Data Analysis in the Health Sciences

Final Exam 2019 – EPIB–621

INSTRUCTIONS This examination consists of 5 questions on 14 pages, including this one. Tables of the normal distribution are provided on the last page. Please write your answers (NEATLY) in the spaces provided. Fully explain all of your answers. Each question is worth 10 points, for a total of 50. 1 2 3 4 5	Student's Name:
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3 4	
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Total (out of 50)	

1. Some researchers report the following: "In our sample of 225 subjects, the blood pressure was lowered by an average of 5 points (p=0.05 using a two sided t-test)." From the information given, provide an approximate 95% confidence interval for the average amount by which blood pressure was lowered.

In a large cross-sectional study of depression, 10,000 subjects are asked if they are currently having symptoms of depression, along with their age and sex. The following logistic regression model is fit to the data collected:

$$logit(Y) = \alpha + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3$$

where:

Y is equal to 1 if the subject was depressed and 0 otherwise, X1 is the subject's sex, with 0 indicating a male and 1 a female, X2 is a centered variable for age in years, centered at age = 50, and X3 is the square of X2.

The following R output is obtained:

Call:

```
glm(formula = Y ~ X1 + X2 + X3, family = "binomial")
```

Deviance Residuals:

```
Min
          1Q
                             3Q
                                     Max
         -0.1873 -0.0416 -0.0034
                                      3.6007
-0.3677
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)
```

```
(Intercept) -3.326355
                       0.172258 -19.310 < 2e-16 ***
X1
                       0.197050
                                   3.194 0.00140 **
             0.629383
X2
             0.050016
                       0.018532
                                   2.699 0.00696 **
ХЗ
                       0.002243 -7.602 2.91e-14 ***
            -0.017050
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1273.5 on 9999 degrees of freedom Residual deviance: 1030.7 on 9996 degrees of freedom

AIC: 1038.7

Number of Fisher Scoring iterations: 11

(a) Provide an interpretation for the intercept in this model, including any applicable implications for the probability of depression for any particular age and sex combination.

(b) Provide the predicted probability that a 55 year old female will be depressed.

(c) Provide the odds ratio for sex and calculate a 95% confidence interval for this odds ratio.

3. Suppose that a prevalence survey on rheumatoid arthritis is conducted across eight different health regions, with 500 subjects studied within each region for a total sample size of 4,000. The following model is run in WinBUGS, using the data and initial values as given below:

```
model
{
 for (i in 1:8)
 {
 x[i] ~ dbin(p[i],n[i])
 logit(p[i]) \leftarrow z[i]
 z[i] ~ dnorm(mu,tau)
  }
   mu ~ dnorm(0,0.001)
   tau <- 1/(sigma*sigma)
   sigma ~ dunif(0, 20)
      ~ dnorm(mu, tau)
   w \leftarrow \exp(y)/(1+\exp(y))
}
# Data
list(n=c(500, 500, 500, 500, 500, 500, 500, 500),
     x=c(10, 12, 5, 8, 15, 18, 13, 7))
# Initial Values
list(mu=0, sigma=1)
```

In this model n[i] provides the sample sizes sampled within each region, and x[i] is the number of subjects with rheumatoid arthritis out of the n[i] subjects sampled within each region. The following WinBUGS output is produced upon running the above program with the above data and initial values:

node	mean	sd	MC error	2.5%	median	97.5%	start	sample
mu	-3.84	0.1842	0.003059	-4.223	-3.835	-3.497	2001	20000
p[1]	0.02088	0.004717	5.925E-5	0.01223	0.02069	0.03107	2001	20000
p[2]	0.02281	0.005024	5.477E-5	0.01422	0.02226	0.03439	2001	20000
p[3]	0.01653	0.00481	1.212E-4	0.00751	0.01652	0.02579	2001	20000
p[4]	0.01906	0.00457	8.075E-5	0.01042	0.01901	0.02822	2001	20000
p[5]	0.02569	0.00583	8.514E-5	0.01662	0.02486	0.03947	2001	20000
p[6]	0.02881	0.007052	1.37E-4	0.01849	0.02768	0.04549	2001	20000
p[7]	0.02379	0.005231	6.256E-5	0.01514	0.02312	0.03596	2001	20000
p[8]	0.01823	0.00459	9.125E-5	0.0095	0.01831	0.02716	2001	20000
sigma	0.3417	0.2296	0.006373	0.0216	0.3094	0.8796	2001	20000
W	0.02317	0.01546	1.373E-4	0.00829	0.02131	0.05048	2001	20000
У	-3.84	0.4472	0.004359	-4.784	-3.827	-2.934	2001	20000

(a) Ignoring the WinBUGS model that was run, and using the data from region [3] only, calculate the point estimate and an approximate 95% confidence interval for the prevalence of rheumatoid arthritis in region [3].

(b) Now using the results from the WinBUGS model, state the prevalence estimate and 95% credible interval that was estimated for region [3]. Comparing the two point estimates from parts (a) and (b), state which is higher and explain why there is a difference between them.

(c) State the point estimate and 95% credible interval for ${\tt w},$ and explain the information given by this parameter in this model.

4. The following WinBUGS program is created and run, with results given below the program:

```
model
{
    for (i in 1:10) {
     mu[i]
             <- alpha + beta*age[i]</pre>
     y[i]
             ~ dnorm(mu[i],tau)
    }
             ~ dnorm(0.0, 0.001)
    alpha
             ~ dnorm(0.0, 0.001)
    beta
    tau
              <- 1/(sigma*sigma)
           ~ dunif(0.001, 100)
 }
#Data
list(age = c(59, 52, 37, 40, 67, 43, 61, 34, 51, 58),
  y = c(143, 132, 88, NA, 177, 102, 154, NA, 131, 150))
# Results
node
        mean
                 sd
                           2.5%
                                   median
                                            97.5%
        -17.36
                           -38.06
                                   -17.69
                                            5.066
alpha
                 10.6
                            2.434
                                    2.847
                                             3.221
beta
        2.842
                 0.1957
        5.201
                 2.045
                            2.804
                                    4.727
                                             10.32
sigma
y [4]
        96.25
                 6.397
                            83.84
                                    96.16
                                             109.7
y [8]
        79.27
                 7.074
                            65.56
                                    79.14
                                             93.84
```

(a) Explain what the output for y[8] represents, and provide an interpretation for its mean and 95% credible interval.

(b) The same program as in 4 (a) is run again in WinBUGS, but now using the slightly different data set below. The program now terminates with an error message. Explain why.

#Data

```
list(age = c(59, 52, 37, 40, 67, 43, NA, 34, 51, 58),
y = c(143, 132, 88, NA, 177, 102, 154, NA, 131, 150))
```

5. Researchers are deriving an equation to best predict the probability a subject will have osteoporosis based on five predictors, including age (in years), sex (1=female, 0 = male), smoking status (1 indicates a smoker, 0 indicates a non-smoker), family history of osteoporosis (1 = yes, 0 = no) and previous fracture (1 = yes, 0 = no). The researchers collect data on 2000 subjects, and run the big.glm program using a binomial family (i.e., they are running logistic regression).

They find the following results based on their sample of 2000 subjects:

> summary(osteo.dat)

osteo	smoke	age	sex	fam_hist
Min. :0.00	Min. :0.0000	Min. :50.0	Min. :0.000	Min. :0.000
1st Qu.:0.00	1st Qu.:0.0000	1st Qu.:57.0	1st Qu.:0.000	1st Qu.:0.000
Median :1.00	Median :0.0000	Median:65.0	Median :0.000	Median :0.000
Mean :0.52	Mean :0.2955	Mean :65.1	Mean :0.484	Mean :0.255
3rd Qu.:1.00	3rd Qu.:1.0000	3rd Qu.:73.0	3rd Qu.:1.000	3rd Qu.:1.000
Max. :1.00	Max. :1.0000	Max. :80.0	Max. :1.000	Max. :1.000

```
prev_frac
Min. :0.000
1st Qu.:0.000
Median :0.000
Mean :0.202
3rd Qu.:0.000
Max. :1.000
```

> output <- bic.glm(osteo ~ age + sex + smoke + fam_hist + prev_frac,
glm.family = binomial, data=osteo.dat, OR = 2000)</pre>

> summary(output)

Call:

bic.glm.formula(f = osteo ~ age + sex + smoke + fam_hist + prev_frac,
data = osteo.dat, glm.family = binomial, OR = 2000)

4 models were selected Best 4 models (cumulative posterior probability = 1):

	p!=0	EV	SD	model 1	model 2	model 3	model 4
Intercept	100	-3.898	0.3708	-3.924e+00	-3.850e+00	-3.765e+00	-3.689e+00
age	100.0	0.048	0.0054	4.860e-02	4.865e-02	4.809e-02	4.810e-02
sex	100.0	0.973	0.0960	9.760e-01	9.677e-01	9.590e-01	9.507e-01
smoke	100.0	0.669	0.1062	6.714e-01	6.689e-01	6.404e-01	6.380e-01
fam_hist	97.7	0.418	0.1263	4.280e-01	4.271e-01		
prev_frac	69.8	0.252	0.1940	3.623e-01		3.612e-01	
nVar				5	4	4	3
BIC				-1.263e+04	-1.263e+04	-1.262e+04	-1.262e+04
post prob				0.682	0.295	0.016	0.007

(a) Using the coefficients optimized for future predictions, provide the predicted probability of osteoporosis for a female subject aged 65, who smokes, and has both a family history of osteoporosis and had a previous fracture.

(b) State the average rate of osteoporosis in this sample, and compare it to the predicted probability you calculated in part (a). Explain why one of these numbers is higher than the other.

Normal Density Table

	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986
3.0	0.9987	0.9987	0.9987	0.9988	0.9988	0.9989	0.9989	0.9989	0.9990	0.9990

Table of standard normal distribution probabilities. Each number in the table provides the probability that a standard normal random variable will be less than the number indicated.