Computer Lab 4

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Assignment 1

a,

We do a glm() fit and obtain the maximum liklihood estimator of β in the Poisson regression model for the eBay data.

[1] "The maximum liklihood estimator of beta coefficients:"

```
## PowerSeller VerifyID Sealed Minblem MajBlem LargNeg
## 0.71896702 -0.13883985 0.71681438 0.37324148 0.01400166 0.64177063
## LogBook MinBidShare
## -0.22946925 -3.23450853
```

Through the summary.glm() function, we can say that it looks like VerifyID, Sealed, MajBlem, LogBook and MinBidShare are significant predictors in this model.

Next, we do a Bayesian analysis of the Poisson regression with prior distribution $\beta \sim \mathcal{N}(\mathbf{0}, 100 \cdot (X'X)^{-1})$. We know that we can use the optim() function to numerically fund the posterior mode $\tilde{\beta}$ and the Hessian $-J_{y,\tilde{\beta}}$ at that posterior mode. With these values we can approximate the posterior distribution as a multivariate normal distribution, $\beta|y \sim \mathcal{N}(\tilde{\beta}, J_{y,\tilde{\beta}}^{-1})$.

```
## [1] "The posterior mode beta coefficients: "
```

```
## [1] 0.72028328 -0.13469969 0.71574282 0.37329126 0.01443572 0.64240277 ## [7] -0.22694135 -3.21879090
```

[1] "The hessian at the posterior mode: "

```
##
               [,1]
                           [,2]
                                         [,3]
                                                        [,4]
                                                                      [,5]
## [1,] -1576.80764
                     -68.65215 -3.818716e+02 -1.378046e+02 -9.355086e+01
          -68.65215
                    -128.73447 -5.880810e+01 -1.554736e+01 0.000000e+00
## [3,]
         -381.87165
                     -58.80810 -5.057035e+02 -5.684342e-08 -5.684342e-08
## [4,]
         -137.80459
                     -15.54736 -5.684342e-08 -3.091830e+02 0.000000e+00
## [5,]
          -93.55086
                       0.00000 -5.684342e-08 0.000000e+00 -1.270093e+02
          -57.11834
## [6,]
                       0.00000 5.684342e-08 -4.660851e+01 -5.684342e-08
## [7,]
         -103.04136
                     -81.57867 -1.618865e+02 -6.939475e+01 -3.779654e+01
          304.58623
##
  [8,]
                      40.44637 1.193921e+02
                                              7.699874e+01 3.972684e+01
                              [,7]
##
                 [,6]
                                         [,8]
## [1,] -5.711834e+01
                       -103.04136
                                    304.58623
                        -81.57867
## [2,]
         0.000000e+00
                                     40.44637
## [3,]
         5.684342e-08
                       -161.88649
                                    119.39211
## [4,] -4.660851e+01
                        -69.39475
                                     76.99874
## [5,] -5.684342e-08
                        -37.79654
                                     39.72684
## [6,] -3.858052e+02
                       -215.89732
                                    122.08186
## [7,] -2.158973e+02 -1565.97023
                                    444.17956
## [8,]
        1.220819e+02
                        444.17956 -354.63256
```

[1] "A posterior draw of beta: "

```
## [,1] [,2] [,3] [,4] [,5] [,6]
## [1,] 0.7111967 -0.2890974 0.8119556 0.35097 -0.01587766 0.6786866
## [,7] [,8]
## [1,] -0.2393631 -3.331718
```

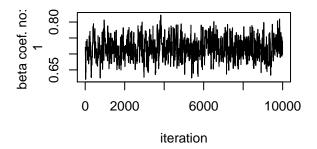
Now we simulate from the actual posterior of β using the random walk Metropolis-Hanstings algorithm. We are going to use a multivariate normal density, $\theta_p|\theta_c \sim \mathcal{N}(\theta_c, \tilde{c} \cdot \Sigma)$ as proposal density where $\Sigma = J_{y,\tilde{\beta}}^{-1}$ and \tilde{c} is equal to 2.4 divided by the squre root of the number of parameters.

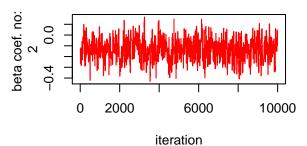
[1] "last iteration of M-H algorithm: "

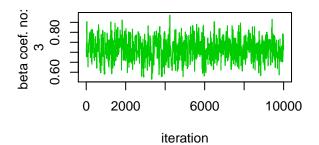
```
## [1] 0.69845754 -0.01422904 0.75913661 0.28230703 0.05263073 0.64118339
## [7] -0.29970627 -3.28996248
```

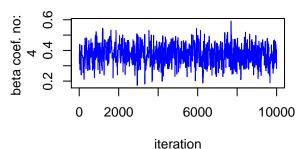
As it can be seen the draw from the M-H algorithm is very similar to the one found before.

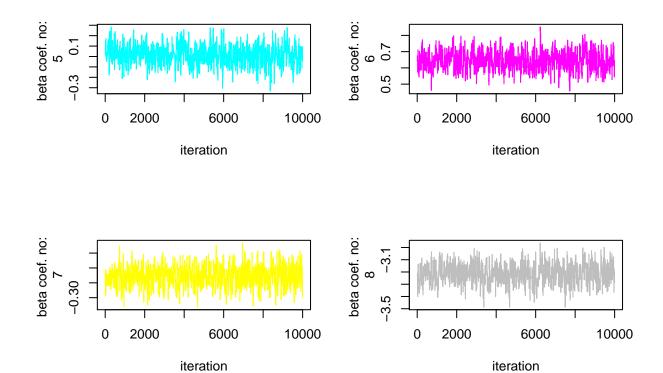
We plot the traceplots for each beta coefficient and then we plot the histograms for the posterior distributions of $\phi_j = \exp \beta_j$.





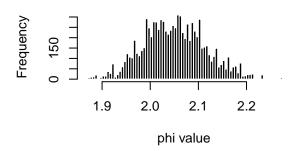




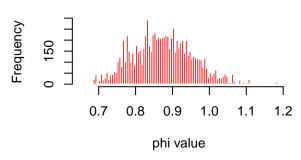


var1 var2 var3 var4 var5 var6 var7 var8 ## 344.2819 370.3795 419.2641 413.6408 397.9017 355.0700 396.7633 374.3520

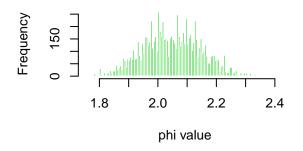
Posterior dist. of phi coef. no:



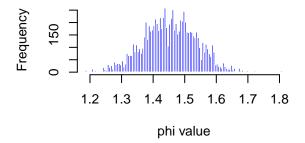
Posterior dist. of phi coef. no: 2

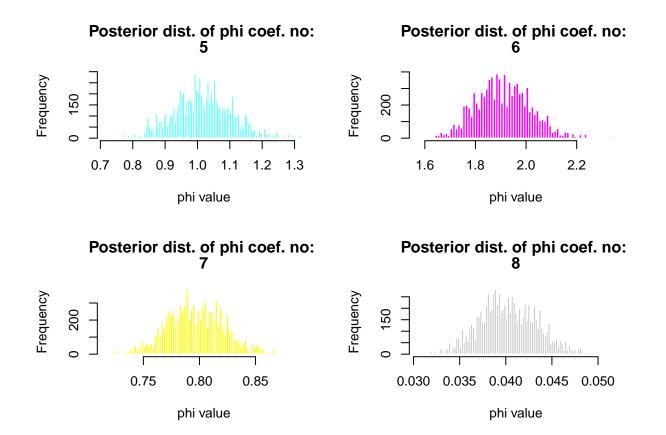


Posterior dist. of phi coef. no: 3



Posterior dist. of phi coef. no:

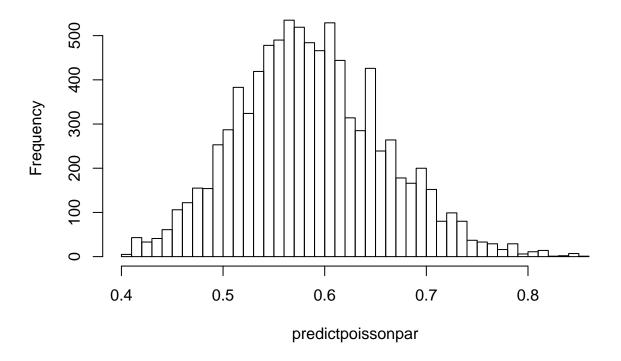




From the trace plots we can see that all the β values seem to converge. Anyways the effective samples size is only about 3.5%.

Finally we plot a histogram over the predictive distribution of poisson parameter λ for the auction given in the lab instructions and calculate the probability that that auction will have zero bids.

Histogram of predictpoissonpar



[1] "Mean probability for no bids on the new auction: 0.558"

We can tell that the coin object and the seller is of good quality, but that the MinBidShare and LogBook values were rather higher than average. These factors balance each other out and we believe that the result is reasonable.