

LOGISTIC REGRESSION

In logistic regression we assume that we sample units from a population. For each unit we have the values of independent variables and the value of the response variable which is binary, 0 and 1 say. The model assumes that the values of the variables are created as independent draws from a distribution (Y, X_1, \dots, X_m) where we assume that

$$P(Y = 1 | X_1 = x_1, \dots, X_m = x_m) = \frac{\exp(\beta_0 + \sum_{k=1}^n \beta_k x_k)}{1 + \exp(\beta_0 + \sum_{k=1}^n \beta_k x_k)},$$

The parameters $\beta_0, \beta_1, \dots, \beta_m$ are assumed unknown.

The file `logistic.dat` you have the data. Below is the printout from S-PLUS where the parameters are estimated.

```
*** Import Data ***
```

```
Import Successful
```

```
File name: /valjhun/mihael/teaching/istat/verstat/data/logit.dat
```

```
Data name: logit
```

```
Number of rows: 1000
```

```
Number of columns: 3
```

```
Columns:
```

```
  Name    Type
```

```
1 Col1 numeric
```

```
2 Col2 numeric
```

```
3 Col3 numeric
```

```
*** Generalized Linear Model ***
```

```
Call: glm(formula = Col1 ~ Col2 + Col3,  
          + family = binomial(link = logit),  
          + data = logit, na.action = na.exclude,  
          + control = list(epsilon = 0.0001, maxit = 50, trace = F))
```

Deviance Residuals:

	Min	1Q	Median	3Q	Max
	-2.061219	-1.047069	0.5614318	1.049224	2.122089

Coefficients:

	Value	Std. Error	t value
(Intercept)	0.05832206	0.06777698	0.8604996
Col2	0.41705971	0.08130599	5.1295077
Col3	0.56273870	0.07989416	7.0435527

(Dispersion Parameter for Binomial family taken to be 1)

Null Deviance: 1385.81 on 999 degrees of freedom

Residual Deviance: 1252.243 on 997 degrees of freedom

Number of Fisher Scoring Iterations: 3

Correlation of Coefficients:

	(Intercept)	Col2
Col2	0.0623449	
Col3	-0.0199886	-0.3277031

Analysis of Deviance Table

Binomial model

Response: Col1

Terms added sequentially (first to last)

	Df	Deviance	Resid. Df	Resid. Dev
NULL			999	1385.810
Col2	1	79.92338	998	1305.887
Col3	1	53.64399	997	1252.243

- For the given data y_1, y_2, \dots, y_n in $\mathbf{x}_1, \dots, \mathbf{x}_n$ write the (conditional) likelihood function.
- Convince yourself (and me) that the parameters are estimated by max-

imum likelihood.

- c. Compute the Fisher information matrix.
- d. Convince yourself that the standard errors given in the printout are obtained from the Fisher information matrix.
- e. How would you test the hypothesis $H_0: \beta_1 = \beta_2 = 0$?