

```

# small area Poisson model
date()
set.seed(1980)
path<-getwd()
library("coda")
library("splines")
library("rbugs")
library("rjags")
library("R2WinBUGS")
library("dclone")
library("cluster")
library("nlme")
library("MASS")
R=1
      s1=0
      s=s1+1

K=1
I=50
T=5
Nit=3
rho=0.4

Nsp=min(40,floor(I*T/4))

alpha <-0.1
beta  <-0.01          # X coefficient

sigma2.e=1            #Error
tau.e=1/(sigma2.e)

sigma2.nu=1           #area
tau.nu=1/(sigma2.nu)

sigma2.sp=1           #area
tau.sp=1/(sigma2.sp)

# generate fixed parts
mean.x=0
x <- rnorm(I*T, mean.x, 1)

sp<-rnorm(Nsp,0,sqrt(sigma2.sp))

degree=1
nbetas <- degree+1

x.quan<-as.integer((I*T)/(Nsp+1)*c(1:Nsp))
x.sort<-sort(x)
knot<-x.sort[x.quan]

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pow <- function(a,b) a^b
step <- function(a) (a>0)*1

X <- matrix(NA, I*T, degree+1)
Z <- matrix(NA, I*T, Nsp)
u <- matrix(NA, I*T, Nsp)
MUZ <- matrix(NA, I*T, Nsp)
MUX<-matrix(NA, I*T, nbetas)

for (i in 1:(I*T))
{
  for (l in 1:(degree+1)){
    X[i,l]<-pow(x[i],l-1)
  }
}
for (i in 1:(I*T))
{
  for (k in 1:Nsp)
  {
    u[i,k]<-(x[i]-knot[k])*step(x[i]-knot[k])
    Z[i,k]<-pow(u[i,k],degree)
  }
}
for (i in 1:(I*T))
{
  #MUZ[i,]<-colMeans(Z[(1+(i-1)*m):(i*m),])
  MUX[i,]<-c(1,mean.x)
}
MUZ=Z

parameters.t.hb <- c("beta", "mu", "rho", "sigma2.e", "sigma2.nu",
"sigma2.sp")
parameters.t.dc <- c("beta", "rho", "sigma2.e", "sigma2.nu",
"sigma2.sp")
parameters.t.p <- c("mu")

model.t.hb <- function() {
  for(k in 1:K){
    for(i in 1:I){
      for(t in 1:T){
        y[(i+I*(k-1)),t] ~ dpois(Nit*mu[(i+I*(k-
1)),t])
        log[mu[(i+I*(k-1)),t]] <- mmu[(i+I*(k-1)),t]
        mmu[(i+I*(k-1)),t]<-inprod(X[(i+(t-
1)*I),],beta[1,])+inprod(Z[(i+(t-1)*I),],sp[,k])+nu[i,k]
      }
    }
  }

  for(l in 1:nbetas){beta[1,l]~dnorm(0.0,1.0E-6)}
}

```

```

rho ~ dunif(-1,1)
tau.e ~ dgamma(1.0E-3,1.0E-3)
tau.nu~ dgamma(1.0E-3,1.0E-3)
tau.sp~ dgamma(1.0E-3,1.0E-3)

sigma2.e <-1/tau.e
sigma2.nu<-1/tau.nu
sigma2.sp<-1/tau.sp

for(k in 1:K){
  for(i in 1:I){
    nu[i,k]~dnorm(0,tau.nu)
  }
}

for(k in 1:K){
  for(i in 1:Nsp){
    sp[i,k] ~ dnorm(0,tau.sp)
  }
}

for(i in 1:T){tauE[i,i]<-1}
for(i in 1:T-1){tauE[i,i+1]<-0
  tauE[i+1,i]<-0}
for(i in 1:T-2){tauE[i,i+2]<-0
  tauE[i+2,i]<-0}
for(i in 1:T-3){tauE[i,i+3]<-0
  tauE[i+3,i]<-0}
tauE[1,T]<-0
tauE[T,1]<-0

for(i in 1:T){gam[i,i]<-(1/(1-(rho*rho)))}
for(i in 1:T-1){gam[i,i+1]<-rho/(1-rho*rho)
  gam[i+1,i]<-rho/(1-rho*rho)}
for(i in 1:T-2){gam[i,i+2]<-pow(rho,2)/(1-rho*rho)
  gam[i+2,i]<-pow(rho,2)/(1-rho*rho)}
for(i in 1:T-3){gam[i,i+3]<-pow(rho,3)/(1-rho*rho)
  gam[i+3,i]<-pow(rho,3)/(1-rho*rho)}
gam[1,T]<-pow(rho,4)/(1-rho*rho)
gam[T,1]<-pow(rho,4)/(1-rho*rho)

sigmaG<-sigma2.e*gam
invG<-inverse(sigmaG)

for(i in 1:T){tauG[i,i]<-invG[i,i]}
for(i in 1:T-1){tauG[i,i+1]<-(invG[i,i+1]+invG[i+1,i])/2
  tauG[i+1,i]<-(invG[i,i+1]+invG[i+1,i])/2}
for(i in 1:T-2){tauG[i,i+2]<-(invG[i,i+2]+invG[i+2,i])/2
  tauG[i+2,i]<-(invG[i,i+2]+invG[i+2,i])/2}
for(i in 1:T-3){tauG[i,i+3]<-(invG[i,i+3]+invG[i+3,i])/2
  tauG[i+3,i]<-(invG[i,i+3]+invG[i+3,i])/2}
tauG[1,T]<-(invG[1,5]+invG[5,1])/2
tauG[T,1]<-(invG[1,5]+invG[5,1])/2
}

```

```

model.t.p <- function() {
  for(k in 1:K){
    for(i in 1:I){
      for(t in 1:T){
        y[(i+I*(k-1)),t] ~ dpois(Nit*mu[(i+I*(k-1)),t])
        log[mu[(i+I*(k-1)),t]] <- mmu[(i+I*(k-1)),t]
        mmu[(i+I*(k-1)),t]<-inprod(X[(i+(t-
1)*I),],beta[1,])+inprod(Z[(i+(t-1)*I),],sp[,k])+nu[i,k]
      }
    }
  }

  tmp[1:(nbetas+4)] ~ dmnorm(param[], prec[,])
  for(l in 1:nbetas){beta[1,l] <- tmp[l]}
  rho <- (1-step(tmp[(nbetas+1)]+1))*(-
0.9999)+step(tmp[(nbetas+1)]-1)*(0.9999)+(1-step(abs(tmp[(nbetas+1)])-
1))*(tmp[(nbetas+1)])
  sigma2.e <- abs(tmp[(nbetas+2)])
  sigma2.nu <- abs(tmp[(nbetas+3)])
  sigma2.sp <- abs(tmp[(nbetas+4)])

  tau.e <- 1/sigma2.e
  tau.nu <- 1/sigma2.nu
  tau.sp <- 1/sigma2.sp

  for(k in 1:K){
    for(i in 1:I){
      nu[i,k] ~ dnorm(0,tau.nu)
    }
  }

  for(k in 1:K){
    for(i in 1:Nsp){
      sp[i,k] ~ dnorm(0,tau.sp)
    }
  }

  for(i in 1:T){tauE[i,i]<-1}
  for(i in 1:T-1){tauE[i,i+1]<-0
    tauE[i+1,i]<-0}
  for(i in 1:T-2){tauE[i,i+2]<-0
    tauE[i+2,i]<-0}
  for(i in 1:T-3){tauE[i,i+3]<-0
    tauE[i+3,i]<-0}
  tauE[1,T]<-0
  tauE[T,1]<-0

  for(i in 1:T){gam[i,i]<-(1/(1-(rho*rho)))}
  for(i in 1:T-1){gam[i,i+1]<-rho/(1-rho*rho)
    gam[i+1,i]<-rho/(1-rho*rho)}
  for(i in 1:T-2){gam[i,i+2]<-pow(rho,2)/(1-rho*rho)

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        gam[i+2,i]<-pow(rho,2)/(1-rho*rho)}
for(i in 1:T-3){gam[i,i+3]<-pow(rho,3)/(1-rho*rho)
        gam[i+3,i]<-pow(rho,3)/(1-rho*rho)}
gam[1,T]<-pow(rho,4)/(1-rho*rho)
gam[T,1]<-pow(rho,4)/(1-rho*rho)

sigmaG<-sigma2.e*gam
invG<-inverse(sigmaG)

for(i in 1:T){tauG[i,i]<-invG[i,i]}
for(i in 1:T-1){tauG[i,i+1]<-(invG[i,i+1]+invG[i+1,i])/2
        tauG[i+1,i]<-(invG[i,i+1]+invG[i+1,i])/2}
for(i in 1:T-2){tauG[i,i+2]<-(invG[i,i+2]+invG[i+2,i])/2
        tauG[i+2,i]<-(invG[i,i+2]+invG[i+2,i])/2}
for(i in 1:T-3){tauG[i,i+3]<-(invG[i,i+3]+invG[i+3,i])/2
        tauG[i+3,i]<-(invG[i,i+3]+invG[i+3,i])/2}
tauG[1,T]<-(invG[1,5]+invG[5,1])/2
tauG[T,1]<-(invG[1,5]+invG[5,1])/2
}

stats.t.hb <-list()
stats.t.dc<-list()
stats.t.p<-list()

modelout.sim<-modelout.mcmc<-list()

mean.t.hb<-MSE.t.hb<-mse.t.hb<-Bias.t.hb<-sd.t.hb<-rep(0,nbetas+I*T+4)
mean.t.dc<-MSE.t.dc<-mse.t.dc<-Bias.t.dc<-sd.t.dc<-rep(0,nbetas+4)
mean.t.p<-MSE.t.p<-mse.t.p<-Bias.t.p<-sd.t.p<-rep(0,I*T)
covp99.t.hb<-covp98.t.hb<-covp95.t.hb<-covp90.t.hb<-
rep(0,nbetas+I*T+4)
covp99.t.dc<-covp98.t.dc<-covp95.t.dc<-covp90.t.dc<-rep(0,nbetas+4)
covp99.t.p<-covp98.t.p<-covp95.t.p<-covp90.t.p<-rep(0,I*T)
domain99.t.hb<-domain98.t.hb<-domain95.t.hb<-domain90.t.hb<-
rep(0,nbetas+I*T+4)
domain99.t.dc<-domain98.t.dc<-domain95.t.dc<-domain90.t.dc<-
rep(0,nbetas+4)
domain99.t.p<-domain98.t.p<-domain95.t.p<-domain90.t.p<-rep(0,I*T)

paramvalue.t.hb<-rep(0,nbetas+I*T+4)
paramvalue.t.dc<-rep(0,nbetas+4)
paramvalue.t.p<-rep(0,I*T)

outbound.hb<-rep(0,R)
outbound.dc<-rep(0,R)

```

```

for(n in 1:R){
  ar<-matrix(0,I,T)
  for(j in 1:I){
    ar[j,]<-
arima.sim(model=list(ar=c(rho)),n=T,sd=sqrt(sigma2.e))
  }
  ar2<-c(ar)
  area<-rnorm(I,0,sqrt(sigma2.nu))
  logtp<-matrix(alpha+beta*X[, -
1]+Z%*%sp+rep(area,T)+ar2,I,T,byrow=F)
  MU<-exp(logtp)

  y<-matrix(0,I,T)

  for(i in 1:I)
    for(j in 1:T){
      {y[i,j]<-rpois(1,Nit*exp(logtp[i,j])
      }}

  if(n>s1){
    # -----
HB for true model
    data.sim<-list(y=y, X=X, Z=Z, I=I, T=T,Nit=Nit,
Nsp=Nsp, nbetas=nbetas, K=1)
    print(paste("----- HB model",n,"-----
-"))
    modelout.sim <- jags.fit(data=data.sim,
parameters.t.hb, model.t.hb, n.chains=2, thin = 5, n.iter=10000,
n.adapt=10000)
    update(updated.model(modelout.sim),10000)
    modelout.mcmc<-
coda.samples(updated.model(modelout.sim), parameters.t.hb,
n.iter=10000, thin=5)
    num<-0
    while( (max ((gelman.diag(modelout.mcmc))$psrf[,1])
>1.05) & num<20){
      update(updated.model(modelout.sim),10000)
      modelout.mcmc<-
coda.samples(updated.model(modelout.sim), parameters.t.hb,
n.iter=10000, thin=5)
      num<-num+1
    }
    print(paste("model",n))
    print(paste("num",num))
    print(paste("Gelman"))
    print(max((gelman.diag(modelout.mcmc))$psrf[,1]))
    stats.t.hb<-
summary(modelout.mcmc,quantiles=c(0.005,0.01,0.025,0.05,0.95,0.975,0.9
9,0.995))
    print(stats.t.hb)

```

```

print(paste("Lambda for true model",n))
lambda.t.hb<-as.numeric(lambdamax.diag(modelout.mcmc))
print(paste("lambda.t.hb",lambda.t.hb))

paramvalue.t.hb<-c(alpha, beta, rho, MU, sigma2.e,
sigma2.nu, sigma2.sp)
if(max(abs((stats.t.hb[[1]][,1])-c(alpha,beta, rho,
sigma2.e, sigma2.nu, sigma2.sp)))>4){
  outbound.hb[n]<-1
}
print(paste("RB, MSE and coverage for true model",n))
mean.t.hb<-mean.t.hb+stats.t.hb[[1]][,1]
covp99.t.hb<-
covp99.t.hb+1*((stats.t.hb[[2]][,1]<paramvalue.t.hb &
paramvalue.t.hb<stats.t.hb[[2]][,8] ))
covp98.t.hb<-
covp98.t.hb+1*((stats.t.hb[[2]][,2]<paramvalue.t.hb &
paramvalue.t.hb<stats.t.hb[[2]][,7] ))
covp95.t.hb<-
covp95.t.hb+1*((stats.t.hb[[2]][,3]<paramvalue.t.hb &
paramvalue.t.hb<stats.t.hb[[2]][,6] ))
covp90.t.hb<-
covp90.t.hb+1*((stats.t.hb[[2]][,4]<paramvalue.t.hb &
paramvalue.t.hb<stats.t.hb[[2]][,5] ))
domain99.t.hb<-domain99.t.hb+((stats.t.hb[[2]][,8]-
stats.t.hb[[2]][,1] ))
domain98.t.hb<-domain98.t.hb+((stats.t.hb[[2]][,7]-
stats.t.hb[[2]][,2] ))
domain95.t.hb<-domain95.t.hb+((stats.t.hb[[2]][,6]-
stats.t.hb[[2]][,3] ))
domain90.t.hb<-domain90.t.hb+((stats.t.hb[[2]][,5]-
stats.t.hb[[2]][,4] ))

MSE.t.hb<-MSE.t.hb+((stats.t.hb[[1]][,1]-
paramvalue.t.hb)^2)
Bias.t.hb<-Bias.t.hb+sqrt((stats.t.hb[[1]][,1]-
paramvalue.t.hb)^2)
mse.t.hb<-mse.t.hb+((stats.t.hb[[1]][,2])^2)
sd.t.hb<-sd.t.hb+((stats.t.hb[[1]][,2]))
#----- DC for true model

K1<-0 #initial number of clones
r2<-1
mse<-1
lambda.ratio<-1
r.hat=2
num<-0
while((((r2>0.01 & mse>0.01) | lambda.ratio>=0.05 |
r.hat>1.05) & num<5) ){
  K1<-K1+80
  #cl<-makeCluster(2,type="SOCK")

```

```

        dcmo<-dc.fit(list(y=dclone(y,K1), X=X, Z=Z,
I=I, T=T, Nsp=Nsp,Nit=Nit,nbetas=nbetas, K=1), parameters.t.dc,
model.t.hb,
n.clones=c(K1),multiply=c("K"),unchanged=c("I","T","nbetas","X","Z",
"Nsp", "Nit"), n.chains=2, n.adapt=10000, n.update=10000,n.iter=10000,
thin=10)

        dcd<-dcdiag(dcmo)
        r2<-dcd[1,4]
        r.hat<-dcd[1,5]
        mse<-dcd[1,3]
        lambda.ratio<-dcd[1,2]/lambda.t.hb
        #stopCluster(cl)
        num<-num+1
        print(dcd)
    }
    paramvalue.t.dc<-c(alpha,beta, rho,
sigma2.e,sigma2.nu, sigma2.sp)
    print(paste("num",num))
    stats.t.dc<-
summary(dcmo,quantiles=c(0.005,0.01,0.025,0.05,0.95,0.975,0.99,0.995)
)
    print(stats.t.dc)

    if(max(abs((stats.t.dc[[1]][,1])-c(alpha,beta, rho,
sigma2.e,sigma2.nu, sigma2.sp)))>4){
        outbound.dc[n]<-1
    }
    print(paste("outbound for true DC",n, "is",
outbound.dc[n]))

    mean.t.dc<-mean.t.dc+stats.t.dc[[1]][,1]
    MSE.t.dc<-MSE.t.dc+((stats.t.dc[[1]][,1]-
paramvalue.t.dc)^2)
    mse.t.dc<-mse.t.dc+((stats.t.dc[[1]][,3])^2)
    sd.t.dc<-sd.t.dc+(stats.t.dc[[1]][,3])
    Bias.t.dc<-Bias.t.dc+sqrt((stats.t.dc[[1]][,1]-
paramvalue.t.dc)^2)

#-----Prediction of random effects:
true model
model=" ,n))
    prec <- make.symmetric(K1*solve(vcov(dcmo)))
    prdat <-list(y=y, X=X, Z=Z, I=I, T=T, Nsp=Nsp,
Nit=Nit,nbetas=nbetas, param = coef(dcmo), prec = prec, K=1)
    mod1.1.pr <- jags.fit(prdat, parameters.t.p,
model.t.p, n.chains=2, n.adapt=10000, n.update=10000, n.iter=10000,
thin=5)

    update(updated.model(mod1.1.pr), 10000)

```



```

modl.2.pr <- coda.samples(updated.model(modl.1.pr),
parameters.t.p, n.iter=10000, thin = 5)

num<-0
while( (max ((gelman.diag(modl.2.pr))$psrf[,1]) >1.05)
& num<20){
    update(updated.model(modl.1.pr), 10000)
    modl.2.pr <-
coda.samples(updated.model(modl.1.pr), parameters.t.p, n.iter=10000,
thin = 5)
    num<-num+1
}
print(paste("num",num))
stats.t.p<-
summary(modl.2.pr,quantiles=c(0.005,0.01,0.025,0.05,0.95,0.975,0.99,0.
995))
    paramvalue.t.p<-c(MU)
    print(paste("-----RB, MSE and
covarage for true model",n,"-----"))
    mean.t.p<-mean.t.p+stats.t.p[[1]][,1]
    covp99.t.p<-
covp99.t.p+1*((stats.t.p[[2]][,1]<paramvalue.t.p &
paramvalue.t.p<stats.t.p[[2]][,8] ))
    covp98.t.p<-
covp98.t.p+1*((stats.t.p[[2]][,2]<paramvalue.t.p &
paramvalue.t.p<stats.t.p[[2]][,7] ))
    covp95.t.p<-
covp95.t.p+1*((stats.t.p[[2]][,3]<paramvalue.t.p &
paramvalue.t.p<stats.t.p[[2]][,6] ))
    covp90.t.p<-
covp90.t.p+1*((stats.t.p[[2]][,4]<paramvalue.t.p &
paramvalue.t.p<stats.t.p[[2]][,5] ))
    domain99.t.p<-domain99.t.p+((stats.t.p[[2]][,8]-
stats.t.p[[2]][,1] ))
    domain98.t.p<-domain98.t.p+((stats.t.p[[2]][,7]-
stats.t.p[[2]][,2] ))
    domain95.t.p<-domain95.t.p+((stats.t.p[[2]][,6]-
stats.t.p[[2]][,3] ))
    domain90.t.p<-domain90.t.p+((stats.t.p[[2]][,5]-
stats.t.p[[2]][,4] ))
    MSE.t.p<-MSE.t.p+((stats.t.p[[1]][,1]-
paramvalue.t.p)^2)
    Bias.t.p<-Bias.t.p+sqrt((stats.t.p[[1]][,1]-
paramvalue.t.p)^2)
    mse.t.p<-mse.t.p+((stats.t.p[[1]][,2])^2)
    sd.t.p<-sd.t.p+(stats.t.p[[1]][,2])

}
}
date()
outbound.hb[R]
outbound.dc[R]

```

