

Generating Data

```
rm(list=ls())
setwd("~/Dropbox/CRC BOOK/Code for online supplement/Chapter 5/dynr")
set.seed(12345678)
library(mvtnorm)
# setting up matrices
time <- 1000
npad <- 0 #Occasions to throw out to wash away the effects of initial condition
ne <- 2 #Number of latent variables
ny <- 6 #Number of manifest variables
psi <- matrix(c(2.77, 2.47, # Residual variance-covariance matrix
               2.47, 8.40),
              ncol = ne, byrow = T)
lambda <- matrix(c(1, 0, # Lambda matrix containing contemporaneous relations among
                  2, 0, # observed variables and 2 latent variables.
                  1, 0,
                  0, 1,
                  0, 2,
                  0, 1),
                 ncol = ne, byrow = TRUE)
theta <- diag(.5, ncol = ny, nrow = ny) # Measurement error variances.
epsilon <- rmvnorm(time, mean = c(0, 0, 0, 0, 0, 0), sigma = theta) # Measurement errors.
#t(chol(theta)%%matrix(rnorm(ny*time),ncol=time)) #
beta <- matrix(c(0.5, 0, # Lagged directed relations among variables.
                0.4, 0.5),
               ncol = ne, byrow = TRUE)
zeta <- mvtnorm::rmvnorm(time+npad, mean = c(0, 0), sigma = psi)# Latent variable residuals.
#t(chol(psi)%%matrix(rnorm(ne*time),ncol=time))
a0 <- c(0,0)
etaC <- matrix(0, nrow = ne, ncol = time + npad) # Set up matrix for contemporaneous variables.
etaL <- matrix(0, nrow = ne, ncol = time + npad + 1) # Set up matrix for lagged variables.

etaL[,1] <- a0
# generate factors
for (i in 1:(time+npad)){
  etaC[,i] <- beta %%% etaL[,i] + zeta[i, ]
  etaL[,i+1] <- etaC[,i]
}
etaC <- etaC[, (npad+1):(npad+time)]
etaL <- etaL[, (npad+1):(npad+time)]
eta <- t(etaC)

# generate observed series
y <- matrix(0, nrow = time, ncol = ny)
for (p in 1:nrow(y)){
  y[p, ] <- lambda %%% eta[p, ] + epsilon[p, ]
}

write.csv(y, "temp.csv", row.names=FALSE)
```

```

#Write out mkfm6 data files
filey = "ch5mkfm6.dat"
write(time,file=filey,append=F)
write(a0,file=filey,ncolumn=ne,append=T)
write(t(y),file=filey,ncolumn=ny,append=T)

```

```

# Load packages
require(dynr)

```

Specify dynr recipes

```

#Define the dynamic model
# Define the dynamic model - the matrix B
dynamics <- prep.matrixDynamics(
  values.dyn=matrix(c(.5, 0, .3, .5), ncol=2,byrow=T),
  params.dyn=matrix(c('phi11', 'fixed', 'phi21', 'phi22'), ncol=2,byrow=T),
  isContinuousTime=FALSE)

meas <- prep.measurement(
  values.load=matrix(c(1,0,
                      2,0,
                      1,0,
                      0,1,
                      0,2,
                      0,1),ncol=2,byrow=T), #Starting values for entries in Lambda
  params.load=matrix(c('fixed',0,
                      'lambda21',0,
                      'lambda31',0,
                      0,'fixed',
                      0,'lambda52',
                      0,'lambda62'),
                    ncol=2,byrow=T), #Labels for fixed and freed parameters in Lambda
  #values.int = rep(0.1,6),
  #params.int = paste0('int',1:6),
  state.names=c("eta1","eta2"), #Labels for latent variables in eta(t)
  obs.names=paste0('V',1:6) #Labels for observed variables in y(t)
)

```

#Note that in dynr, prep.initial sets the structure of $E(\eta(1|0))$ and $Cov(\eta(1|0))$

```

initial <- prep.initial(
  values.inistate=c(0, 0),
  params.inistate=c('fixed', 'fixed'),#initial means fixed to a vector of zeros.
  values.inicov=matrix(c(2,0,0,2),ncol=2),
  params.inicov=matrix(c('fixed','fixed','fixed','fixed'),ncol=2)) #initial covariance fixed to a
#diagonal matrix of 2s for compatibility with mkfm6 output. Could also be freely estimated #with mu

```

#Process and measurement noise covariance matrices

```

mdcov <- prep.noise(
  values.latent=matrix(c(2,.5,
                      .5,6),ncol=2,byrow=T),
  params.latent=matrix(c('psi_11','psi_12',
                      'psi_12','psi_22'),ncol=2,byrow=T),

```

```

values.observed=diag(rep(.5,6),6),
params.observed=diag(paste0('var_e',1:6),6)
)

```

Read in data and set up data structure in dynr

```

ch5 = read.table('temp.csv',header=TRUE,sep=",")
ch5$ID = rep(1,dim(ch5)[1]) #Add subject ID to the data set
ch5$Time = rep(1:dim(ch5)[1])
# Data
ch52 <- dynr.data(ch5, id="ID", time="Time", observed=paste0("V",1:6))

```

Cook it!

```

#Put recipes and data together to prepare the full model
model <- dynr.model(dynamics=dynamics, measurement=meas,
                    noise=mdcov, initial=initial, data=ch52,
                    outfile="PFA3.c")
#Use same upper and lower bounds as in mkfm6

model@ub[!model$param.names %in% c('psi_11','psi_12','psi_22')] = c(rep(2,3), rep(5,4), #rep(3,6),
rep(log(10),6))
model@lb[!model$param.names %in% c('psi_11','psi_12','psi_22')] = c(rep(-2,3), rep(-5,4), #rep(-3,6),
rep(log(1e-10),6))

res <- dynr.cook(model)

```

```
coef(res)
```

```

##      phi11      phi21      phi22  lambda21  lambda31  lambda52
## 0.49610174 0.36953507 0.56883784 1.87850247 0.84347089 1.97849688
##  lambda62  psi_11  psi_12  psi_22  var_e1  var_e2
## 0.96346830 1.06866856 0.77970301 3.67441110 0.21825724 0.05442611
##  var_e3  var_e4  var_e5  var_e6
## 0.23165679 0.25272313 0.27297952 0.16239815

```

```
summary(res)
```

```

## Coefficients:
##      Estimate Std. Error t value ci.lower ci.upper Pr(>|t|)
## phi11      0.49610    0.03216  15.426  0.43307  0.55914 <2e-16 ***
## phi21      0.36954    0.06901   5.355  0.23428  0.50479 <2e-16 ***
## phi22      0.56884    0.03596  15.817  0.49835  0.63933 <2e-16 ***
## lambda21  1.87850    0.03985  47.139  1.80040  1.95661 <2e-16 ***
## lambda31  0.84347    0.02604  32.392  0.79243  0.89451 <2e-16 ***
## lambda52  1.97850    0.02154  91.872  1.93629  2.02071 <2e-16 ***
## lambda62  0.96347    0.01201  80.226  0.93993  0.98701 <2e-16 ***
## psi_11    1.06867    0.09892  10.803  0.87479  1.26255 <2e-16 ***
## psi_12    0.77970    0.12853   6.066  0.52780  1.03161 <2e-16 ***
## psi_22    3.67441    0.31659  11.606  3.05390  4.29492 <2e-16 ***
## var_e1    0.21826    0.02267   9.629  0.17383  0.26269 <2e-16 ***

```

```
## var_e2    0.05443    0.04781    1.138 -0.03928    0.14813    0.128
## var_e3    0.23166    0.02178   10.637    0.18897    0.27434   <2e-16 ***
## var_e4    0.25272    0.02642    9.567    0.20095    0.30450   <2e-16 ***
## var_e5    0.27298    0.06421    4.251    0.14713    0.39883    1e-05 ***
## var_e6    0.16240    0.01965    8.266    0.12389    0.20091   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## -2 log-likelihood value at convergence = 17526.66
## AIC = 17558.66
## BIC = 17637.19
```