

R script code for the performed latent class analysis:

```
5P3T_dep_default_with_priors.
```

```
Model
```

```
{
```

```
# multinomial models
```

```
for (k in 1:K){
```

```
  y[k,1:2,1:2,1:2] ~ dmulti(p[k,1:2,1:2,1:2], n[k])
```

```
}
```

```
# cell probabilities expressed in terms of se, sp and pi
```

```
for (k in 1:K){
```

```
  p[k,1,1,1] <- pi[k]*se[1]*(se[2]*se[3]+cov23[1]) + (1-pi[k])*(1-sp[1])*((1-sp[2])*(1-sp[3])+cov23[2])
```

```
  p[k,1,1,2] <- pi[k]*se[1]*(se[2]*(1-se[3])-cov23[1]) + (1-pi[k])*(1-sp[1])*((1-sp[2])*sp[3]-cov23[2])
```

```
  p[k,1,2,1] <- pi[k]*se[1]*((1-se[2])*se[3]-cov23[1]) + (1-pi[k])*(1-sp[1])*(sp[2]*(1-sp[3])-cov23[2])
```

```
  p[k,1,2,2] <- pi[k]*se[1]*((1-se[2])*(1-se[3])+cov23[1]) + (1-pi[k])*(1-sp[1])*(sp[2]*sp[3]+cov23[2])
```

```
  p[k,2,1,1] <- pi[k]*(1-se[1])*(se[2]*se[3]+cov23[1]) + (1-pi[k])*sp[1]*((1-sp[2])*(1-sp[3])+cov23[2])
```

```
  p[k,2,1,2] <- pi[k]*(1-se[1])*(se[2]*(1-se[3])-cov23[1]) + (1-pi[k])*sp[1]*((1-sp[2])*sp[3]-cov23[2])
```

```
  p[k,2,2,1] <- pi[k]*(1-se[1])*((1-se[2])*se[3]-cov23[1]) + (1-pi[k])*sp[1]*(sp[2]*(1-sp[3])-cov23[2])
```

```
  p[k,2,2,2] <- pi[k]*(1-se[1])*((1-se[2])*(1-se[3])+cov23[1]) + (1-pi[k])*sp[1]*(sp[2]*sp[3]+cov23[2])
```

```
}
```

```
# prior distributions
```

```
se[1] ~ dbeta(1,1)
```

```
sp[1] ~ dbeta(1,1)
```

```
se[2] ~ dbeta(1,1)
```

```
sp[2] ~ dbeta(1,1)
```

```
se[3] ~ dbeta(30.665,9.367)
```

```
sp[3] ~ dbeta(100,1.497)
```

```
for (k in 1:K){
```

```
  pi[k] ~ dbeta(1,1)
```

```
}
```

```
# Covariance note: cov12[1] ~ sensitivity, cov12[2] ~ specificity
```

```
cov23_low[1] <- max(-(1-se[2])*(1-se[3]),-se[2]*se[3])
```

```
cov23_upp[1] <- min(se[2]*(1-se[3]),(1-se[2])*se[3])
```

```
cov23_low[2] <- max(-(1-sp[2])*(1-sp[3]),-sp[2]*sp[3])
```

```
cov23_upp[2] <- min(sp[2]*(1-sp[3]),(1-sp[2])*sp[3])
```

```
cov23[1] ~ dunif(cov23_low[1],cov23_upp[1])
```

```
cov23[2] ~ dunif(cov23_low[2],cov23_upp[2])
```

```
#Step function
```

```
sediff[1] <- se[1]- se[2]
```

```
sediff[2] <- se[1]- se[3]
```

```
sediff[3] <- se[2]- se[1]
```

```
sediff[4] <- se[2]- se[3]
```

```
sediff[5] <- se[3]- se[1]
```

```
sediff[6] <- se[3]- se[2]
```

```
spdiff[1] <- sp[1]- sp[2]
```

```
spdiff[2] <- sp[1]- sp[3]
```

```
spdiff[3] <- sp[2]- sp[1]
```

```
spdiff[4] <- sp[2]- sp[3]
```

```
spdiff[5] <- sp[3]- sp[1]
```

```
spdiff[6] <- sp[3]- sp[2]
```

```
Prsediff[1] <- step(sediff[1])
```

```
Prsediff[2] <- step(sediff[2])
```

```
Prsediff[3] <- step(sediff[3])
```

```
Prsediff[4] <- step(sediff[4])
```

```
Prsediff[5] <- step(sediff[5])
```

```
Prsediff[6] <- step(sediff[6])
```

```

Prspdiff[1] <- step(spdiff[1])
Prspdiff[2] <- step(spdiff[2])
Prspdiff[3] <- step(spdiff[3])
Prspdiff[4] <- step(spdiff[4])
Prspdiff[5] <- step(spdiff[5])
Prspdiff[6] <- step(spdiff[6])

# 3-test model: Bayesian p-value for Pearson X2
for (k in 1:K){
  s[k,1:2,1:2,1:2] ~ dmulti(p[k,1:2,1:2,1:2],n[k])
  for (i1 in 1:2){
    for (i2 in 1:2){
      for (i3 in 1:2){
        x2[k,i1,i2,i3] <- pow(y[k,i1,i2,i3]-p[k,i1,i2,i3]*n[k],2)/(p[k,i1,i2,i3]*n[k])
        ref[k,i1,i2,i3] <- pow(s[k,i1,i2,i3]-p[k,i1,i2,i3]*n[k],2)/(p[k,i1,i2,i3]*n[k])
      }
    }
  }
  x2total <- sum(x2[,,,])
  reftotal <- sum(ref[,,,])
  bayesp <- step(reftotal-x2total)

}

# note: required order of cell counts:
# (test1=+,test2=+,test3=+), (+,+,-), (+,-,+), (+,-,-), (-,+,+), (-,+,-), (-,-,+), (-,-,-)
# for popul 1,2,...

# note: first number in c() list for .Dim is the number of populations (K)
list(K=5, n=c(93,178,180,92,51),
y=structure(.Data=c(0,0,0,0,0,5,0,88,2,0,0,1,0,1,0,173,0,0,0,4,3,10,1,162,1,1,0,7,7,13,0,63,0,0,0,0,0,0,0,
51),.Dim=c(5,2,2,2)))

```