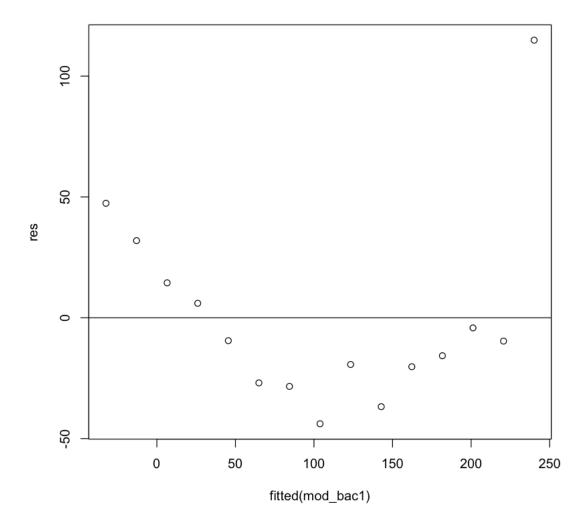
P2

November 6, 2021

```
[11]: #2
      (a)
[12]: data_bac <- read.table("Table6.2.txt",</pre>
                             head = TRUE,
                             sep = "\t")
      #data bac
[13]: mod_bac1 <- lm(N_t ~ t, data_bac)
      print(summary(mod_bac1))
     Call:
     lm(formula = N_t ~ t, data = data_bac)
     Residuals:
         Min
                  1Q Median
                                  3Q
                                         Max
     -43.867 -23.599 -9.652 10.223 114.883
     Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
     (Intercept)
                   259.58
                              22.73 11.420 3.78e-08 ***
                   -19.46
                               2.50 -7.786 3.01e-06 ***
     t
     ___
     Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
     Residual standard error: 41.83 on 13 degrees of freedom
     Multiple R-squared: 0.8234, Adjusted R-squared: 0.8098
     F-statistic: 60.62 on 1 and 13 DF, p-value: 3.006e-06
[14]: res <- resid(mod_bac1)</pre>
      plot(fitted(mod_bac1), res)
      abline(0,0)
```



No. Generally, when the regression model satisfies the "L" assumption, the points on the residual plot are irregular and randomly distributed around the horizontal line 0. However, in situation problem 2 (1), the points on the residual plot are curved.

(b)

```
[15]: mod_bac2 <- lm(log(N_t) ~ t, data_bac)
print(summary(mod_bac2))</pre>
```

```
Call:
lm(formula = log(N_t) ~ t, data = data_bac)
```

Residuals:

Min 1Q Median 3Q Max -0.18445 -0.06189 0.01253 0.05201 0.20021

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 5.973160 0.059778 99.92 < 2e-16 ***
t -0.218425 0.006575 -33.22 5.86e-14 ***

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

Residual standard error: 0.11 on 13 degrees of freedom Multiple R-squared: 0.9884, Adjusted R-squared: 0.9875

F-statistic: 1104 on 1 and 13 DF, p-value: 5.86e-14

Regression line equation:

$$\log\left(n_t\right) = \beta_0 + \beta_1 t + \varepsilon$$

That is

$$Log(n_t) = 5.97 - 0.22t + \varepsilon$$

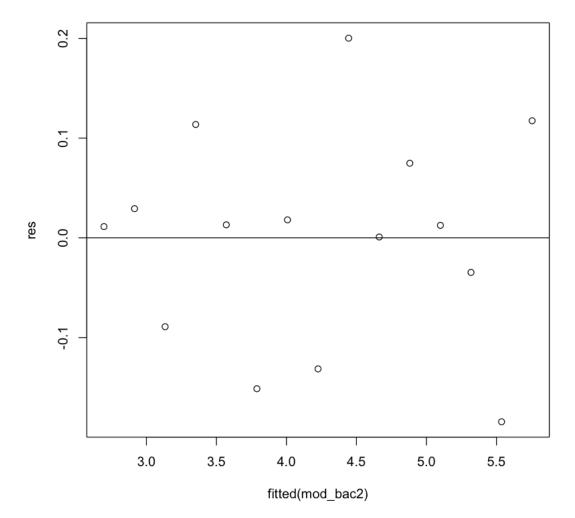
OR

$$n_t = \exp\left(\beta_0 + \beta_1 t\right) + \varepsilon$$

That is

$$n_t = \exp(5.97 - 0.22t) + \varepsilon$$

[10]: res <- resid(mod_bac2)
 plot(fitted(mod_bac2), res)
 abline(0,0)</pre>



The "L" assumption is no longer violated.