

## Solution - Sheet 03

Jaya Bharatam 235291

Agatha Anna Baby 235293

Amrutha Manoharan 236892

Nishant Madan 230370

4/21/2022

```
library(nlme)
load("SimulatedTreatmentEffect-3pLL.RData")
```

```
##Part a)
##In 3pLL model the first parameter theta_1=0 for the entire dataset.
model.3pLL.001 <-
  gnls(
    resp ~ (0 + (th4-0) / (1 + (exp((conc - th2) * th3))))),
    data = conc.resp.df,
    params = list(th2 + th3 + th4 ~ 1),
    start = c(1, 2, 100),
    control = gnlsControl(nlsTol = 0.1)
  )
model.3pLL.001
```

```
## Generalized nonlinear least squares fit
## Model: resp ~ (0 + (th4 - 0)/(1 + (exp((conc - th2) * th3))))
## Data: conc.resp.df
## Log-likelihood: -228.6547
##
## Coefficients:
##      th2      th3      th4
## 0.959338 2.305889 99.081280
##
## Degrees of freedom: 68 total; 65 residual
## Residual standard error: 7.143413
```

The lower asymptote,  $\theta_1$ , is 0 and the upper asymptote,  $\theta_4$  is 99.081280, which is nearly close to 100. At  $\theta_2=0.959338$ , we got the half-maximal response which is  $(0.959338+99.081280)/2 = 50.02$ . The parameter  $\theta_3=2.305889 > 0$ , which signifies a decreasing profile for increasing concentrations.

```
##Part b)
##In 3pLL model the first parameter theta_11,theta_12,theta_13=0 for the entire dataset.
dum1 <- ifelse(conc.resp.df$treat == "T1", 1, 0)
dum2 <- ifelse(conc.resp.df$treat == "T2", 1, 0)
dum3 <- ifelse(conc.resp.df$treat == "T3", 1, 0)

model.3pLL.002 <- gnls(
  resp ~ th4 * dum3 +
```

```

      (0 + (th4-0) / (1 + (exp((conc - th2) * th3))))),
data = conc.resp.df,
params = list(th2 + th3 + th4 ~ 1),
start = c(1, 2, 100),
control = gnlsControl(nlsTol = 0.1)
)
model.3pLL.002

```

```

## Generalized nonlinear least squares fit
##   Model: resp ~ th4 * dum3 + (0 + (th4 - 0)/(1 + (exp((conc - th2) * th3))))
##   Data: conc.resp.df
##   Log-likelihood: -300.7103
##
## Coefficients:
##      th2      th3      th4
## 1.271901  3.373962 80.102577
##
## Degrees of freedom: 68 total; 65 residual
## Residual standard error: 20.61115

```

The lower asymptote,  $\theta_1$ , is 0 and the upper asymptote,  $\theta_4$  is 80.102577, which is actually not really close to 100. At  $\theta_2=1.271901$ , we got the half-maximal response which is  $(1.271901+80.102577)/2 = 40.69$ . The parameter  $\theta_3=3.373962 > 0$ , which signifies a decreasing profile for each increasing concentrations.

```

##Part c)
model.3pLL.003 <- gnls(
  resp ~ (0 + (th4-0) / (1 + (exp((conc - th21) * th31)))) * (dum1)
    + (0 + (th4-0) / (1 + (exp((conc - th22) * th32 )))) * (dum2),
  data = conc.resp.df,
  params = list(th21 + th22 + th31 + th32 + th4 ~ 1),
  start = c(1, 1, 2, 2, 100),
  control = gnlsControl(nlsTol = 0.1, apVar = TRUE)
)
model.3pLL.003

```

```

## Generalized nonlinear least squares fit
##   Model: resp ~ (0 + (th4 - 0)/(1 + (exp((conc - th21) * th31)))) * (dum1) +      (0 + (th4 - 0)/(1 +
##   Data: conc.resp.df
##   Log-likelihood: -313.5675
##
## Coefficients:
##      th21      th22      th31      th32      th4
## 0.7304149  1.1886932  2.3739178  2.4848554  99.2201308
##
## Degrees of freedom: 68 total; 63 residual
## Residual standard error: 25.29317

```

The model use parameter  $\theta_4$  is shared across all three treatments and,  $\theta_2$  and  $\theta_3$  are estimated separately for the first and the second treatment. Lower asymptote  $\theta_1$  is zero.  $\theta_2$  0.730,  $\theta_2$  1.189 and the upper asymptote is 99.22, which is nearly close to 100. The parameter  $\theta_{31}=2.3477416 > 0$  and  $\theta_{32}=2.4848554 > 0$ , a higher value for the slope parameter, demonstrates a steeper decrease of the curve.

```
#Part d)
anova(model.3pLL.001, model.3pLL.002)
```

```
##           Model df      AIC      BIC    logLik
## model.3pLL.001    1  4 465.3094 474.1874 -228.6547
## model.3pLL.002    2  4 609.4207 618.2987 -300.7103
```

Both AIC and BIC value increases after adding the treatment 3 in the second model, which further increases p-value but not enough to cross the significance level of 5%, hence we can clearly reject the null hypothesis that both models are equivalent. This indicates that the second model has a significantly better fit and thus, we can difference between all the three treatment groups can be easily seen.