

Comlex Data - HW #5

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Exercise 1

- 1) In our case $\beta_3 + \beta_5$ represents the log ratio of odds of success (remission) for the new drug versus the standard drug at time = 1, for a particular individual. Namely:

$$\beta_3 + \beta_5 = \log \left(\frac{p_{ND}(1 - p_{ND})}{p_S(1 - p_S)} \right)$$

Where p_{ND} is the probability of remission on the new drug, and p_S is the probability of remission on the standard drug.

Taking the exponent we get that $e^{\beta_3 + \beta_5}$ is the odds ratio. The treatment with the new drug is more effective than the treatment with the standard drug, if that ratio is greater than 1. Since $\beta_3 = -0.05967$, and $\beta_5 = 1.01817$, we have $e^{-0.05967 + 1.01817} = e^{0.9585} = 2.607782$, and so we conclude that the odds of remission on the new drug are 2.607782 times higher than on the standard drug at time = 1.

- 2) Similarly the ratio of odds at time = 2, is equal to $e^{\beta_3 + 2\beta_5} = e^{-0.05967 + 2 \cdot 1.01817} = 7.218665$. So the odds of remission on the new drug are 7.218665 higher than on the standard drug.

- 3) For time = 0, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + b_{i1}$$

For time = 1, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + \beta_4 + b_{i1}$$

For time = 2, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + 2\beta_4 + b_{i1}$$

Therefore for individuals with the same random effect and the same severity, the log ratio of odds is β_4 . So the odds of remission are $e^{0.48274} = 1.620509$ times higher at time = 1, than at time = 0, and 1.620509 times higher at time = 2, than at time = 1.

We can also calculate the values of the logit function for all individuals, using the random effect estimates provided by the function `ranef`.

The table contains the logit function for each time point for groups of individuals with the same random effect and severity:

##	ids	drug	severe	time0	time1	time2
## 1	1 - 16	0	0	-0.024177	0.458563	0.941303
## 2	17 - 38	0	0	-0.027401	0.455339	0.938079
## 3	39 - 41	0	0	-0.030624	0.452116	0.934856
## 4	42 - 55	0	0	-0.027401	0.455339	0.938079
## 5	56 - 74	0	0	-0.030624	0.452116	0.934856
## 6	75 - 80	0	0	-0.033848	0.448892	0.931632
## 10	151 - 152	0	1	-1.336115	-0.853375	-0.370635
## 11	153 - 162	0	1	-1.339339	-0.856599	-0.373859

```
## 12 163 - 171    0    1 -1.342563 -0.859823 -0.377083
## 13 172 - 180    0    1 -1.339339 -0.856599 -0.373859
## 14 181 - 222    0    1 -1.342563 -0.859823 -0.377083
## 15 223 - 250    0    1 -1.345788 -0.863048 -0.380308
```

4) For time = 0, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + \beta_3 + b_{i1}$$

For time = 1, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + \beta_3 + \beta_4 + \beta_5 + b_{i1}$$

For time = 2, we have:

$$\text{logit}\{Pr(Y_{ij} = 1|b_{i1})\} = \beta_1 + \beta_2 \text{severe}_i + \beta_3 + 2(\beta_4 + \beta_5) + b_{i1}$$

For individuals with the same random effect and the same severity, the log ratio of odds is $\beta_4 + \beta_5$. So the odds of remission are $e^{0.48274+1.01817} = 4.485769$ times higher at time = 1, than at time = 0, and 4.485769 times higher at time = 2, than at time = 1.

Values of the logit function for groups of individuals with the same random effect and severity:

```
##          ids drug severe    time0    time1    time2
## 7      81 - 111    1      0 -0.085159 1.415751 2.916661
## 8     112 - 139    1      0 -0.088385 1.412525 2.913435
## 9     140 - 150    1      0 -0.091611 1.409299 2.910209
## 16    251 - 257    1      1 -1.397856 0.103054 1.603964
## 17    258 - 264    1      1 -1.401081 0.099829 1.600739
## 18    265 - 266    1      1 -1.404306 0.096604 1.597514
## 19    267 - 297    1      1 -1.401081 0.099829 1.600739
## 20    298 - 334    1      1 -1.404306 0.096604 1.597514
## 21    335 - 339    1      1 -1.407531 0.093379 1.594289
```

Exercise 2

Let $\text{severe}_i = \text{severe}_k = s$, and $b_{i1} = b_{k1} = b$, then:

- time = 0

Patient on a standard treatment: $\text{logit}\{Pr(Y_{ij} = 1)|b_{i1}\} = \beta_1 + \beta_2 s + b$

Patient on a new treatment: $\text{logit}\{Pr(Y_{kj} = 1)|b_{k1}\} = \beta_1 + \beta_2 s + \beta_3 + b$

The log odds ratio is equal to β_3 . The 95% confidence interval for the odds ratio is:

$$e^{\beta_3 \pm 1.96 SE_{\beta_3}} = e^{-0.05967 \pm 1.96 * 0.22240} = (0.6092211, 1.456788)$$

- time = 1

The log odds ratio is equal to $\beta_3 + \beta_5$. The standard error is equal to: $\sqrt{SE_{\beta_3}^2 + SE_{\beta_5}^2 + 2\text{cov}(\beta_3, \beta_5)}$.

```
vcov(depress.glmr)
```

```
## 5 x 5 Matrix of class "dpoMatrix"
##          (Intercept)      severe      drug      time  drug:time
## (Intercept)  0.026917970 -0.0097407782 -0.0224107779 -0.012769015  0.01452528
## severe      -0.009740778  0.0232897759 -0.0001755429 -0.002178549  -0.00353706
## drug        -0.022410778 -0.0001755429  0.0494598077  0.013477141  -0.03161830
## time        -0.012769015 -0.0021785490  0.0134771412  0.013377365  -0.01245581
## drug:time    0.014525280 -0.0035370599 -0.0316182968 -0.012455814  0.03667169
```

The 95% confidence interval for the odds ratio:

$$e^{\beta_3 + \beta_5 \pm 1.96 SE_{\beta_3 + \beta_5}} = e^{-0.05967 + 1.01817 \pm 1.96 * \sqrt{(-0.05967)^2 + 1.01817^2 - 2 * 0.03162}} = (0.3757544, 18.09833)$$

- time = 2

The log odds ratio is equal to $\beta_3 + 2\beta_5$. The standard error is equal to: $\sqrt{SE_{\beta_3}^2 + 4SE_{\beta_5}^2 + 4cov(\beta_3, \beta_5)}$.

The 95% confidence interval for the odds ratio:

$$e^{\beta_3 + 2\beta_5 \pm 1.96 SE_{\beta_3 + 2\beta_5}} = e^{-0.05967 + 2 * 1.01817 \pm 1.96 * \sqrt{(-0.05967)^2 + 4 * 1.01817^2 - 4 * 0.03162}} = (0.1415707, 368.0783)$$