NLS_Report

Censoring criteria

```
db %>% filter(CENSOR != "NO") %>% distinct(CMPD, SITE, TIME, CENSOR) %>%
 pivot wider(names from=SITE, values from=TIME, values fn=function(x){
   paste(x, collapse=", ")
 })
## # A tibble: 10 x 6
    CMPD
            CENSOR INJ PC INJ IP `IMM + INJ PC (Combination)`
##
    <fct>
            <chr> <chr> <chr> <chr>
                                                             <chr>
## 1 S-CY3A AGGREG 0.25
                          0.25
                                  0.25
                                                             <NA>
## 2 S-CY5.5A BLOQ <NA> <NA> <NA>
                                                             1, 3
## 3 S-CY5.5A AGGREG 0.25 0.25 0.25
                                                             <NA>
## 4 S-CY5A BLOQ <NA>
                           <NA>
                                  <NA>
                                                             1, 3, 6
## 5 S-CY5A UNEXPL 0.25, 1 0.25, 1 0.25, 1
                                                             <NA>
## 6 FAMA UNEXPL 0.25, 1 0.25, 1 0.25, 1
                                                             <NA>
## 7 TAMRA
            AGGREG 0.25
                        0.25
                                0.25
                                                             <NA>
## 8 R6GA
            BLOQ <NA>
                        <NA>
                                  <NA>
                                                             1, 3, 6
## 9 R6GA AGGREG 0.25 0.25 0.25
                                                             <NA>
## 10 CY3A
            AGGREG 0.25, 1 0.25, 1 0.25, 1
                                                             <NA>
```

Model fit

```
m2 <- lapply(seq along(cmpdDb), function(i) {</pre>
  cat("===== COMPOUND ",cmpdDb[[i]]$CMPD[1], "\n")
  fitDb <- filter(cmpdDb[[i]], CENSOR == "NO")</pre>
  #fitDb <- cmpdDb[[i]]</pre>
  m <- nls(formula,</pre>
     data=fitDb,
      start=start,
      # lower=start*0,
      # upper=start*100,
      # algorithm = "port",
      control=nls.control(maxiter=1000*100, minFactor=1E-24))
 nlstools::overview(m)
} )
## ===== COMPOUND 1
##
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
##
      TIME)))
##
## Parameters:
         Estimate Std. Error t value Pr(>|t|)
        2.318e-10 1.075e-10 2.156 0.0323 *
## Ke 1.208e-02 1.842e-03 6.556 4.79e-10 ***
## FLUOR 8.293e+09 2.472e+08 33.551 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.041 on 196 degrees of freedom
## Number of iterations to convergence: 5
## Achieved convergence tolerance: 9.729e-07
##
## -----
## Residual sum of squares: 213
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       1.975150e-11 4.438884e-10
## Q
## Ke 8.443300e-03 1.570865e-02
## FLUOR 7.805764e+09 8.780731e+09
##
## -----
## Correlation matrix:
                           Ke
                                    FLUOR
       1.000000000 0.3765945 -0.004486099
## Q
        0.376594468 1.0000000 0.561522269
## FLUOR -0.004486099 0.5615223 1.000000000
##
## ===== COMPOUND 2
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        3.426e-10 7.470e-11 4.585 8.57e-06 ***
## Ke 2.291e-02 2.249e-03 10.186 < 2e-16 ***
## FLUOR 2.632e+09 6.954e+07 37.854 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2763 on 176 degrees of freedom
## Number of iterations to convergence: 6
## Achieved convergence tolerance: 6.075e-06
##
## -----
## Residual sum of squares: 13.4
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       1.951194e-10 4.899830e-10
## Q
## Ke 1.846802e-02 2.734361e-02
## FLUOR 2.495141e+09 2.769625e+09
##
## -----
## Correlation matrix:
                          Ke
                                  FLUOR
      1.000000000 0.4476229 0.005580052
## Q
## Ke 0.447622906 1.0000000 0.527399997
## FLUOR 0.005580052 0.5274000 1.000000000
##
## ===== COMPOUND 3
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        1.608e-09 1.520e-10 10.580 < 2e-16 ***
## Ke 1.375e-02 1.927e-03 7.137 5.3e-11 ***
## FLUOR 1.088e+09 3.764e+07 28.918 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.1251 on 135 degrees of freedom
## Number of iterations to convergence: 5
## Achieved convergence tolerance: 9.937e-07
##
## -----
## Residual sum of squares: 2.11
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       1.307518e-09 1.908732e-09
## Q
       9.939665e-03 1.755985e-02
## Ke
## FLUOR 1.013955e+09 1.162826e+09
##
## -----
## Correlation matrix:
                         Ke FLUOR
## Q 1.0000000 0.4368652 -0.1191826
        0.4368652 1.0000000 0.6131015
## FLUOR -0.1191826 0.6131015 1.0000000
##
## ===== COMPOUND 4
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        1.708e-09 2.259e-10 7.560 3.01e-12 ***
## Ke 1.207e-02 2.240e-03 5.391 2.50e-07 ***
## FLUOR 4.352e+09 1.892e+08 23.006 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6056 on 159 degrees of freedom
## Number of iterations to convergence: 5
## Achieved convergence tolerance: 8.29e-06
##
## -----
## Residual sum of squares: 58.3
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       1.261944e-09 2.154404e-09
## Q
## Ke 7.649905e-03 1.649689e-02
## FLUOR 3.978084e+09 4.725239e+09
##
## -----
## Correlation matrix:
                         Ke FLUOR
## Q 1.0000000 0.3709325 -0.1193837
        0.3709325 1.0000000 0.6398915
## FLUOR -0.1193837 0.6398915 1.0000000
##
## ===== COMPOUND 5
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        1.453e-09 2.525e-10 5.754 3.41e-08 ***
## Ke 5.194e-02 4.969e-03 10.452 < 2e-16 ***
## FLUOR 1.970e+10 5.773e+08 34.131 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.891 on 192 degrees of freedom
## Number of iterations to convergence: 6
## Achieved convergence tolerance: 5.478e-06
##
## -----
## Residual sum of squares: 686
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       9.545567e-10 1.950435e-09
## Q
## Ke 4.213601e-02 6.173743e-02
## FLUOR 1.856534e+10 2.084269e+10
##
## -----
## Correlation matrix:
                       Ke
                              FLUOR
## Q 1.0000000 0.5363328 0.1057257
## Ke 0.5363328 1.0000000 0.5779940
## FLUOR 0.1057257 0.5779940 1.0000000
##
## ===== COMPOUND 6
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        3.226e-09 2.945e-10 10.96 < 2e-16 ***
## Ke 1.077e-02 2.373e-03 4.54 1.08e-05 ***
## FLUOR 1.388e+09 5.574e+07 24.90 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2421 on 165 degrees of freedom
## Number of iterations to convergence: 4
## Achieved convergence tolerance: 2.091e-06
##
## -----
## Residual sum of squares: 9.67
##
```

```
## -----
## t-based confidence interval:
                2.5%
                           97.5%
       2.644912e-09 3.807841e-09
## Q
## Ke 6.088411e-03 1.546020e-02
## FLUOR 1.277797e+09 1.497898e+09
##
## -----
## Correlation matrix:
                         Ke FLUOR
      1.0000000 0.5211084 -0.1753926
## Q
        0.5211084 1.0000000 0.5317858
## FLUOR -0.1753926 0.5317858 1.0000000
##
## ==== COMPOUND 7
## -----
\#\# Formula: RFU ~ FLUOR * (Dose * exp(-Ke * TIME) + M * Q/Ke * (1 - exp(-Ke
## TIME)))
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
        8.365e-09 9.226e-10 9.067 4.37e-16 ***
## Ke 4.981e-02 7.564e-03 6.586 6.28e-10 ***
## FLUOR 5.444e+09 3.497e+08 15.565 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9914 on 159 degrees of freedom
## Number of iterations to convergence: 55
## Achieved convergence tolerance: 9.637e-06
##
## -----
## Residual sum of squares: 156
##
```

```
## t-based confidence interval:

## 2.5% 97.5%

## Q 6.542843e-09 1.018712e-08

## Ke 3.487355e-02 6.475101e-02

## FLUOR 4.753014e+09 6.134517e+09

##

##

## -----

## Correlation matrix:

## Q Ke FLUOR

## Q 1.0000000 0.7178605 -0.2070472

## Ke 0.7178605 1.0000000 0.4475268

## FLUOR -0.2070472 0.4475268 1.0000000
```

Coefficients

```
lapply(m2, coef) %>%
bind_rows %>%
mutate(T12 = log(2)/Ke, CMPDName = as.character(unique(db$CMPD))) %>%
pander::pander()
```

Q	Ke	FLUOR	T12	CMPDName
2.318e-10	0.01208	8.293e+09	57.4	S-CY3A
3.426e-10	0.02291	2.632e+09	30.26	S-CY5.5A
1.608e-09	0.01375	1.088e+09	50.41	S-CY5A
1.708e-09	0.01207	4.352e+09	57.41	FAMA
1.452e-09	0.05194	1.97e+10	13.35	TAMRA
3.226e-09	0.01077	1.388e+09	64.33	R6GA
8.365e-09	0.04981	5.444e+09	13.92	CY3A

Figures

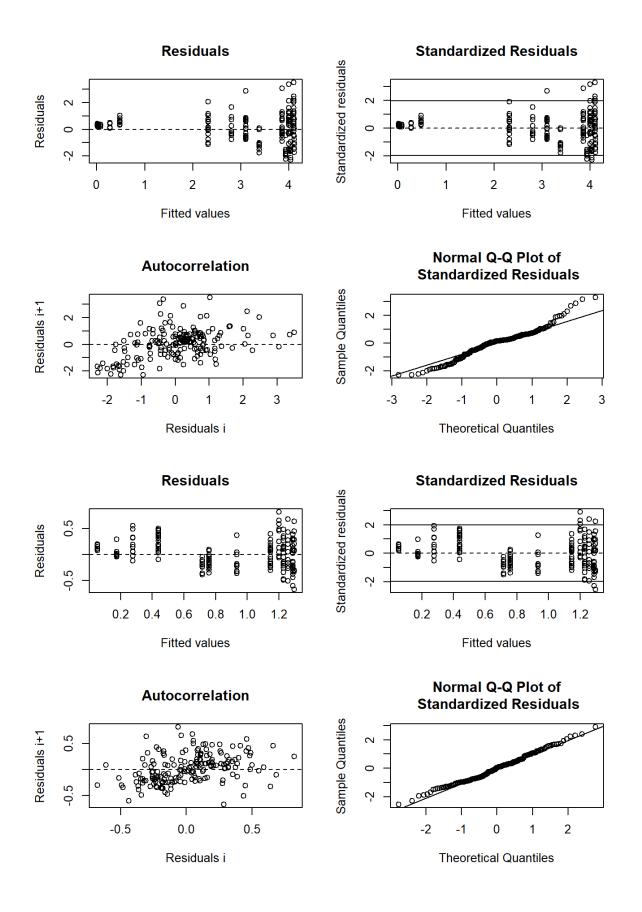
```
for(i in seq_along(cmpdDb)) {
    #cmpdDb[[i]]$IRES <- (cmpdDb[[i]]$IPRED - cmpdDb[[i]]$RFU)

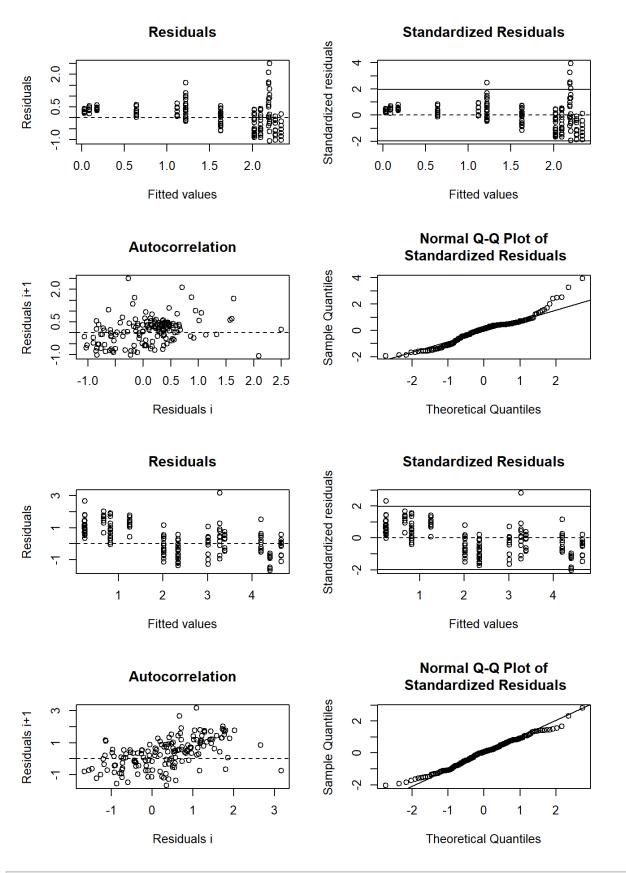
    j <- cmpdDb[[i]]$CENSOR == "NO"

    res <- nlstools::nlsResiduals(m2[[i]])

    cmpdDb[[i]]$IRES <- cmpdDb[[i]]$RFU - predict(m2[[i]], newdata=cmpdDb[[i]])

    cmpdDb[[i]]$IWRES <- NA
    cmpdDb[[i]]$IWRES[j] <- res$resi2[,2]
    cmpdDb[[i]]$ARES[j] <- res$resi3[,2]
    cmpdDb[[i]]$ARES[j] <- res$resi3[,2]
    cmpdDb[[i]]$MODEL <- list(m2[[i]])
    plot(res)
}</pre>
```

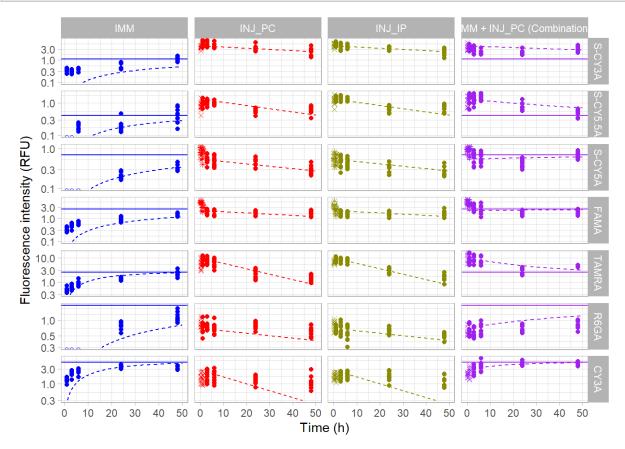




predictDb <- lapply(seq_along(m2), function(i) {
 predictDb <- cmpdDb[[i]] %>%

```
distinct(CMPD, MW, SITE, Dose, M, HAS IMM) %>%
    expand grid (TIME=seq (0, 50, by=0.01))
  predictDb$RFU <- predict(m2[[i]], newdata=predictDb)</pre>
  limitDb <- cmpdDb[[i]] %>% filter(HAS IMM) %>%
    distinct(CMPD, MW, SITE, Dose, M, HAS IMM) %>%
   mutate(TIME=99999)
  limitDb$RFU LIM <- predict(m2[[i]], newdata=limitDb)</pre>
  predictDb %>% left join(limitDb %>% select(-TIME))
}) %>% bind rows %>%
  left join(data.frame(
    CMPD=factor(c("S-CY3A", "S-CY5.5A", "S-CY5A", "FAMA", "TAMRA", "R6GA",
"CY3A")),
    MIN=c(0.10, 0.10, 0.10, 0.10, 0.3, 0.3, 0.3)
  )) %>% filter(RFU > MIN)
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = c("CMPD", "MW", "SITE", "HAS IMM", "Dose", "M")
## Joining, by = "CMPD"
result <- bind rows(cmpdDb)</pre>
resultIres <- result %>%
 group by (CMPD) %>%
  filter(CENSOR == "NO") %>%
  summarize(SIGMA = summary(MODEL[1][[1]])$sigma,
         M IRES = mean(IRES, na.rm=TRUE))
## `summarise()` ungrouping output (override with `.groups` argument)
# ggplot(result, aes(x=TIME, y=RFU, color=SITE)) +
# geom point() +
  geom line(aes(y=PRED)) +
# facet grid(CMPD~SITE) +
# scale y log10()
```

```
ggplot(result, aes(x=TIME, y=RFU, color=SITE)) +
  geom hline(data=predictDb, aes(yintercept=RFU LIM, color=SITE), linetype=
1) +
  geom point(aes(shape=CENSOR)) +
  geom line(data=predictDb, linetype=2) +
  facet grid(CMPD~SITE, scales="free") +
  #facet wrap(CMPD~., scales="free") +
  scale y log10() +
  #scale x continuous(minor breaks=c(1, 3, 6, 24, 48, 0.25)) +
  scale shape manual(values=c(NO=16, AGGREG=4, UNEXPL=8, BLOQ=1, `IMM+INJ`=
3)) +
  labs(x="Time (h)", y="Fluorescence intensity (RFU)") +
  theme light() +
  theme(legend.position="none") +
  scale color manual (values=c(IMM="blue", INJ PC="red", INJ IP="yellow4",
IMM + INJ PC (Combination) `="purple"))
## Warning: Transformation introduced infinite values in continuous y-axis
## Warning: Removed 68868 rows containing missing values (geom hline).
```



```
ggplot(result %>% filter(CENSOR != "BLOQ"), aes(x=TIME, y=IRES, color=SITE)
) +
```

```
geom_point(aes(shape=CENSOR)) +
  facet_grid(CMPD~SITE, scales="free_y") +
  geom_hline(yintercept=0) +
  geom_hline(data=resultIres, aes(yintercept=M_IRES), linetype=2) +
  geom_hline(data=resultIres, aes(yintercept=M_IRES+1.96*SIGMA), linetype=2) +
  geom_hline(data=resultIres, aes(yintercept=M_IRES-1.96*SIGMA), linetype=2) +
  #geom_smooth(se=F) +
  labs(x="Time (h)", y="Residuals") +
  scale_shape_manual(values=c(NO=16, AGGREG=4, UNEXPL=8, BLOQ=1, `IMM+INJ`=3)) +
  theme_light() +
  theme(legend.position="none") +
  scale_color_manual(values=c(IMM="blue", INJ_PC="red", INJ_IP="yellow4", `IMM + INJ_PC (Combination)`="purple"))
```

AUC_0,48h calculation

```
predictDb <- lapply(seq along(m2), function(i) {</pre>
  predictDb <- cmpdDb[[i]] %>%
    distinct(CMPD, MW, SITE, Dose, M, HAS IMM) %>%
    expand grid(TIME=seq(0, 48, by=0.01))
  predictDb$RFU <- predict(m2[[i]], newdata=predictDb)</pre>
  predictDb
}) %>% bind rows
sumDb <- predictDb %>% group by(CMPD, SITE) %>%
  summarize(AUC = sum(RFU*0.01)) %>%
  pivot wider(names from=SITE, values from=AUC)
## `summarise()` regrouping output by 'CMPD' (override with `.groups` argum
ent)
sumDb$RE 0 48 = sumDb$IMM / sumDb$INJ PC
sumDb$Fcom 0 48 = sumDb$`IMM + INJ PC (Combination)` / sumDb$IMM
sumDb$CMPD <- 1:7</pre>
sumDb1 <- sumDb
```

<pre>pander::pander(sumDb)</pre>	
ranaci. Fanaci (camza)	

CMPD	IMM	INJ_PC	INJ_IP	IMM + INJ_PC (Combination)]
1	12.75	151.1	151.1	163.8	
2	7.839	38.33	38.33	46.17	
3	8.97	19.13	19.13	28.1	
4	29.45	79.28	79.28	108.7	
5	78.17	174.1	174.1	252.2	
6	20.23	26.01	26.01	46.25	
7	144.3	49.66	49.66	193.9	

AUC_0,3h calculation

```
predictDb <- lapply(seq_along(m2), function(i) {
  predictDb <- cmpdDb[[i]] %>%
    distinct(CMPD, MW, SITE, Dose, M, HAS_IMM) %>%
    expand_grid(TIME=seq(0, 3, by=0.01))
  predictDb$RFU <- predict(m2[[i]], newdata=predictDb)
  predictDb
}) %>% bind_rows

sumDb <- predictDb %>% group_by(CMPD, SITE) %>%
  summarize(AUC = sum(RFU*0.01)) %>%
  pivot_wider(names_from=SITE, values_from=AUC)

## `summarise()` regrouping output by 'CMPD' (override with `.groups` argument)

sumDb$RE_0_3 = sumDb$IMM / sumDb$INJ_PC
  sumDb$Fcom_0_3 = sumDb$`IMM + INJ_PC (Combination)` / sumDb$IMM

sumDb$CMPD <- 1:7</pre>
```

```
sumDb2 <- sumDb
pander::pander(sumDb)</pre>
```

CMPD	IMM	INJ_PC	INJ_IP	IMM + INJ_PC (Combination)
1	0.05934	12.26	12.26	12.32
2	0.04196	3.829	3.829	3.871
3	0.0427	1.605	1.605	1.647
4	0.1371	6.432	6.432	6.569
5	0.5739	27.46	27.46	28.03
6	0.09252	2.055	2.055	2.148
7	1.038	7.61	7.61	8.648

Multiple Linera Regression

```
## -1.736e-09 -1.149e-09 -3.845e-10 4.796e-10 3.460e-09
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.005e-09 6.919e-10 2.898 0.0274 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.831e-09 on 6 degrees of freedom
m1 \leftarrow step(lm(RE 0 3 \sim 0, data=fitSet),
    scope= ~ `Log P` + MW + HBA + HBD + TPSA + rotor + MR,
    direction="both",
    k=2
    trace=9999 )
## Start: AIC=-40.23
## RE 0 3 ~ 0
##
##
           Df Sum of Sq RSS AIC
## + MR
           1 0.0091346 0.013223 -41.902
## + rotor
            1 0.0072523 0.015105 -40.971
## + MW 1 0.0071810 0.015176 -40.938
           1 0.0056798 0.016677 -40.277
## + HBD
## <none>
                       0.022357 -40.226
## + `Log P` 1 0.0051271 0.017230 -40.049
## + TPSA
           1 0.0019726 0.020385 -38.872
## + HBA
            1 0.0014877 0.020869 -38.708
##
## Step: AIC=-41.9
## RE 0 3 ~ MR - 1
##
           Df Sum of Sq RSS AIC
##
## + `Log P` 1 0.0099352 0.0032873 -49.645
## + HBA 1 0.0053132 0.0079094 -43.499
## + TPSA
            1 0.0050338 0.0081888 -43.256
## + MW
            1 0.0043494 0.0088732 -42.694
## <none>
                        0.0132226 -41.902
## + rotor 1 0.0014744 0.0117482 -40.730
```

```
## - MR 1 0.0091346 0.0223572 -40.226
## + HBD 1 0.0001184 0.0131042 -39.965
## Step: AIC=-49.65
## RE 0 3 \sim MR + `Log P` - 1
           Df Sum of Sq RSS AIC
##
                      0.0032873 -49.645
## <none>
## + rotor 1 0.0007943 0.0024930 -49.581
           1 0.0001523 0.0031351 -47.977
## + HBA
           1 0.0000730 0.0032144 -47.802
## + TPSA
## + HBD 1 0.0000411 0.0032462 -47.733
           1 0.0000175 0.0032698 -47.683
## + MW
## - `Log P` 1 0.0099352 0.0132226 -41.902
## - MR 1 0.0139427 0.0172301 -40.049
summary(m1)
##
## Call:
## lm(formula = RE \ 0 \ 3 \sim MR + \ Log \ P \ - 1, data = fitSet)
##
## Residuals:
                  2 3 4 5 6
         1
## 0.015642 -0.003146 -0.001514 -0.003694 -0.029571 -0.026485 0.037960
## Coefficients:
         Estimate Std. Error t value Pr(>|t|)
## MR 2.699e-04 5.861e-05 4.605 0.00581 **
## `Log P` 3.035e-02 7.808e-03 3.887 0.01156 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.02564 on 5 degrees of freedom
## Multiple R-squared: 0.853, Adjusted R-squared: 0.7941
## F-statistic: 14.5 on 2 and 5 DF, p-value: 0.00829
```