An Examination of Quantitative Traits in *Brassica*

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Outline

• Brassica study

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- linkage maps
- classical vs. mapmaker QTL
- testing with likelihood ratios
- confidence interval for major QT gene
- major and minor QT genes
- stepwise location of QT loci

Brassica study at UW–Madison Tom Osborn and Keming Song, Agronomy

95 families297 RFLP markers24 (20) quantitative traits

- Song, Suzuki, Slocum, Williams and Osborn (1990)
- Song, Slocum and Osborn (1990)

Testing for join in Linkage Maps

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two strands vs. one?

LOD(one) - LOD(two) \sim \chi_1^2/2log(10)

\frac{\text{level} \quad \text{LOD}}{.05 \quad 0.834}

.01 1.441

.001 2.351

.0001 3.287

.00001 4.237
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multiple comparisons problem: doing many tests with 100 markers there are about 4950 tests possible exptwise error rate = $1 - (1 - \alpha)^{4950}$ = .39 for α = .0001

= .05 for α = .00001

one actually looks for MLE of whole map

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Testing for QTLs

$$trait = \begin{cases} mean & alleles \\ \mu + a & AA \\ \mu + d & if & ABorBA \\ \mu - a & BB \end{cases} + error$$
$$y_i = \mu + ax_i + d(1 - |x_i|) + \epsilon_i$$

$$\begin{array}{l} \mu = \text{reference mean} \\ a = \text{additive effect} \\ d = \text{dominance effect} \\ x_i = +1 \ (\text{AA}), \ 0 \ (\text{AB}), \ -1 \ (\text{BB}) \\ \epsilon_i \sim N(0, \sigma^2) \end{array}$$

null hypothesis: no QTL (a = d = 0)

nuisance parameter: position along chromosome t recombinant frequency btw. flanking markers

$$\hat{y}_{i} = \hat{\mu} + \hat{a}x_{i} + \hat{d}(1 - |x_{i}|)$$

$$SSModel = \sum_{i}(\bar{y} - \hat{y}_{i})^{2}$$

$$SSError = \sum_{i}(y_{i} - \hat{y}_{i})^{2} = (n - 3)\hat{\sigma}^{2}$$

$$F = SSModel/2\hat{\sigma}^{2} \sim F_{2,n-3}$$

Distribution of LODs

$$L(\hat{a}, \hat{d}) = \log_{10} Pr\{trait | m \hat{o} del\} \\= \frac{1}{2\hat{\sigma}^2} SSError / \log(10) + terms in n, \sigma^2$$

$$LOD = L(0,0) - L(\hat{a},\hat{d})$$

= $\frac{1}{2}SSModel/(\hat{\sigma}^2 \log(10))$
 $\approx \chi_2^2/(2\log(10))$

luck! LOD=2 for 1% level test

| level | LOD |
|--------|-------|
| .05 | 1.301 |
| .01 | 2.000 |
| .001 | 3.000 |
| .0001 | 4.000 |
| .00001 | 5.000 |

counts and generalized linear models: use $LOD = deviance / \log(10)$

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Confidence Interval for Major QT locus

pointwise estimate at ML peak must have LOD > 2 at peak to be significant at 1% level 99% confidence interval

{t; $\max_t(LOD(\hat{a}, \hat{d})) - LOD(\hat{a}, \hat{d}|t) > 2$ }

equivalent to normal interval in additive case based on profile likelihood need locally quadratic LOD

profile actually quite bumpy less (?) information at markers

CI may include regions on other chromosomes

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Confidence Intervasl for Major and Minor QT loci

Major Gene

confidence interval for major gene

twin genes close together: difficult to distinguish could use χ^2 test with df=4

Minor Gene on another chromosome

add LODs for both chromosomes 2-D contour map 1-D look conditional on major gene minor gene must add signif. amount test with χ^2 at df=2

problem:

chromosomes not independent due to nature of sample

Stepwise Location of QTLs

Finding loci

remove effect of major gene locate max conditional LOD across genome repeat in stepwise fashion

when to stop? same issues as stepwise regression

Effect on Estimates

additive vs. dominance near MLE very nonlinear note sensitivity to loci on other chromosomes