

HOS 6236 Molecular Marker Assisted Plant Breeding Fall 2017

Last Class:

QTL analysis methods

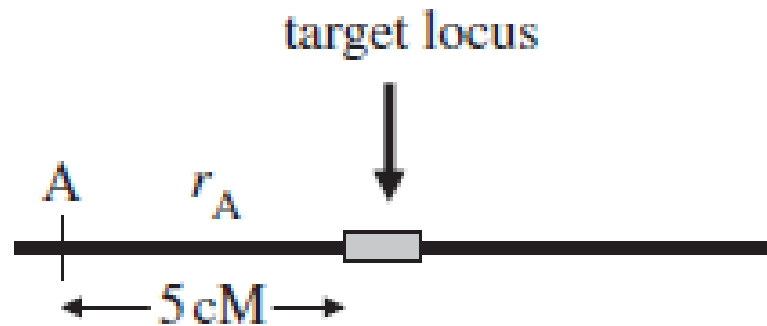
Today's Class:

More about QTL analysis, hands-on activity

Steps for a QTL analysis

1. Create or find a suitable population
2. Genotype with molecular markers
3. Use markers to build a linkage map
4. Phenotype for trait of interest (and more)
5. **Use the linkage map with the phenotypic data to determine whether markers are correlated to traits;** Number of QTLs, the amount of variation and position on genome

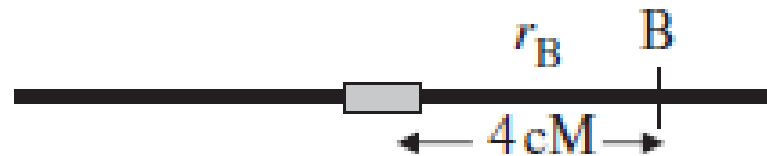
Single vs. Flanking Markers



reliability for selection

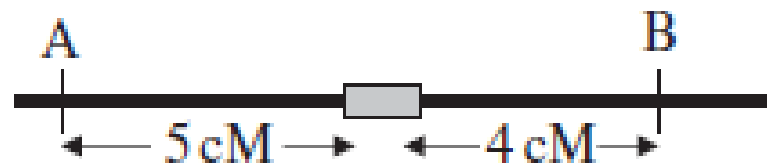
using marker A only:

$$1 - r_A = \sim 95\%$$



using marker B only:

$$1 - r_B = \sim 96\%$$



using both marker A and B:

$$1 - 2 r_A r_B = \sim 99.6\%$$

Permutation Testing

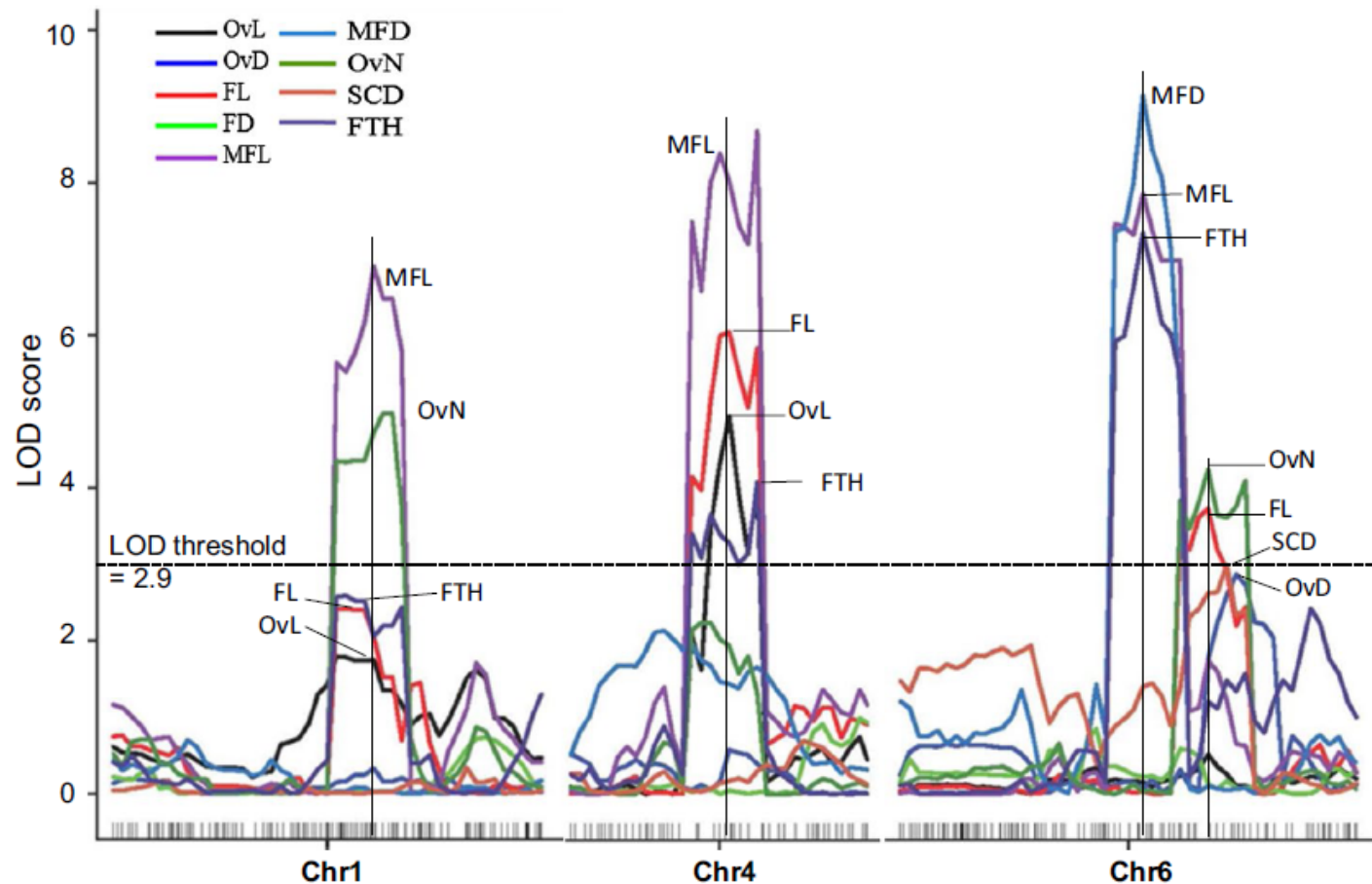
- LOD threshold for QTL is commonly determined by permutation testing
- Shuffle phenotypic data
- Keep genotypic data constant
- Repeat 1000 or more times
- Determine a significance level (α)

Permutation Testing

- If $\alpha = 0.05$, then with 1000 permutation tests, we would falsely declare a QTL-marker association 50 times
- Genome-wide LOD score is calculated for each random iteration of phenotypic data
- For 1000 permutation, the 950th largest LOD score becomes the LOD threshold value

QTL Map

Fig. 4 LOD profiles of fruit size-related QTLs detected with MQM model in the RIL population and high-density SNP maps in cucumber chromosomes 1 (left), 4 (middle), and 6 (right). The dashed horizontal line is LOD threshold for all QTLs (LOD = 2.9). *OvL* ovary length, *OvD* ovary diameter, *FL* immature fruit length, *FD* immature fruit diameter, *MFL* mature fruit length, *MFD* mature fruit diameter. *FTH* flesh thickness, *SCD* seed cavity size and *OvN* ovule number in mature fruit

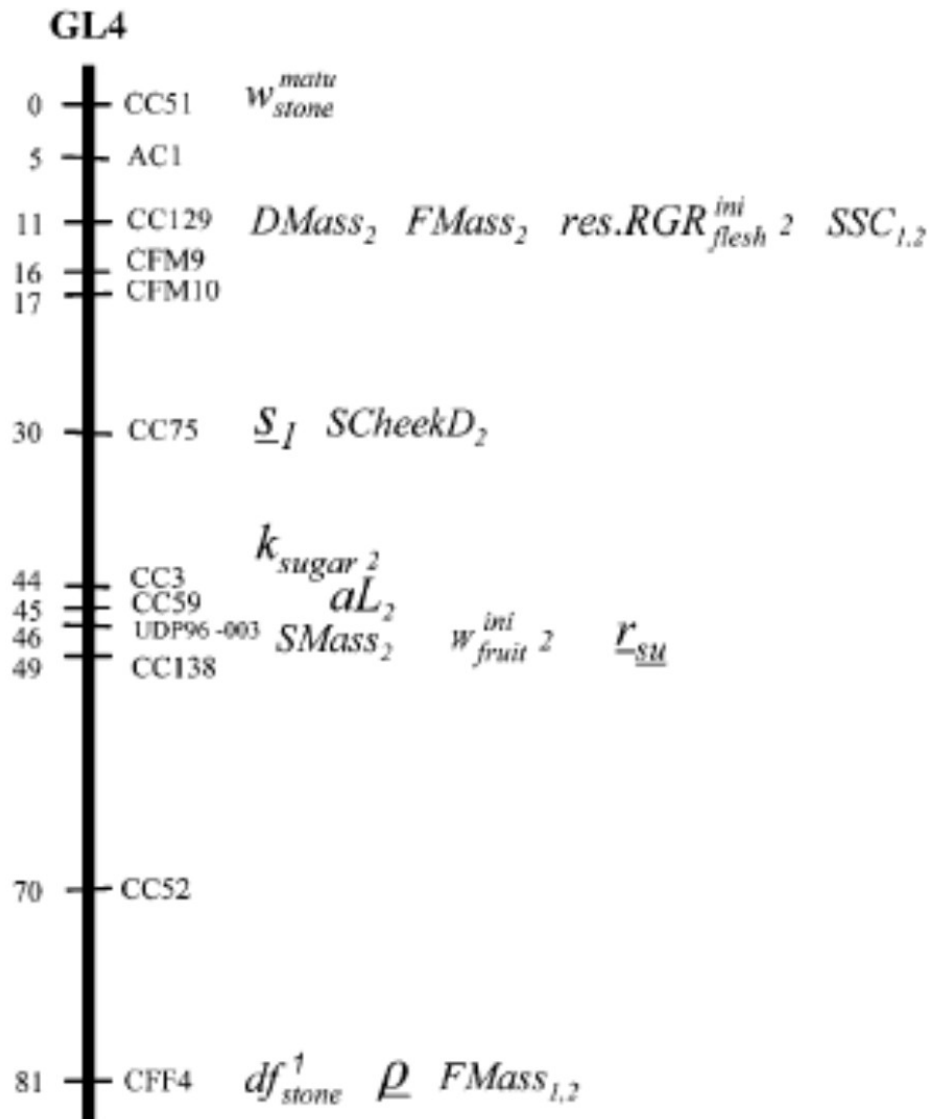


Weng et al. (2015)
Theor. Appl Genet.
128:1747



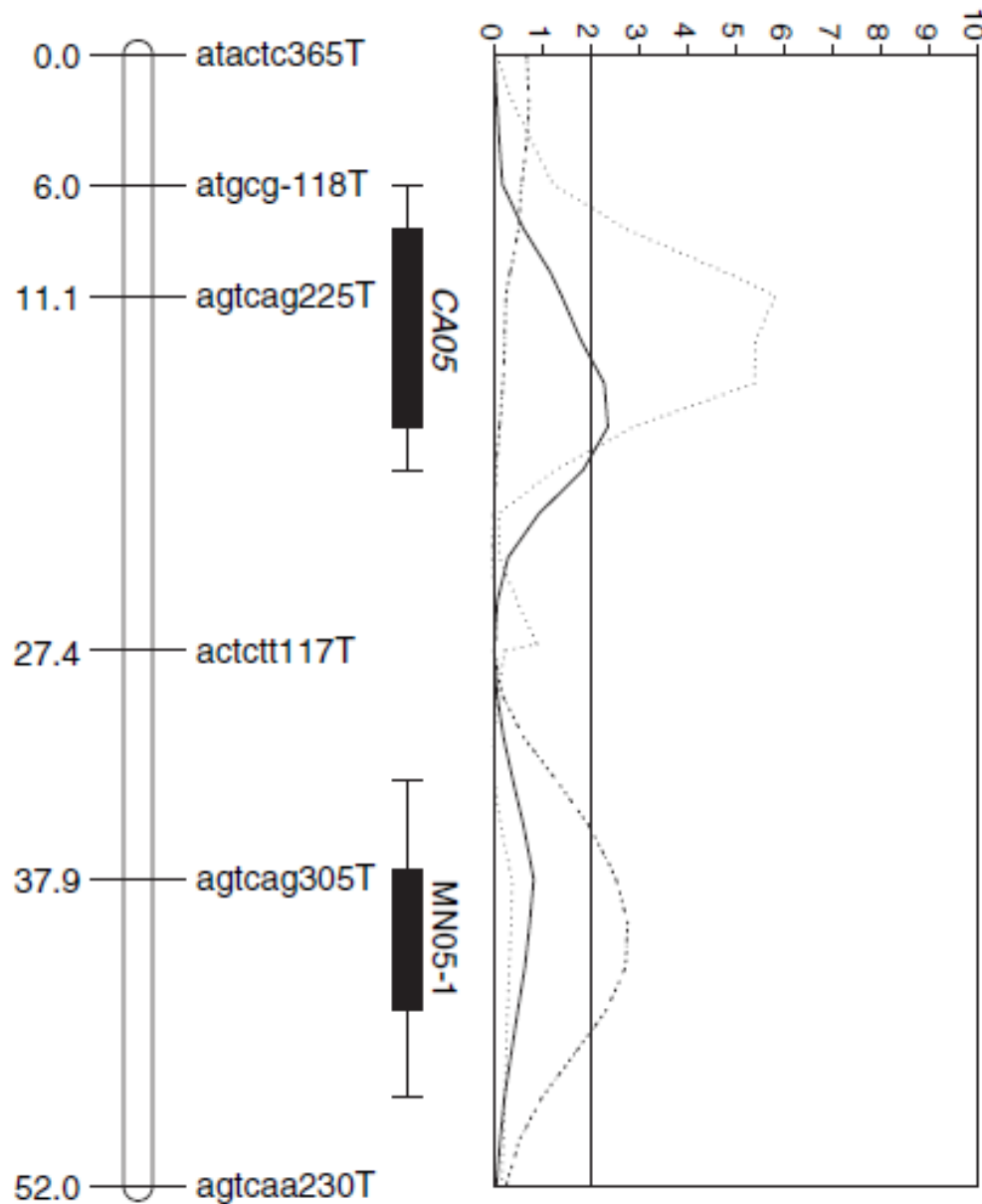
28.5 cM interval: Width of interval – marker density; 2 LOD confidence interval (Lynch and Walsh 1998)

Reporting a QTL



Reporting a QTL

Weebadde et al., 2008



Reporting a QTL

Table 4 Characteristics of the detected QTLs for each trait measured in different years

Trait	Year	Linkage group ^a	Parent ^b	QTL ^c	Nearest marker	LOD score	LOD threshold ^d	R ² (%) ^e
Number of clusters/vine	2003	19	I	<i>Cn3.1</i>	<i>mCTC eATC9</i>	3.28	2.7	7
	2004	8	BP	<i>Cn4.1</i>	<i>VVS4</i>	4.23	2.5	10
		8	BP	<i>Cn4.2</i>	<i>VMC7H2</i>	4.17	2.5	10
Cluster weight	2002	12	I	<i>Cw2.1</i>	<i>mCAC eACA7</i>	3.19	2.6	2.3
	2003	5	BP	<i>Cw3.1</i>	<i>mCAT eAAG13</i>	3.01	2.8	1.3
		16	I	<i>Cw3.2</i>	<i>VMC1E11</i>	2.98	2.5	4
	2004	5	BP	<i>Cw4.1</i>	<i>mCAT eAAG13</i>	3.32	2.7	6.7
		17	BP	<i>Cw4.2</i>	<i>mCTC eATG12</i>	3.02	2.7	4
Number of berries/cluster	2002	8	BP	<i>Bn2.1</i>	<i>mCAT eAAG4</i>	3.72	2.6	5
		12	BP	<i>Bn2.2</i>	<i>mCTG eATT3</i>	3.12	2.6	7
		17	I	<i>Bn2.3</i>	<i>mCTG eATC8</i>	3.02	2.8	6
		2	I	<i>Bn2.4</i>	<i>VVI055</i>	3.19	2.8	1.2
	2003	5	BP	<i>Bn3.1</i>	<i>mCAT eAAG137</i>	4.09	2.7	4
		7	I	<i>Bn3.2</i>	<i>mCAT eATT1</i>	3.01	2.7	4.5
	2004	7	BP	<i>Bn4.1</i>	<i>VMC7A4</i>	4.25	2.6	9
		7	BP	<i>Bn4.2</i>	<i>mCAT eATG15</i>	4.25	2.6	1.5
		5	I	<i>Bn4.3</i>	<i>mCAG eATG15</i>	3.32	2.2	2.8
		7	I	<i>Bn4.4</i>	<i>VVMD7</i>	3.02	2.2	5.7
		7	I	<i>Bn4.5</i>	<i>VMC16F3</i>	2.83	2.2	4
Berry weight	2002	5	I	<i>Bw2.1</i>	<i>mCAT eATT2</i>	3.31	3	10
		16	I	<i>Bw2.2</i>	<i>mCTA eAAG5</i>	3.35	3	3
		5	BP	<i>Bw2.3</i>	<i>VMC3B9</i>	3.2	2.7	19
	2003	4	BP	<i>Bw3.1</i>	<i>VMC7H3</i>	3.13	2.5	5
		13	BP	<i>Bw3.2</i>	<i>mCAG eAAG13</i>	2.95	2.5	2
	2004	20	BP	<i>Bw4.1</i>	<i>mCAT eAAG14</i>	3.19	2.5	5.8
		20	I	<i>Bw4.2</i>	<i>mCTG eAAG3</i>	3.24	2.8	4.8

Estimating the QTL effect

Sample size plays a significant role in the QTL effect.

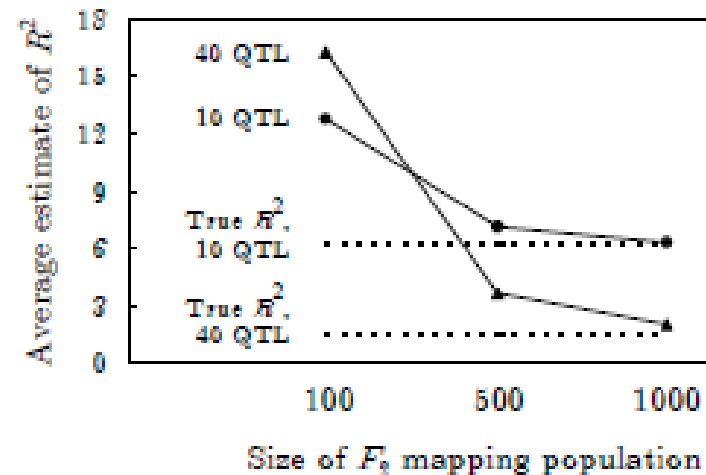
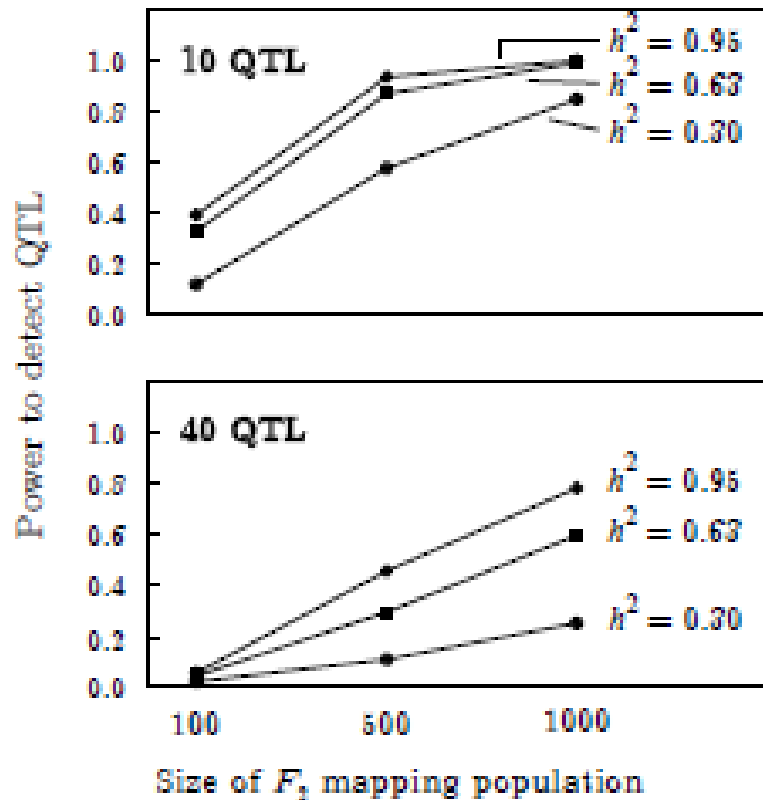
For example a small sample size may miss QTLs with small effect thus will overestimate the QTL effect of the ones that is able to identify, the “Beavis effect” (Beavis 1994, 1997)

Also, a small detected QTL effect could be either:

- A tightly-linked QTL with a small effect

- A loose linkage QTL with large effect

The Beavis Effect



- Phenotypic variance associated with QTL are overestimated when small population sizes are used
- Number of QTL are underestimated with small population sizes

Multiple Interval Mapping (MIM)

All previous methods tested by a single QTL at the time

Alternatively the test could be for multiple QTLs simultaneously. MIM consider multiple QTLs simultaneously.

Has higher precision

Can test for QTL-QTL interactions (i.e. epistasis)

Better description of the genetic architecture of a trait:
number, position, effect size, and interactions.

QTL analysis in outcrossing species

For some species inbred lines are just not possible

QTLs detected by:

- inbred-line crossing – fixed difference between lines
- outbred crosses – differences within population variation

Within population variation has less power to detect QTLs. In inbred all F1 are identical, so all are informative, and LD is maximized.

Use of population of relatives is common (i.e. half-sibs, full-sibs, etc.)

Inbred vs outbred

Not all outbred parents are **informative** (double heterozygous)

Number of alleles may be larger than 2 in outbred

Linkage phase may be different in different parents. So analysis by parent are needed.

Problems using QTLs in Breeding

1. QTLs are not the causal variant
2. QTLs are in LD with the causal variant
3. Different families can have different Marker allele – causal variant allele phase
4. Different families may have recombination among the marker and QTL
5. Different families may have epistatic effects that are not taken in account by QTLs
6. Breeding population may not have the allele discovered in the mapping population
7. There may be QTL-by-environment interaction

ALL THESE WILL BE REVIEWED IN THE MAS SECTION!