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#####  
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#####  
#####  
# GGAMM in R-INLA to identify the pattern and determinants of  
#malnutrition among under-five age children in Ethiopia ###  
#####  
setwd("D:/PhDnurtion/papers/Paper2/finalpaper2/BMCRM_revision")###seting the working directory  
### Activating INLA library and its dependence  
#install.packages("INLA", repos=c(getOption("repos"), INLA="https://inla.r-inla-  
download.org/R/stable"), dep=TRUE)  
# Packages required  
library(rgdal)  
library(splines)  
library(sp)  
library(maptools)  
library(spdep)  
library(INLA)  
library(foreign)  
library(shapefiles)  
library(BayesX)  
library(lattice)  
library(R2BayesX)  
library(sf)  
inla.version()# we can see the INLA versions  
## Importing the shape file
```

```
### Generating spatial weight matrix in R-InLA
```

```
shapef <-
```

```
st_read("D:/PhDnurtion/papers/paper2/finalpaper2/BMCRM_revision/shepefile/ETH_adm2.shp")
```

```
temp <- poly2nb(shapef)
```

```
nb2INLA("Ethiopia.graph", temp)
```

```
Ethiopia_shape <- paste(getwd(), "/Ethiopia.graph", sep="")
```

```
#call the map
```

```
H=("D:/PhDnurtion/papers/paper2/finalpaper2/BMCRM_revision/Ethiopia.graph")
```

```
library(readr)
```

```
finaldata <- read_csv("D:/PhDnurtion/papers/paper2/finalpaper2/BMCRM_revision/finaldata.csv")
```

```
data=data.frame(finaldata)
```

```
data$CIAF=as.factor(data$CIAF)
```

```
attach(data)
```

```
data$CIAFRw[data$CIAF=="A"]=0
```

```
data$CIAFRw[data$CIAF=="CIAF"]=1
```

```
attach(data)
```

```
head(data)
```

```
#####Potential models#####
```

```
##### to add the weight, first we need to enable the weighted function in R inla
```

```
inla.setOption(enable.inla.argument.weights=TRUE)
```

```
GLMM<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+  
factor(DDS)+factor(type_birth)+  
factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+facto  
r(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year),family="binomi  
al",data=data,control.compute=list(dic=TRUE,mlik=TRUE,cpo=TRUE,waic=TRUE),weights =weight)
```

```
GAMM<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+  
factor(DDS)+factor(type_birth)+  
factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+facto
```

```
r(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(BMIMother)+
f(bord)+f(HHsize)+
```

```
(aridity)+f(lst)+f(ur)+f(agechild)+f(evi),family="binomial",data=data,control.compute=list(dic=TRUE,mlik=
TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weights = weight)
```

```
# including the spatial effect
```

```
data$zone.unstruct = data$zone
```

```
GGAMM<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+
factor(DDS)+factor(type_birth)+
factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+facto
r(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(BMIMother)+
f(bord)+f(HHsize)+
```

```
f(aridity)+f(lst)+f(ur)+f(agechild)+f(evi)+ f(zone,model="besag",graph.file=H, param=c(1,0.01)) +
```

```
f(zone.unstruct,model="iid",param=c(1,0.01)),family="binomial",data=data,control.compute=list(dic=TR
UE,mlik=TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weights = weights)
```

```
#####generating both the fixed and random
effects#####
```

```
####prior sensitivity analysis for the selected model (GGAMM) #####
```

```
##### IG(a,b) IG(a,b) IG(a,b) IG(a,b)
```

```
#structured (0.001, 0.001) (0.01, 0.01) (0.05, 0.0005) (1, 0.026)
```

```
#unstructured (0.001, 0.001) (0.01, 0.01) (0.05, 0.0005) (1, 0.026)
```

```
###IG (0.001, 0.001)
```

```
GGAMM1<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+
factor(DDS)+factor(type_birth)+
factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+facto
r(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(BMIMother)+
f(bord)+f(HHsize)+
```

```
f(aridity)+f(lst)+f(ur)+f(agechild)+f(evi)+ f(zone,model="besag",graph.file=H,
param=c(0.001,0.001)) +
```

```
f(zone.unstruct,model="iid",param=c(0.001,0.001)),family="binomial",data=data,control.compute=list(dic=TRUE,mlik=TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weights = weight)
```

```
##IG (0.01, 0.01)
```

```
GGAMM2<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+factor(DDS)+factor(type_birth)+factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+factor(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(inla.group(BMIMother),model="rw2")+f(inla.group(bord),model="rw2")+f(inla.group(HHsize),model="rw2")+
```

```
f(inla.group(aridity),model="rw2")+f(inla.group(lst),model="rw2")+f(inla.group(ur),model="rw2")+f(inla.group(agechild),model="rw2")+f(inla.group(evi),model="rw2")+ f(zone,model="besag",graph.file=H, param=c(0.01,0.01)) +
```

```
f(zone.unstruct,model="iid",param=c(0.01,0.01)),family="binomial",data=data,control.compute=list(dic=TRUE,mlik=TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weights = weight)
```

```
###IG (0.5, 0.0005)
```

```
GGAMM3<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+factor(DDS)+factor(type_birth)+factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+factor(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(BMIMother)+f(bord)+f(HHsize)+
```

```
f(aridity)+f(lst)+f(ur)+f(agechild)+f(evi)+ f(zone,model="besag",graph.file=H, param=c(0.05,0.0005)) +
```

```
f(zone.unstruct,model="iid",param=c(0.05,0.0005)),family="binomial",data=data,control.compute=list(dic=TRUE,mlik=TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weights = weight)
```

```
###(1, 0.026)
```

```
GGAMM4<-inla(CIAFRw~factor(sex_child)+factor(breast_feeding)+factor(comorbidity) +factor(size_c)+factor(DDS)+factor(type_birth)+factor(place_R)+factor(edu_mother)+factor(edu_father)+factor(autonomy)+factor(work_mother)+factor(wateri)+factor(toileti)+factor(Ufive)+factor(media)+factor(wealth_i)+factor(DHS_year)+f(BMIMother)+f(bord)+f(HHsize)+
```

```

      f(aridity)+f(lst)+f(ur)+f(agechild)+f(evi)+ f(zone,model="besag",graph.file=H, param=c(1,0.026))
+

f(zone.unstruct,model="iid",param=c(1,0.026)),family="binomial",data=data,control.compute=list(dic=T
RUE,mlik=TRUE,cpo=TRUE,waic=TRUE),verbose=TRUE,weight=weight)

summary(GGAMM1)

GGAMM1$dic

```

```

#####AUC computation for the given models#####

#install.packages(c("pROC","ROCit")) # Install the `pROC` and "ROCit" packages.

#summaries (mean, standard deviation and quantiles) of the fitted values

###a prediction (on the response scale) associated with each row of the input data

```

```

install.packages(c("pROC","ROCit"))

library(ROCit)

library(pROC)# Load the caret package into this R session.

pred_glm=GLMM$summary.fitted.values$mean

pred_gamm=GAMM$summary.fitted.values$mean

pred_ggamm=GGAMM$summary.fitted.values$mean

Rocfinal=data.frame(pred_glm,pred_gamm,pred_ggamm,CIAFw)

attach(Rocfinal)

####We can take the rounds

###Add this to code1

library(ROCit)

```

```

### for references, we will replace by y_test

roc1 <- plot.roc(as.vector(CIAFw),as.vector(pred_glm), main="ROC comparison")
roc2 <- lines.roc(as.vector(CIAFw), as.vector(pred_gamm), percent=TRUE, col="blue")
roc3 <- rocit(as.vector(CIAFw),as.vector(pred_ggamm))

## Plot ROC curve

plot(roc1 , col = 1,legend = FALSE, YIndex = FALSE)

lines(roc2$TPR~roc2$FPR,col = 15, lwd = 0.5)

lines(roc3$TPR~roc3$FPR,col = 4, lwd = 0.5)

legend("bottomright", col = c(1,15,4),c("GLMM:AUC=0.5137",
"GAMM:AUC=0.5502","GGAMM:AUC=0.7218"), lwd = 0.5)

###Exact summary of the AUC

auc_glm = auc(roc_glm)#Calculate the area under the ROC curve for GLMM

auc_gamm = auc(roc_gamm)#Calculate the area under the ROC curve for GAMM

auc_ggamm = auc(roc_ggamm)#Calculate the area under the ROC curve for GGAMM

#### summary statistics for AUC

auc_glm;auc_gamm;auc_ggamm#we can see the Area under the curve values

```