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Math 611
Homework 10

1. For the Beta Blockers data set, consider the observations from center (clinical site) 1 only. Use the appropriate transformations to the data frame to create a binary random variable (1=Death, 0=No Death). Use 'MCMCpack' to generate a sample from the posterior distribution of a logistic regression model using a random-walk Metropolis Hastings algorithm and a multivariate normal prior of your choice. Present the plots of this model.

```
bb<-read.csv("BetaBLOCKERS.csv",header=T)
bblist<-list()

for( i in 1:1){ #ℓ

  CD<-bb[which(bb$center == i & bb$trt== "C" & bb$value== "Death" ) , ]
  CD<-CD[1,][rep(seq_len(nrow(CD[1,])), each=sum(CD[,1])),]

  TD<-bb[which(bb$center == i & bb$trt== "T" & bb$value== "Death" ) , ]
  TD<-TD[1,][rep(seq_len(nrow(TD[1,])), each=sum(TD[,1])),]

  CT<-bb[which(bb$center == i & bb$trt== "C" & bb$value== "Total" ) , ]
  CT<-CT[1,][rep(seq_len(nrow(CT[1,])), each=(sum(CT[,1]))-
    nrow(CD)),)]

  TT<-bb[which(bb$center == i & bb$trt== "T" & bb$value== "Total" ) , ]#ℓ
  TT<-TT[1,][rep(seq_len(nrow(TT[1,])), each=(sum(TT[,1]))-
    nrow(TD)),)]

  bblist[[i]]<-rbind(CD,TD,CT,TT)}
bblist<- do.call("rbind", bblist);rownames(bblist)<-NULL
bblist$death<-ifelse(bblist$value == "Death" , 1,0)
table(bblist[,5])

##
##    0    1
## 508  98
```

```
require("MCMCpack")
posterior <- MCMClogit(death~as.factor(trt) , b0=0, B0=.001,data=bblist)
summary(posterior)
```

```
Iterations = 1001:11000
Thinning interval = 1
Number of chains = 1
Sample size per chain = 10000
```

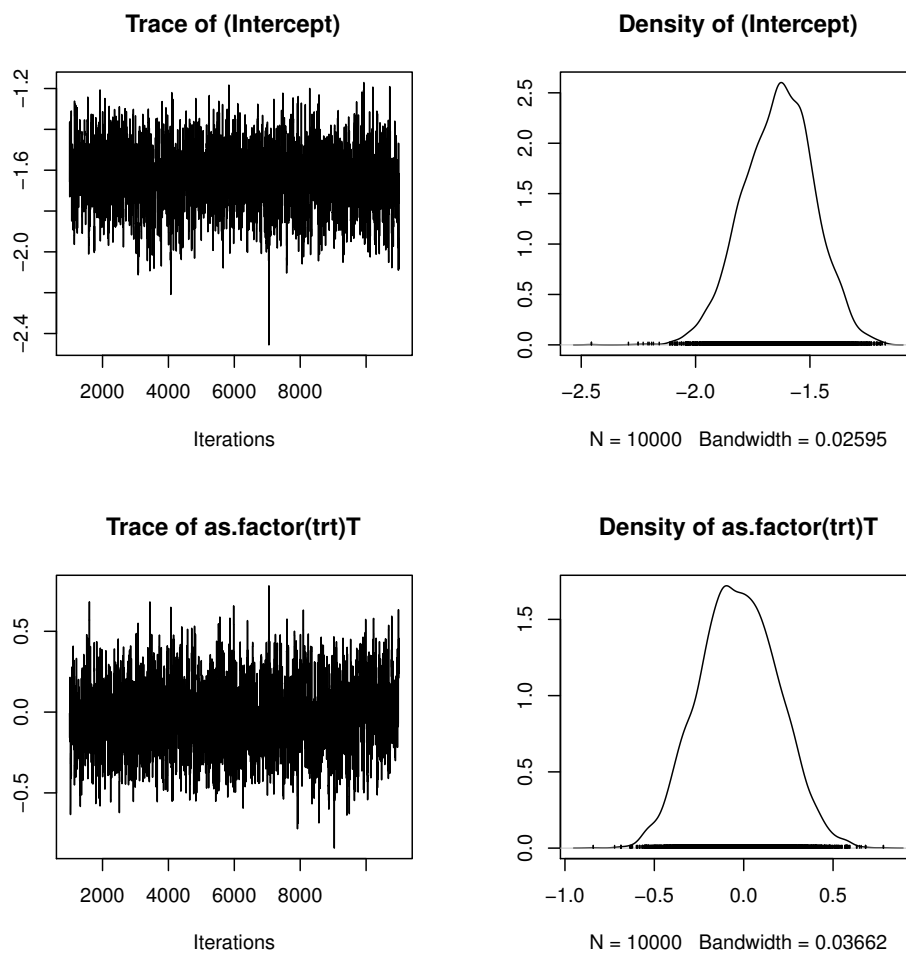
1. Empirical mean and standard deviation for each variable,
plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
(Intercept)	-1.63735	0.1544	0.001544	0.004622
as.factor(trt)T	-0.02719	0.2180	0.002180	0.006417

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
(Intercept)	-1.9484	-1.743	-1.63246	-1.5327	-1.3470
as.factor(trt)T	-0.4363	-0.179	-0.03035	0.1245	0.4002

```
plot(posterior)
```



2. The following R-function uses the Metropolis-Hastings algorithm and an exponential proposal density to draw from the logarithm of the density of interest. Modify this function such that it uses the random walk Metropolis Hastings and a normal proposal density, to draw observations from the logarithm of a density of interest (only the code is needed-no output).

```
MHexp<-function(logf,currentvalue,B)
  {# B: number of iterations
    S<-rep(0,B)
    n_accept<-0
    for(i in 1:B)
      {proposal<-rexp(1,currentvalue)
        probacc<-exp(logf(proposal)-logf(currentvalue))
        dexp(currentvalue,proposal,log=TRUE)-
          dexp(proposal,currentvalue,log=TRUE))
        accept<-ifelse(runif(1)<probacc,1,0)
        currentvalue<-ifelse(accept==1,proposal,currentvalue)
        S[i]<-currentvalue
        n.accept<-n.accept+(accept==1)}
    c(S,n.accept/B)
  }
```

```
MHexp<-function(logf,currentvalue,B,SD=1)
  {# B: number of iterations
    S<-rep(0,B)
    n_accept<-0

    for(i in 1:B)
      {proposal<-rnorm(1,currentvalue,SD)

        probacc<-exp( (logf(proposal) )-(logf(currentvalue) ) )

        accept<-ifelse(runif(1)<probacc,1,0)

        currentvalue<-ifelse(accept==1,proposal,currentvalue)
        S[i]<-currentvalue

        n.accept<-n.accept+(accept==1)}
    c(S,n.accept/B)
  }
```

3. The number of failures $N(t)$, which occur in a computer network over the time interval $[0, t)$, can be described by a homogeneous Poisson process $\{N(t), t \geq 0\}$. On an average, there is a failure after every 4 hours, i.e. the intensity (rate) of the process per hour-interval is equal to $\lambda = 0.25$.

- a. What is the probability of at most 1 failure in $[0, 8)$, at least 2 failures in $[8, 16)$, and at most 1 failure in $[16, 24)$ (time unit: hour)?

Let $N(t) \sim \lambda t$, where $\lambda = 0.25$

$$\begin{aligned} & P\left(\text{at most 1 failure in } [0, 8) \text{ and at least 2 failures in } [8, 16) \text{ and at most 1 failure in } [16, 24) \right) \\ &= P\left(N(8) - N(0) \leq 1, N(16) - N(8) \geq 2, N(24) - N(16) \leq 1\right) \\ &= P\left(N(8) \leq 1\right) P\left(N(8) \geq 2\right) P\left(N(8) \leq 1\right) \end{aligned}$$

$$P\left(N(8) \leq 1\right) = \sum_{x=0}^1 \frac{2^x e^{-2}}{x!} = 0.4060058$$

$$P\left(N(8) \geq 2\right) = 1 - P\left(N(8) \leq 1\right) = 1 - \left[\sum_{x=0}^1 \frac{2^x e^{-2}}{x!} \right] = 0.5939942$$

$$\begin{aligned} & P\left(N(8) \leq 1\right) P\left(N(8) \geq 2\right) P\left(N(8) \leq 1\right) \\ &= (0.4060058) (0.5939942) (0.4060058) \\ &= 0.09791443 \end{aligned}$$

- b. What is the probability that the third failure occurs after 8 hours?

Let T_3 be the time of the third failure. We want to know the probability that the third failure occurs after 8 hours. Alternatively, we want to know the probability of 0, 1, or 2 failures within the first 8 hours.

$$P(T_3 > 8) = P\left(N(8) \leq 2\right) = \sum_{x=0}^2 \frac{2^x e^{-2}}{x!} = 0.6766764$$