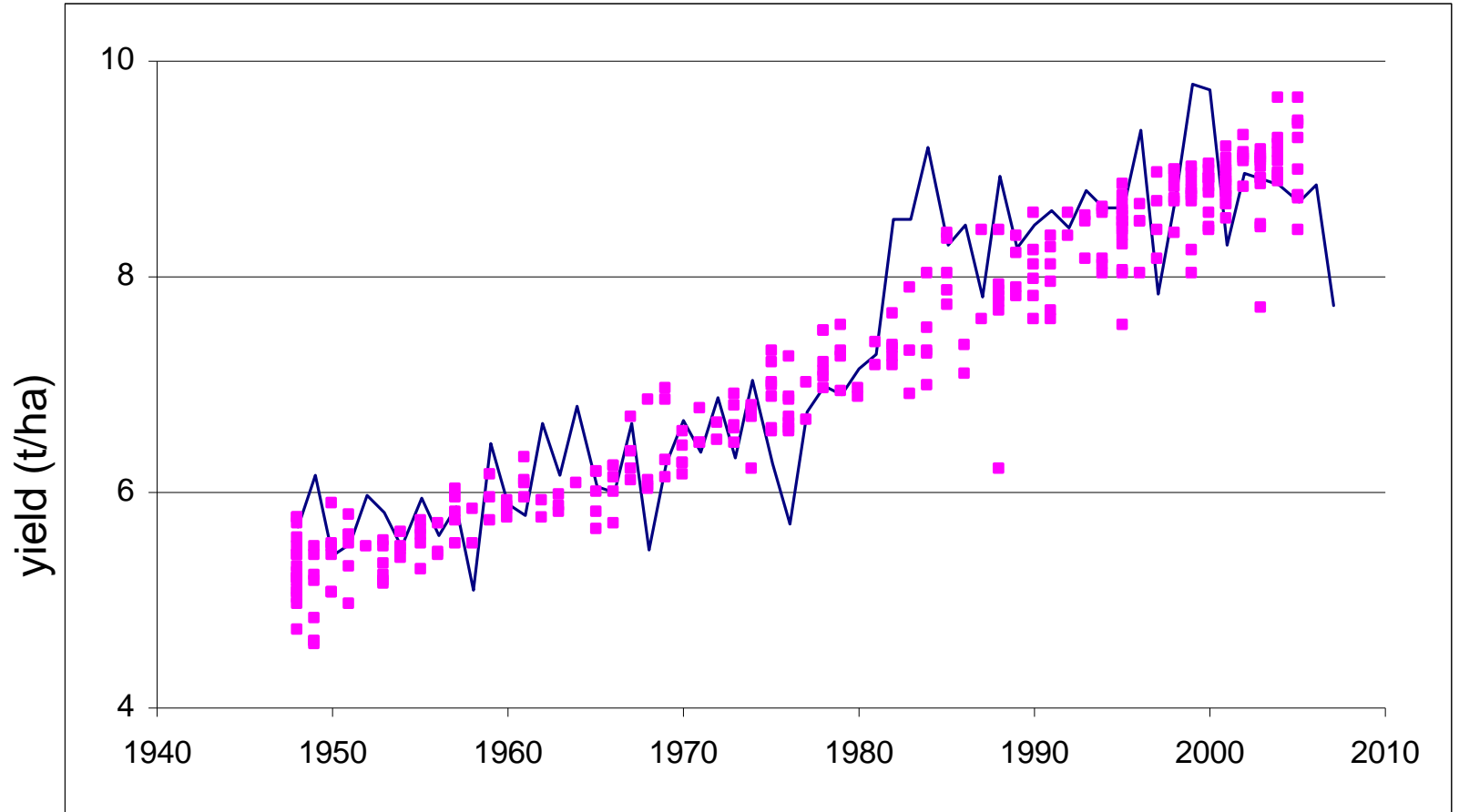
No linking of trials over years or analysis by region

Cannot simply treat varieties or years as random:

trends

relationships

# winter wheat



first year in trial



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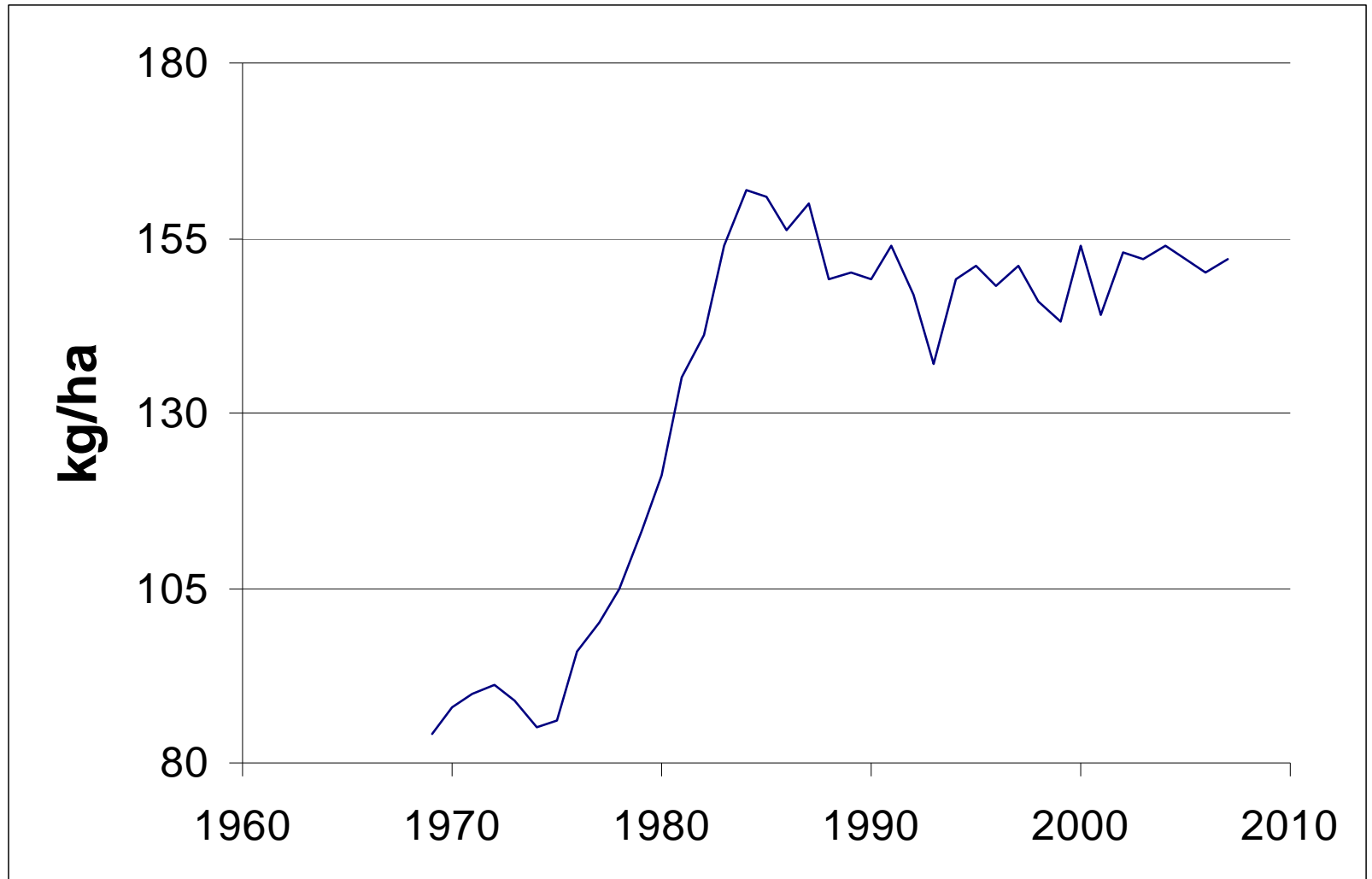
# Linear trends in yield

|             | G     | E     |
|-------------|-------|-------|
| 1948 - 2007 | 0.071 | 0.069 |
| 1948 - 1981 | 0.061 | 0.041 |
| 1982 - 2007 | 0.074 | 0.010 |

Improvements in both agronomy and genetics brought about big improvements in yield for ~ 30 years from 1948.

However, for the last 25 years, yields have increased primarily from plant breeding alone.

# N use for tillage crops: England & Wales



# treated and untreated trials

Since 1982 trials have been conducted as two series:

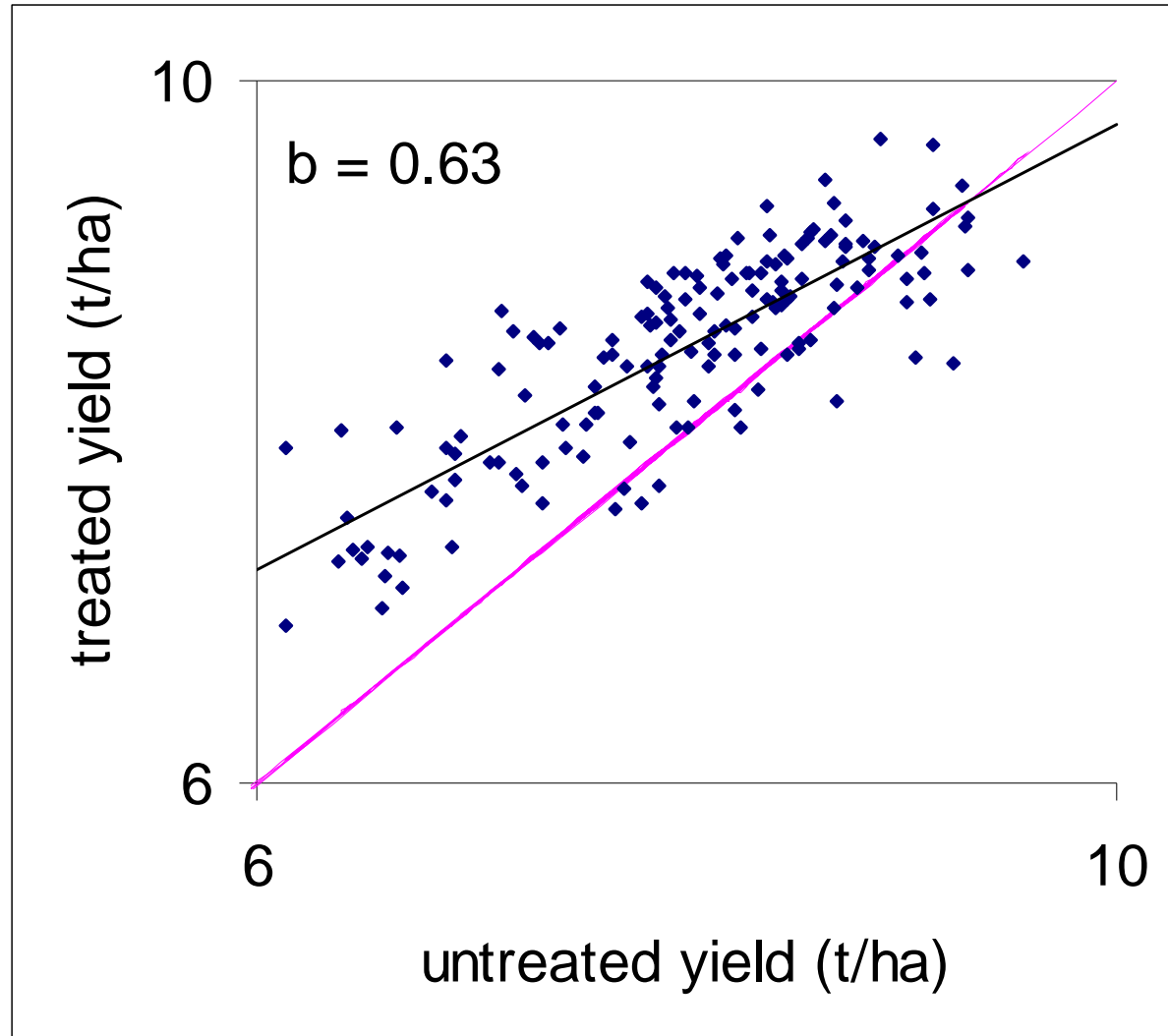
Treated: fungicide treated.

Untreated: no fungicide.

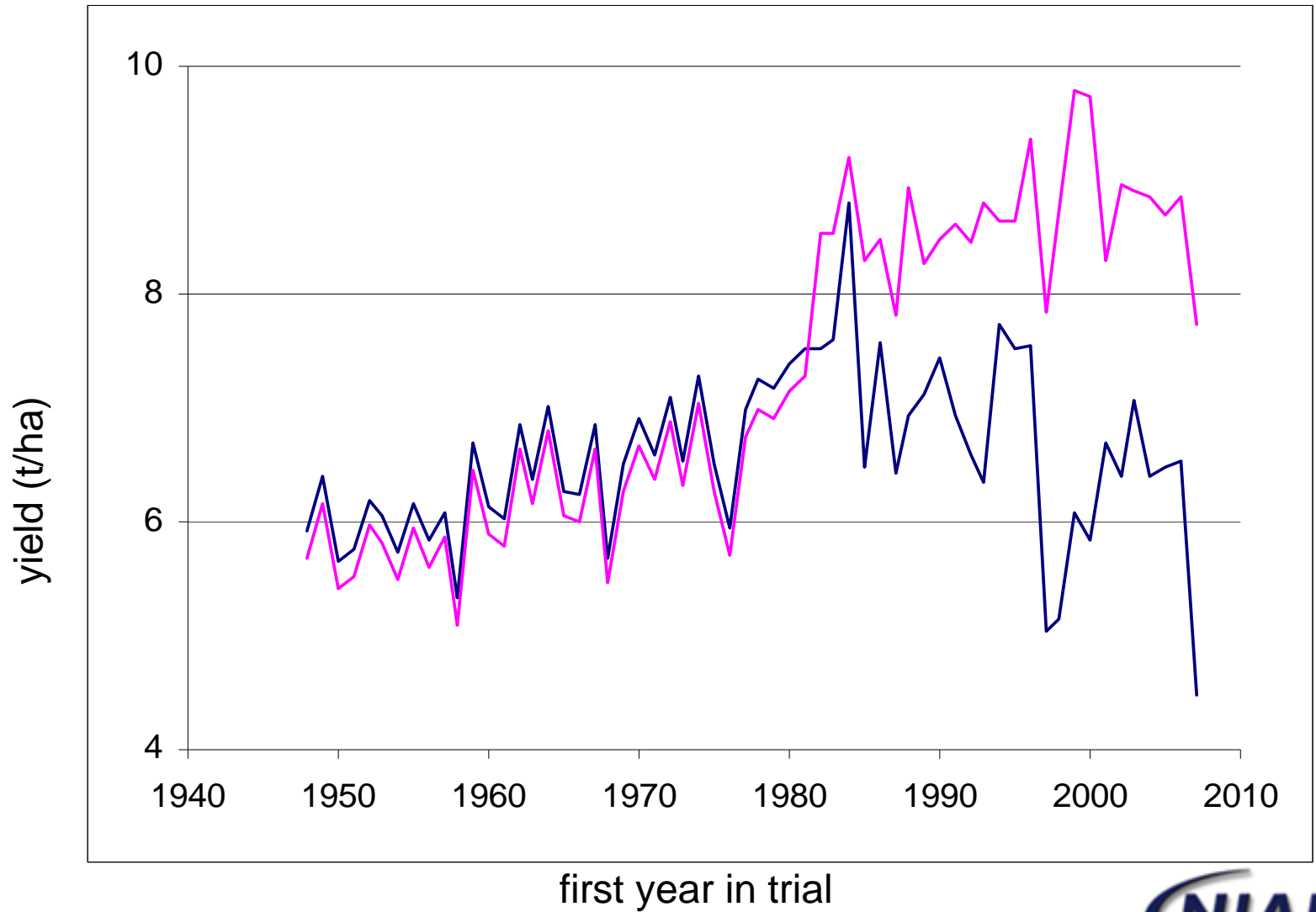
Series analysed separately and results compared

# Genetic effects correlate between the two series:

(analysis 1982-2007 only)



# but year effects diverge

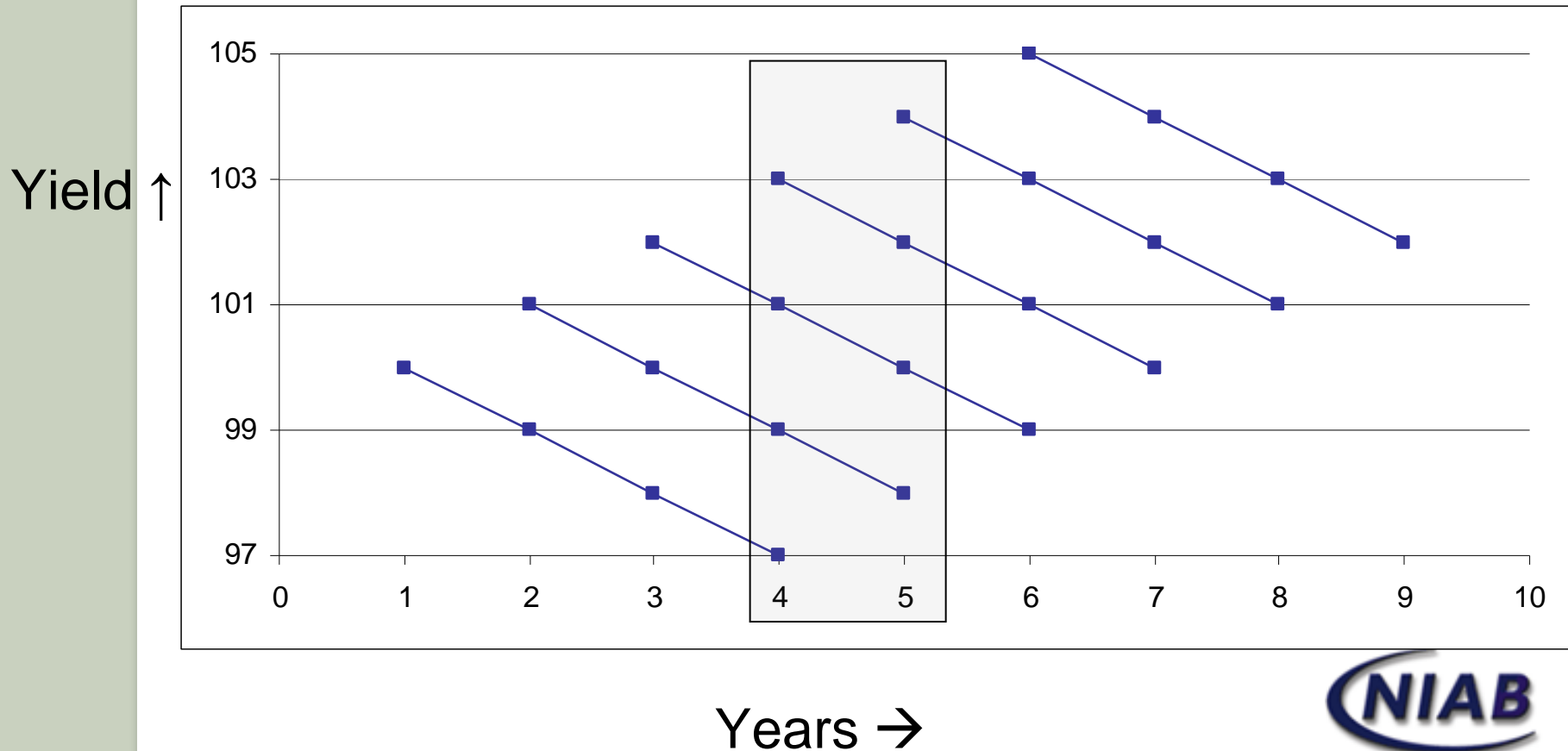


Can the apparent decline in the environment in untreated trials be explained by breakdown of disease resistance?

Simple model:

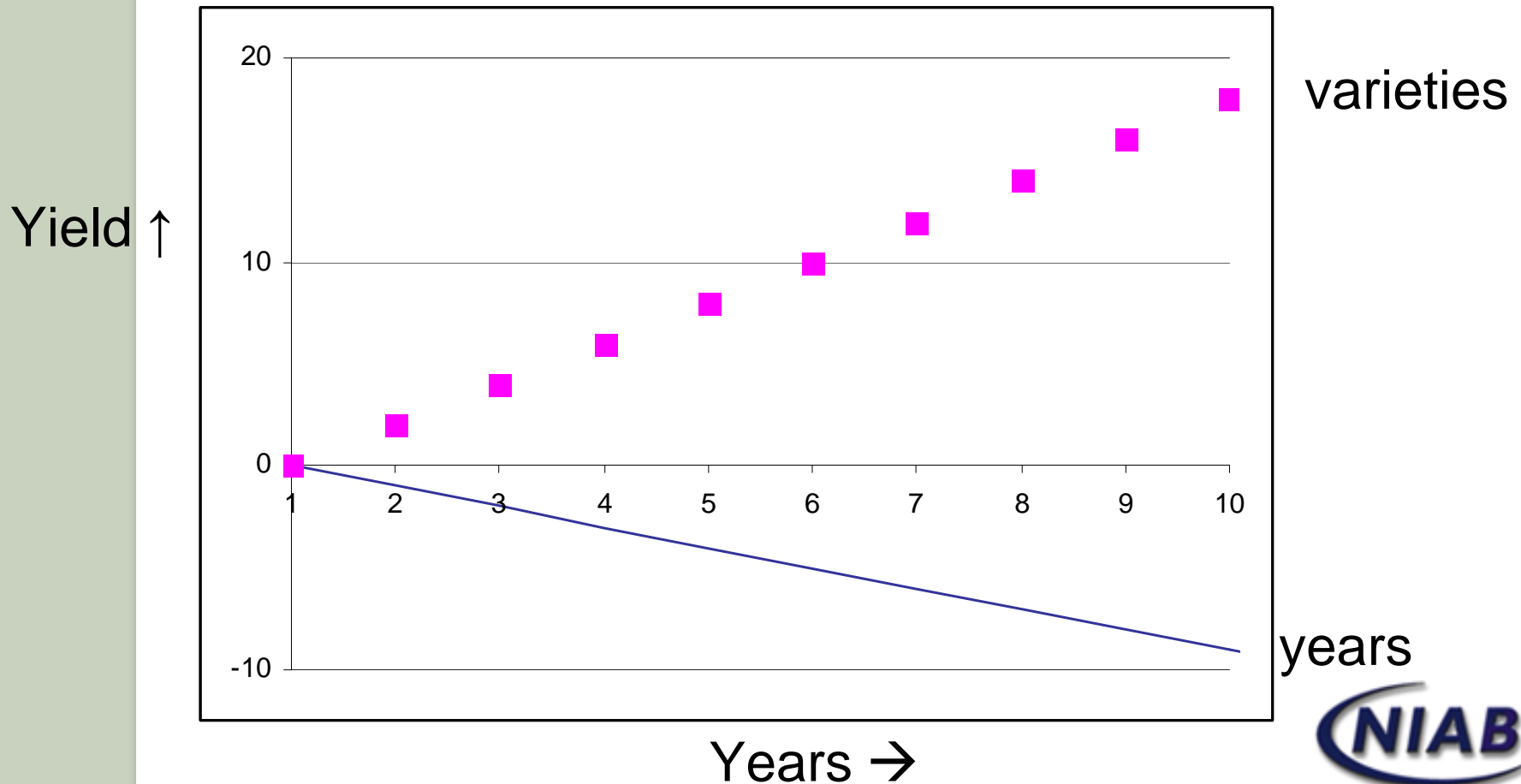
no years effect

yields increase while resistance breaks down.



# Estimated effects

Variety effect is overestimated but compensated by a progressive decline in the year effect.



# A matched set of T & U winter wheat trials.

748 paired trial locations (T and U)

Analyse differences in variety means at each paired site.

Common environment effects (inc. year) should cancel.

Estimate the effect of “trial year” on the difference.

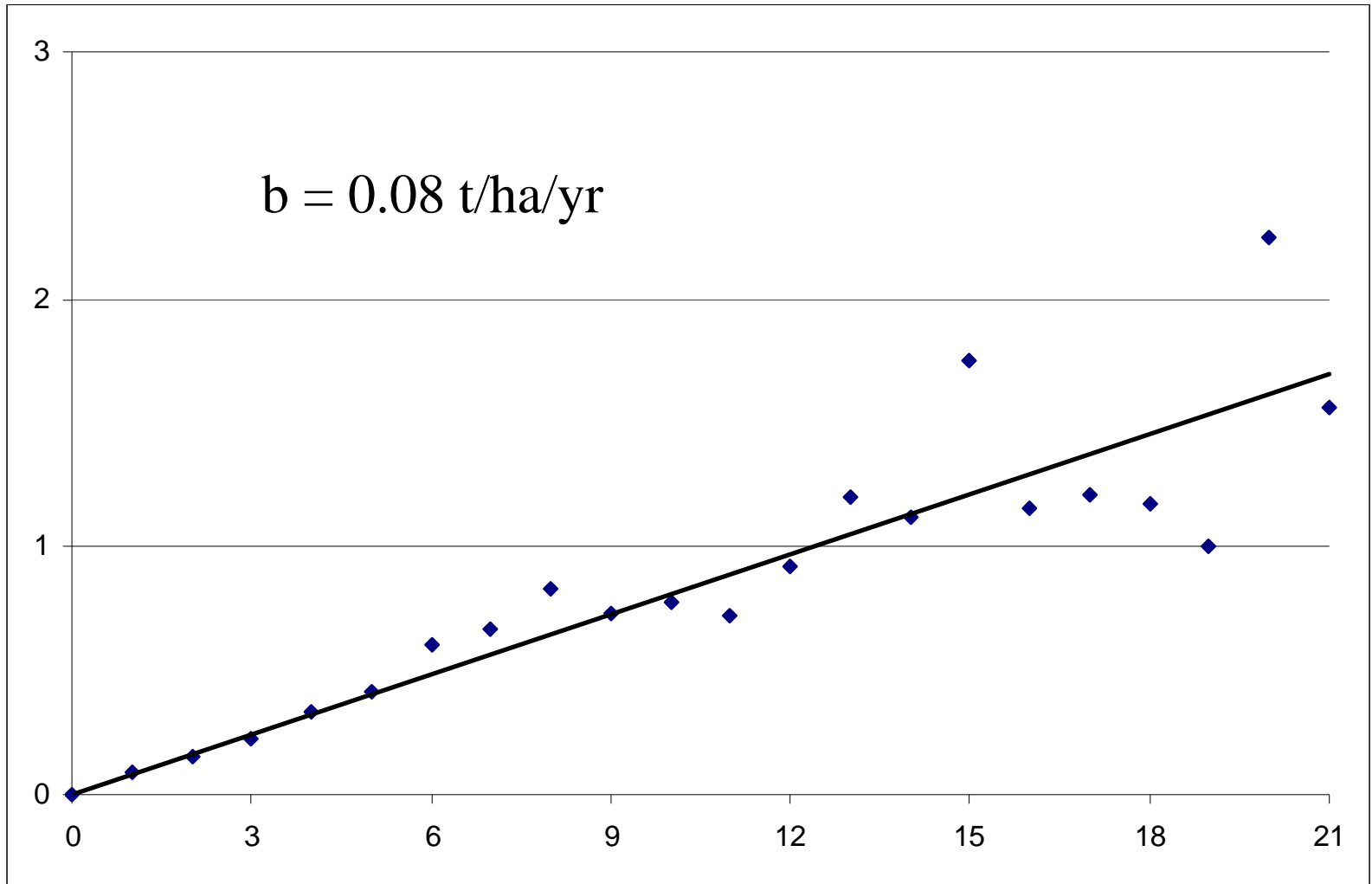
Is there a progressive increase?

Fixed                    varieties + trial year

Random                varieties.trial year + sites + error

# Estimated effect of trial year

T-U ↑  
effect



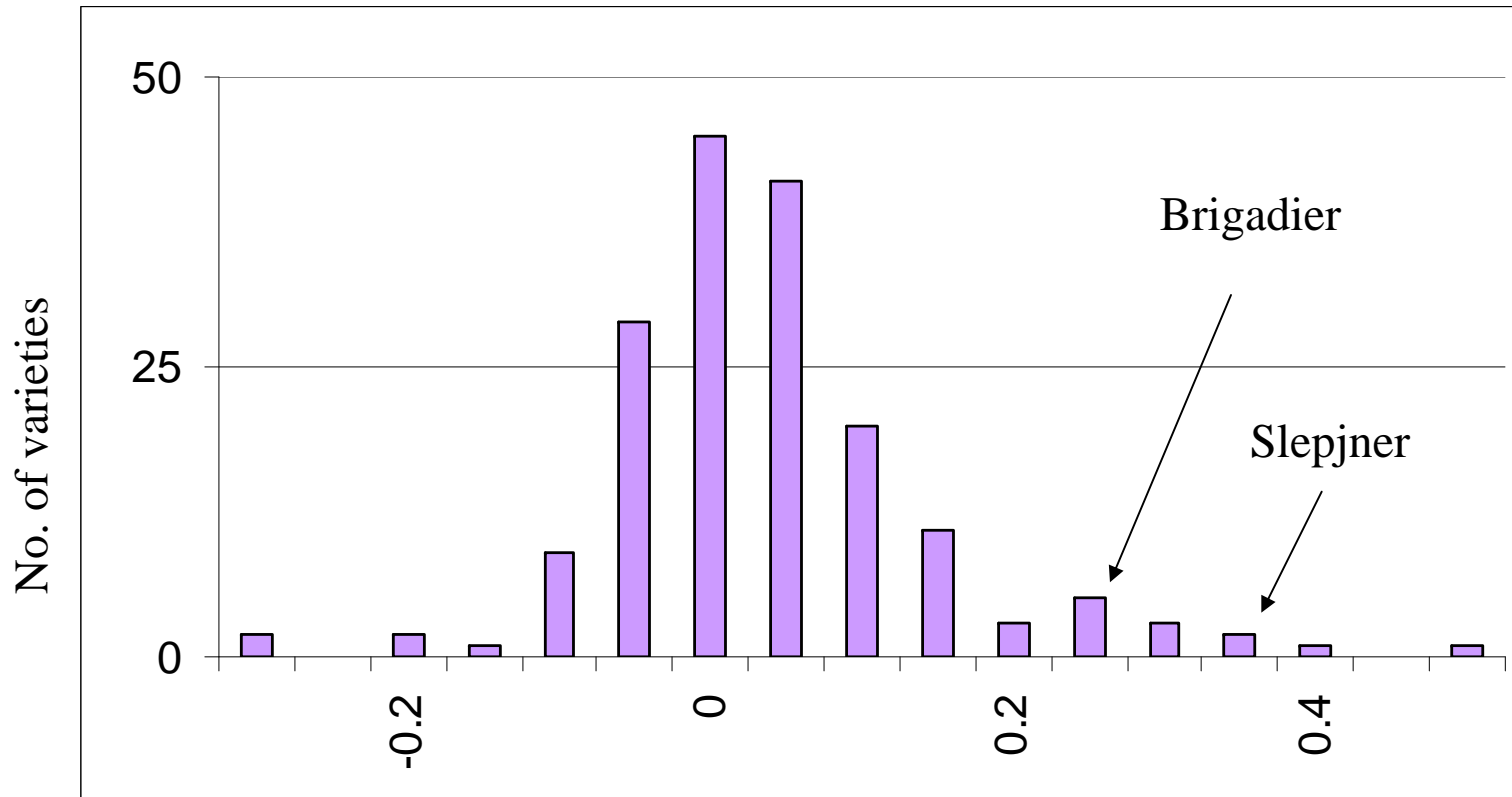
Years in trial →



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# vars.trial year regressed on trial year

distribution of regression coefficients



Regression coefficient of T-U on years in trial

# GEI and climate

WW 1982- 2007 treated trials

8 climatic variables from the UK met office:

Central England av temp                      Aut, Wint, Spr, Sum

England and Wales rainfall                      Aut, Wint, Spr, Sum

Correlate var x years effects with climate

# Number of varieties with q-value <0.5

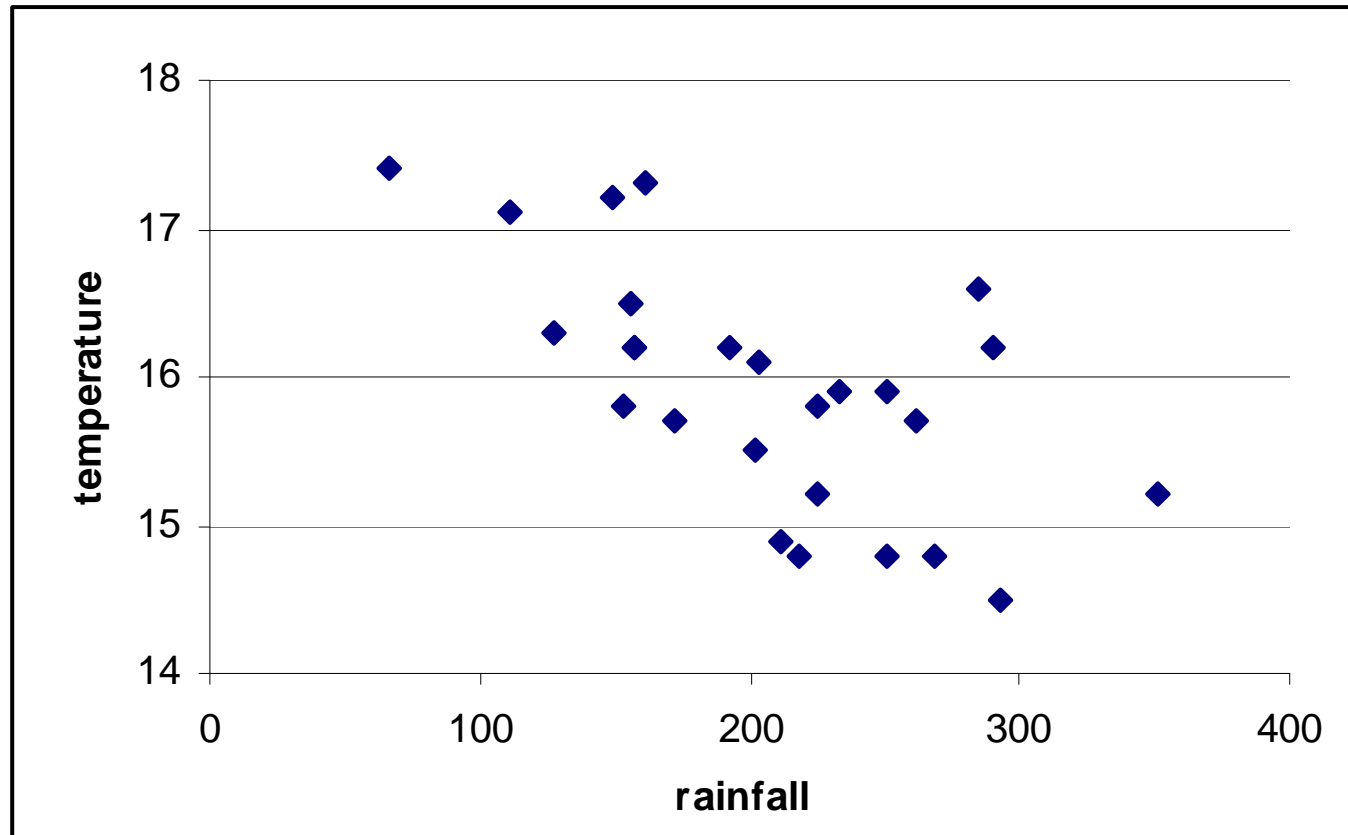
|      | Autumn | Winter | Spring | Summer |
|------|--------|--------|--------|--------|
| Rain | 0      | 0      | 1      | 26     |
| Temp | 0      | 27     | 8      | 26     |

Greatest sensitivity is:

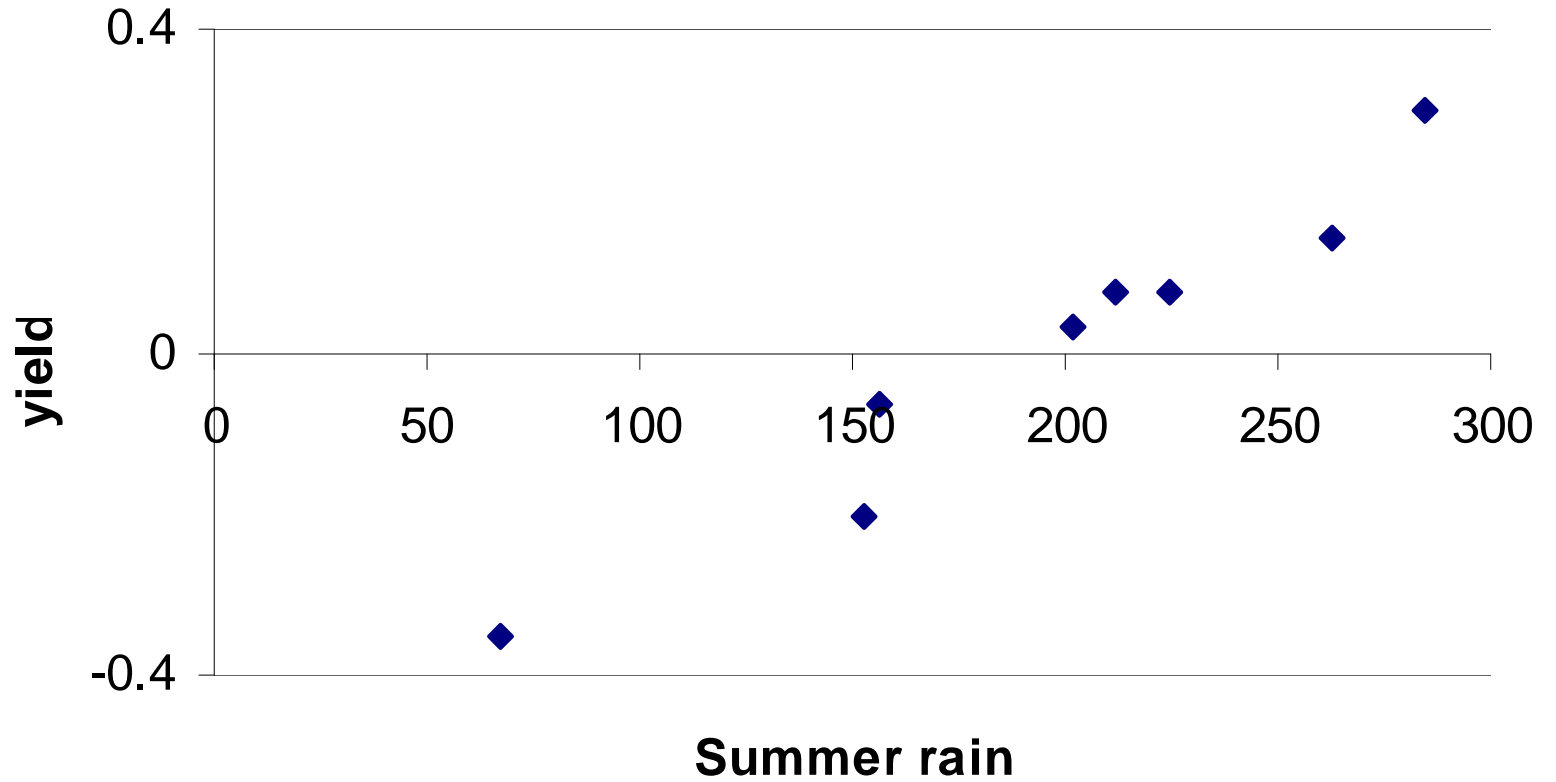
cold & wet summers

wet winters

# Wet summers are cold



# Cadenza



# Genomic selection in Winter wheat

159 varieties: 1949 - 2004

217 mapped DArT markers

# Method

Split data into a training set and a test set:

- a) uniformly w.r.t. age
- b) on age

Regress yield on all markers in the training set.

More markers than varieties: use ridge regression (easy)

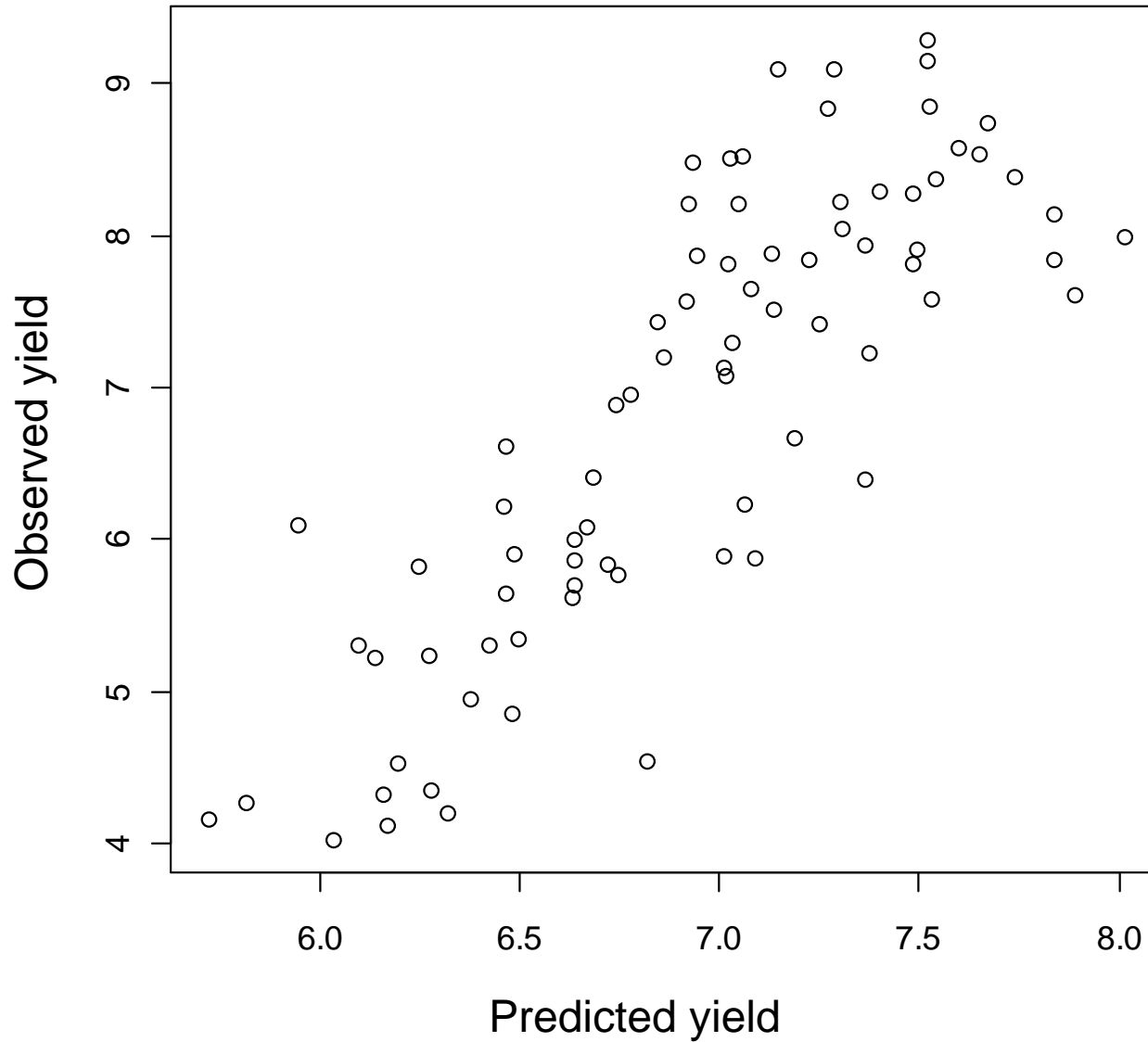
Predict yield in the test set.

Compare observed and predicted

# WW uniform split

$r = 0.83$

## Winter Wheat

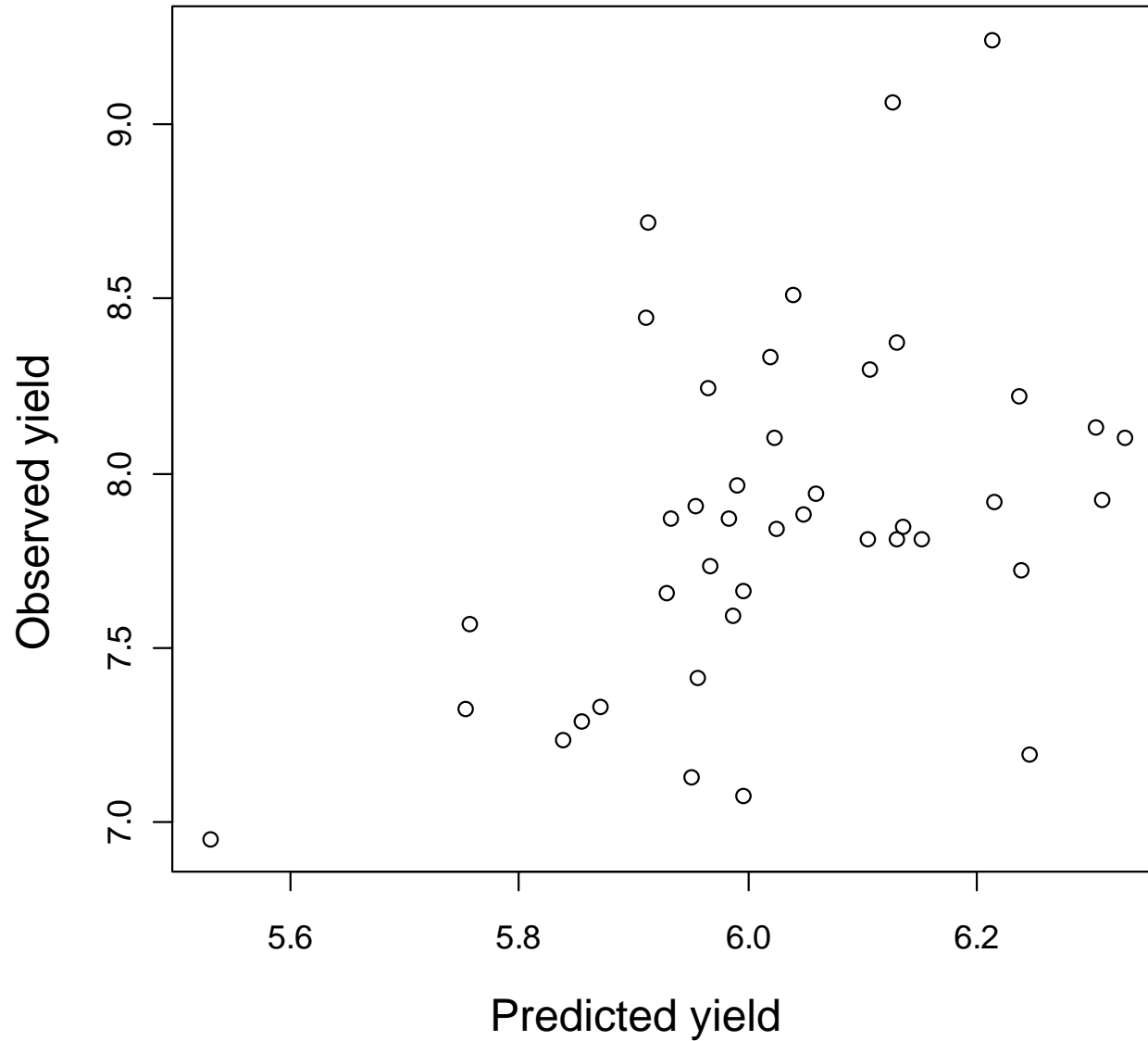




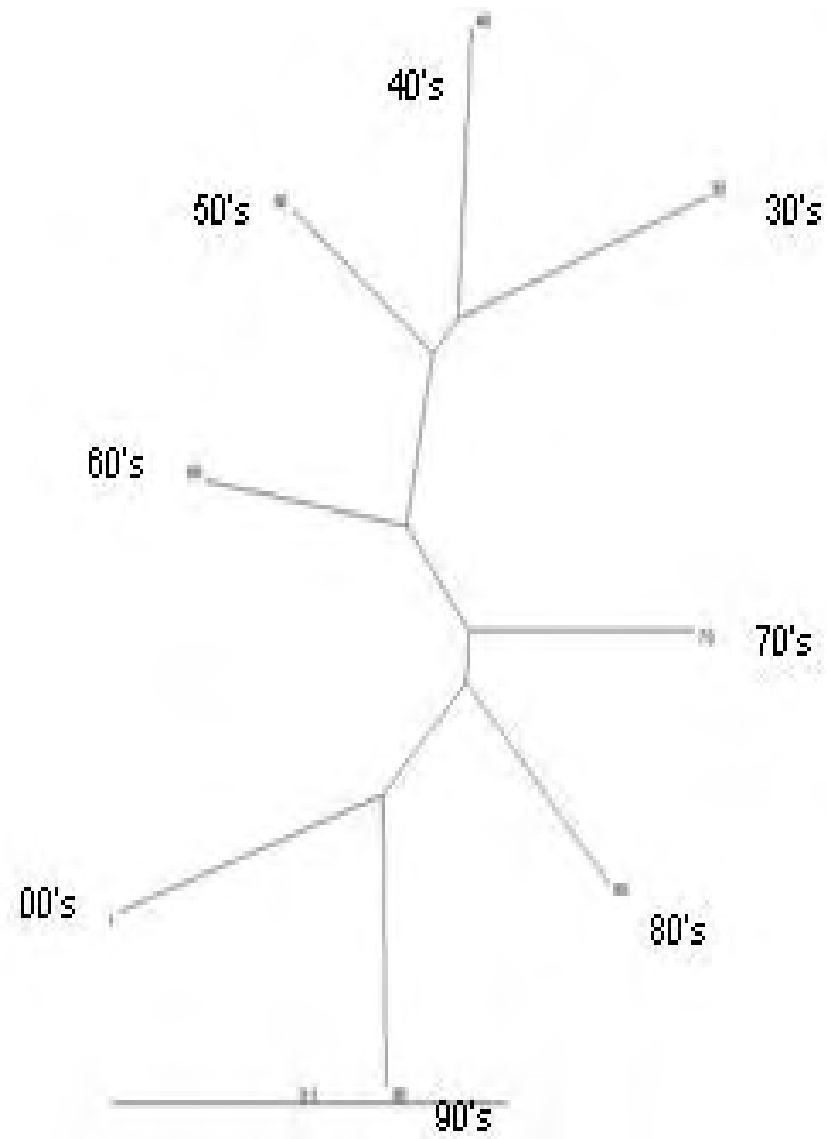
WW split on age –  
10 years only

$r = 0.44$

### Winter Wheat



# Changes in marker genotypes over time (UK)



# Results

Can predict yield

but correlation not great when predicting forward in time

predicting on kinship

Breeders already predict on kinship

cross the best with the best

Historical datasets are suitable to test methods

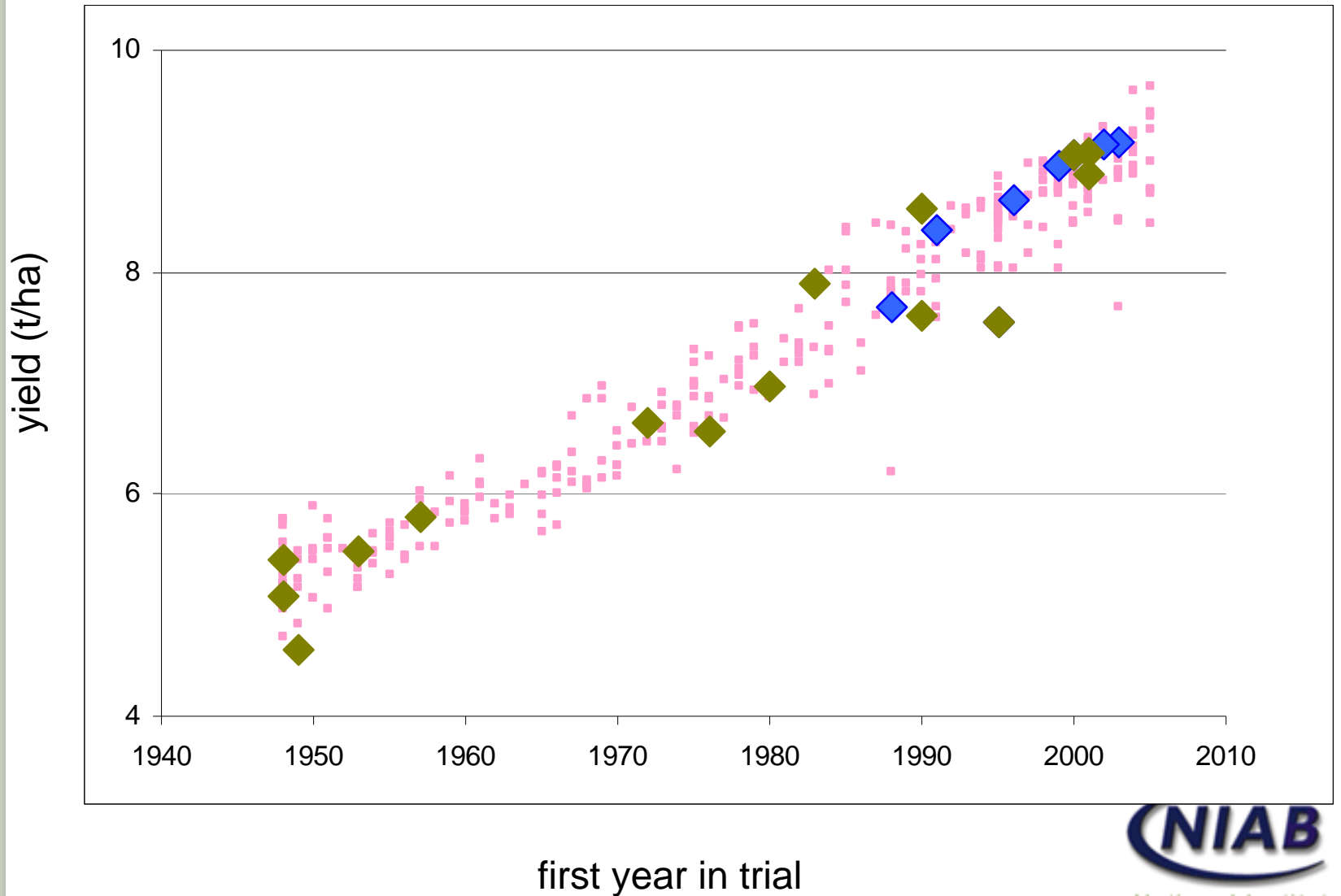
but mustn't get carried away by positive results.

Require experimental populations too

MAGIC



# Founders of WW MAGIC populations



# Conclusion

There is more to historical datasets than just a source of free phenotypes for association mapping.

“Retrospective analysis of genetic gain and variance can be useful in designing strategies to manage genetic variation for target traits in breeding programs.” Condóna et al *Crop Sci* 2009

Strength will come from exploiting links with experimental populations.

# Acknowledgements



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