

qtl2_test

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1. Start your engine

```
library(qtl2)
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --

## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.6      v dplyr  1.0.7
## v tidyr   1.1.4      v stringr 1.4.0
## v readr   2.1.1      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter()   masks stats::filter()
## x dplyr::lag()      masks stats::lag()
## x readr::read_csv() masks qtl2::read_csv()
```

write control file and read

```
control_test <- file.path("C:/Users/Bora Kim/Desktop/Bora Kim/6. RIL_QTL/RIL_data_analysis", "control_test.yaml")
write_control_file(control_test,
  crosstype="riself",
  geno_file="genotypes.csv",
  gmap_file="physical_map_mb.csv", #physical map in Mb
  pmap_file="physical_map_mb.csv",
  pheno_file="phenotypes_median_test.csv",
  geno_codes=c(A=1L, B=2L),
  alleles=c("A", "B"),
  na.strings=c("-", "NA"))
```

```
test <- read_cross2("C:/Users/Bora Kim/Desktop/Bora Kim/6. RIL_QTL/RIL_data_analysis/control_test.yaml")
```

2. QTL analysis

Since we don't have a genetic map that is used for probability calculation, here genetic map was estimated.

```
est<-est_map(test, cores=parallel::detectCores()-1)
test$gmap<-est #replace gmap object
```

Insert pseudomarkers into estimated genetic map

```
map <- insert_pseudomarkers(map=test$gmap, step=0.1)
```

Compares the number of markers between the estimated genetic map and the pseudomarkers-inserted map

```
sapply(test$gmap, length)
```

```
## ch01 ch02 ch03 ch04 ch05 ch06 ch07 ch08 ch09 ch10 ch11 ch12  
## 508 412 438 398 349 329 327 392 447 332 304 279
```

```
sapply(map, length)
```

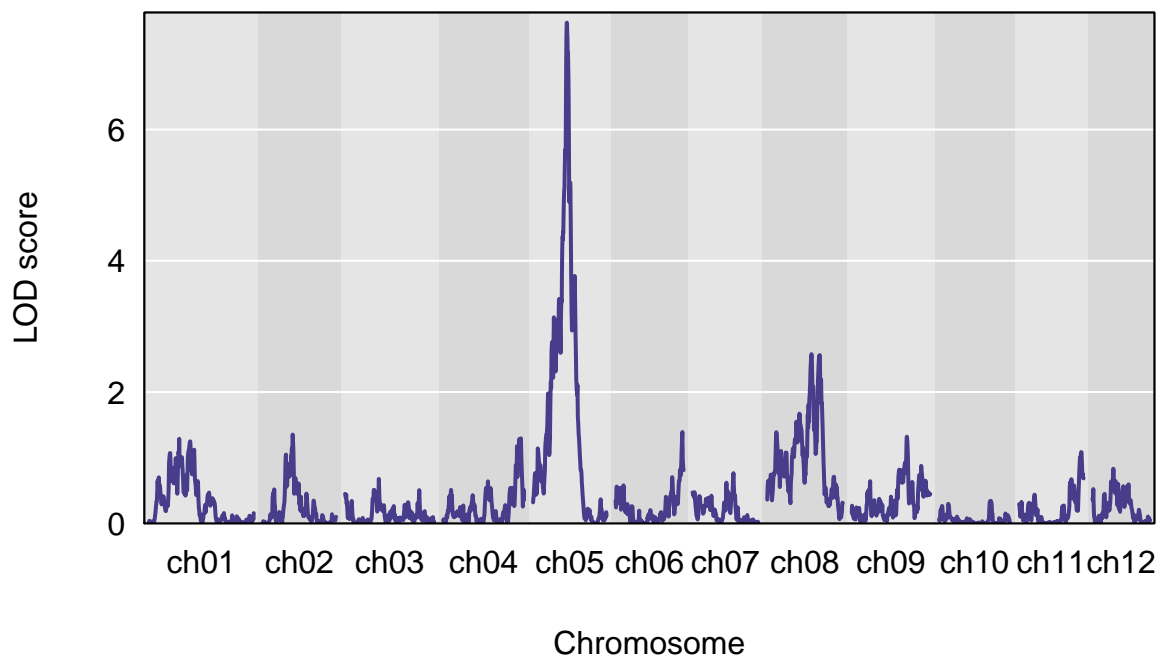
```
## ch01 ch02 ch03 ch04 ch05 ch06 ch07 ch08 ch09 ch10 ch11 ch12  
## 1827 1343 1566 1417 1278 1194 1160 1348 1441 1229 1119 1003
```

Calculate genotype probability based on pseudomarkers-inserted map

```
pr <- calc_genoprob(cross=test, map=map, error_prob=0.001)
```

Finding LOD peaks

```
out <- scan1(genoprob = pr, pheno = test$pheno, cores=parallel::detectCores()-1)  
plot_scan1(out, map = map, lodcolumn = "fresh_shoot_g")
```



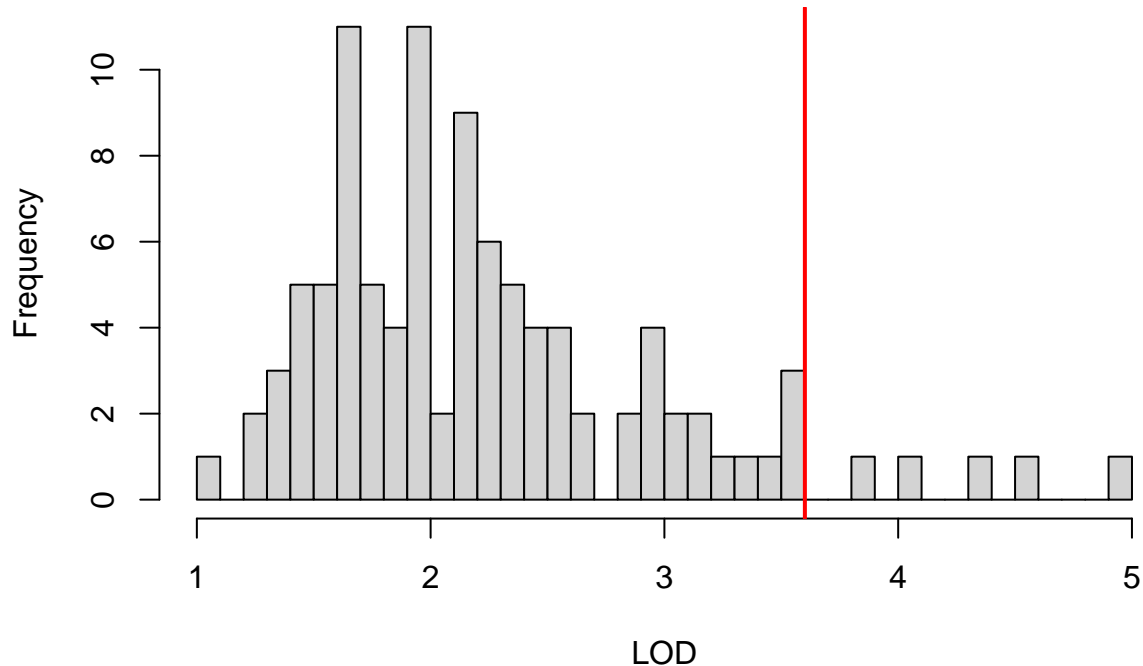
permutation test; is our LOD peak statistically significant comparing to randomly generated LOD peaks?

```
operm <- scan1perm(genoprob = pr, pheno = test$pheno, n_perm = 100, cores=parallel::detectCores()-1)  
summary(operm, alpha=c(0.2, 0.05))
```

```
## LOD thresholds (100 permutations)  
##      fresh_shoot_g fresh_root_g total_fresh_biomass_g  
## 0.2           2.86           2.74           2.82  
## 0.05          3.60           3.20           3.43
```

```
hist(operm[, 'fresh_shoot_g'], breaks = 50, xlab = "LOD", main = "LOD scores for fresh_shoot_g scan with
abline(v = summary(operm)[, 'fresh_shoot_g'], col = 'red', lwd = 2)
```

LOD scores for fresh_shoot_g scan with threshold in red



Find peaks that has higher LOD threshold that is determined from permutation test

```
thr <- summary(operm)
find_peaks(scan1_output = out, map = map, threshold = thr, prob = 0.95, expand2markers = T)
```

##	lodindex	lodcolumn	chr	pos	lod	ci_lo	ci_hi
## 1	1	fresh_shoot_g	ch05	44.1000	7.629985	43.05117	46.12834
## 2	2	fresh_root_g	ch04	101.3153	3.974327	76.74435	105.39196
## 3	3	total_fresh_biomass_g	ch05	44.0000	7.088931	42.54355	46.12834

Look at the markers in the range of interval (between ci_lo and ci_hi) where it showed significant LOD score for the phenotype 'fresh_shoot_g' in chromosome 5.

```
map$ch05 %>%
  as.data.frame()%>%
  filter(between(., 43.05117, 46.12834)) # Questionhere: how to get physical position of them?
```

```
##
## snp1912      43.05117
## snp1913      43.05117
## cch05.loc43.1 43.10000
## cch05.loc43.2 43.20000
## cch05.loc43.3 43.30000
## cch05.loc43.4 43.40000
## cch05.loc43.5 43.50000
```

```

## snp1914      43.55879
## snp1915      43.55879
## cch05.loc43.6 43.60000
## cch05.loc43.7 43.70000
## cch05.loc43.8 43.80000
## cch05.loc43.9 43.90000
## cch05.loc44   44.00000
## cch05.loc44.1 44.10000
## cch05.loc44.2 44.20000
## cch05.loc44.3 44.30000
## cch05.loc44.4 44.40000
## cch05.loc44.5 44.50000
## snp1916      44.58975
## snp1917      44.58975
## snp1918      44.58975
## snp1919      44.58975
## snp1920      44.58975
## snp1921      44.58975
## snp1922      44.58976
## snp1923      44.58976
## snp1924      44.58976
## snp1925      44.58976
## cch05.loc44.6 44.60000
## cch05.loc44.7 44.70000
## cch05.loc44.8 44.80000
## cch05.loc44.9 44.90000
## cch05.loc45   45.00000
## snp1926      45.09737
## snp1927      45.09737
## snp1928      45.09737
## snp1929      45.09737
## cch05.loc45.2 45.20000
## cch05.loc45.3 45.30000
## cch05.loc45.4 45.40000
## cch05.loc45.5 45.50000
## cch05.loc45.6 45.60000
## cch05.loc45.7 45.70000
## cch05.loc45.8 45.80000
## cch05.loc45.9 45.90000
## cch05.loc46   46.00000
## cch05.loc46.1 46.10000
## snp1930      46.12834
## snp1931      46.12834

```

Estimate QTL effects on chromosome 5 for the fresh_shoot_g

```

c2eff <- scan1coef(pr[, "ch05"], test$pheno[, "fresh_shoot_g"]) #getting coefficient
dim(c2eff)

```

```

## [1] 1278    3

```

```

head(c2eff)

```

```

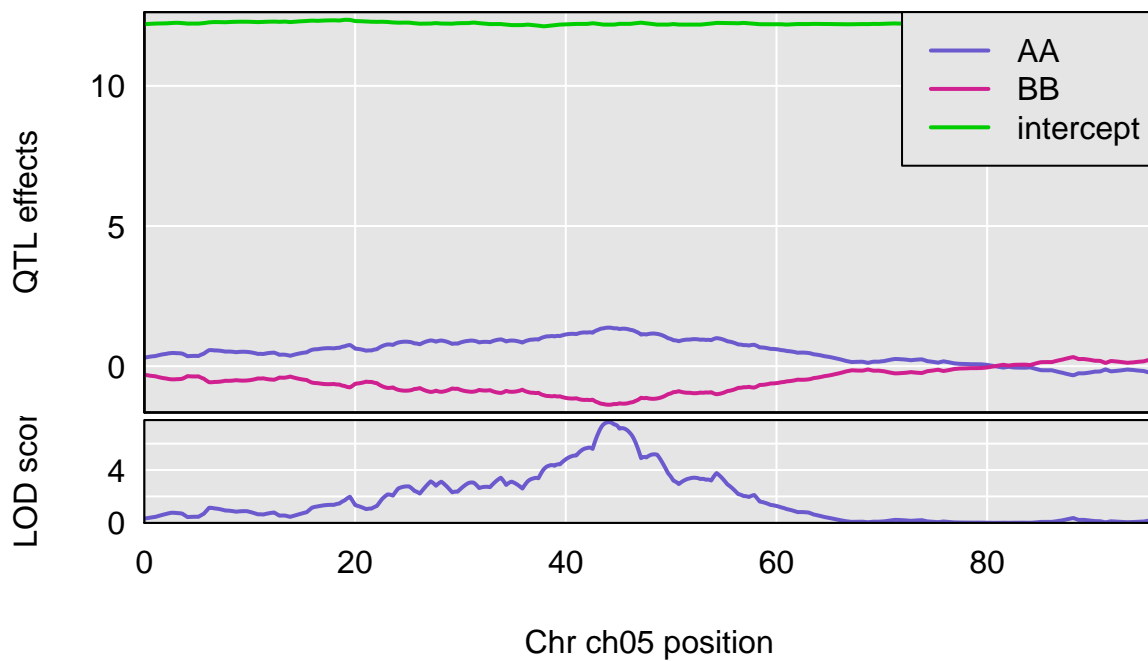
##                AA          BB intercept
## snp1757         0.3027348 -0.3027348  12.20301
## snp1758         0.3027348 -0.3027348  12.20301

```

```
## cch05.loc0.1 0.3104026 -0.3104026 12.20468
## cch05.loc0.2 0.3177067 -0.3177067 12.20637
## cch05.loc0.3 0.3246068 -0.3246068 12.20810
## cch05.loc0.4 0.3310653 -0.3310653 12.20985
```

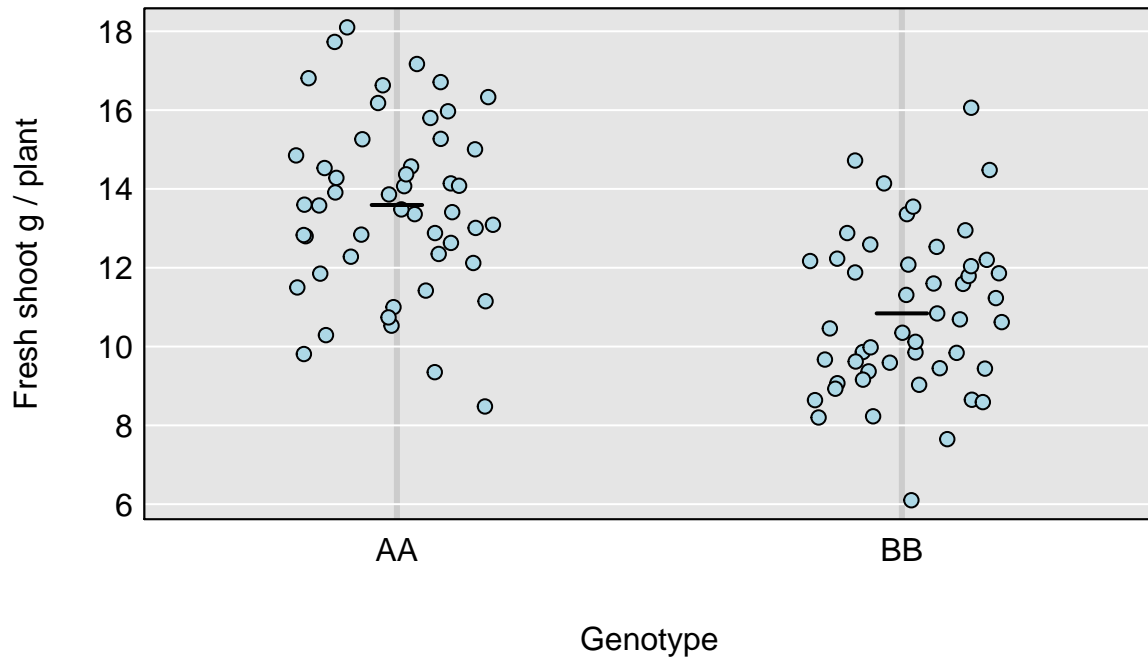
```
#plot_coef(c2eff, map, legend = "topright")
plot_coef(c2eff, map, scan1_output = out,
          main = "Chromosome 5 QTL effects and LOD scores",
          legend = "topright")
```

Chromosome 5 QTL effects and LOD scores



Plot the raw phenotypes against the genotypes at a single putative QTL position

```
g <- as.factor(maxmarg(pr, map, chr="ch05", pos=44.1000, return_char=TRUE))
plot_pxg(g, test$pheno[,"fresh_shoot_g"], ylab="Fresh shoot g / plant")
```



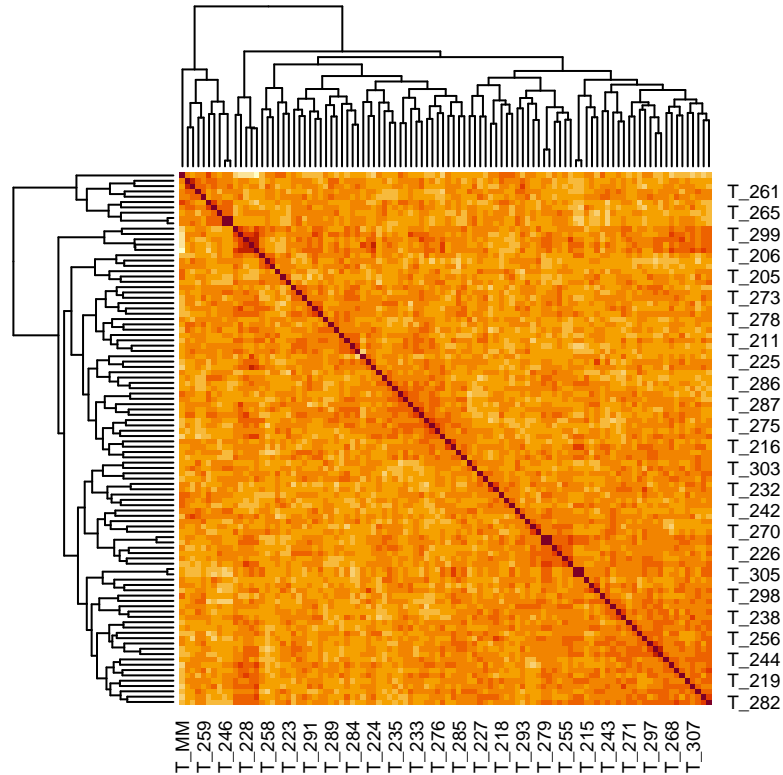
There are other methods that incorporate kinship and find LOD peak by either linear mixed model or leave one chromosome out (LOCO) method.

Calculate Kinship matrix and perform a genome scan using a linear mixed model

```
kinship <- calc_kinship(probs = pr)
kinship[1:5, 1:5]
```

```
##           T_MM      T_251      T_205      T_206      T_207
## T_MM  9.999268e-01  7.317771e-05  0.6111815  0.3825838  0.3486678
## T_251  7.317771e-05  9.999268e-01  0.3888185  0.6174162  0.6513322
## T_205  6.111815e-01  3.888185e-01  0.9964588  0.5299421  0.5616939
## T_206  3.825838e-01  6.174162e-01  0.5299421  0.9963969  0.6047783
## T_207  3.486678e-01  6.513322e-01  0.5616939  0.6047783  0.9970126
```

```
heatmap(kinship, symm = TRUE)
```



```
out_pg <- scan1(pr, test$pheno, kinship=kinship, cores=parallel::detectCores()-1)
```

Use the “leave one chromosome out” (LOCO) method

```
kinship_loco <- calc_kinship(pr, "loco")
out_pg_loco <- scan1(pr, test$pheno, kinship_loco, cores=parallel::detectCores()-1)
```

Compare LOD peaks from three different methods

```
plot_scan1(out_pg_loco, map = map, lodcolumn = "fresh_shoot_g", col = "black")
plot_scan1(out, map = map, lodcolumn = "fresh_shoot_g", col = "green", add = TRUE)
plot_scan1(out_pg, map = map, lodcolumn = "fresh_shoot_g", col = "blue", add = TRUE)
```

