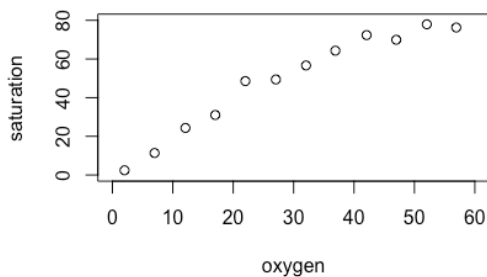


```
oxygen <- c(2.01, 6.98, 12.09, 17.03, 22.01, 27.06, 32.06, 36.91, 42.08, 46.99,
52.05, 56.92)
saturation <- c(2.42, 11.37, 24.33, 31.03, 48.57, 49.41, 56.66, 64.29, 72.36,
69.94, 77.94, 76.28)
d <- data.frame(oxygen, saturation)
```

#1. plotting the data

```
plot(saturation ~ oxygen, data = d, xlim = c(0, 60), ylim = c(0, 80))
```

it may have cooperativity. The curve seems to increase faster than expected when the oxygen concentration is high.



#2. Fit the data to Michaelis-Menten model

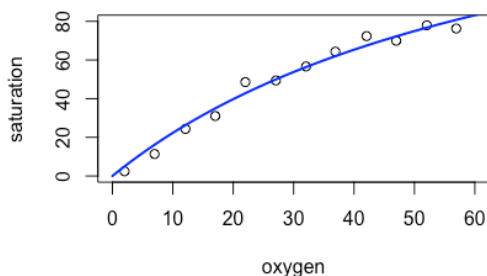
```
m0.oxytrans <- nls(saturation ~ Vmax * oxygen / (Km + oxygen),
start = list(Vmax = 80, Km = 20),
data = d)
```

```
summary(m0.oxytrans)
```

plot the regressed curve against the original data points

```
x <- 0:80
```

```
lines(x, predict(m0.oxytrans, newdata = data.frame(oxygen = x)), col = "blue",
lwd = 2)
```

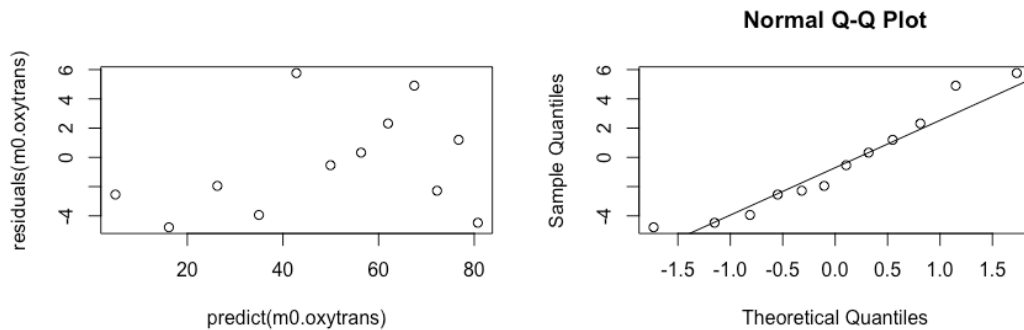


do the QC of residuals

```
plot(residuals(m0.oxytrans) ~ predict(m0.oxytrans))
```

```
qqnorm(residuals(m0.oxytrans))
```

```
qqline(residuals(m0.oxytrans))
```



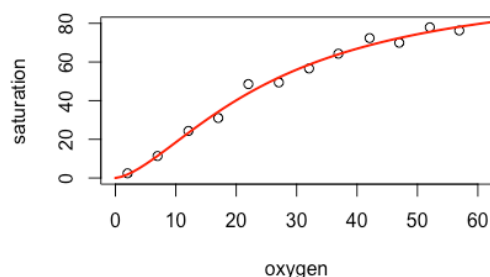
```
shapiro.test(residuals(m0.oxytrans))
# p-value = 0.4136 # it seems the residuals follow normal distribution
```

```
# compute the 95% CI for the paramters
confint(m0.oxytrans)
##      2.5%    97.5%
#Vmax 137.23915 284.6987
#Km    44.58951 134.9181
```

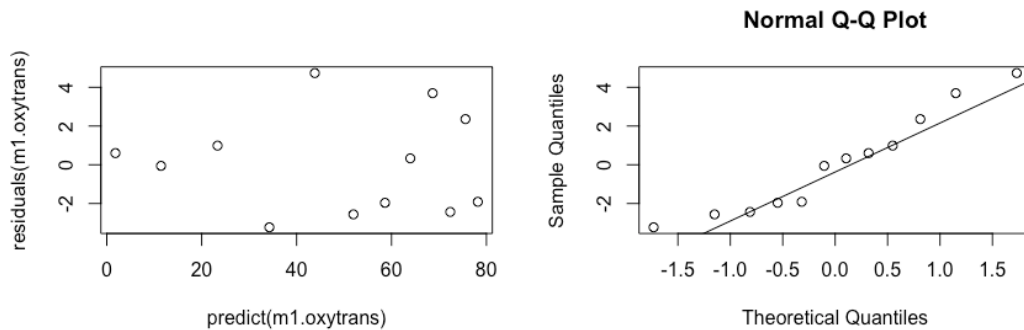
#Comment: it seems that the model underestimates the saturation when oxygen level is low, and the CI ranges are wide. The model is not good enough.

```
#3. Fit the data to a Hill model
m1.oxytrans <- nls(saturation ~ Vmax * oxygen ^ n / (Km ^ n + oxygen ^ n),
                  start = list(Vmax = 80, Km = 20, n = 1),
                  data = d)
summary(m1.oxytrans)

# plot the regressed curve against the original data points
plot(saturation ~ oxygen, data = d, xlim = c(0, 60), ylim = c(0, 80))
x <- 0:80
lines(x, predict(m1.oxytrans, newdata = data.frame(oxygen = x)), col = "red", lwd
      = 2)
```



```
# do the QC of residuals
plot(residuals(m1.oxytrans) ~ predict(m1.oxytrans))
qqnorm(residuals(m1.oxytrans))
qqline(residuals(m1.oxytrans))
```



```
shapiro.test(residuals(m1.oxytrans))
```

```
# p-value = 0.4063 # it seems the residuals follow normal distribution
```

```
# compute the 95% CI for the paramters
```

```
confint(m1.oxytrans)
```

```
##          2.5%      97.5%
#Vmax 83.848271 156.016477
#Km   20.131075  53.139889
#n     1.101171   2.103657
```

#Comment: The data fits the Hill model better. $n > 1$ shows a positive cooperativity. The CI ranges are narrower than the previous one.

```
#4. F-test to compare the two models
```

```
anova(m0.oxytrans, m1.oxytrans)
```

```
#   Res.Df Res.Sum Sq Df Sum Sq F value    Pr(>F)
```

```
#1      10    138.382
```

```
#2       9     73.513  1 64.869  7.9417 0.02011 *
```

F value > 1, $p < 0.05$, showing Hill model explains better than Michaelis-Menten model

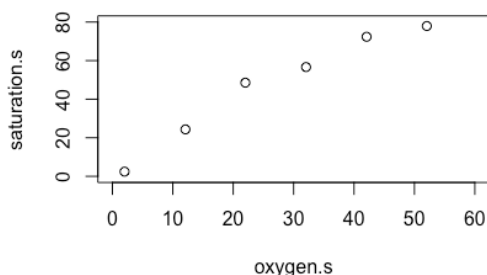
```
#5. repeat the analysis with a subset of the data
```

```
oxygen.s <- c(2.01, 12.09, 22.01, 32.06, 42.08, 52.05)
```

```
saturation.s <- c(2.42, 24.33, 48.57, 56.66, 72.36, 77.94)
```

```
d.s <- data.frame(oxygen.s, saturation.s)
```

```
plot(saturation.s ~ oxygen.s, data = d.s, xlim = c(0, 60), ylim = c(0, 80))
```



```
# Fit the data to Michaelis-Menten model
```

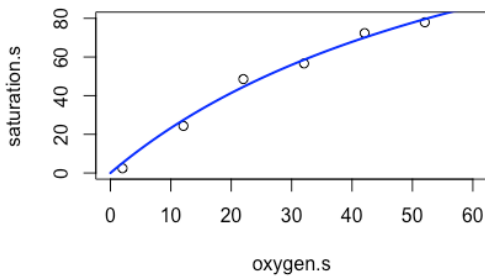
```
m0.oxytrans.s <- nls(saturation.s ~ Vmax * oxygen.s / (Km + oxygen.s),
```

```

start = list(Vmax = 80, Km = 20),
data = d.s)
summary(m0.oxytrans.s)

# plot the regressed curve against the original data points
x <- 0:80
lines(x, predict(m0.oxytrans.s, newdata = data.frame(oxygen.s = x)), col =
"blue", lwd = 2)

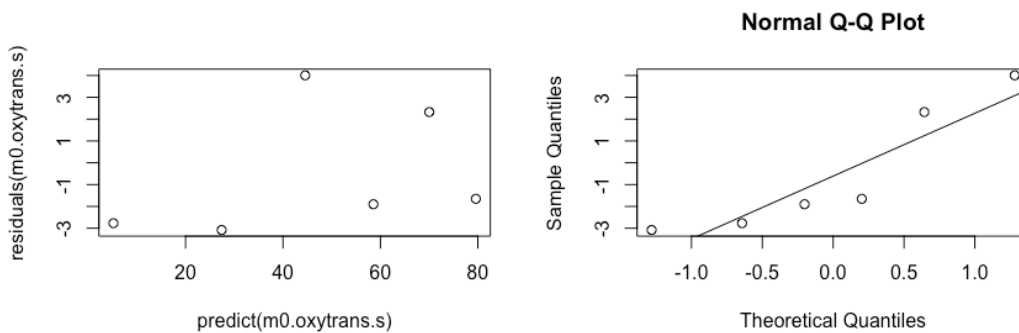
```



```

# do the QC of residuals
plot(residuals(m0.oxytrans.s) ~ predict(m0.oxytrans.s))
qqnorm(residuals(m0.oxytrans.s))
qqline(residuals(m0.oxytrans.s))

```



```

shapiro.test(residuals(m0.oxytrans.s))
# p-value = 0.1121

```

```

confint(m0.oxytrans.s)
##          2.5%    97.5%
#Vmax 122.78224 494.7246
#Km    34.40765 252.1068

```

```

#Fit the data to a Hill model
m1.oxytrans.s <- nls(saturation.s ~ Vmax * oxygen.s ^ n / (Km ^ n + oxygen.s ^
n),
start = list(Vmax = 80, Km = 20, n = 1),
data = d.s)
summary(m1.oxytrans.s)

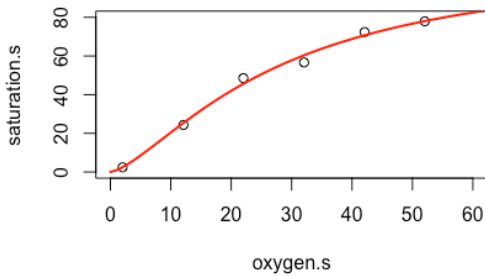
```

```

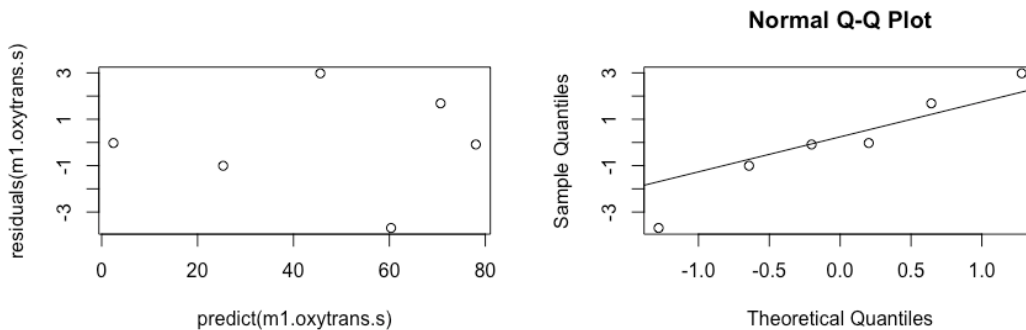
# plot the regressed curve against the original data points
plot(saturation.s ~ oxygen.s, data = d.s, xlim = c(0, 60), ylim = c(0, 80))

```

```
x <- 0:80
lines(x, predict(m1.oxytrans.s, newdata = data.frame(oxygen.s = x)), col = "red",
      lwd = 2)
```



```
# do the QC of residuals
plot(residuals(m1.oxytrans.s) ~ predict(m1.oxytrans.s))
qqnorm(residuals(m1.oxytrans.s))
qqline(residuals(m1.oxytrans.s))
```



```
shapiro.test(residuals(m1.oxytrans.s))
# p-value = 0.876
```

```
# compute the 95% CI for the parameters
confint(m1.oxytrans.s)
#Waiting for profiling to be done...
#Error in prof$getProfile() :
#step factor 0.000488281 reduced below 'minFactor' of 0.000976562
```

```
# switch the algorithm: don't do much help here.
m1.oxytrans.s1 <- nls(saturation.s ~ Vmax * oxygen.s ^ n / (Km ^ n + oxygen.s ^
n),
                     start = list(Vmax = 80, Km = 20, n = 1),
                     data = d.s, algorithm = "plinear") # show an error
summary(m1.oxytrans.s1)
confint(m1.oxytrans.s1)

m1.oxytrans.s2 <- nls(saturation.s ~ Vmax * oxygen.s ^ n / (Km ^ n + oxygen.s ^
n),
                     start = list(Vmax = 80, Km = 20, n = 1),
                     data = d.s, algorithm = "port")
summary(m1.oxytrans.s2)
```

```
confint(m1.oxytrans.s2)
# 2.5% 97.5%
#Vmax 81.69619 NA
#Km 20.51137 NA
#n NA 2.792693
```

```
# F-test to compare the two models
anova(m0.oxytrans.s, m1.oxytrans.s)
# Res.Df Res.Sum Sq Df Sum Sq F value Pr(>F)
#1 4 44.972
#2 3 26.341 1 18.631 2.1219 0.2412
```

Comment: F value is not significant larger than 1. P value > 0.05, suggesting the Michelis-Menten model works better

It's difficult to estimate the CI because there are not enough degrees of freedom (df = 3 for the Hill Model). In t test, with few degrees of freedom, the values of t distribution are much higher than the corresponding values for a normal distribution.

My interpretation: the data is noisy, and t is too large to the CI. We need to collect more data points to get a (narrower) CI.