



Appendix C

Computing codes

This appendix provides codes for several models presented in this textbook. The codes are provided in SAS, R, and/or WinBUGS. The codes include the following.

Negative binomial model

SAS code

```
PROC GENMOD DATA=texasdata;  
  MODEL crashes=logaadt/DIST=NEGBIN OFFSET=loglength  
LINK=LOG;  
RUN;
```

R code

```
dat<-read.csv('texasdata.csv')  
summary(dat)  
glm.nb(crashes~logaadt+offset(log(length)),data=dat)
```

WinBUGS code

```
model  
{  
  for(i in 1:N) {  
    crashes[i] ~ dnegbin(p[i],r)  
    p[i] <- r/(r+mu[i])  
    mu[i] <-exp(beta0 +beta1*logaadt[i]+loglength[i])  
  }  
  beta0~dnorm(0.0,0.001)  
  beta1~dnorm(0.0,0.001)  
  r~dgamma(0.001,0.001)  
}  
INITIALIZATION of parameters
```

```
list(
beta0=-10,
beta1=0,
r=1,
)
Alternative specification:
model
{
for(i in 1:N) {
crashes[i] ~ dpois(mu[i])
mu[i] <- exp(beta0 +beta1*logaadt[i]+loglength[i])*e[i]
}
for (i in 1:N) { e[i] ~ dgamma(phi, phi) }
beta0 ~ dnorm(0.0,0.001)
beta1 ~ dnorm(0.0,0.001)
phi ~ dgamma(0.001,0.001)
}
INITIALIZATION of parameters
list(
beta0=-10,
beta1=0,
phi=1,
)

```

Negative binomial model with varying dispersion parameter

SAS code

```
PROC NLMIXED DATA=texasdata COVB;
PARMS beta0 1 beta1 1 rho 1;
mu = exp(loglength + beta0 + beta1*logaadt);
eta_p = rho+loglength;
phi = exp(eta_p);
alpha=1/phi;
loglike = (lgamma(crashes+(1/alpha)) - lgamma(crashes+1) -
lgamma(1/alpha) + crashes*log(alpha*mu) - (crashes+(1/alpha))
*log(1+alpha*mu));
MODEL crashes ~ general(loglike);
RUN;

```

WinBUGS code

```

model
{
  for(i in 1:N) {
    Crashes[i] ~ dpois(mu[i])
    mu[i] <- exp(beta0 +beta1*logaadt[i]+loglength[i])*e[i]
  }
  for (i in 1:N) { e[i] ~ dgamma(phi[i], phi[i]) }
  phi[i] <-exp(rho+loglength[i])
  beta0 ~ dnorm(0.0,0.001)
  beta1 ~ dnorm(0.0,0.001)
  rho ~ dgamma(0.001,0.001)
}
INITIALIZATION for parameters
list(
  beta0=-10,
  beta1=0,
  rho=1,
)

```

Radom effects negative binomial model

SAS code

```

PROC NLMIXED DATA=texasdata COVB;
PARMS beta0 1 beta1 1 alpha 1 s2u 0;
mu = exp(loglength + beta0 + beta1*logaadt + u);
loglike = (lgamma(crashes+(1/alpha)) - lgamma(crashes+1) -
lgamma(1/alpha) + crashes*log(alpha*mu) - (crashes+(1/alpha))
*log(1+alpha*mu));
MODEL crashes ~ general(loglike);
RANDOM u ~ normal(0,s2u) subject=subject;
RUN;

```

WinBUGS code

```

model
{
  for(i in 1:N) {
    Crashes[i] ~ dpois(mu[i])
    mu[i] <- exp(beta0 +beta1*logaadt[i]+loglength[i]+u[i])*e[i]
  }
}

```

```

}
for (i in 1:N) { e[i] ~ dgamma(phi, phi)
                u[i] ~ dnorm(u0, v0)}
beta0 ~ dnorm(0.0,0.001)
beta1 ~ dnorm(0.0,0.001)
u0 ~ dnorm(0.0,0.001)
phi ~ dgamma(0.001,0.001)
v0 ~ dgamma(0.001,0.001)
}
INITIALIZATION for parameters
list(
beta0=-10,
beta1=0,
u0=0,
phi=1,
v0=0
)

```

Random parameters negative binomial model

SAS code

```

PROC NLMIXED DATA=texasdata COVB;
PARMS beta0 1 beta1 1 alpha 1 s2u 0 s2v 0 cov 0;
mu = exp(loglength + beta0 + beta1*logaadt);
loglike = (lgamma(crashes+(1/alpha)) - lgamma(crashes+1) -
lgamma(1/alpha) +
          crashes*log(alpha*mu) - (crashes+(1/alpha))
*log(1+alpha*mu));
MODEL crashes ~ general(loglike);
RANDOM beta0 beta1 ~ normal([0,0],[s2u,cov,s2v]) subject=subject;
RUN;

```

WinBUGS code

```

model
{
for(i in 1:N) {
Crashes[i] ~ dpois(mu[i])
mu[i] <- exp(beta0[i] + beta1[i]*logaadt[i]+loglength[i])*e[i]
}
for (i in 1:N) {
beta0[i] ~ dnorm(u0, v0)

```

```

        beta1[i] ~ dnorm(u1, v1)
        e[i] ~ dgamma(phi, phi)
    }
    u0 ~ dnorm(0.0, 0.001)
    u1 ~ dnorm(0.0, 0.001)
    phi ~ dgamma(0.001, 0.001)
    v0 ~ dgamma(0.001, 0.001)
    v1 ~ dgamma(0.001, 0.001)
}
INITIALIZATION for parameters
list(
  u0=-10,
  u1=0,
  v0=1,
  v1=1,
  phi=1,
)

```

Poisson-lognormal model

WinBUGS code

```

model
{
  for( i in 1 : N) {
    crashes[i] ~ dpois(mu[i])
    mu[i] <- exp(beta0 + beta1 * logaadt[i] + lambda[i])
    lambda[i] ~ dnorm(0.0, tau)
  }
  alpha ~ dnorm(0.0, 0.001)
  beta ~ dnorm(0.0, 0.001)
  tau <- 1 / (sigma * sigma)
  sigma ~ dunif(0, 100)
}
Alternative specification:
model
{
  for( i in 1 : N) {
    crashes[i] ~ dpois(mu[i])
    log(mu[i]) <- alpha + beta * x[i] + lambda[i]
    lambda[i] ~ dnorm(0.0, tau)
  }
  alpha ~ dnorm(0.0, 0.001)
}

```

```

beta ~ dnorm(0.0,0.001)
tau <- 1/(sigma*sigma)
sigma ~ dgamma(0.001, 0.001)
}

```

Negative binomial-Lindley model

WinBUGS code

```

model
{
  for(i in 1:N) {
    crashes[i] ~ dnegbin(p[i],r)
    p[i] <- r/(r+a[i]*mu[i])
    mu[i] <- exp(beta0 + beta1*logaadt[i]+loglength[i])
    a[i] ~ dgamma(f[i],t)
    f[i] <- -1+z[i]
    z[i] ~ dbern(k)
  }
  r <- 1/alpha
  t <- (1-k)/k
  beta0 ~ dnorm(0.0,0.001)
  beta1 ~ dnorm(0.0,0.001)
  alpha ~ dunif(0.1,10)
  k ~ dbeta(N/3,N/2)
}
INITIALIZATION of parameters
list(
  beta0=-10,
  beta1=0,
  alpha=0.1,
  k=0.5
)

```

Conway—Maxwell—Poisson distribution

R code

The complete code is located at:

<https://cran.r-project.org/web/packages/COMPoissonReg/COM-PoissonReg.pdf>

Multinomial logit model

SAS code

```

PROC NLMIXED DATA=crashdata;
PARMS k0=0 a0=0 b0=0 c0=0
      k1=0 a1=0 b1=0 c1=0
      k2=0 a2=0 b2=0 c2=0
      k3=0 a3=0 b3=0 c3=0
      k4=0 a4=0 b4=0 c4=0 ;
k= exp(k0 + k1*variable1 + k2* variable2 + k3* variable3 + k4*
variable4 );
a= exp(a0 + a1* variable1 + a2* variable2 + a3* variable3 + a4*
variable4 );
b= exp(b0 + b1* variable1+ b2* variable2 + b3* variable3 + b4*
variable4 );
c= exp(c0 + c1* variable1 + c2* variable2 + c3* variable3 + c4*
variable4 );
denominator= 1 + k + a + b + c;
probability1 = k / denominator;
probability2 = a / denominator;
probability3 = b / denominator;
probability4 = c / denominator;
probability5 = 1 / denominator;
if severity=1 then loglike=log(probability1);
if severity=2 then loglike=log(probability2);
if severity=3 then loglike=log(probability3);
if severity=4 then loglike=log(probability4);
if severity=5 then loglike=log(probability5);
model severity ~ general(loglike);
RUN;

```

Nested logit model

SAS code

This is a 2-level model. Level 1 at the bottom has alternatives that are nested at level 2 in two nests (first nest has three alternatives and second nest has two alternatives). The nests are joined at the top of the tree.

```

PROC MDC DATA=crashdata COVEST=hess;
MODEL decision = variable1 variable2 variable3 variable4/
      TYPE=nlogit
      CHOICE=(alt);

```

```
ID id;  
UTILITY u(1, ) = variable1 variable2 variable3,  
          u(2, 1 2@1) = variable4;  
NEST level(1) = (1 2 3@ 1, 4 5 @ 2),  
               level(2) = (1 2 @ 1);  
RUN;
```