

Supplementary Text 2. WinBUGS models.

The following is the WinBUGS codes for Bayesian latent-class model with conditional dependence between results of IgM IFA and IgM ELISA. The same model structure is applied to each and every dataset created

```

model{

for (i in 1:135){
    status[i]~dbern(prev)

##### Likelihood #####
for (j in 1:5){
    y[i,j]~dbern(p[i,j])
    ypred[i,j]~dbern(p[i,j])
}
}

for (i in 1:135){
for (j in 1:2){
    logit(p[i,j])<-status[i]*alpha[j]+(1-status[i])*beta[j]
}
for (j in 3:4){
    logit(p[i,j])<-status[i]*alpha[j]+(1-status[i])*beta[j]+status[i]*re[i]*IgM
    logit(pstatus[i,j])<-status[i]*alpha[j]+status[i]*re[i]*IgM+(1-status[i])*(-1000)
}
    logit(p[i,5])<-status[i]*alpha[5]+(1-status[i])*beta[5]
    re[i]~dnorm(0,1)
}

#####
Prior #####
prev~dbeta(0.5,0.5)
IgM~dnorm(0.0,0.1)I(0,6)
alpha[1]~dnorm(0.0,0.1)
logit(s[1])<-alpha[1]
beta[1]<--1000
for (j in 2:5){
    alpha[j]~dnorm(0.0,0.1)
    beta[j]~dnorm(0.0,0.1)I(,1)
    logit(s[j])<-alpha[j]
    logit(x[j])<--beta[j]
}
se[1]<-s[1]
se[2]<-s[2]
se[5]<-s[5]
for (j in 3:4){
    se[j]<-sum(pstatus[,j])/sum(status[])
}

#### Unbiased estimation of sensitivity, specificity and accuracy #####
for (i in 1:135){
for (j in 1:5){
    diseasetestpos[i,j]<-status[i]*y[i,j]
    nondiseasetestneg[i,j]<-(1-status[i])*(1-y[i,j])
}
}
for (j in 1:5){
    sen[j]<-sum(diseasetestpos[,j])/sum(status[])
    spec[j]<-sum(nondiseasetestneg[,j])/(135-sum(status[]))
}

for (j in 1:5){
accuracy[j]<-(sum(diseasetestpos[,j])+ sum(nondiseasetestneg[,j]))/135
}

#####
Prediction #####
for (i in 1:135){
for (k in 1:32){
for (j in 1:5){
    arraymatched[i,k,j]<-equals(ypred[i,j],pattern[k,j])
}
nmatched[i,k]<-sum(arraymatched[i,k,])
}
}

```

```

matchedpattern[i,k]<-equals(nmatched[i,k],5)
}
}
for (k in 1:32){
  freqpred[k]<-sum(matchedpattern[,k])
}

##### Bayesian p value #####
for (i in 1:135){
for (k in 1:32){
for (j in 1:5){
  arraymatchedobs[i,k,j]<-equals(y[i,j],pattern[k,j])
}
nmatchedobs[i,k]<-sum(arraymatchedobs[i,k,])
matchedpatternobs[i,k]<-equals(nmatchedobs[i,k],5)
}
}
for (k in 1:32){
  freqobs[k]<-sum(matchedpatternobs[,k])
}
for (k in 1:32){
pvalue[k]<-step(freqpred[k]-freqobs[k])
}
pvalue34<-step(sen[3]-sen[4]-0.005)+0.5*equals(sen[4],sen[3])
}

```