# Bayesian Hierarchical Weibull Regression

JQT paper with 200 drivers - Model 3

Miao Cai miao.cai@slu.edu

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### 1 Weibull Distribution

The probability density function of a Weibull distribution is

$$f(t) = \begin{cases} \frac{\kappa}{\theta} \left(\frac{t}{\theta}\right)^{\kappa - 1} e^{(t/\theta)^{\kappa}}, & t > 8\\ 0, & t \le 0 \end{cases}$$

Then the survival function is:

$$S(t) = P(T > t) = 1 - P(T < t) = e^{-(t/\theta)^{\kappa}}$$

The hazard function is the first order derivative of the survival function:

$$h(t) = \frac{dS(t)}{dt} = \frac{\kappa}{\theta} \left(\frac{t}{\theta}\right)^{\kappa - 1}$$

## 2 Weibull regression

We assume  $Y_{i,d(i),s(i)}$  is the time until the first critical event from the start of a trip, then we have

$$Y_{i,d(i),s(i)} \sim \text{WEIBULL}(\kappa, \theta)$$
  
 $\theta = \exp(\beta_{0,d(i)} + \beta_{1,d(i)} \cdot \text{CT}_i + \xi \cdot \mathbf{W} + \nu \cdot \mathbf{D_i})$ 

Where the  $\theta$  is the scale parameter and  $\kappa$  is the shape parameter. When  $\kappa > 1$ , the hazard of having critical event is increasing as cumulative driving time gets longer, which indicates fatigued driving. When  $\kappa = 1$ , the Weibull distribution becomes an exponential distribution with constant hazard  $\frac{1}{\theta}$ , which indicates no fatigue. When  $0 < \kappa < 1$ , the hazard of having critical events is decreasing with longer cumulative driving time, which indicates anti-fatigue, or burn-in.

## 3 Examples

This is an example of Weibull regression provided by the StAN forum.

### 3.1 Example code 1

```
rm(list=ls(all=TRUE))
require(foreign)
require(rstan)
require(MASS)
require(VGAM)
require(loo)
```

```
time <- c(6, 6, 6, 6, 7, 9, 10, 10, 11, 13, 16, 17, 19, 20, 22, 23, 25, 32, 32, 34, 35,
        1, 1, 2, 2, 3, 4, 4, 5, 5, 8, 8, 8, 8, 11, 11, 12, 12, 15, 17, 22, 23)
N uncensored <- sum(status == 1)
N_censored <- sum(status == 0)</pre>
group_uncensored <- drug[status == 1] + 1</pre>
group_censored <- drug[status == 0] + 1</pre>
t_uncensored <- time[status == 1]</pre>
censor_time <- time[status == 0]</pre>
M = 2
# THE MODEL.
modelString = "
data {
int<lower=0> N uncensored;
int<lower=0> N_censored;
int<lower=0> M;
int<lower=1,upper=M> group_uncensored[N_uncensored];
int<lower=1,upper=M> group_censored[N_censored];
real<lower=0> censor_time[N_censored];
real<lower=0> t_uncensored[N_uncensored];
parameters {
real<lower=0> r;
real beta[M];
// t_censored / censor_time
real<lower=1> t2_censored[N_censored];
model {
r ~ exponential(0.001);
beta ~ normal(0, 100);
for (n in 1:N_uncensored) {
t_uncensored[n] ~ weibull(r, exp(-beta[group_uncensored[n]] / r));
for (n in 1:N_censored) {
t2_censored[n] ~ weibull(r, exp(-beta[group_censored[n]] / r) / censor_time[n]);
}
}
generated quantities {
real median[M];
real drug;
// real veh_control;
```

```
for (m in 1:M)
median[m] = pow(log(2) * exp(-beta[m]), 1/r);
drug = beta[2] - beta[1];
}
" # close quote for modelstring
writeLines(modelString,con="model.txt")
#-----
# THE DATA.
dataList = list(
 N_censored = N_censored ,
 N_uncensored = N_uncensored , # BUGS does not treat 1-column mat as vector
 M = M
 group_uncensored = group_uncensored ,
  group_censored = group_censored,
 censor_time = censor_time,
 t_uncensored = t_uncensored
# INTIALIZE THE CHAINS.
# glmInfo = lm( dataList$y ~ dataList$x) # R func.
# show( glmInfo ); flush.console() # display in case glm() has troubles
# b0Init = glmInfo$coef[1]
\# bInit = qlmInfo$coef[-1]
# should use those from ordered logit to start with
if (FALSE) {
 threshInit = myologit$zeta
 bInit = myologit$coefficients
 initsList = list(
   b = bInit,
   thresh = threshInit )
}
# RUN THE CHAINS
parameters = c("beta", "drug") # The parameter(s) to be monitored.
                         # Number of steps to "tune" the samplers.
adaptSteps = 500
burnInSteps = 500
                          # Number of steps to "burn-in" the samplers.
nChains = 3
                           # Number of chains to run.
initsChains <- list()</pre>
if (FALSE) {
 for (i in 1:nChains) {
   initsChains[[i]] <- initsList</pre>
 }
}
numSavedSteps=6000
                          # Total number of steps in chains to save.
thinSteps=1
                            # Number of steps to "thin" (1=keep every step).
```

```
nPerChain = ceiling( (numSavedSteps * thinSteps ) / nChains ) # Steps per chain.
# Burn-in:
cat( "Burning in the MCMC chain...\n" )
# The saved MCMC chain:
cat( "Sampling final MCMC chain...\n" )
time.used <- proc.time()</pre>
mcmcSamples <- stan(model_code=modelString, data=dataList, seed = 47306,
                    pars=parameters, chains=nChains, # init=initsChains,
                    iter=nPerChain,
                    warmup=burnInSteps ) # init=initsChains
summary(mcmcSamples)
print(mcmcSamples)
# mcmcChain = as.matrix(mcmcSamples)
# Stop the clock
proc.time() - time.used
if (FALSE) {
 log_lik <- extract_log_lik(mcmcSamples)</pre>
 waic(log_lik)
 loo(log_lik)
  save.image("bayesGologitMod01aStan.Rdata")
  sink()
}
```

## 3.2 Example code 2

```
rm(list=ls(all=TRUE))
require(foreign)
require(rstan)
require(MASS)
require(VGAM)
require(loo)
time \leftarrow c(6, 6, 6, 6, 7, 9, 10, 10, 11, 13, 16, 17, 19, 20, 22, 23, 25, 32, 32, 34, 35,
         1, 1, 2, 2, 3, 4, 4, 5, 5, 8, 8, 8, 8, 11, 11, 12, 12, 15, 17, 22, 23)
status \leftarrow c(0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0,
         N_uncensored <- sum(status == 1)
N_censored <- sum(status == 0)</pre>
group_uncensored <- drug[status == 1]</pre>
group_censored <- drug[status == 0]</pre>
t uncensored <- time[status == 1]</pre>
censor_time <- time[status == 0]</pre>
M = 1
```

```
# THE MODEL.
modelString = "
data {
 int<lower=0> N_uncensored;
 int<lower=0> N_censored;
 int<lower=0> M;
 int<lower=0,upper=1> group_uncensored[N_uncensored];
  int<lower=0,upper=1> group_censored[N_censored];
 real<lower=0> censor_time[N_censored];
 real<lower=0> t_uncensored[N_uncensored];
}
parameters {
 real<lower=0> r;
 vector[M] beta;
 // t_censored / censor_time
 real<lower=1> t2_censored[N_censored];
}
model {
 r ~ exponential(0.001);
 beta ~ normal(0, 100);
 for (n in 1:N_uncensored) {
   t_uncensored[n] ~ weibull(r, exp(-beta*group_uncensored[n] / r));
 for (n in 1:N_censored) {
    t2_censored[n] ~ weibull(r, exp(-beta*group_censored[n] / r) / censor_time[n]);
generated quantities {
 real median[M];
 real drug;
 // real veh_control;
 for (m in 1:M)
  median[m] = pow(log(2) * exp(-beta[m]), 1/r);
" # close quote for modelstring
writeLines(modelString,con="model.txt")
# THE DATA.
dataList = list(
  N_censored = N_censored ,
  N_uncensored = N_uncensored , # BUGS does not treat 1-column mat as vector
  group_uncensored = group_uncensored ,
  group_censored = group_censored,
  censor_time = censor_time,
```

```
t_uncensored = t_uncensored
# INTIALIZE THE CHAINS.
# glmInfo = lm( dataList$y ~ dataList$x) # R func.
# show( glmInfo ) ; flush.console() # display in case glm() has troubles
# b0Init = glmInfo$coef[1]
# bInit = glmInfo$coef[-1]
# should use those from ordered logit to start with
if (FALSE) {
  threshInit = myologit$zeta
  bInit = myologit$coefficients
  initsList = list(
    b = bInit,
    thresh = threshInit )
}
# RUN THE CHAINS
parameters = c("beta") # The parameter(s) to be monitored.
{\tt adaptSteps = 500} \qquad \qquad {\tt \# Number \ of \ steps \ to \ "tune" \ the \ samplers}.
burnInSteps = 500
                            # Number of steps to "burn-in" the samplers.
nChains = 3
                              # Number of chains to run.
initsChains <- list()</pre>
if (FALSE) {
for (i in 1:nChains) {
  initsChains[[i]] <- initsList</pre>
}
}
numSavedSteps=6000
                              # Total number of steps in chains to save.
thinSteps=1
                               # Number of steps to "thin" (1=keep every step).
nPerChain = ceiling( (numSavedSteps * thinSteps ) / nChains ) # Steps per chain.
# Burn-in:
cat( "Burning in the MCMC chain...\n" )
# The saved MCMC chain:
cat( "Sampling final MCMC chain...\n" )
time.used <- proc.time()</pre>
mcmcSamples <- stan(model_code=modelString, data=dataList, seed = 47306,
                    pars=parameters, chains=nChains, # init=initsChains,
                    iter=nPerChain,
                    warmup=burnInSteps ) # init=initsChains
summary(mcmcSamples)
print(mcmcSamples)
# mcmcChain = as.matrix(mcmcSamples)
# Stop the clock
proc.time() - time.used
```

```
if (FALSE) {
  log_lik <- extract_log_lik(mcmcSamples)
  waic(log_lik)
  loo(log_lik)
  save.image("bayesGologitModO1aStan.Rdata")
  sink()
}</pre>
```