Stats 1: Effect of digestion on composition and more

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Vars

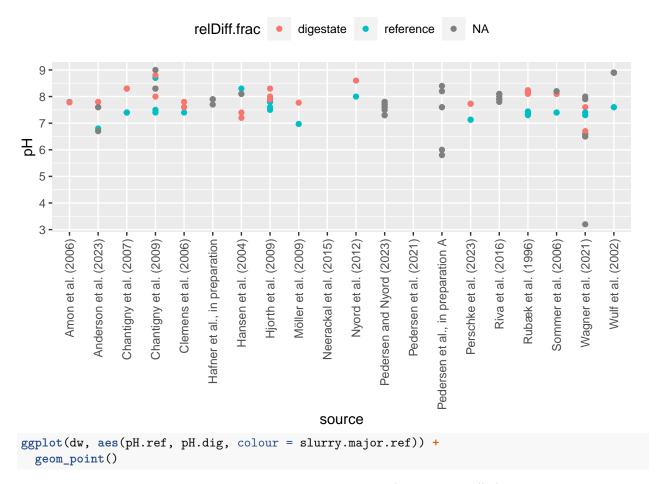
```
Difference in pH due to digestion.
```

```
dw$dpH <- dw$pH.dig - dw$pH.ref</pre>
dw$dDM <- dw$DM.dig - dw$DM.ref</pre>
table(dw$source)
##
##
                             Anderson et al. (2023) Chantigny et al. (2007)
        Amon et al. (2006)
##
##
   Chantigny et al. (2009)
                              Clemens et al. (2006)
                                                         Hansen et al. (2004)
##
##
      Hjorth et al. (2009)
                                Möller et al. (2009) Neerackal et al. (2015)
##
                             Pedersen et al. (2021)
##
       Nyord et al. (2012)
                                                       Perschke et al. (2023)
##
##
       Rubæk et al. (1996)
                                Sommer et al. (2006)
                                                         Wagner et al. (2021)
##
##
        Wulf et al. (2002)
##
```

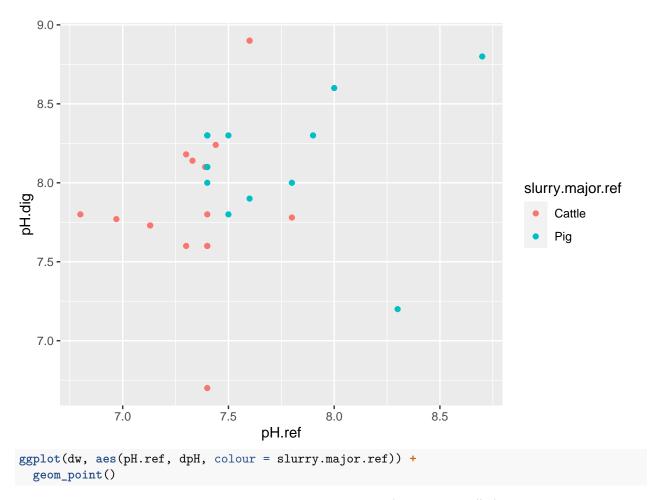
Plots

```
ggplot(dat, aes(source, pH, colour = relDiff.frac)) +
  geom_point() +
  theme(legend.position = 'top', axis.text.x = element_text(angle = 90, vjust = 0.5, hjust = 1))
```

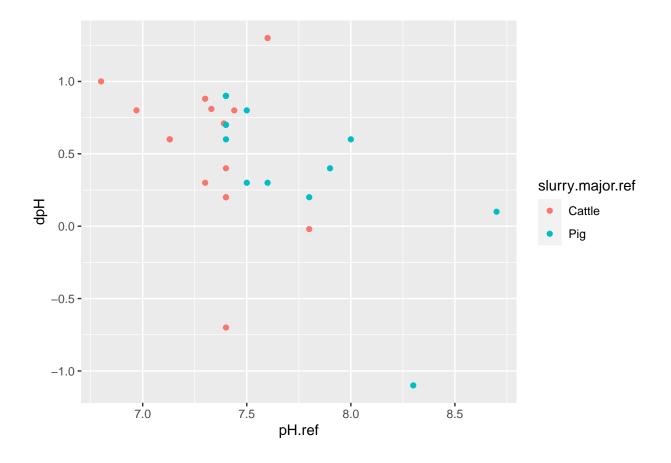
Warning: Removed 14 rows containing missing values (`geom_point()`).



Warning: Removed 8 rows containing missing values (`geom_point()`).



Warning: Removed 8 rows containing missing values (`geom_point()`).



Stats

```
m1 <- lm(pH.dig ~ pH.ref + slurry.major.ref, data = dw)</pre>
summary(m1)
##
## lm(formula = pH.dig ~ pH.ref + slurry.major.ref, data = dw)
##
## Residuals:
                     Median
       Min
                 1Q
                                   ЗQ
                                            Max
## -1.22736 -0.20649 0.03326 0.23480 0.91961
## Coefficients:
                      Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                        5.9652
                                   1.9543
                                            3.052 0.00505 **
## pH.ref
                        0.2652
                                   0.2659
                                            0.997 0.32753
## slurry.major.refPig
                        0.1378
                                   0.1939
                                            0.711 0.48317
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
\#\# Residual standard error: 0.4623 on 27 degrees of freedom
     (8 observations deleted due to missingness)
## Multiple R-squared: 0.09634, Adjusted R-squared: 0.0294
## F-statistic: 1.439 on 2 and 27 DF, p-value: 0.2547
```

```
m2 <- lm(pH.dig ~ pH.ref, data = dw)
summary(m2)
##
## Call:
## lm(formula = pH.dig ~ pH.ref, data = dw)
##
## Residuals:
                 1Q Median
##
       Min
                                   3Q
                                           Max
## -1.28218 -0.17555 0.07506 0.27243 0.84633
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                5.3369
                          1.7277
                                    3.089
                                            0.0045 **
                0.3575
                           0.2300
                                    1.554
## pH.ref
                                            0.1314
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4582 on 28 degrees of freedom
     (8 observations deleted due to missingness)
                                 Adjusted R-squared: 0.04654
## Multiple R-squared: 0.07942,
## F-statistic: 2.416 on 1 and 28 DF, p-value: 0.1314
Try robust regression.
m3 <- MASS::rlm(pH.dig ~ pH.ref + slurry.major.ref, data = dw)
summary(m3)
## Call: rlm(formula = pH.dig ~ pH.ref + slurry.major.ref, data = dw)
## Residuals:
                   1Q
                         Median
                                       3Q
## -1.226190 -0.255524 0.008057 0.208057 0.902945
##
## Coefficients:
                      Value Std. Error t value
## (Intercept)
                      5.3042 1.3889
                                        3.8189
## pH.ref
                      0.3543 0.1890
                                        1.8750
## slurry.major.refPig 0.1658 0.1378
                                        1.2030
## Residual standard error: 0.3501 on 27 degrees of freedom
     (8 observations deleted due to missingness)
m4 <- MASS::rlm(pH.dig ~ pH.ref, data = dw)
summary(m4)
##
## Call: rlm(formula = pH.dig ~ pH.ref, data = dw)
## Residuals:
                 1Q
                     Median
                                   3Q
## -1.30077 -0.19951 0.07298 0.22510 0.80426
## Coefficients:
              Value Std. Error t value
## (Intercept) 4.4867 1.2497
                                3.5902
## pH.ref
              0.4749 0.1664
                                2.8544
```

```
##
## Residual standard error: 0.3313 on 28 degrees of freedom
     (8 observations deleted due to missingness)
```

Digestate pH does seem correlated with raw pH but only with robust regression. Issue seems to be a decrease in change in pH at higher raw pH. So post digestion pH seems to be the same regardless of raw pH. Seems plausible. Say low raw pH is caused by a lot of VFAs, which then have no effect on digestate pH.

```
Simpler question: how does digestion change pH and DM?
t.test(dw$dpH)
##
##
    One Sample t-test
##
## data: dw$dpH
## t = 5.5511, df = 29, p-value = 5.492e-06
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 0.3258861 0.7061139
## sample estimates:
## mean of x
       0.516
##
Clearly pH does increase, according to a one-sample t-test. Mixed-effects model more appropriate.
m5pH <- lmer(dpH ~ (1|source), data = dw)
summary(m5pH)
## Linear mixed model fit by REML ['lmerMod']
## Formula: dpH ~ (1 | source)
##
      Data: dw
##
## REML criterion at convergence: 32.6
##
## Scaled residuals:
##
        Min
                  1Q
                      Median
                                     3Q
                                             Max
  -2.40843 -0.24053 0.08254 0.28675
##
##
## Random effects:
## Groups
                          Variance Std.Dev.
             Name
##
   source
             (Intercept) 0.30699 0.5541
                          0.05266 0.2295
## Residual
## Number of obs: 30, groups: source, 14
##
## Fixed effects:
               Estimate Std. Error t value
## (Intercept)
                 0.4670
                             0.1554
                                      3.005
confint(m5pH)
## Computing profile confidence intervals ...
##
                   2.5 %
                             97.5 %
## .sig01
               0.3426621 0.8435668
## .sigma
               0.1673777 0.3470989
## (Intercept) 0.1502382 0.7812450
```

Compare among animal types.

```
m6pH <- lmer(dpH ~ slurry.major.ref + (1|source), data = dw)</pre>
summary(m6pH)
## Linear mixed model fit by REML ['lmerMod']
## Formula: dpH ~ slurry.major.ref + (1 | source)
##
      Data: dw
##
## REML criterion at convergence: 32.5
## Scaled residuals:
               1Q Median
##
       Min
                                 3Q
                                        Max
## -2.4381 -0.2670 0.1201 0.2397 1.9352
##
## Random effects:
## Groups
             Name
                         Variance Std.Dev.
## source
             (Intercept) 0.32301 0.5683
## Residual
                         0.05229 0.2287
## Number of obs: 30, groups: source, 14
##
## Fixed effects:
##
                       Estimate Std. Error t value
## (Intercept)
                         0.5678
                                     0.2108
                                              2.693
## slurry.major.refPig -0.2347
                                     0.3211 -0.731
## Correlation of Fixed Effects:
##
               (Intr)
## slrry.mjr.P -0.657
Not clearly smaller for pig.
But we don't care about standard deviation estimates, so simpler to explain mean by study.
dws <- aggregate2(dw, c('dpH', 'dDM'), by = 'source', FUN = list(mean = mean, n = function(x) sum(!is.n</pre>
t.test(dws$dpH.mean)
##
   One Sample t-test
##
##
## data: dws$dpH.mean
## t = 4.558, df = 11, p-value = 0.0008194
\#\# alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 0.2907323 0.8337121
## sample estimates:
## mean of x
## 0.562222
And repeat for DM:
t.test(dw$dDM)
##
##
    One Sample t-test
##
## data: dw$dDM
## t = -6.6101, df = 35, p-value = 1.224e-07
```

```
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -2.909077 -1.542034
## sample estimates:
## mean of x
## -2.225556
Clearly DM does increase, according to a one-sample t-test. Mixed-effects model more appropriate.
m5DM <- lmer(dDM ~ (1|source), data = dw)
summary(m5DM)
## Linear mixed model fit by REML ['lmerMod']
## Formula: dDM ~ (1 | source)
##
      Data: dw
##
## REML criterion at convergence: 130.5
##
## Scaled residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
## -3.9984 -0.0775 -0.0079 0.1895 1.5780
##
## Random effects:
## Groups Name
                         Variance Std.Dev.
## source
           (Intercept) 2.7817
## Residual
                         0.9728
                                  0.9863
## Number of obs: 36, groups: source, 16
##
## Fixed effects:
               Estimate Std. Error t value
## (Intercept) -2.1976
                            0.4555 -4.824
confint(m5DM)
## Computing profile confidence intervals ...
##
                    2.5 %
                             97.5 %
## .sig01
                1.0683971 2.496403
## .sigma
                0.7464718 1.379687
## (Intercept) -3.1143616 -1.277186
Compare among animal types.
m6DM <- lmer(dDM ~ slurry.major.ref + (1|source), data = dw)</pre>
summary(m6DM)
## Linear mixed model fit by REML ['lmerMod']
## Formula: dDM ~ slurry.major.ref + (1 | source)
##
      Data: dw
##
## REML criterion at convergence: 125.8
##
## Scaled residuals:
##
       Min
               1Q Median
                                3Q
                                       Max
## -4.0063 -0.1773 0.0730 0.1991 1.6032
##
## Random effects:
## Groups
                         Variance Std.Dev.
           Name
```

```
## source (Intercept) 2.3977 1.5485
## Residual
                      0.9613 0.9805
## Number of obs: 36, groups: source, 16
##
## Fixed effects:
##
                    Estimate Std. Error t value
## (Intercept)
                    -2.7972 0.5416 -5.164
## slurry.major.refPig 1.5894
                             0.8828 1.800
##
## Correlation of Fixed Effects:
             (Intr)
## slrry.mjr.P -0.614
```

Some evidence change is smaller for pigs.

We don't care about standard deviation estimates, so simpler to explain mean by study.

t.test(dws\$dDM.mean)

```
##
## One Sample t-test
##
## data: dws$dDM.mean
## t = -5.4297, df = 13, p-value = 0.0001152
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -3.400268 -1.464613
## sample estimates:
## mean of x
## -2.43244
```