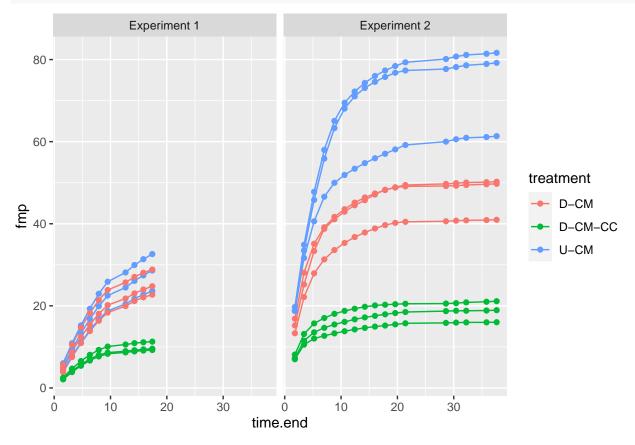
Data analysis for 4-methyl phenol from digestate experiments

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Plots

```
ggplot(dat, aes(time.end, fmp, colour = treatment, group = interaction(tunnel, treatment))) +
  geom_line() +
  geom_point() +
  facet_wrap(~ experiment)
```



Stats

Set reference to untreatmented cattle manure.

```
dat$treatment <- factor(dat$treatment, levels = c('U-CM', 'D-CM', 'D-CM-CC'))
dat$experiment <- factor(dat$experiment)</pre>
```

Unit of analysis will be wind tunnel plot.

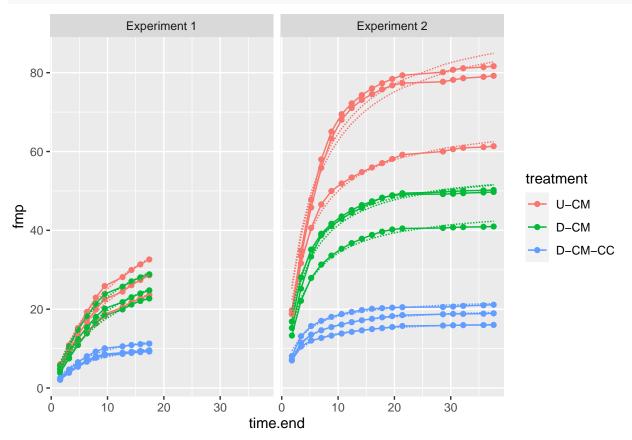
First fit model to each wind tunnel to get the ks.

Add calculated values to data frame.

```
dat <- merge(dat, lmods, by = c('experiment', 'treatment', 'tunnel'))
dat$fmp.calc <- dat$mm * dat$time.end / (dat$time.end + dat$km)</pre>
```

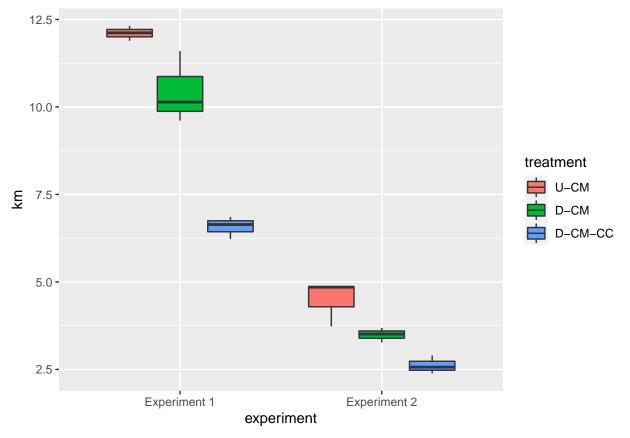
Look at fit.

```
ggplot(dat, aes(time.end, fmp, colour = treatment, group = interaction(tunnel, treatment))) +
geom_line() +
geom_point() +
geom_line(aes(time.end, fmp.calc), lty = '11') +
facet_wrap(~ experiment)
```



Take a look at km values.

```
ggplot(lmods, aes(experiment, km, fill = treatment)) +
  geom_boxplot()
```



Now analysis.

```
modk1 <- lm(km ~ treatment * experiment, data = lmods)</pre>
summary.aov(modk1)
##
                        Df Sum Sq Mean Sq F value
                                                   Pr(>F)
                                    21.10
                                          72.25 2.03e-07 ***
## treatment
                         2 42.19
## experiment
                        1 171.80
                                  171.80 588.38 1.45e-11 ***
                                           19.59 0.000166 ***
## treatment:experiment 2 11.44
                                     5.72
## Residuals
                        12
                             3.50
                                     0.29
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary(modk1)
##
## Call:
## lm(formula = km ~ treatment * experiment, data = lmods)
## Residuals:
       Min
                 1Q
                     Median
##
## -0.84027 -0.22806 0.01571 0.26198 1.15075
##
## Coefficients:
##
                                           Estimate Std. Error t value Pr(>|t|)
                                                        0.3120 38.807 5.52e-14
## (Intercept)
                                            12.1068
## treatmentD-CM
                                            -1.6571
                                                        0.4412 -3.756 0.00274
## treatmentD-CM-CC
                                            -5.5330
                                                        0.4412 -12.541 2.95e-08
```

```
## experimentExperiment 2
                                           -7.6197
                                                       0.4412 -17.271 7.68e-10
## treatmentD-CM:experimentExperiment 2
                                            0.6594
                                                       0.6239
                                                               1.057 0.31141
## treatmentD-CM-CC:experimentExperiment 2
                                            3.6635
                                                       0.6239
                                                                5.872 7.58e-05
##
## (Intercept)
## treatmentD-CM
                                           **
## treatmentD-CM-CC
## experimentExperiment 2
                                           ***
## treatmentD-CM:experimentExperiment 2
## treatmentD-CM-CC:experimentExperiment 2 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.5404 on 12 degrees of freedom
## Multiple R-squared: 0.9847, Adjusted R-squared: 0.9783
## F-statistic: 154.4 on 5 and 12 DF, p-value: 1.848e-10
Interactions complicated. Let's look by experiment. First experiment 1.
modexp1 <- aov(km ~ treatment, data = lmods, subset = experiment == 'Experiment 1')</pre>
summary(modexp1)
              Df Sum Sq Mean Sq F value
## treatment
               2 48.38 24.191
                                  59.98 0.000108 ***
## Residuals
               6
                   2.42
                          0.403
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary.lm(modexp1)
##
## Call:
## aov(formula = km ~ treatment, data = lmods, subset = experiment ==
##
       "Experiment 1")
##
## Residuals:
                 1Q
                      Median
       Min
## -0.84027 -0.31048 0.01001 0.20598 1.15075
## Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
##
                                0.3667 33.019 5.13e-08 ***
## (Intercept)
                    12.1068
                                0.5185 -3.196
## treatmentD-CM
                    -1.6571
                                                0.0187 *
## treatmentD-CM-CC -5.5330
                                0.5185 -10.670 4.00e-05 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6351 on 6 degrees of freedom
## Multiple R-squared: 0.9524, Adjusted R-squared: 0.9365
## F-statistic: 59.98 on 2 and 6 DF, p-value: 0.0001081
coef(modexp1)
                      treatmentD-CM treatmentD-CM-CC
##
        (Intercept)
          12.106822
##
                          -1.657076
                                           -5.532987
```

```
confint(modexp1)
                        2.5 %
##
                                 97.5 %
## (Intercept)
                    11.209638 13.0040052
## treatmentD-CM
                    -2.925885 -0.3882665
## treatmentD-CM-CC -6.801796 -4.2641780
model.tables(modexp1, type = 'means')
## Tables of means
## Grand mean
##
## 9.710134
##
## treatment
## treatment
     U-CM
             D-CM D-CM-CC
## 12.107 10.450 6.574
Use this model in paper. Both D-CM and CC-CM have lower k than reference in experiment 1.
Experiment 2 next.
modexp2 <- aov(km ~ treatment, data = lmods, subset = experiment == 'Experiment 2')</pre>
summary(modexp2)
##
              Df Sum Sq Mean Sq F value Pr(>F)
               2 5.250 2.6252
## treatment
                                 14.53 0.00501 **
               6 1.084 0.1806
## Residuals
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary.lm(modexp2)
##
## aov(formula = km ~ treatment, data = lmods, subset = experiment ==
       "Experiment 2")
##
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                       Max
## -0.7572 -0.2184 0.0214 0.2832 0.3935
##
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                     4.4871
                                0.2454 18.286 1.72e-06 ***
## treatmentD-CM
                     -0.9977
                                0.3470 -2.875 0.02825 *
## treatmentD-CM-CC -1.8695
                                0.3470 -5.387 0.00168 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.425 on 6 degrees of freedom
## Multiple R-squared: 0.8289, Adjusted R-squared: 0.7718
## F-statistic: 14.53 on 2 and 6 DF, p-value: 0.00501
coef (modexp2)
```

treatmentD-CM treatmentD-CM-CC

##

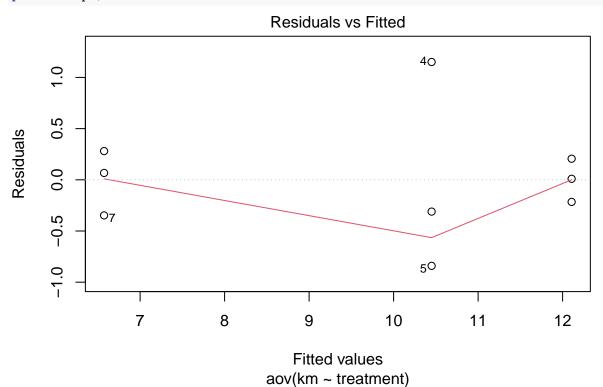
(Intercept)

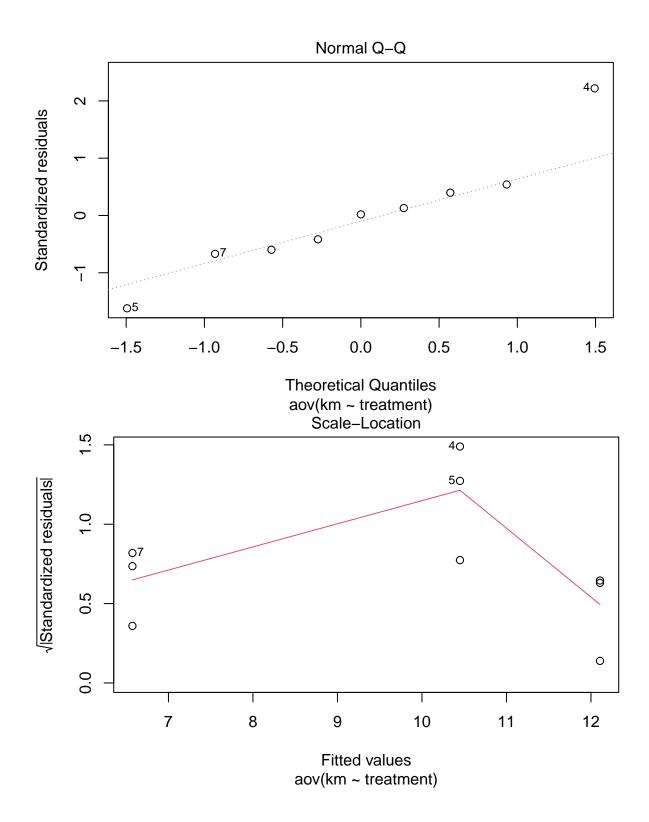
```
##
          4.4870881
                           -0.9976792
                                            -1.8694707
confint(modexp2)
                                   97.5 %
##
                         2.5 %
## (Intercept)
                     3.886645 5.0875311
## treatmentD-CM
                    -1.846834 -0.1485246
## treatmentD-CM-CC -2.718625 -1.0203161
model.tables(modexp2, type = 'means')
## Tables of means
## Grand mean
##
## 3.531371
##
##
    treatment
##
  treatment
      U-CM
              D-CM D-CM-CC
##
##
     4.487
             3.489
                     2.618
```

Both D-CM and CC-CM have lower k than reference in experiment 2. Use this model in paper for experiment 2.

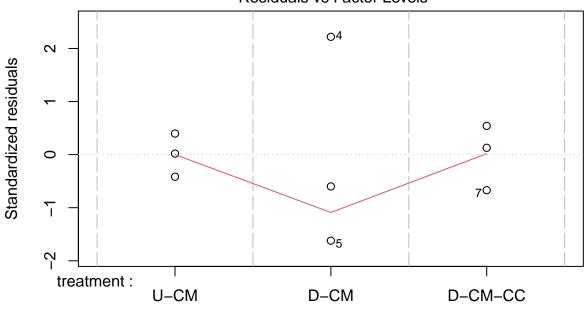
Diagnostic plots.

```
plot(modexp1, ask = FALSE)
```



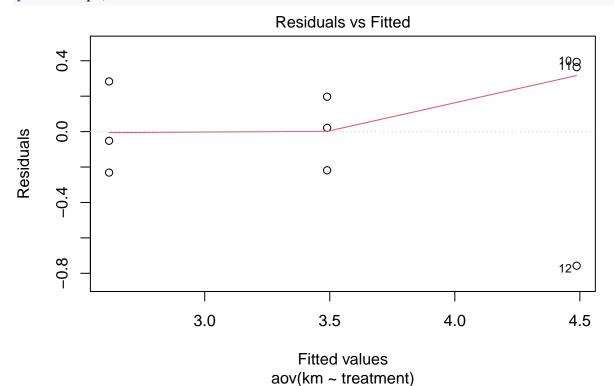


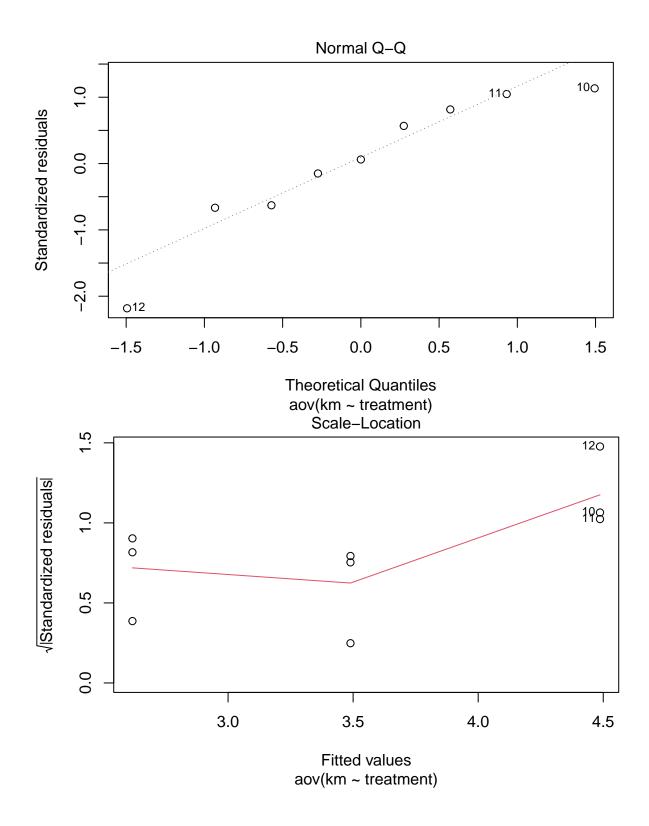
Constant Leverage: Residuals vs Factor Levels



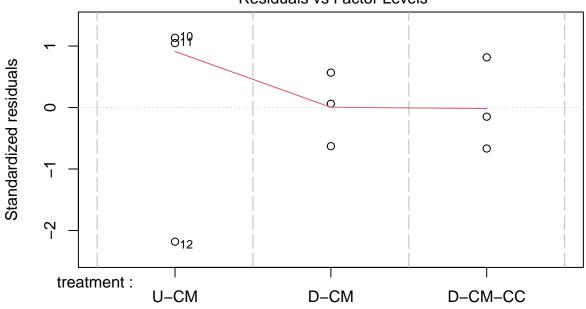
Factor Level Combinations

plot(modexp2, ask = FALSE)





Constant Leverage: Residuals vs Factor Levels



Factor Level Combinations

Not terrible.