



# Multiple QTL mapping in autopolyploids with QTLpoly

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INTERNATIONAL  
POTATO CENTER



CGIAR

# Outline

- 1 Multiple QTL mapping
  - Components of variance
  - Random-effect QTL model
- 2 Applications
  - Potato: 'Atlantic' × B1829-5
  - Sweetpotato: 'Beauregard' × 'Tanzania'
- 3 Final considerations

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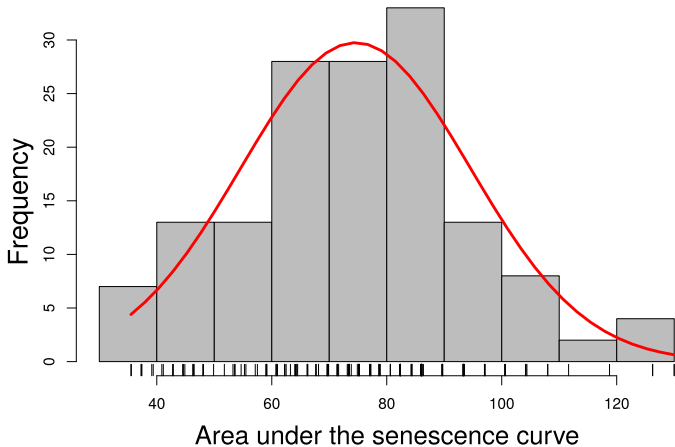
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Potato: 'Atlantic' × B1829-5 ( $N = 154$ )

## Foliage maturity

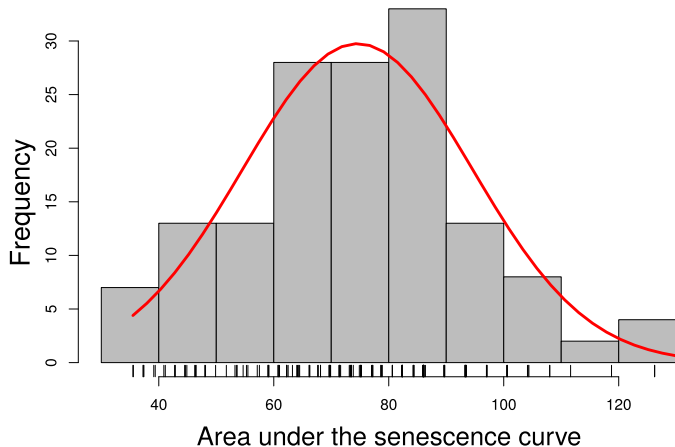


$$y_i \sim (\mu, \sigma_p^2)$$

Phenotypic data is usually described in terms of a mean and a variance

Potato: 'Atlantic' × B1829-5 ( $N = 154$ )

## Foliage maturity



$$y_i \sim (74.57, 399.24)$$

Phenotypic data is usually described in terms of a mean and a variance

# What makes potatoes different from each other?

1<sup>st</sup> case: all individuals are the same (e.g., clones of 'Atlantic')

$$\sigma_p^2 = \sigma_e^2$$

- $\sigma_e^2 = 399.24$ : environmental component

2<sup>st</sup> case: individuals are an F<sub>1</sub> progeny ('Atlantic' × B1829-5)

$$\sigma_p^2 = \sigma_g^2 + \sigma_e^2$$

- $\sigma_g^2 = 183.65$ : genetic component (polygenic effects)
- $\sigma_e^2 = 215.59$ : environmental component

Phenotypic variance ( $\sigma_p^2$ ) can be attributed to different sources

## Genetic variance ( $\sigma_g^2$ )

### Broad-sense heritability ( $H^2$ )

$$H^2 = \frac{\sigma_g^2}{\sigma_p^2}$$

### Additive genetic variance ( $\sigma_a^2$ )

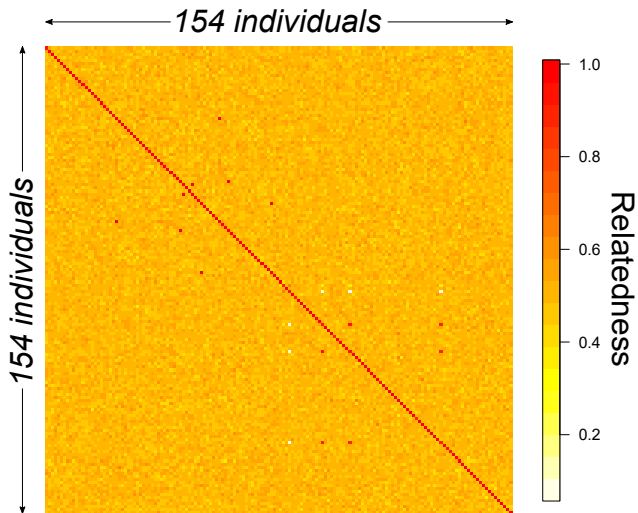
$$\sigma_p^2 = \mathbf{A}^{(i,i')} \sigma_a^2 + \sigma_e^2$$

- $\mathbf{A}^{(i,i')}$  is the genome-wide relationship describing the average covariance between a sib-pair ( $i, i'$ ) if the covariance is caused by heritable effects
- Narrow-sense heritability:  $h^2 = \frac{\sigma_a^2}{\sigma_p^2}$

Genetic variance ( $\sigma_g^2$ ) can be further decomposed into additive ( $\sigma_a^2$ ) and other variances (e.g., dominance, epistasis)



## Realized relationship matrix $\mathbf{G}$ for 4,285 markers



Expected value for full-sibs:  $E(\mathbf{G}^{(i,i')}) = \mathbf{A}^{(i,i')} = \frac{1}{2}$  (the kinship matrix)

# Are there any specific loci associated with the phenotype?

## Quantitative trait loci (QTL)

Region of DNA which contributes to the variation of a particular trait

## Variance due to a single QTL ( $\sigma_q^2$ )

$$\sigma_p^2 = \mathbf{G}_q^{(i,i')} \sigma_q^2 + \sigma_e^2$$

- $\mathbf{G}_q^{(i,i')}$  is the relationship describing the covariance between a sib-pair  $(i, i')$  if the covariance is caused by a QTL  $q$

## In the region of a gene influencing a trait of interest

- Correlated QTL effects among individuals are due to shared alleles identical-by-descent (IBD)
- Higher IBD allele sharing for a QTL is expected among individuals with similar phenotypic values

## Diploid: outcrossing species

(2 gametes each parent)

$ab \times cd$

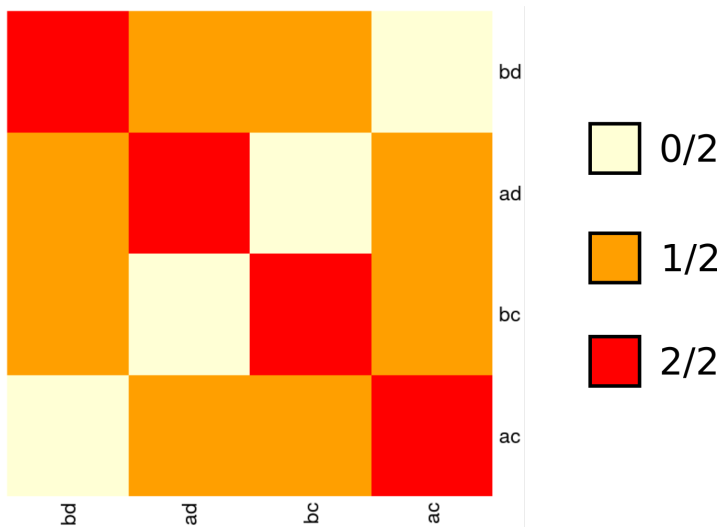
↓

	c	d
a	ac	ad
b	bc	bd

(1 ac : 1 ad : 1 bc : 1 bd)

(4 genotypes)

# Diploid: identity-by-descent



## Autotetraploid: outcrossing species

(6 gametes each parent)

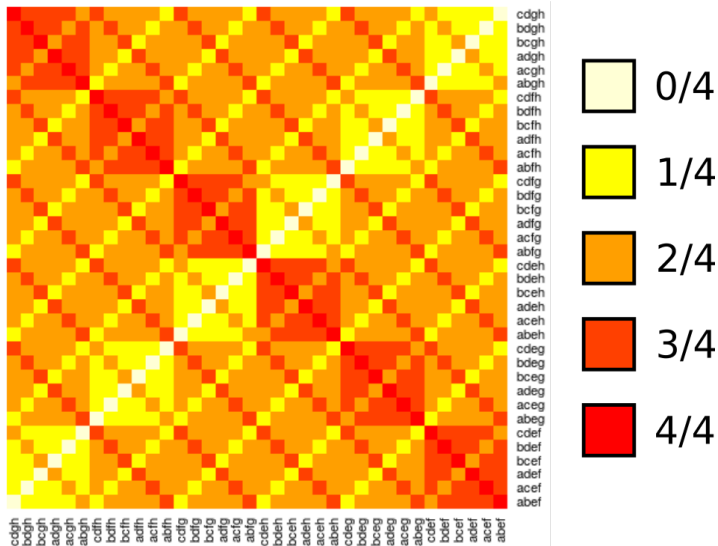
abcd × efgh

↓

	ef	eg	eh	fg	fh	gh
ab	abef	abeg	abeh	abfg	abfh	abgh
ac	acef	aceg	aceh	acfg	acfh	acgh
ad	adef	adeg	adeh	adfg	adfh	adgh
bc	bcef	bceg	bceh	bcfg	bcfh	bcgh
bd	bdef	bdeg	bdeh	bdfg	bdfh	bdgh
cd	cdef	cdeg	cdeh	cdfg	cdfh	cdgh

(36 genotypes)

# Autotetraploid: identity-by-descent



# Autohexaploid: outcrossing species

(20 gametes each parent)

abcdef × ghijkl

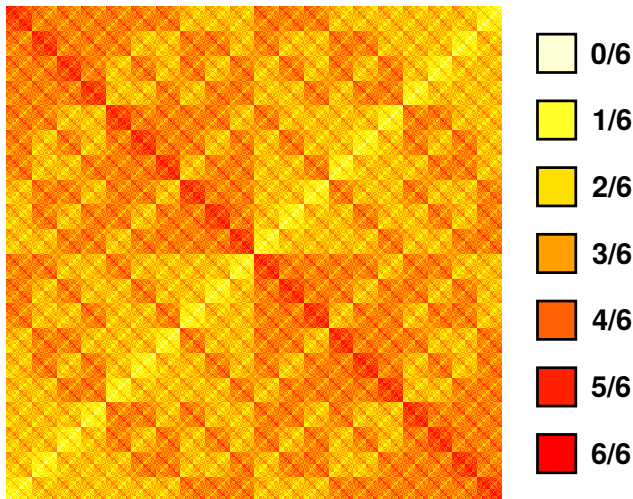


	ghl	ghj	ghk	...	ijl	ikl	jkl
abc	abcghl	abcghj	abcghk	...	abcijl	abcikl	abcjkl
abd	abdghl	abdghj	abdghk	...	abdiyl	abdikl	abdjkl
abe	abeghl	abeghj	abeghk	...	abeiyl	abeikl	abejkl
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
cdf	cdfghl	cdfghj	cdfghk	...	cdfijl	cdfikl	cdfjkl
cef	cefg hl	cefg hj	cefg hk	...	cefijl	cefikl	cefjkl
def	defghl	defghj	defghk	...	defijl	defikl	defjkl

(400 genotypes)

# Autohexaploid: identity-by-descent

abcdef × ghijkl





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# Random-effect multiple QTL model

## From single to multiple QTL analysis

- Random-effect multiple interval mapping (REMIM)

$$\sigma_p^2 = \sum_{q=1}^Q \mathbf{G}_q^{(i,i')} \sigma_q^2 + \sigma_e^2$$

- $\mathbf{G}_q^{(i,i')}$  is the relationship describing the covariance between a sib-pair  $(i, i')$  if the covariance is caused by a QTL  $q$
- There are  $Q$  QTL contributing to the variance of a particular trait

## QTL heritability ( $h_q^2$ )

$$h_q^2 = \frac{\sigma_q^2}{\sigma_p^2}$$

## Random-effect multiple QTL model

$$\mathbf{y} = \mathbf{1}\mu + \sum_{q=1}^Q \mathbf{g}_q + \mathbf{e}$$

- $\mathbf{y}$  is the  $n \times 1$  vector of phenotypic values
- $\mu$  is the fixed effect of the population mean
- $\mathbf{g}_q$  is the  $n \times 1$  random vector of the genotype effects of QTL  $q$  ( $q = 1, \dots, Q$ ) with  $\mathbf{g}_q \sim \mathcal{N}(\mathbf{0}, \mathbf{G}_q \sigma_q^2)$
- $\mathbf{e}$  is the  $n \times 1$  random vector of residual error with  $\mathbf{e} \sim \mathcal{N}(\mathbf{0}, \mathbf{I} \sigma_e^2)$ 
  - ▶  $\mathbf{1}$  and  $\mathbf{I}$  are an  $n \times 1$  vector of 1's and an  $n \times n$  identity matrix, respectively
  - ▶  $\mathbf{G}_q$  is the  $n \times n$  IBD-based relationship matrix at QTL  $q$

## Multiple QTL Mapping in Autopolyploids: A Random-Effect Model Approach with Application in a Hexaploid Sweetpotato Full-Sib Population

Guilherme da Silva Pereira,<sup>\*,†,1,2</sup> Dorcus C. Gemenet,<sup>\*,†</sup> Marcelo Mollinari,<sup>\*,†</sup> Bode A. Olukolu,<sup>§</sup> Joshua C. Wood,<sup>\*\*</sup> Federico Diaz,<sup>††</sup> Veronica Mosquera,<sup>††</sup> Wolfgang J. Gruneberg,<sup>††</sup> Awais Khan,<sup>\*\*</sup> C. Robin Buell,<sup>\*\*</sup> G. Craig Yencho,<sup>†</sup> and Zhao-Bang Zeng<sup>\*,†</sup>

### Genome scanning

- Tests are performed every 1 cM along the whole genome
- Multiple-QTL model: QTLs are added consecutively to the model

Hypothesis testing:  $H_0 : \sigma_q^2 = 0$  ;  $H_1 : \sigma_q^2 > 0$

- Restricted maximum likelihood (REML): estimation
- Score statistics: significance of the variance component
- Resampling method: genome-wide significance level ( $\alpha$ )

## Model selection algorithm

- 1 A model starts with no QTL

$$\sigma_p^2 = \sigma_e^2$$

- 2 *Forward search* adds one QTL  $q$  at a time if significant ( $\alpha = 0.20$ )

$$\sigma_p^2 = \sum_{q=1}^Q \mathbf{G}_q^{(i,i')} \sigma_q^2 + \sigma_e^2$$

- 3 *Model optimization* re-evaluate every QTL  $r$  again conditional to all the other QTL in the model at a higher significance level ( $\alpha = 0.05$ )

$$\sigma_p^2 = \mathbf{G}_r^{(i,i')} \sigma_r^2 + \sum_{q \neq r} \mathbf{G}_q^{(i,i')} \sigma_q^2 + \sigma_e^2$$

- ▶ *Position refining* updates QTL positions
- ▶ *Backward elimination* excludes QTL showing no significant effect

QTL detection consists of a stepwise procedure

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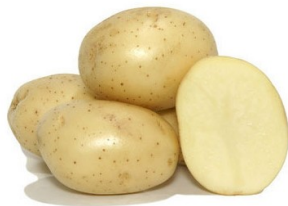
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# Potato: 'Atlantic' × B1829-5 ( $N = 154$ )

- *Solanum tuberosum* ( $2n = 4x = 48$ ,  $x = 12$ )

## Phenotypic data

- 4 years (2006, 2007, 2008, 2014)
- 7 traits
  - 1 Plant yield
  - 2 Internal heat necrosis (IHN) severity
  - 3 IHN incidence
  - 4 Specific Gravity
  - 5 Dry Matter
  - 6 Skin texture
  - 7 Foliage maturity



## The recombination landscape and multiple QTL mapping in a *Solanum tuberosum* cv. 'Atlantic'-derived $F_1$ population

 Guilherme Da Silva Pereira,  Marcelo Mollinari,  Mitchell J Schumann, Mark E Clough, Zhao-Bang Zeng,  Craig Yencho

doi: <https://doi.org/10.1101/2020.08.24.265397>

This article is a preprint and has not been certified by peer review



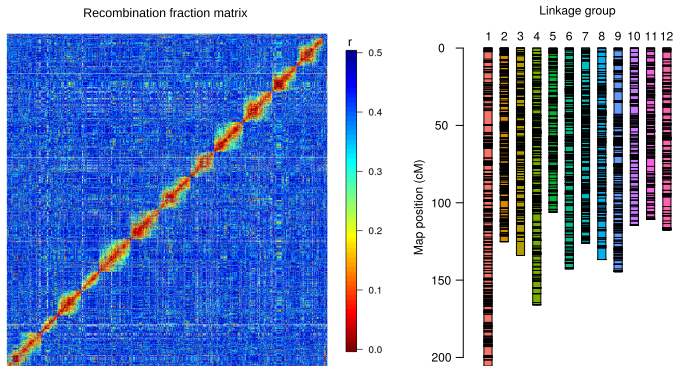
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# Potato: 'Atlantic' × B1829-5

Genotypic data: 8,303 SNPs (Illumina chip array)

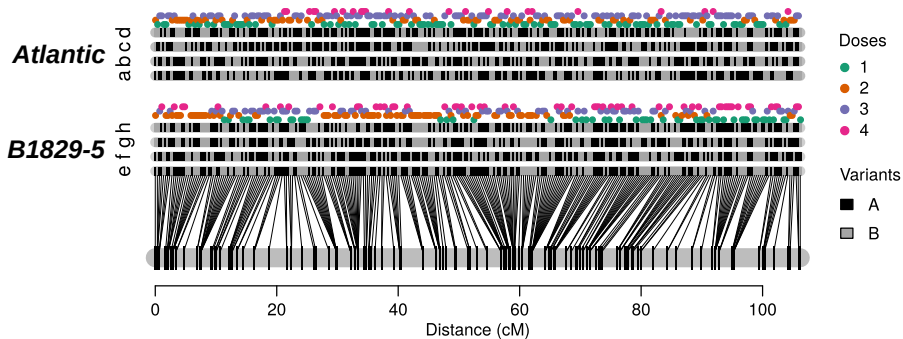
- 12 linkage groups: 4,285 SNPs (1,630 cM)
- Parental homologous haplotypes
- Inheritance of chromosomal segments from parents to progeny



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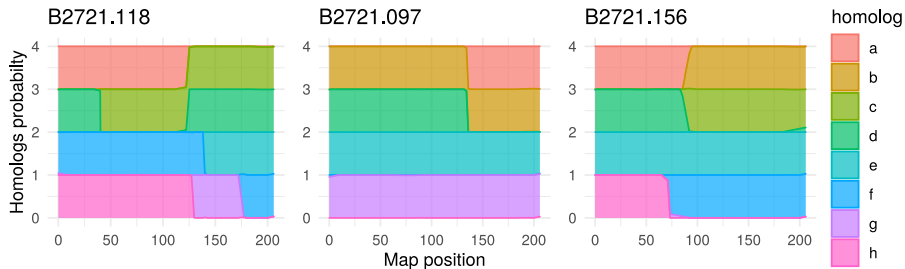
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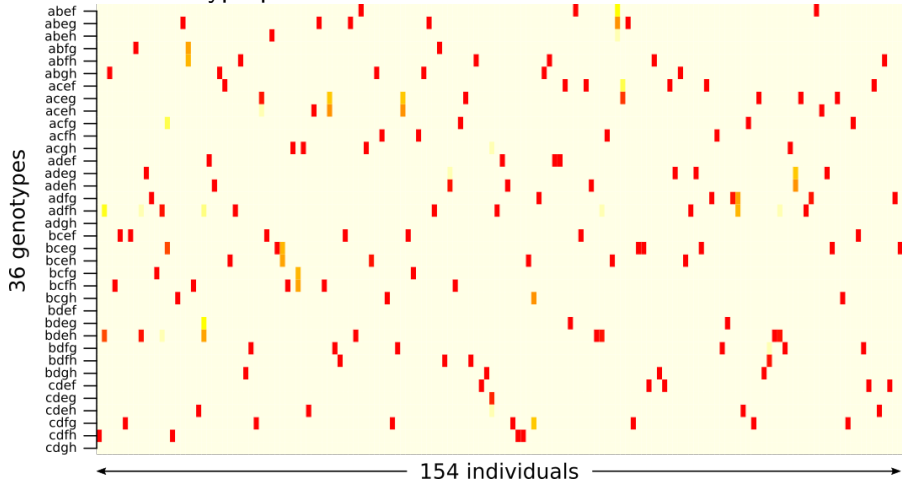
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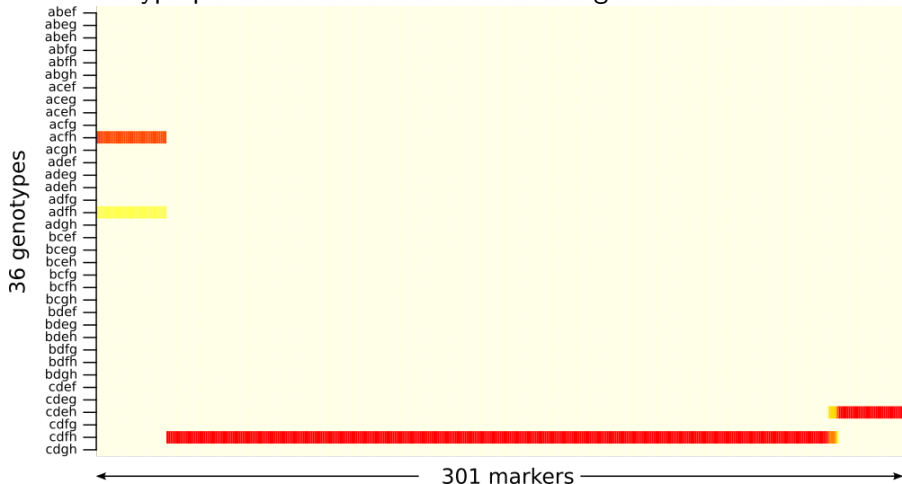
Genotype probabilities for all individuals on a marker



Individuals sharing the same alleles are more related

# Potato: 'Atlantic' × B1829-5

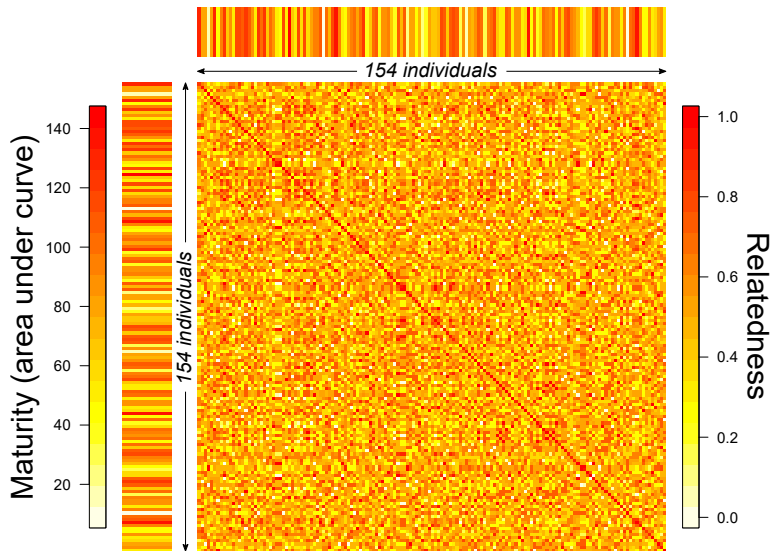
Genotype probabilities of an individual along the chromosome 5



Differences between individuals are due to recombination and segregation

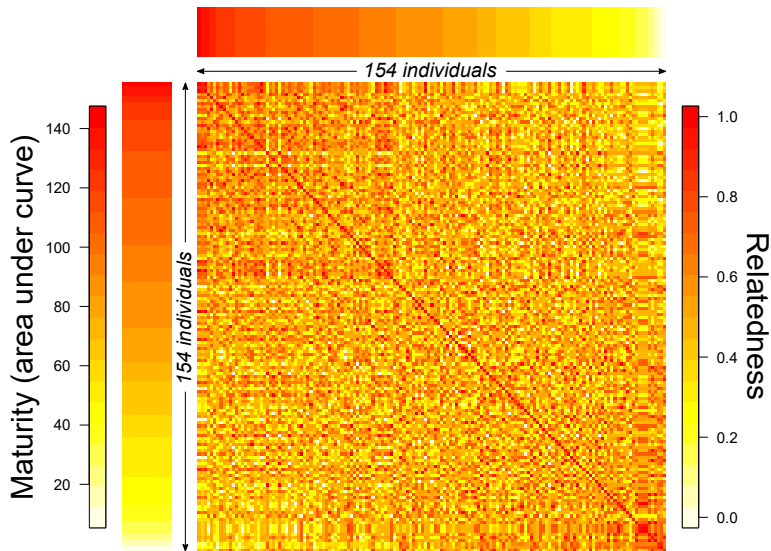
## Potato: 'Atlantic' × B1829-5

- Unordered trait values

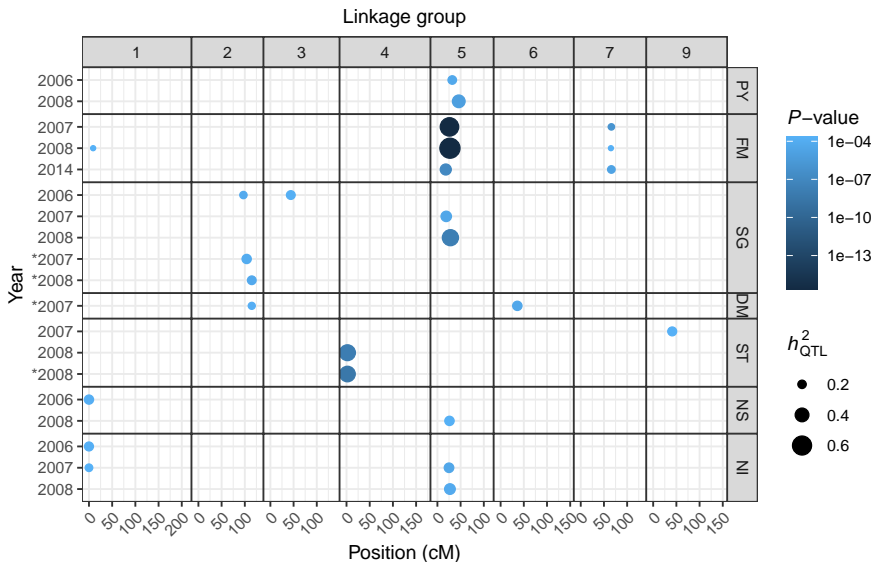


# Potato: 'Atlantic' × B1829-5

- Ordered trait values



# Potato: 'Atlantic' × B1829-5





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# Sweetpotato: Beaugard × Tanzania ( $N = 315$ )

- *Ipomoea batatas* ( $2n = 6x = 90$ ,  $x = 15$ )

## Phenotypic data

- 3 sites, 2 years
- 4 quality traits + 1
  - 1 DM – Dry Matter (%)
  - 2 Starch (%)
  - 3 BC –  $\beta$ -carotene (%)
  - 4 FC\_P – Flesh color in Peru
  - 5 FC\_U – Flesh color in Uganda

## Genotypic data

- 15 linkage groups with 30,684 SNPs (2708.3 cM)



Theoretical and Applied Genetics (2020) 133:23–36  
<https://doi.org/10.1007/s00122-019-03437-7>

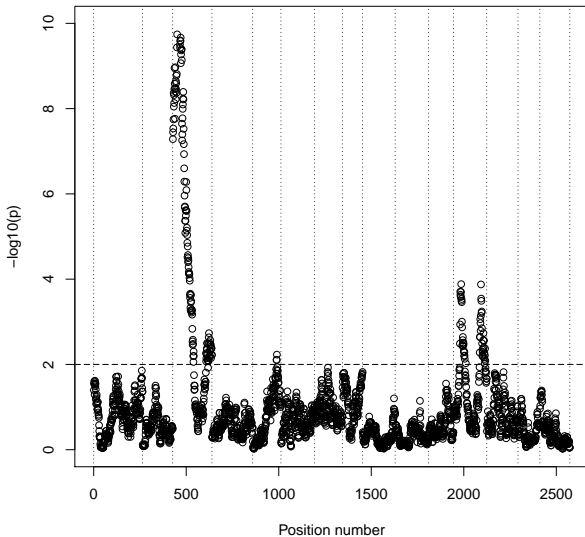
### ORIGINAL ARTICLE

## Quantitative trait loci and differential gene expression analyses reveal the genetic basis for negatively associated $\beta$ -carotene and starch content in hexaploid sweetpotato [*Ipomoea batatas* (L.) Lam.]

Dorcus C. Gemenet<sup>1</sup> · Guilherme da Silva Pereira<sup>2</sup> · Bert De Boeck<sup>3</sup> · Joshua C. Wood<sup>4</sup> · Marcelo Mollinari<sup>2</sup> · Bode A. Olukolu<sup>2,11</sup> · Federico Diaz<sup>3</sup> · Veronica Mosquera<sup>3</sup> · Reuben T. Ssali<sup>5</sup> · Maria David<sup>3</sup> · Mercy N. Kitavi<sup>1</sup> · Gabriela Burgos<sup>3</sup> · Thomas Zum Felde<sup>3</sup> · Marc Ghislain<sup>1</sup> · Edward Carey<sup>6</sup> · Jolien Swanckaert<sup>6</sup> · Lachlan J. M. Coin<sup>7</sup> · Zhangjun Fei<sup>8</sup> · John P. Hamilton<sup>4</sup> · Benard Yada<sup>9</sup> · G. Craig Yencho<sup>2</sup> · Zhao-Bang Zeng<sup>2</sup> · Robert O. M. Mwanga<sup>5</sup> · Awais Khan<sup>3,10</sup> · Wolfgang J. Gruneberg<sup>3</sup> · C. Robin Buell<sup>4</sup>

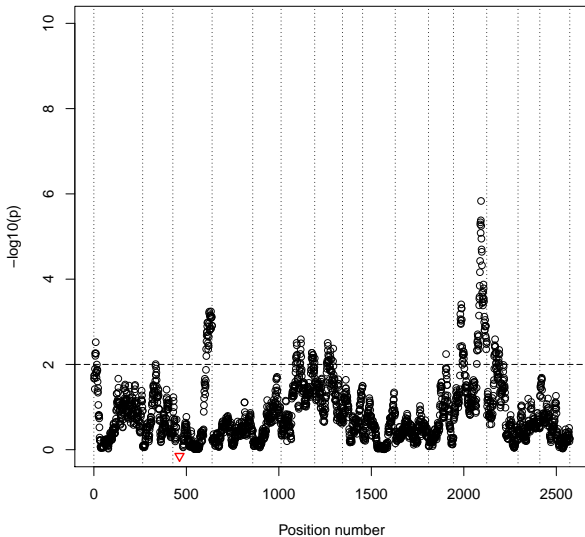
# Dry matter – Forward search

Search round #1



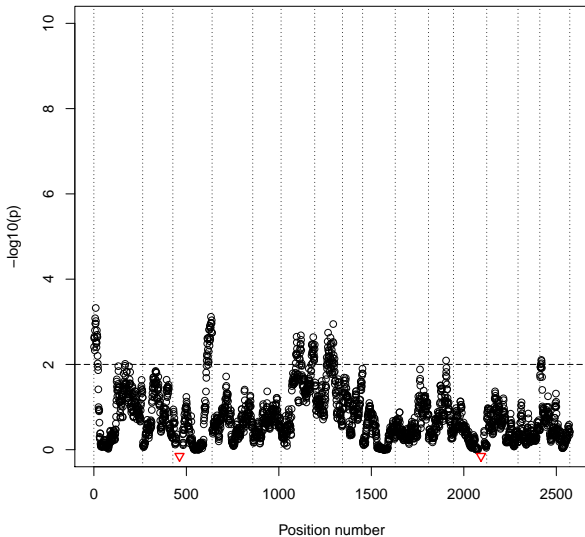
# Dry matter – Forward search

Search round #2



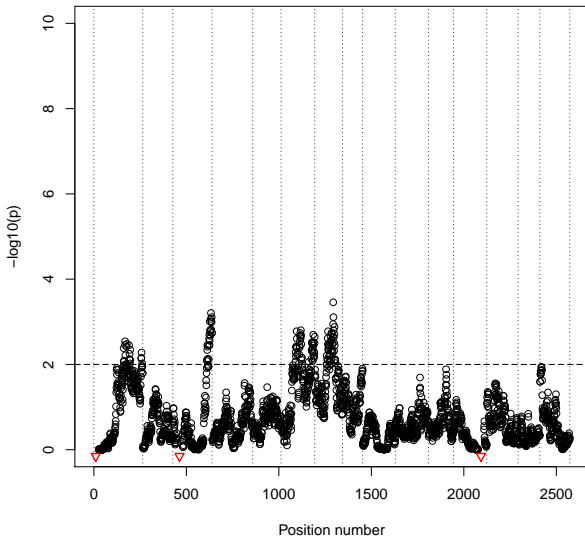
# Dry matter – Forward search

Search round #3



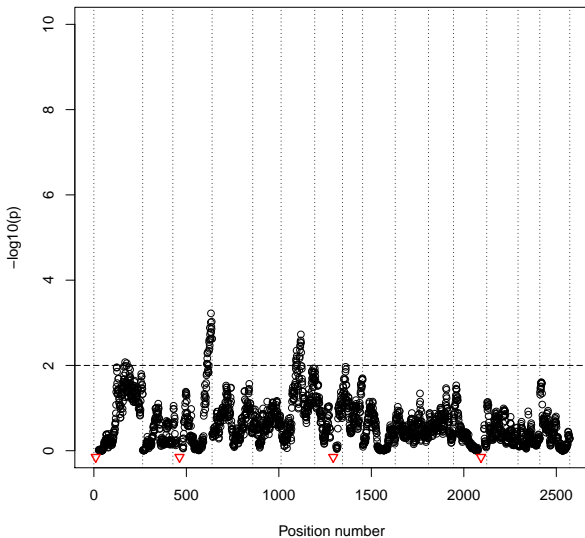
# Dry matter – Forward search

Search round #4



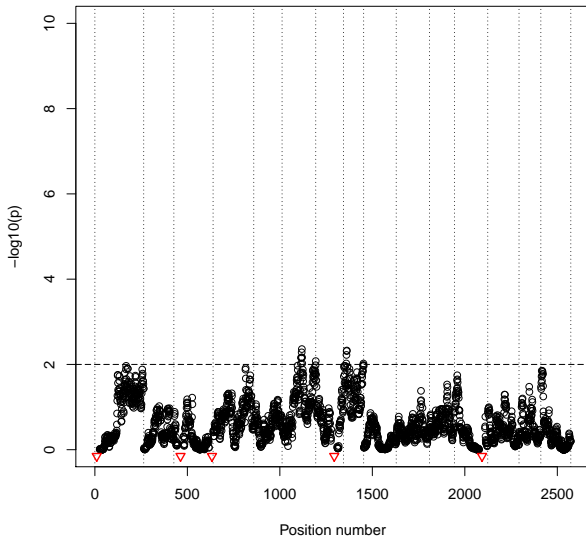
# Dry matter – Forward search

Search round #5



# Dry matter – Forward search

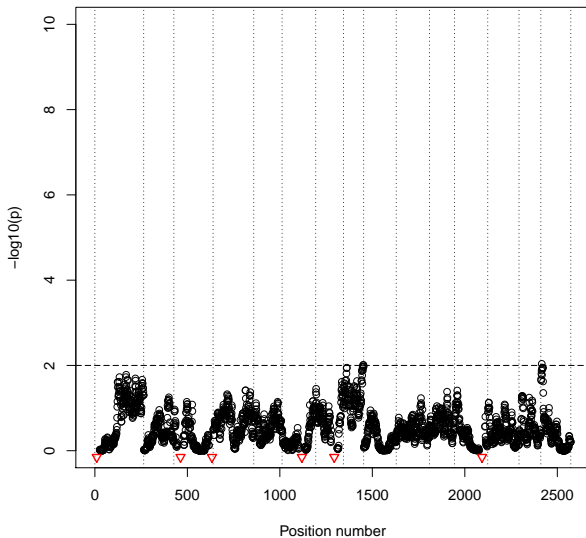
Search round #6





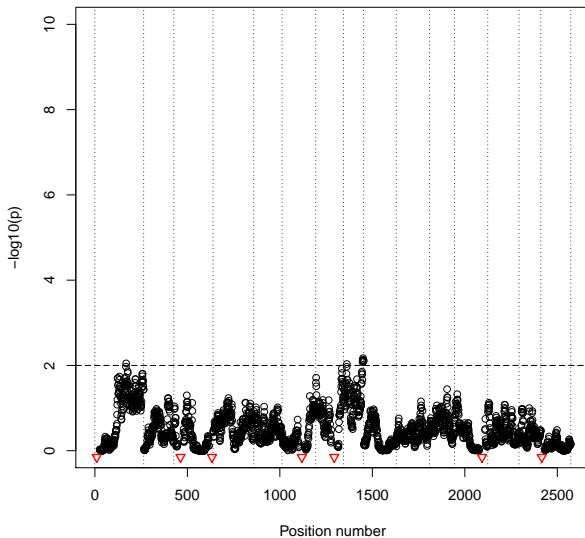
# Dry matter – Forward search

Search round #7



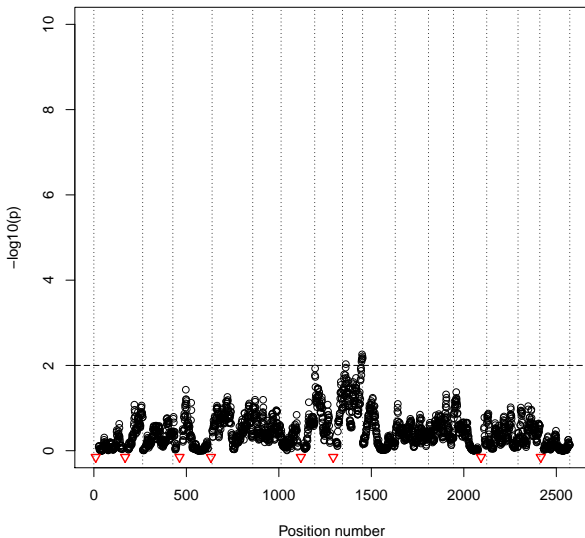
# Dry matter – Forward search

Search round #8



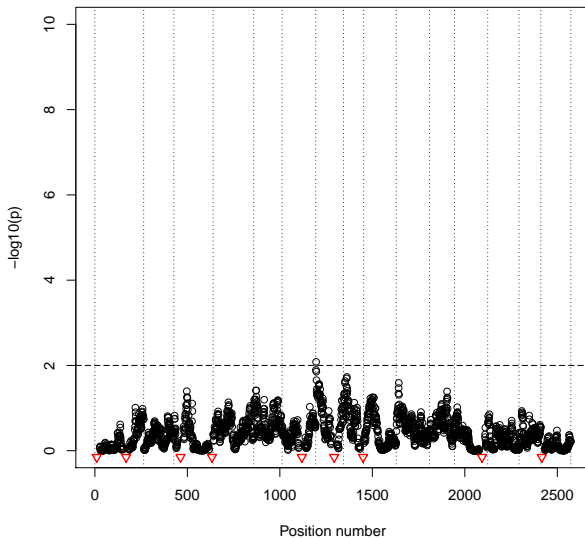
# Dry matter – Forward search

Search round #9



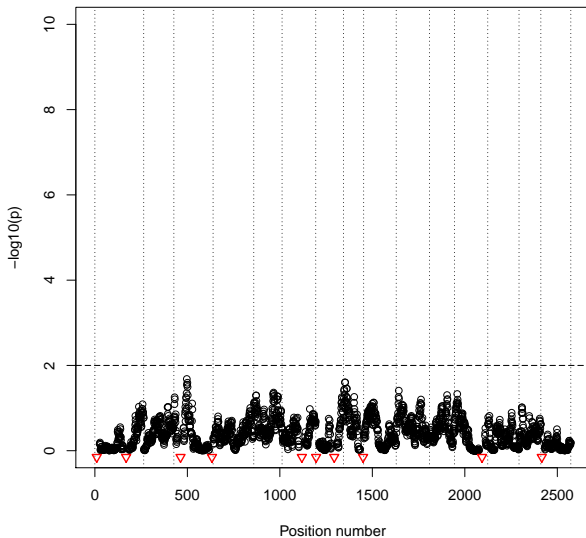
# Dry matter – Forward search

Search round #10



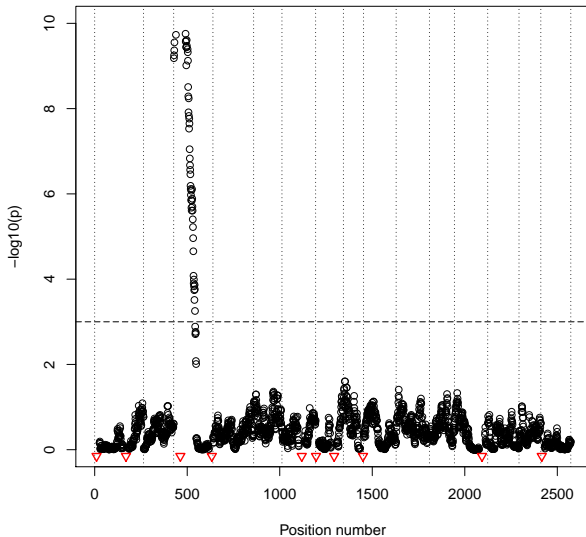
# Dry matter – Forward search

Search round #11



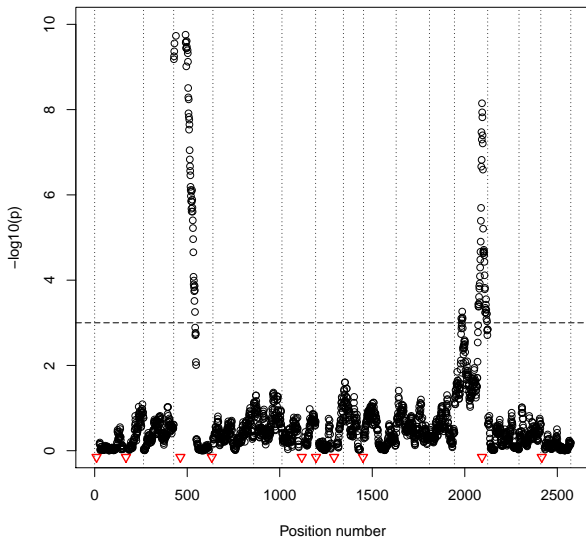
# Dry matter – Model optimization

Refining round #1



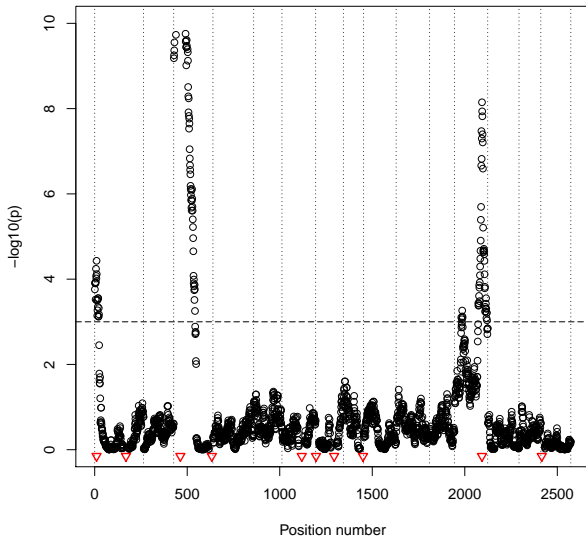
# Dry matter – Model optimization

Refining round #2



# Dry matter – Model optimization

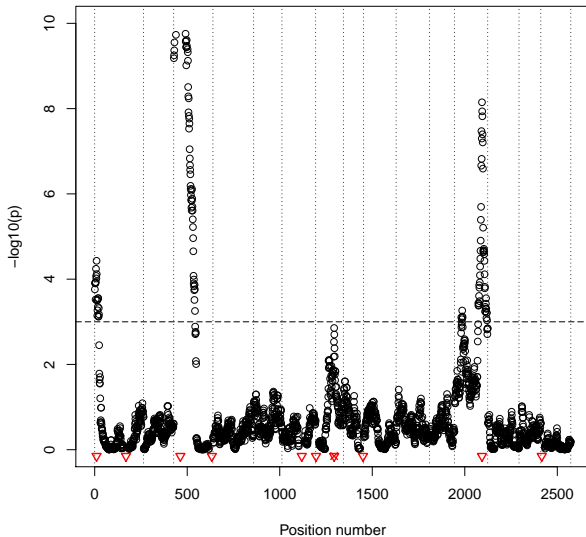
Refining round #3





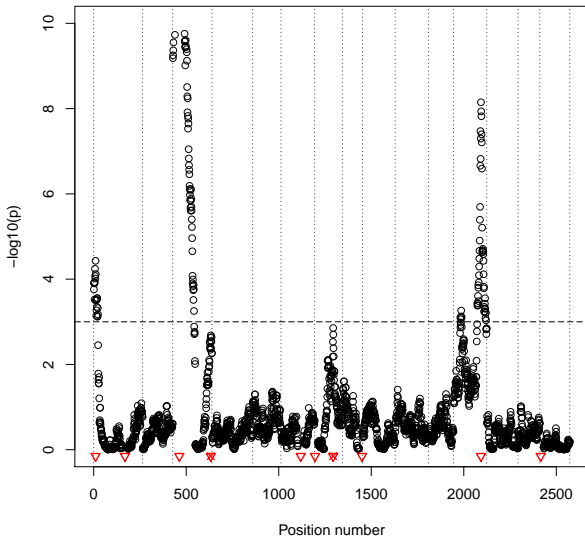
# Dry matter – Model optimization

Refining round #4



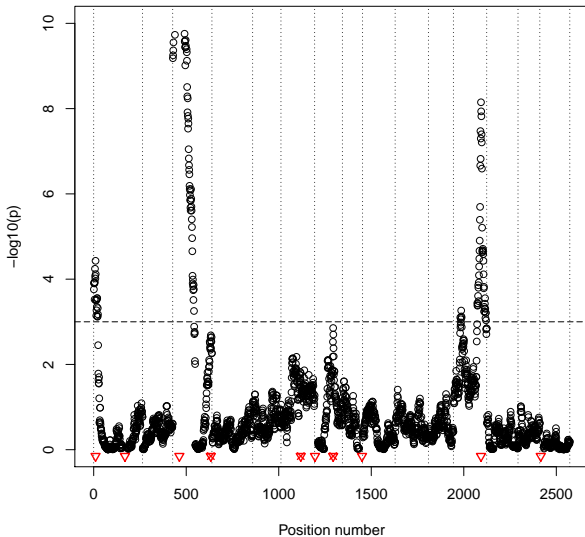
# Dry matter – Model optimization

Refining round #5



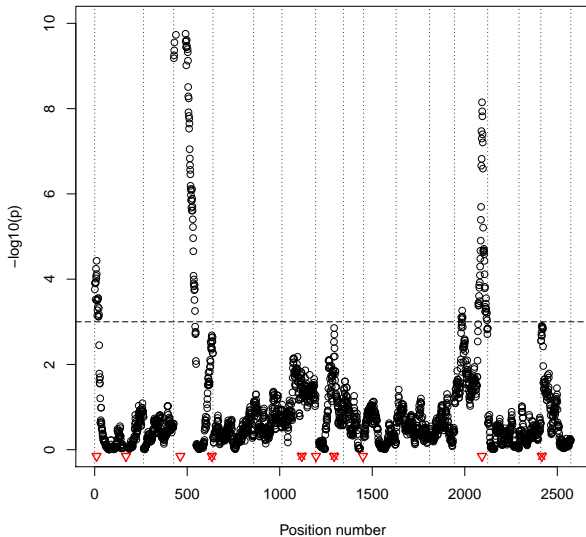
# Dry matter – Model optimization

Refining round #6



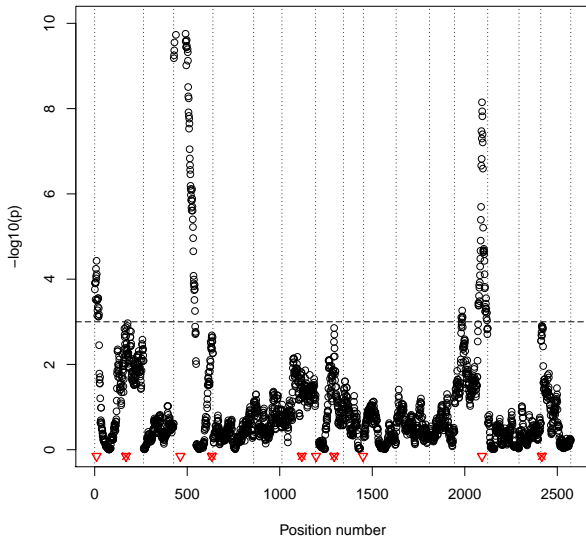
# Dry matter – Model optimization

Refining round #7



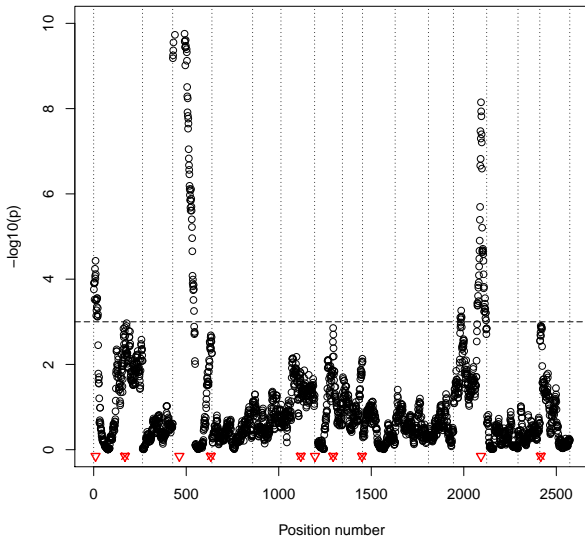
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Refining round #8



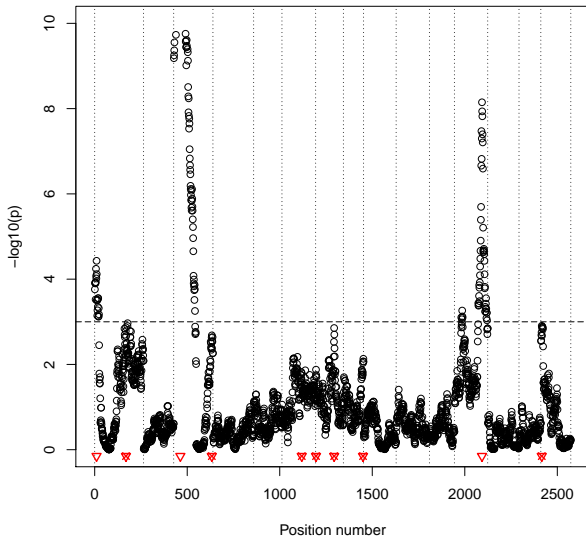
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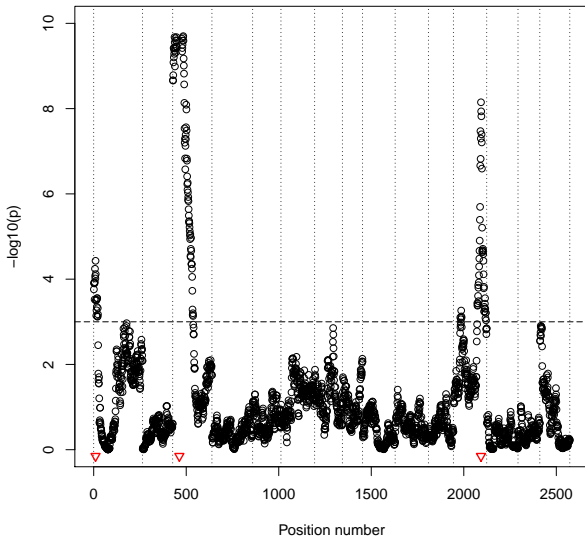
# Dry matter – Model optimization

Refining round #10



# Dry matter – Model optimization

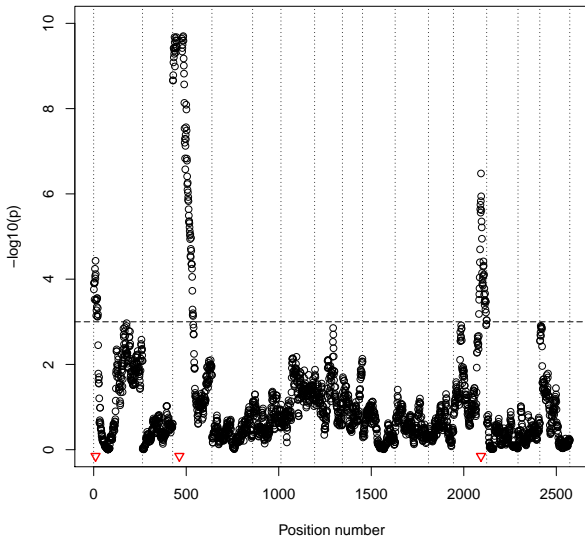
Refining round #1





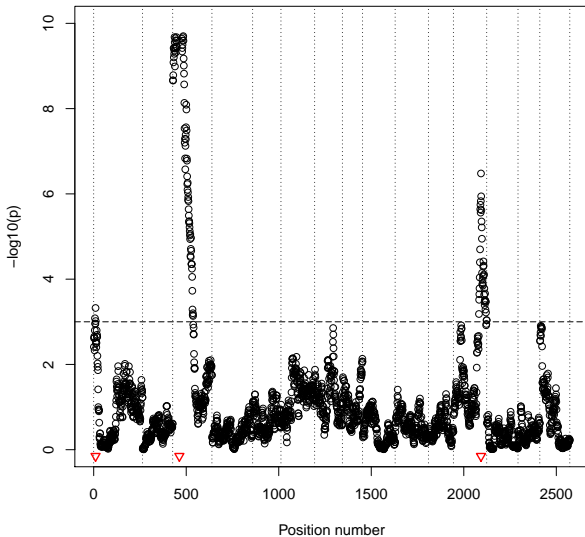
# Dry matter – Model optimization

Refining round #2



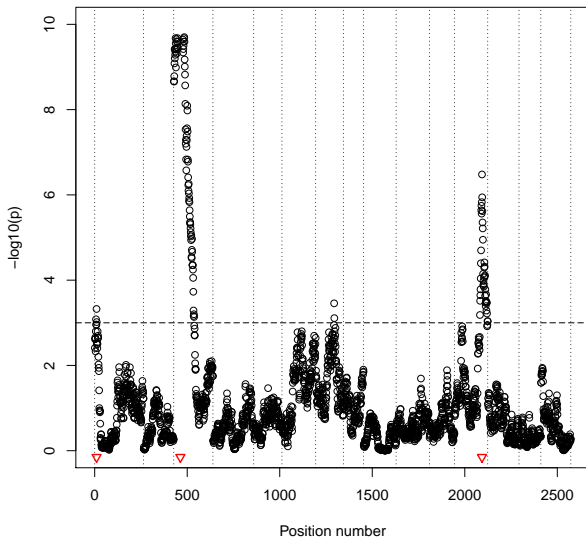
# Dry matter – Model optimization

Refining round #3



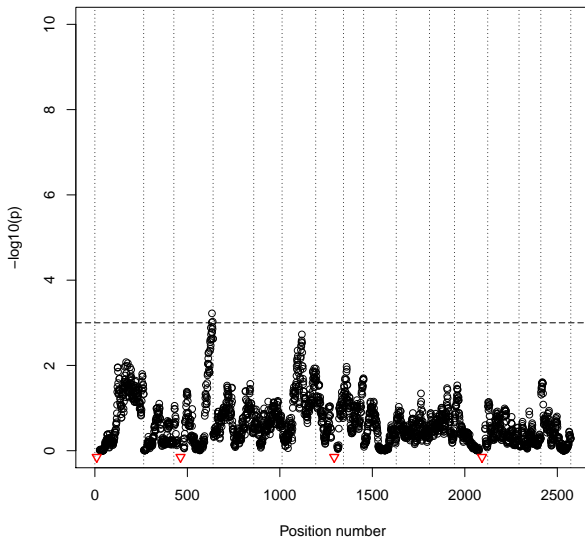
# Dry matter – Model optimization

Completing genome



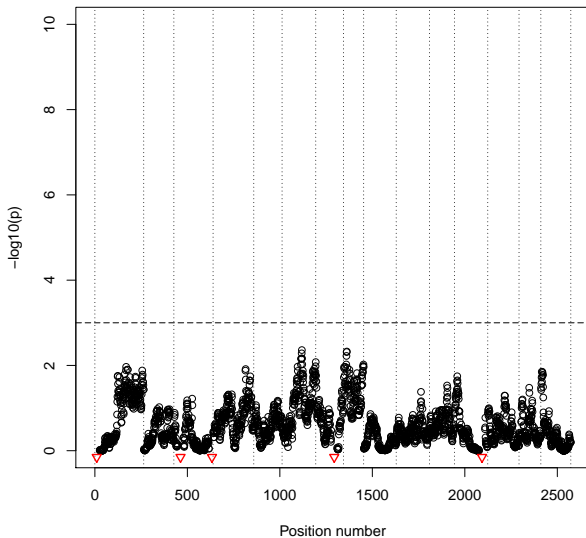
# Dry matter – Forward search

Search round #12

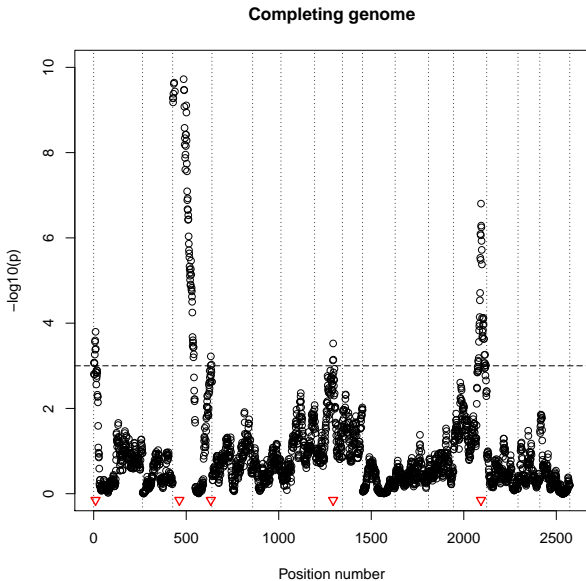


# Dry matter – Forward search

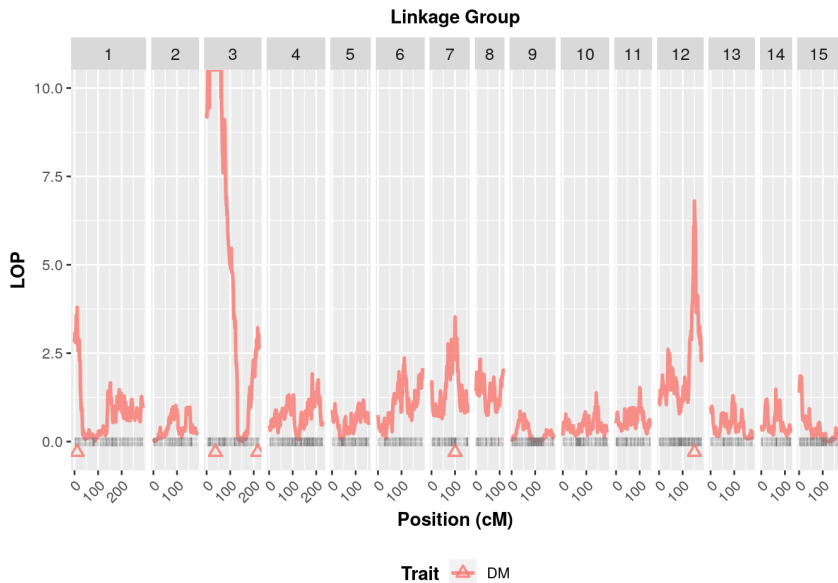
Search round #13



# Dry matter – Model optimization



# Dry matter – Log of $P$ -value (LOP) profile



## Dry matter – Summary

QTL	LG	Pos	Score	<i>P</i> -value	$\mu$	$\sigma_q^2$	$\sigma_e^2$	$h_q^2$
1	1	11.41	158.93	1.59e-04		0.659047		0.0690
2	3	37.44	604.15	7.72e-16		3.571334		0.3738
3	3	217.13	133.93	6.00e-04		0.590448		0.0618
4	7	100.26	151.34	2.99e-04		0.570229		0.0597
5	12	150.05	264.05	1.57e-07		1.065668		0.1115
					27.33		3.096347	0.6759

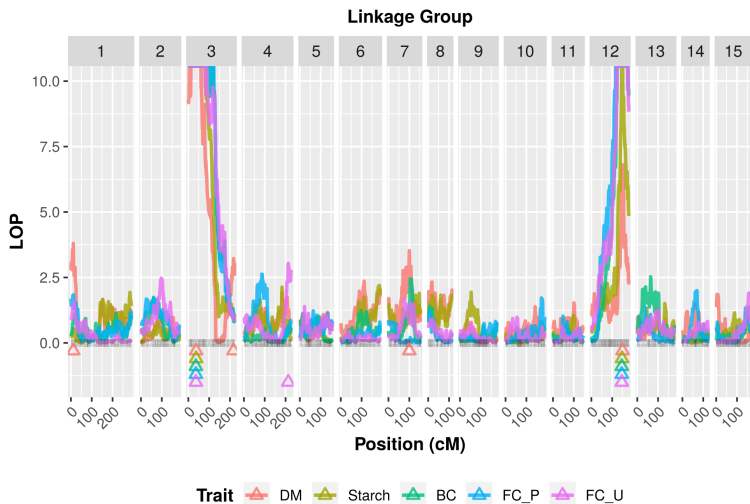
- There are  $Q = 5$  QTL contributing significantly to the variation of Dry Matter in B×T population:

$$\sigma_p^2 = \sum_{q=1}^5 G_q^{(i,i')} \sigma_q^2 + \sigma_e^2$$

- They explain a total of  $h_{\text{total}}^2 = 67.6\%$  of the phenotypic variance



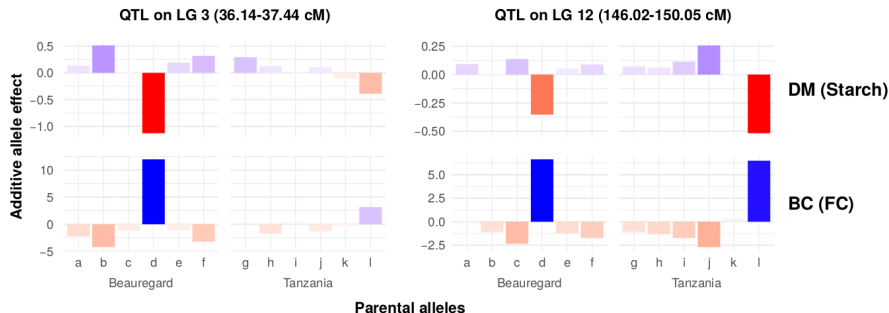
# Quality-related traits – Log of $P$ -value (LOP) profiles



## Total QTL heritability (%) per trait

- DM = 67.6, Starch = 68.2, BC = 79.2, FC\_P = 82.9, FC\_U = 79.7

# Quality-related traits – Allele effects



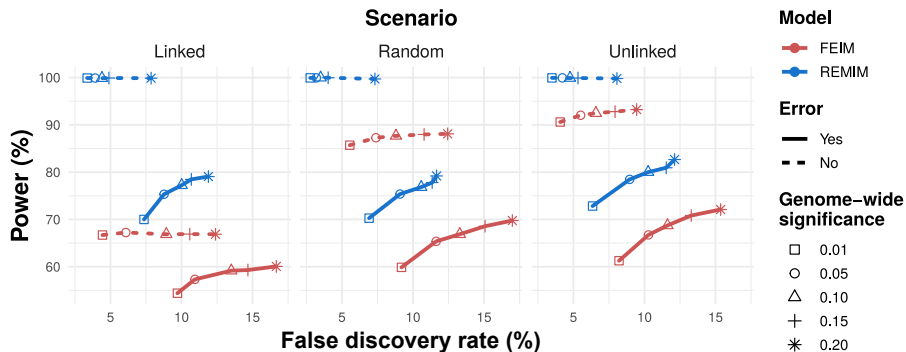
## Genes: RNA-seq and *Ipomoea trifida* genome

- LG 3: *phytoene synthase* and *sucrose synthase* linked genes are involved in carotenoid and starch biosynthesis, respectively
- LG 12: *Orange* gene acts as a molecular switch for chromoplast biogenesis (carotenoid accumulation)

# Fixed vs. random-effect models: simulated traits

1,000 simulations using 'Beauregard'  $\times$  'Tanzania' genetic map

- 3 QTL each simulation:  $h_q^2 = \{0.3, 0.2, 0.1\}$
- 3 scenarios: linked, random, unlinked QTL
- 5 genome-wide significance levels:  $\alpha = \{0.20, 0.15, 0.10, 0.05, 0.01\}$
- Error: yes ( $\sum_{q=1}^3 G_q^{(i,i')} \sigma_q^2 + \sigma_e^2$ ) or no ( $\sum_{q=1}^3 G_q^{(i,i')} \sigma_q^2$ )



# Outline

- 1 Multiple QTL mapping
  - Components of variance
  - Random-effect QTL model
- 2 Applications
  - Potato: 'Atlantic' × B1829-5
  - Sweetpotato: 'Beauregard' × 'Tanzania'
- 3 Final considerations

## Final considerations

### QTLpoly R package

- R package for multiple mapping QTL in autoployploids by using a random-effect model approach
  - ▶ **Repository:** <https://github.com/guilherme-pereira/qtlpoly>
  - ▶ **Tutorial:** <https://guilherme-pereira.github.io/QTLpoly/1-tutorial>
  - ▶ **Issues:** <https://github.com/guilherme-pereira/QTLpoly/issues>
  - ▶ **Paper:** <https://doi.org/10.1534/genetics.120.303080>

### Main functions

- `read_data()` creates object structure with genotypes and phenotypes
- `remim()` performs REMIM in a automatic way
- `fit_model()` fits multiple QTL model with selected regions
- `qtl_effects()` estimates allele effects for each QTL

# Final considerations

## Other QTLpoly features

- It includes functions for fixed-effect model + permutations
- It allows for phenotype weights (more interesting for adjusted means derived from very unbalanced experiments)
- It is flexible such that one can build/modify the QTL model

## To-do list

- Allow for genotypes originated from double-reduction events
- Implement multiple-trait/multiple-environment models
- Increase flexibility to analyze multi-parental populations
- Add experimental design effects (one-step analysis)
- ...

# Thank you!

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