

STAT646 - Homework 6

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1. Context = "SNPs" data from package SNPassoc. These are data on 157 case / control individuals, for which we know gender, arterial blood pressure, "protein levels?", and genotype for 35 SNPs.

- (a) Use the association function to test for association between SNP 1 and case / control status, adjusted for gender and blood pressure. Assume a dominant pattern of inheritance, so that both CC and CT are phenotype one and only TT is phenotype two. Match the results by using the glm function to carry out logistic regression. What are the probabilities of being a case for females with arterial blood pressure = 13, comparing phenotype one and two? That is, use your logistic regression model to compute two estimated probabilities, both for females with b.p. = 13, one for phenotype one and one for phenotype two.

```
> library(SNPassoc)
> data(SNPs)
> # Using the package function
> Y <- setupSNP(SNPs, colSNPs = 6:40, info = SNPs.info.pos, sep = "")
> outAssociation <- association(casco ~ sex + blood.pre + snp10001,
+                               data = Y,
+                               model = "dominant")
> outAssociation

SNP: snp10001  adjusted by: sex blood.pre
              0    % 1    %   OR lower upper p-value   AIC
Dominant
T/T          24 51.1 68 61.8 1.00                0.2286 196.1
C/T-C/C      23 48.9 42 38.2 0.65   0.32   1.31

> # Using the glm model
> Y <- data.frame("casco" = Y$casco, "sex" = Y$sex,
+                 "blood.pre" = Y$blood.pre,
+                 "snp10001" = ifelse(Y$snp10001 == "T/T", 0, 1))
> glmOut <- glm(casco ~ sex + blood.pre + snp10001,
+               family = binomial, data = Y)
> # Comparing the result
> cf <- summary(glmOut)$coefficients
> exp(cf[4, 1])

[1] 0.6504373

> exp(cf[4, 1] + c(-1, 1) * 1.96 * cf[4, 2])

[1] 0.3230523 1.3095981
```

```

> 2 * (1 - pnorm(abs(cf[4, 1] / cf[4, 2])))
[1] 0.2283576
> # Female case
> newData <- data.frame("sex" = factor(rep("Female", 2)),
+                       "blood.pre" = rep(13, 2),
+                       "snp10001" = c(1, 0))
> predict(glmOut, newdata = newData, type = "response")
           1           2
0.6245024 0.7188599

```

- (b) The `tableHWE` function uses an exact test to test whether Hardy Weinberg Equilibrium (HWE) holds with respect to each SNP. Manually implement the exact test and match the `tableHWE` p-value for SNP 1. The formula is given in Wigginton et al (2005). You don't need to use the recursion formulas given in equation (2). Note that you will need to compute lots of factorials to get the p-value. Factorials of big numbers can be problematic due to instability of the resulting numbers. I recommend that you use the `lfactorial` function in R to compute factorials on the log scale.

```

> Y <- setupSNP(SNPs, colSNPs = 6:40, info = SNPs.info.pos, sep = "")
> head(tableHWE(Y))
           HWE (p value)
snp10001  0.281639248
snp10002  0.004944837
snp10003           NA
snp10004           NA
snp10005  0.008019904
snp10006           NA

> testHWE <- function(snp) {
+   N <- length(snp)
+   snp <- as.character(snp)
+   snp_tbl <- table(snp)
+   if(length(snp_tbl) == 3) {
+     if(all(!is.na(snp))) {
+       al <- unlist(strsplit(snp, split = "/"))
+       al_uq <- unique(al)
+       al_tbl <- table(al)
+       oo <- order(al_tbl)
+       A <- names(al_tbl[oo[1]])
+       B <- names(al_tbl[oo[2]])
+       n_A <- al_tbl[oo[1]]
+       n_B <- al_tbl[oo[2]]
+       n_AB <- table(snp)[2]
+       n_AA <- (n_A - n_AB) / 2
+       n_BB <- (n_B - n_AB) / 2
+       even <- floor(n_A / 2) == ceiling(n_A / 2)
+       if(even) {
+         n_AB_seq <- seq(from = 0, to = n_A, by = 2)
+       } else {
+         n_AB_seq <- seq(from = 1, to = n_A, by = 2)
+       }
+     }
+   }
+ }

```

```

+     }
+     k <- length(n_AB_seq)
+     PP <- rep(NA, k)
+     for(i in 1:k) {
+       n_ab <- n_AB_seq[i]
+       n_aa <- (n_A - n_ab) / 2
+       n_bb <- (n_B - n_ab) / 2
+       PP[i] <- exp(n_ab * log(2) + lfactorial(N) +
+                   lfactorial(n_A) + lfactorial(n_B) -
+                   lfactorial(n_aa) - lfactorial(n_ab) -
+                   lfactorial(n_bb) - lfactorial(2 * N))
+     }
+     p_obs <- PP[match(n_AB, n_AB_seq)]
+     p_val <- sum(PP[PP <= p_obs])
+     return(p_val)
+   } else {
+     return(NA) }
+ } else {
+   return(NA)
+ } }
> testHWE(Y$snp10001)
[1] 0.2816392

```

- (c) In the notation of Wigginton et al (2005), n_A and n_B are the allele frequencies for the rarer and more common alleles, respectively; $n_A + n_B = 2N = 2 \times 157$. Also, n_{AA} is the frequency of individuals homozygous for the rarer allele, n_{AB} is the frequency of heterozygous individuals, and n_{BB} is the frequency of individuals homozygous for the more common allele; $n_{AA} + n_{AB} + n_{BB} = N = 157$. What are the allele frequencies n_A and n_B and the genotype frequencies n_{AA} , n_{AB} , and n_{BB} ? Considering n_A and n_B as fixed, and assuming that HWE holds, how many of each genotype would you expect out of our 157 individuals?

```

> snp <- Y$snp10001
> N <- length(snp)
> snp <- as.character(snp)
> snp_tbl <- table(snp)
> al <- unlist(strsplit(snp, split = "/"))
> al_uq <- unique(al)
> al_tbl <- table(al)
> oo <- order(al_tbl)
> A <- names(al_tbl[oo[1]])
> B <- names(al_tbl[oo[2]])
> # Observed alleles
> print(n_A <- al_tbl[oo[1]])
C
77
> print(n_B <- al_tbl[oo[2]])
T
237
> # Observed genotypes
> print(n_AB <- table(snp)[2])

```

```
C/T
53
> print(n_AA <- (n_A - n_AB) / 2)
C
12
> print(n_BB <- (n_B - n_AB) / 2)
T
92
> p_A <- as.numeric(n_A / (2 * N))
> p_B <- as.numeric(1 - p_A)
> # Expected genotypes under HWE
> n_AA_hwe <- ceiling(N * p_A ^ 2)
> names(n_AA_hwe) <- "C/C"
> n_AA_hwe
C/C
10
> n_AB_hwe <- ceiling(2 * N * p_A * p_B)
> names(n_AB_hwe) <- "C/T"
> n_AB_hwe
C/T
59
> n_BB_hwe <- ceiling(N * p_B ^ 2)
> names(n_BB_hwe) <- "T/T"
> n_BB_hwe
T/T
90
```