

# Bayesian Hypothesis Testing

## BIOS719 Generalized Linear Models

### Testing hypothesis

An investigator arrives at a village and after observing the villagers' contact with water, she is 80% sure that infection will be endemic in this village. The investigator takes stool samples from 10 villagers and 7 are positive for the parasite.

H0: Infection is not endemic ( $\theta \leq 0.5$ )

H1: Infection is endemic ( $\theta > 0.5$ )

Table 12.1 Calculations for updating the investigator's priors for  $H_0$  and  $H_1$  to posteriors using the observed data on positive samples for *Schistosoma japonicum*.

$\theta$	Hypothesis	$P(\theta)$ Prior	$P(y \theta)$ Likelihood	$P(y \theta) \times P(\theta)$ Likelihood $\times$ Prior	$P(\theta y)$ Posterior
0.0	$H_0$	0.0333	0.0000	0.0000	0.0000
0.1	$H_0$	0.0333	0.0000	0.0000	0.0000
0.2	$H_0$	0.0333	0.0008	0.0000	0.0002
0.3	$H_0$	0.0333	0.0090	0.0003	0.0024
0.4	$H_0$	0.0333	0.0425	0.0014	0.0114
0.5	$H_0$	0.0333	0.1172	0.0039	0.0315
Sum		0.2000			0.0455
0.6	$H_1$	0.1600	0.2150	0.0344	0.2771
0.7	$H_1$	0.1600	0.2668	0.0427	0.3439
0.8	$H_1$	0.1600	0.2013	0.0322	0.2595
0.9	$H_1$	0.1600	0.0574	0.0092	0.0740
1.0	$H_1$	0.1600	0.0000	0.0000	0.0000
Sum		0.8000		0.1241	0.9545

```
f.table12.1 <- function(theta.cut, prior.H1, bysize, tab=T) {
  theta <- seq(0, 1, by=bysize)
  H1.theta <- ifelse(theta > theta.cut, 1, 0) ## H1

  theta.prior <- ifelse(H1.theta==1,
                        prior.H1/length(H1.theta[H1.theta==1]),
                        (1-prior.H1)/length(H1.theta[H1.theta==0]))
  lik <- dbinom(x=7, size=10, prob=theta)
  lik.theta.prior <- lik*theta.prior
  theta.post <- lik.theta.prior/sum(lik.theta.prior)

  out <- cbind(theta=theta, H1=H1.theta, theta.prior=theta.prior, lik=lik,
               lik.theta.prior=lik.theta.prior, theta.post=theta.post)

  if(tab==T) {
    return(list(tab12.1=round(out, 4),
                H.prior=tapply(theta.prior, H1.theta, sum),
                H.post=tapply(theta.post, H1.theta, sum)))
  } else {
    return(list(H.prior=tapply(theta.prior, H1.theta, sum),
                H.post=tapply(theta.post, H1.theta, sum),
                Equal.tail.interval=
                  c( max(theta[cumsum(theta.post)<=0.05/2]), min(theta[cumsum(theta.post)>1-0.05/2])),

```

```

        theta=theta,
        theta.post=theta.post))
    }
}

test1 <- f.table12.1(theta.cut=0.5, prior.H1=0.8, bysize=0.1, tab=T)
test1

## $tab12.1
##      theta H1 theta.prior    lik lik.theta.prior theta.post
## [1,]  0.0  0      0.0333 0.0000      0.0000      0.0000
## [2,]  0.1  0      0.0333 0.0000      0.0000      0.0000
## [3,]  0.2  0      0.0333 0.0008      0.0000      0.0002
## [4,]  0.3  0      0.0333 0.0090      0.0003      0.0024
## [5,]  0.4  0      0.0333 0.0425      0.0014      0.0114
## [6,]  0.5  0      0.0333 0.1172      0.0039      0.0315
## [7,]  0.6  1      0.1600 0.2150      0.0344      0.2771
## [8,]  0.7  1      0.1600 0.2668      0.0427      0.3439
## [9,]  0.8  1      0.1600 0.2013      0.0322      0.2595
## [10,] 0.9  1      0.1600 0.0574      0.0092      0.0740
## [11,] 1.0  1      0.1600 0.0000      0.0000      0.0000
##
## $H.prior
##    0    1
## 0.2 0.8
##
## $H.post
##          0          1
## 0.04550201 0.95449799

test1$H.prior

##    0    1
## 0.2 0.8

test1$H.post

##          0          1
## 0.04550201 0.95449799

BF1 <- (test1$H.post[2]/test1$H.post[1])/(test1$H.prior[2]/test1$H.prior[1])
BF1

##          1
## 5.244263

test2 <- f.table12.1(theta.cut=0.5, prior.H1=0.5, bysize=0.0001, tab=F)
test2$H.prior

##    0    1
## 0.5 0.5

test2$H.post

##          0          1
## 0.1133256 0.8866744

```

```
BF2 <- (test2$H.post[2]/test2$H.post[1])/(test2$H.prior[2]/test2$H.prior[1])
BF2
```

```
##          1
## 7.82413
```

```
#### Frequentist test
```

```
prop.test(x=7, n=10, p=0.5, alternative=c("greater"), correct=FALSE)
```

```
##
## 1-sample proportions test without continuity correction
##
## data: 7 out of 10, null probability 0.5
## X-squared = 1.6, df = 1, p-value = 0.103
## alternative hypothesis: true p is greater than 0.5
## 95 percent confidence interval:
## 0.4416998 1.0000000
## sample estimates:
## p
## 0.7
```