

Model Coding Language

Rosetta Stone

Nick Holford and Mike K Smith on behalf of the DDMoRe consortium



MDL v5.3 Random Variable Expression	MDL v5.3 Distributions	MLXTRAN	NM-TRAN	BUGS	Phoenix Modeling Language (PML)
<pre>#MODEL OBJECT warfarin_PK_CONC_md1 = mdlobj{ INPUT_VARIABLES{ ID=list(use=id, level=2) TIME=list(use=idv, units="h") WT=list(type=continuous, units="kg") AGE=list(type=continuous, units="") SEX=list(type=categorical(female,male)) AMT=list(use=amt) DVID=list(use=dvid,type=categorical) DV=list(use=dv, level=1) MDV=list(use=mdv) } STRUCTURAL_PARAMETERS{ POP_CL;POP_V;POP_KA;POP_TLAg } VARIABILITY_PARAMETERS{ PPV_CL;PPV_V;PPV_KA;PPV_TLAg;RUV_PROP;RUV_ADD } GROUP_VARIABLES{ GRPCL=POP_CL*(WT/70)^0.75 GRPV=POP_V*WT/70 GRPKA=POP_KA GRPLG=POP_TLAg } } RANDOM_VARIABLE_DEFINITION{ [eta_PPV_CL ~ (type=Normal, mean=0, variance=PPV_CL,level=ID) eta_PPV_V ~ (type=Normal, mean=0, variance=PPV_V,level=ID) eta_PPV_KA ~ (type=Normal, mean=0, variance=PPV_KA,level=ID) eta_PPV_TLAg ~ (type=Normal, mean=0, variance=PPV_TLAg,level=ID) eps_RUV_PROP ~ (type=Normal, mean=0, variance=RUV_PROP,level=DV) eps_RUV_ADD ~ (type=Normal, mean=0, variance=RUV_ADD,level=DV) } INDIVIDUAL-VARIABLES{ CL=GRPCL*exp(eta_PPV_CL) V=GRPV*exp(eta_PPV_V) KA=GRPKA*exp(eta_PPV_KA) ALAG1=GRPLG*exp(eta_PPV_TLAg) } MODEL_PREDICTION{ LIBRARY{ amount=PK(input=first-order, distribution=1, elimination=first-order, parameterization=vc1-t, param=list(cl=CL, v=V, # depot compartment is 0 DCHT=0, cKa=ln(2)/KA, tlag0=ALAG1)) CONC=amount.AL/V } OBSERVATION{ Y = CONC*(1+eps_RUV_PROP)+eps_RUV_ADD }</pre>	<pre>#MODEL OBJECT warfarin_PK_CONC_md1 = mdlobj{ INPUT_VARIABLES{ ID;TIME;WT;AMT;DV;MDV } STRUCTURAL_PARAMETERS{ POP_CL;POP_V;POP_KA;POP_TLAg } VARIABILITY_PARAMETERS{ PPV_CL;PPV_V;PPV_KA;PPV_TLAg;RUV_PROP;RUV_ADD } GROUP_VARIABLES{ GRPCL=POP_CL*(WT/70)^0.75 GRPV=POP_V*WT/70 } } RANDOM_VARIABLE_DEFINITION{ ln(CL) ~ (type=Normal, mean=ln(GRPCL),variance=PPV_CL,level=ID) ln(V) ~ (type=Normal, mean=ln(GRPV), variance=PPV_V,level=ID) ln(KA) ~ (type=Normal, mean=ln(POP_KA), variance=PPV_KA,level=ID) ln(TLAg) ~ (type=Normal, mean=ln(POP_TLAg), variance=PPV_TLAg,level=ID) } INDIVIDUAL-VARIABLES{ CL=exp(ln(CL)) V=exp(ln(V)) KA=exp(ln(KA)) ALAG1=exp(ln(TLAg)) } MODEL_PREDICTION{ LIBRARY{ amount=list(library=nmadvan,model=2,trans=2,param=1 list(V,CL,KA,S2,ALAG1,F,A)) } CONC=amount.param.AL/V } OBSERVATION{ ln(Y) ~ (type=Normal, mean=ln(CONC), variance=CONC^2*(RUV_PROP^2 + RUV_ADD^2) }</pre>	<pre>[INDIVIDUAL] input=(Tlag_pop, omega_Tlag,ka_pop,omega_ka,V_pop,omega_V,Cl_pop, omega_Cl,weight,beta_V,beta_Cl) ;GROUP VARIABLES EQUATION: lw70=log(weight/70) DEFINITION: ; plus INDIVIDUAL VARIABLES Cl = {distribution=lognormal, reference=Cl_pop,sd=omega_Tlag,covariate=lw70,coeff icient=beta_Cl} V = {distribution=lognormal, reference=V_pop,sd=omega_Tlag, covariate=lw70,coefficient=beta_V} ka = {distribution=lognormal, reference=ka_pop,sd=omega_ka} Tlag = {distribution=lognormal, reference=Tlag_pop,sd=omega_Tlag}</pre>	<pre>;VARIABLES \$PK ;GROUP VARIABLES GRPCL=THETA(1)*(WT/70)**0.75 GRPV=THETA(2)*WT/70 GRPKA=THETA(3) GRPLG=THETA(4) ;Implicit definition ;INDIVIDUAL VARIABLES CL=GRPCL*EXP(ETA(1)) V=GRPV*EXP(ETA(2)) KA=GRPKA*EXP(ETA(3)) ALAG1=GRPLG*EXP(ETA(4)) S2=V ;LIBRARY \$SUBR ADVAN2 TRANS2 \$ERROR EQUATION: C=C(A(2)/V ; OBSERVATION Y=CONC*(1+ERR(1))+ERR(2)</pre>	<pre>#GROUP VARIABLES for(i in 1:nsub){ #CL logthetaMean[i, 1] <- logCLHat +0.75*log(weight[start(i)]/70) #V logthetaMean[i, 2] <- logVHat + log(weight[start(i)]/70) # ka - alpha logthetaMean[i, 3] <- logDkaHat # tlag logthetaMean[i, 4] <- logtlagHat theta[i,5] <- 1 # F1 theta[i,6] <- 1 # F2 theta[i,7] <- 0 # tlag2 # RANDOM (INDIVIDUAL) VARIABLE DEFINITION logtheta[i, 1:2] ~ dnorm(logthetaMean[i, 1:2], omega.inv[1:2, 1:2]) logtheta[i,3] ~ dnorm(logthetaMean[i,3],tau.Ka) logtheta[i,4] ~ dnorm(logthetaMean[i,4],tau.tlag1) }logCLHat ~ dnorm(0,1.0E-6) logVHat ~ dnorm(0,1.0E-6) logDkaHat ~ dnorm(0,1.0E-6) log(tlAg) <- logCLHat log(VHat) <- logVHat log(DkaHat) <- logDkaHat tau.Ka <- 1/(sigma.Ka*sigma.Ka) sigma.Ka~dunif(0,1000) tau.tlag1 <- 1/(sigma.tlag1*sigma.tlag1) sigma.tlag1 ~ dunif(0, 1000) tau <- 1/(sigma*sigma) sigma ~ dunif(0,1000) omega.inv[1:2, 1:2] ~ dwish(omega.inv.prior[1:2, 1:2], 2) omega[1:2, 1:2] <- inverse(omega.inv[1:2, 1:2]) eps <- 1.0E-6 ## INDIVIDUAL PREDICTIONS for(i in 1:nsub){ logthetaPred[i, 1:3] ~ dnorm(logthetaMean[i, 1:3], omega.inv[1:3, 1:3]) thetaPred[i,5] <- 1 # F1 thetaPred[i,6] <- 1 # F2 thetaPred[i,7] <- 0 # tlag2 for(j in 1:4){ log(theta[i,j]) <- logtheta[i,j] log(thetaPred[i,j]) <- logthetaPred[i,j] } ## Define structural model xhat[start(i):end(i),1:2] <- OneCptModel(time[start(i):end(i)], amt[start(i):end(i)], rate[start(i):end(i)], ii[start(i):end(i)], evid[start(i):end(i)], cmt[start(i):end(i)], addl[start(i):end(i)], ss[start(i):end(i)], theta[i,j]) xhatPred[start(i):end(i),1:2] <- OneCptModel(time[start(i):end(i)], amt[start(i):end(i)], rate[start(i):end(i)], ii[start(i):end(i)], evid[start(i):end(i)], cmt[start(i):end(i)], addl[start(i):end(i)], ss[start(i):end(i)], thetaPred[i,j]) #end subject specific loop # OBSERVATION for(i in 1:nobs){ logCobs[i] ~ dnorm(logChat[i],tau) logCobsCond[i] ~ dnorm(logChat[i],tau) CChat[i] <- 1000*xhat[i,2]/theta[subject[i],3] logChat[i] <- log(max(CChat[i],eps)) logCobsPred[i] ~ dnorm(logChatPred[i],tau) CChatPred[i] <- 1000*xhatPred[i,2]/thetaPred[subject[i],3] logChatPred[i] <- log(max(CChatPred[i],eps)) } } xdata <- read.csv("warfarin_conc.csv", header=T,ignore="") ## create WinBUGS data bugsdta <- list(nobs = nobs, nsub = length(unique(xdata\$subject)), start = start, end = c(start[-1]-1,nobs), subject = xdata\$subject, weight = xdata\$weight, time = xdata\$time, amt = xdata\$dose, rate = rep(0,nobs), ii = xdata\$ii, evid = xdata\$evid, cmt = ifelse(xdata\$evid > 0, 1, 2), addl = xdata\$addl, ss = rep(0,nobs), logCobs = ifelse(xdata\$cobs <= 0, NA, log(xdata\$cobs)), omega.inv.prior=diag(rep(0.05,3))) #DATA #in warfarin_conc.csv #ID,time,wt,age,sex,amt,rate,dvid,dv,mdv 0,0,66.7,50,1,100,-2,0,,1 0,0.5,66.7,50,1,,1,0,0 #in Cols.txt (column mapping file): id(id) time(time) covr(wt<-wt) covr(age<-age) covr(sex<-sex) dose(a<-amt) obs(cObs<-dv) mdv(mdv)</pre>	<pre>#VARIABLES covariate(wt) grpV=popV*wt/70 #Implicit definition #RANDOM (INDIVIDUAL) VARIABLE DEFINITION logtheta[i, 1:2] ~ dnorm(logthetaMean[i, 1:2], omega.inv[1:2, 1:2]) logtheta[i,3] ~ dnorm(logthetaMean[i,3],tau.Ka) logtheta[i,4] ~ dnorm(logthetaMean[i,4],tau.tlag1) }logCLHat ~ dnorm(0,1.0E-6) logVHat ~ dnorm(0,1.0E-6) logDkaHat ~ dnorm(0,1.0E-6) log(tlAg) <- logCLHat log(VHat) <- logVHat log(DkaHat) <- logDkaHat tau.Ka <- 1/(sigma.Ka*sigma.Ka) sigma.Ka~dunif(0,1000) tau.tlag1 <- 1/(sigma.tlag1*sigma.tlag1) sigma.tlag1 ~ dunif(0, 1000) tau <- 1/(sigma*sigma) sigma ~ dunif(0,1000) omega.inv[1:2, 1:2] ~ dwish(omega.inv.prior[1:2, 1:2], 2) omega[1:2, 1:2] <- inverse(omega.inv[1:2, 1:2]) eps <- 1.0E-6 ## INDIVIDUAL PREDICTIONS for(i in 1:nsub){ logthetaPred[i, 1:3] ~ dnorm(logthetaMean[i, 1:3], omega.inv[1:3, 1:3]) thetaPred[i,5] <- 1 # F1 thetaPred[i,6] <- 1 # F2 thetaPred[i,7] <- 0 # tlag2 for(j in 1:4){ log(theta[i,j]) <- logtheta[i,j] log(thetaPred[i,j]) <- logthetaPred[i,j] } ## Define structural model xhat[start(i):end(i),1:2] <- OneCptModel(time[start(i):end(i)], amt[start(i):end(i)], rate[start(i):end(i)], ii[start(i):end(i)], evid[start(i):end(i)], cmt[start(i):end(i)], addl[start(i):end(i)], ss[start(i):end(i)], theta[i,j]) xhatPred[start(i):end(i),1:2] <- OneCptModel(time[start(i):end(i)], amt[start(i):end(i)], rate[start(i):end(i)], ii[start(i):end(i)], evid[start(i):end(i)], cmt[start(i):end(i)], addl[start(i):end(i)], ss[start(i):end(i)], thetaPred[i,j]) #end subject specific loop # OBSERVATION for(i in 1:nobs){ logCobs[i] ~ dnorm(logChat[i],tau) logCobsCond[i] ~ dnorm(logChat[i],tau) CChat[i] <- 1000*xhat[i,2]/theta[subject[i],3] logChat[i] <- log(max(CChat[i],eps)) logCobsPred[i] ~ dnorm(logChatPred[i],tau) CChatPred[i] <- 1000*xhatPred[i,2]/thetaPred[subject[i],3] logChatPred[i] <- log(max(CChatPred[i],eps)) } } #MODEL PREDICTION #LIBRARY cMicro(AL, CL / V, first = (Aa = Ka)) dosepoint(Aa, tlag = Tlag)</pre>
<pre>#DATA OBJECT warfarin_PK_CONC_dat = dataobj{ FILE{ data=list(source="warfarin_conc_pca.csv", ignore="#", inputformat="NONMEM") } HEADER{ ID=list(type=categorical) TIME=list(type=continuous) WT=list(type=continuous, units="kg") AGE=list(type=continuous, units="") SEX=list(type=categorical(0="female",1="male")) AMT=list(type=continuous) DVID=list(type=categorical) DV=list(type=continuous) MDV=list(type=categorical) } } #PARAMETER OBJECT warfarin_PK_CONC_par = parobj{ STRUCTURAL{ POP_CL=list(value=0.1,lo=0.001) POP_V=list(value=8,lo=0.001) POP_KA=list(value=2,lo=0.001) POP_TLAg=list(value=1,lo=0.001) } VARIABILITY{ matrix(type="VAR"){PPV_CL=0.1,0.01, PPV_V=0.1} diag(type="VAR"){PPV_KA=0.1,PPV_TLAg=0.1} RUV_PROP=list(type="VAR",value=0.01) RUV_ADD=list(type="VAR",value=0.05) } } #TASK PROPERTIES OBJECT warfarin_PK_CONC_task = taskobj{ DATA{IGNORE=if(DVID=2)} myEST=function(t,m,p,d) { ESTIMATE{ target=t model=m parameter=p data=d algo=list("COND INTER") max=9990 sig=3 cov="y" } } }</pre>	<pre>#DATA OBJECT warfarin_PK_CONC_dat = dataobj{ FILE{ data=list(source="warfarin_conc_pca.csv", ignore="#", inputformat="NONMEM") } HEADER{ ID;TIME;WT;AGE;SEX;AMT;DVID;DV;MDV } } #PARAMETER OBJECT warfarin_PK_CONC_par = parobj{ STRUCTURAL{ POP_CL=0.1 POP_V=8 POP_KA=2 POP_TLAg=1 } VARIABILITY{ matrix(type="VAR"){PPV_CL=0.1,0.01, PPV_V=0.1} diag(type="VAR"){PPV_KA=0.1,PPV_TLAg=0.1} RUV_PROP=list(type="VAR",value=0.01) RUV_ADD=list(type="VAR",value=0.05) } } #TASK PROPERTIES OBJECT warfarin_PK_CONC_task = taskobj{ DATA{IGNORE=if(DVID=2)} myEST=function(t,m,p,d) { ESTIMATE{ target=t model=m parameter=p data=d algo=list("COND INTER") max=9990 sig=3 cov="y" } } }</pre>	<pre>[DATA] datafile = 'D:/MLXTRAN/warfarin_data.txt' header = {id, time, amt, wt, sex, age, dv, dvid} id = {use = group} time = {use = time} amt = {use = amount} weight = {use = variable, level = id, type = continuous} sex = {use = variable, level = id, type = categorical} age = {use = ignore} dv = {use = observation, name = { 'concentration', 'pca'}} dvid = {use = observationType, categories = { 1, 2}}</pre>	<pre>;DATA \$INPUT ID TIME WT AGE SEX AMT DVID DV MDV \$DATA warfarin_conc_pca.csv IGNORE=# ; ignore PCA observations IGNORE (DVID,RQ.2) ;PARAMETERS \$THETA (0.001,0.1) ; POP_CL L/h/70kg (0.001,8) ; POP_V L/70kg (0.001,2) ; POP_KA h-1 (0.001,1) ; POP_TLAg h \$OMEGA BLOCK(2) 0.1 ; PPV_CL 0.01 0.1 ; PPV_V \$OMEGA 0.1 ; PPV_KA 0.1 ; PPV_TLAg \$SIGMA 0.01 ; RUV_PROP 0.05 ; RUV_ADD mg/L</pre>	<pre>xdata <- read.csv("warfarin_conc.csv", header=T,ignore="") ## create WinBUGS data bugsdta <- list(nobs = nobs, nsub = length(unique(xdata\$subject)), start = start, end = c(start[-1]-1,nobs), subject = xdata\$subject, weight = xdata\$weight, time = xdata\$time, amt = xdata\$dose, rate = rep(0,nobs), ii = xdata\$ii, evid = xdata\$evid, cmt = ifelse(xdata\$evid > 0, 1, 2), addl = xdata\$addl, ss = rep(0,nobs), logCobs = ifelse(xdata\$cobs <= 0, NA, log(xdata\$cobs)), omega.inv.prior=diag(rep(0.05,3))) #DATA #in warfarin_conc.csv #ID,time,wt,age,sex,amt,rate,dvid,dv,mdv 0,0,66.7,50,1,100,-2,0,,1 0,0.5,66.7,50,1,,1,0,0 #in Cols.txt (column mapping file): id(id) time(time) covr(wt<-wt) covr(age<-age) covr(sex<-sex) dose(a<-amt) obs(cObs<-dv) mdv(mdv)</pre>	<pre>#PARAMETERS #Structural fixef{ popCl = c(0.001, 0.1,) popV = c(0.001, 8,) popKa = c(0.001, 2,) popTlag = c(0.001, 1,) RUVcv = c(0, 0.1,) } #Variability ranef{ block(nV, nCl) = c(0.1,0.01, 0.1) nKa = 0.1 nTlag = 0.1 } error(CRps = 0.1)</pre>
<pre>#PARAMETER OBJECT warfarin_PK_CONC_par = parobj{ STRUCTURAL{ POP_CL=list(value=0.1,lo=0.001) POP_V=list(value=8,lo=0.001) POP_KA=list(value=2,lo=0.001) POP_TLAg=list(value=1,lo=0.001) } VARIABILITY{ matrix(type="VAR"){PPV_CL=0.1,0.01, PPV_V=0.1} diag(type="VAR"){PPV_KA=0.1,PPV_TLAg=0.1} RUV_PROP=list(type="VAR",value=0.01) RUV_ADD=list(type="VAR",value=0.05) } } #TASK PROPERTIES OBJECT warfarin_PK_CONC_task = taskobj{ DATA{IGNORE=if(DVID=2)} myEST=function(t,m,p,d) { ESTIMATE{ target=t model=m parameter=p data=d algo=list("COND INTER") max=9990 sig=3 cov="y" } } }</pre>	<pre>#PARAMETER OBJECT warfarin_PK_CONC_par = parobj{ STRUCTURAL{ POP_CL=0.1 POP_V=8 POP_KA=2 POP_TLAg=1 } VARIABILITY{ matrix(type="VAR"){PPV_CL=0.1,0.01, PPV_V=0.1} diag(type="VAR"){PPV_KA=0.1,PPV_TLAg=0.1} RUV_PROP=list(type="VAR",value=0.01) RUV_ADD=list(type="VAR",value=0.05) } } #TASK PROPERTIES OBJECT warfarin_PK_CONC_task = taskobj{ DATA{IGNORE=if(DVID=2)} myEST=function(t,m,p,d) { ESTIMATE{ target=t model=m parameter=p data=d algo=list("COND INTER") max=9990 sig=3 cov="y" } } }</pre>	<pre>;PARAMETERS (as part of an estimation task) estimatePopulationParameters(initialValues={ pop_Cl = 0.1, pop_V = 8, pop_Ka = 2, pop_tlag = 1, omega2_Cl = 0.1, omega2_V = 0.1, omega2_ka = 0.1, omega2_tlag = 0.1, a_Conc = 0.225, b_Conc = 0.1 })</pre>	<pre>;PARAMETERS \$THETA (0.001,0.1) ; POP_CL L/h/70kg (0.001,8) ; POP_V L/70kg (0.001,2) ; POP_KA h-1 (0.001,1) ; POP_TLAg h \$OMEGA BLOCK(2) 0.1 ; PPV_CL 0.01 0.1 ; PPV_V \$OMEGA 0.1 ; PPV_KA 0.1 ; PPV_TLAg \$SIGMA 0.01 ; RUV_PROP 0.05 ; RUV_ADD mg/L</pre>	<pre>## create initial ##estimates bugsinit <- function() { list(logCLHat = rnorm(1,log(10),0.2), logVHat = rnorm(1,log(70),0.2), logDkaHat = rnorm(1,log(1),0.2), logDtlagHat = rnorm(1,log(1),0.2), omega.inv = rnorm(1,log(1),0.2), solve(diag(exp(2*norm(2,log(0.25),0.5))))), sigma.Ka = runif(1,0.1,2), sigma.tlag=runif(1,0.1,2), sigma = runif(1,0.1,2)) } #TASK PROPERTIES # within R parameters = c("CLHat","VHat", "DkaHat","omega", "sigma","logCobsCond", "logCobsPred") n.chains = 3 n.iter = 10000 n.burnin = 4000 n.thin = 6</pre>	<pre>#TASK PROPERTIES rem Windows Command line using PML set method=3 set iterations=200 set model=warfarin_PK.mdl set map=Cols.txt set data=warfarin_conc.csv</pre>

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