Experimental Data Analysis

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Contents

Packages

Data

Morphometrics & Hydration

Treatment Groups

variables: - individual lizard ID - temp_tmt_C = temperature treatment - humidity_tmt_percent = humidity treatment (high/low, not actually %) - trial_number = which set of lizards that individual was from - conclusion = how that individual's experiment ended (died, canceled, or complete)

```
tmts <- read.csv("./data/exp_tmt_assignment.csv")
```

Capture Data

variables: - date = date of capture & baseline measurements - individual lizard ID - mass_g = mass in grams - hematocrit_percent = % of blood sample that's red blood cells - osmolality_mmol_kg = concentration of solutes in blood plasma - type = when the measurements were taken along the course of the experiment (all on capture day)

```
capture_hydration <- read.csv("./exported_data/capture_hydration.csv",
                              na.strings=c("","NA") # fix empty cells
                              ) %>%
  mutate(# correctly format date-only variable
         date = as.Date(date, format = "%Y-%m-%d")
         ) %>%
  # select only relevant variables
  dplyr::select(date, individual_ID,
                mass_g, hematocrit_percent, osmolality_mmol_kg
                ) %>%
  dplyr::filter(individual_ID %in% tmts$individual_ID) %>%
  mutate(type = as.factor("capture"))
summary(capture_hydration)
```

```
##       date            individual_ID        mass_g      hematocrit_percent
##  Min.   :2021-04-19   Min.   : 31.00   Min.   : 8.20   Min.   :16.00
##  1st Qu.:2021-04-26   1st Qu.: 57.00   1st Qu.:11.10   1st Qu.:32.75
##  Median :2021-04-26   Median : 78.00   Median :12.65   Median :36.00
##  Mean   :2021-04-29   Mean   : 77.46   Mean   :12.18   Mean   :36.08
##  3rd Qu.:2021-05-03   3rd Qu.: 98.25   3rd Qu.:13.32   3rd Qu.:39.00
##  Max.   :2021-05-10   Max.   :122.00   Max.   :15.00   Max.   :54.00
##  osmolality_mmol_kg      type
##  Min.   :319.0      capture:52
##  1st Qu.:354.2
##  Median :373.0
##  Mean   :373.7
##  3rd Qu.:392.2
##  Max.   :423.0
```

extract SVL data separately from capture data:

```
SVL <- read.csv("./exported_data/capture_hydration.csv",
                              na.strings=c("","NA") # fix empty cells
                              ) %>%
  dplyr::select(individual_ID, SVL_mm) %>%
  dplyr::filter(individual_ID %in% tmts$individual_ID)
summary(SVL)
```

```
##  individual_ID        SVL_mm
##  Min.   : 31.00   Min.   :59.00
##  1st Qu.: 57.00   1st Qu.:65.00
##  Median : 78.00   Median :68.00
##  Mean   : 77.46   Mean   :67.62
##  3rd Qu.: 98.25   3rd Qu.:70.00
##  Max.   :122.00   Max.   :73.00
```

extract capture CEWL cloacal temperature separately:

```
cap_CT <- read.csv("./exported_data/capture_hydration.csv",
                              na.strings=c("","NA") # fix empty cells
```

```
                              ) %>%
  dplyr::select(individual_ID, cloacal_temp_C) %>%
  dplyr::filter(individual_ID %in% tmts$individual_ID)
summary(cap_CT)
```

```
##  individual_ID    cloacal_temp_C
##  Min.   : 31.00   Min.   :20.00
##  1st Qu.: 57.00   1st Qu.:22.00
##  Median : 78.00   Median :24.00
##  Mean   : 77.46   Mean   :23.68
##  3rd Qu.: 98.25   3rd Qu.:25.00
##  Max.   :122.00   Max.   :28.00
##                   NA's   :2
```

Experiment Data

variables: - date = date of measurements - individual lizard ID - mass_g = mass in grams - hematocrit_percent = % of blood sample that's red blood cells - osmolality_mmol_kg = concentration of solutes in blood plasma - type = when the measurements were taken along the course of the experiment (either during experimental treatment or after rehab)

```
exp_dat <- read.csv("./data/experimental_data.csv",
                              na.strings=c("","NA") # fix empty cells
                              ) %>%
                # format date
  dplyr::mutate(date = as.Date(date, format = "%m/%d/%y"),
                type = as.factor(type)
                ) %>%
  # select only variables to be analyzed
  dplyr::select(date, individual_ID, mass_g,
                hematocrit_percent, type,
                osmolality_mmol_kg = osmolality_mmol_kg_replicate_mean)
summary(exp_dat)
```

```
##       date            individual_ID        mass_g       hematocrit_percent
##  Min.   :2021-04-21   Min.   : 31.00   Min.   : 6.700   Min.   :12.0
##  1st Qu.:2021-04-28   1st Qu.: 51.25   1st Qu.: 9.875   1st Qu.:23.0
##  Median :2021-05-07   Median : 87.50   Median :11.250   Median :28.0
##  Mean   :2021-05-06   Mean   : 77.85   Mean   :11.076   Mean   :27.8
##  3rd Qu.:2021-05-14   3rd Qu.:101.25   3rd Qu.:12.225   3rd Qu.:33.0
##  Max.   :2021-05-20   Max.   :122.00   Max.   :14.700   Max.   :43.0
##                                                         NA's   :19
##     type    osmolality_mmol_kg
##  exp  :98   Min.   :298.0
##  rehab:34   1st Qu.:342.0
##             Median :355.0
##             Mean   :360.1
##             3rd Qu.:374.8
##             Max.   :441.0
##             NA's   :22
```

Join Dataframes

Now, attach all the dataframes, only use individuals whose treatment was completed, and add a "day" variable for what day of treatment each lizard/observation was on. I also calculate SMI using the equation created in

capture_analysis.

```
all_dat <- exp_dat %>%
  # join data
  rbind(capture_hydration) %>%
  # add tmt group info
  left_join(tmts, by = "individual_ID") %>%
  dplyr::select(-notes) %>%
  # add SVL value for each obs of each indiv.
  # for computing BCI and scaled mass indices
  left_join(SVL, by = "individual_ID") %>%
  # only use completed experiment runs
  dplyr::filter(conclusion == "complete") %>%
  group_by(individual_ID) %>%
  # reformat a lot of variables
  mutate(capture_date = min(date),
         day = as.numeric(date - capture_date),
         humidity_tmt_percent = as.factor(humidity_tmt_percent),
         individual_ID = as.factor(individual_ID),
         temp_tmt_C = as.factor(temp_tmt_C),
         trial_number = as.factor(trial_number),
         conclusion = as.factor(conclusion),
         SMI = mass_g * ((65.02158/SVL_mm) ^ (3.09059/sqrt(0.8944)))
         )

summary(all_dat)
```

```
##       date            individual_ID     mass_g      hematocrit_percent
##  Min.   :2021-04-19   37     :  6   Min.   : 6.70   Min.   :12.00
##  1st Qu.:2021-04-30   39     :  6   1st Qu.:10.20   1st Qu.:24.00
##  Median :2021-05-07   40     :  6   Median :11.50   Median :30.00
##  Mean   :2021-05-06   49     :  6   Mean   :11.27   Mean   :29.58
##  3rd Qu.:2021-05-13   52     :  6   3rd Qu.:12.60   3rd Qu.:35.00
##  Max.   :2021-05-20   47     :  5   Max.   :15.00   Max.   :54.00
##                       (Other):116                   NA's   :12
##       type    osmolality_mmol_kg temp_tmt_C humidity_tmt_percent trial_number
##  exp    :82   Min.   :298.0      25:151     dry  :74             1:35
##  rehab  :34   1st Qu.:342.8                 humid:77             2:24
##  capture:35   Median :359.0                                      3:44
##               Mean   :362.5                                      4:48
##               3rd Qu.:379.0
##               Max.   :441.0
##               NA's   :15
##     conclusion      SVL_mm       capture_date             day
##  complete:151   Min.   :59.00   Min.   :2021-04-19   Min.   : 0.000
##                 1st Qu.:66.00   1st Qu.:2021-04-26   1st Qu.: 2.000
##                 Median :68.00   Median :2021-05-03   Median : 4.000
##                 Mean   :67.45   Mean   :2021-04-30   Mean   : 5.424
##                 3rd Qu.:70.00   3rd Qu.:2021-05-10   3rd Qu.: 9.000
##                 Max.   :73.00   Max.   :2021-05-10   Max.   :11.000
##
##       SMI
##  Min.   : 7.343
##  1st Qu.: 8.990
##  Median :10.011
```

```
##  Mean   : 9.983
##  3rd Qu.:10.751
##  Max.   :13.970
##
```

re-order some factors:

```
all_dat$humidity_tmt_percent <- factor(all_dat$humidity_tmt_percent,
                                       levels = c("humid", "dry"),
                                       labels = c("Humid", "Dry"))
```

make a sub-dataframe without rehab data to prevent any mix-ups:

```
all_dat_no_rehab <- all_dat %>%
  dplyr::filter(type != "rehab")
```

Checks

Dates:

```
# check that capture dates are valid
unique(all_dat$capture_date)
```

```
## [1] "2021-04-19" "2021-04-26" "2021-05-03" "2021-05-10"
```

Check that each lizard only has an accurate number of measurements.

```
all_dat %>%
  group_by(individual_ID, type) %>%
  summarise(n = n()) %>%
  arrange(type)
```

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

```
## # A tibble: 104 x 3
## # Groups:   individual_ID [35]
##    individual_ID type      n
##    <fct>         <fct> <int>
##  1 37            exp       4
##  2 39            exp       4
##  3 40            exp       4
##  4 47            exp       4
##  5 49            exp       4
##  6 52            exp       4
##  7 54            exp       2
##  8 61            exp       2
##  9 66            exp       2
## 10 73            exp       2
## # ... with 94 more rows
```

That all looks good, experimental measurements are either 4 (first trial) or 2 (other trials). I am excluding lizards that died in treatment from the analysis.

CEWL

Capture CEWL

variables: - date = date of capture & baseline measurements - individual lizard ID - region = which body area the measurement was taken from - TEWL_g_m2h = evaporative water loss - cloacal_temp_C = taken

at measurement; influences CEWL

```
cap_CEWL <- read.csv("./exported_data/capture_CEWL.csv") %>%
  dplyr::select(date, individual_ID, region, TEWL_g_m2h) %>%
  mutate(#individual_ID = as.factor(individual_ID), # do later
         date = as.Date(date, format = "%Y-%m-%d"),
         region = as.factor(region),
         day = as.factor("before"), # might change... tbd
         n_day = 0
         ) %>%
  dplyr::filter(individual_ID %in% all_dat$individual_ID) %>%
  left_join(cap_CT, by = 'individual_ID')
summary(cap_CEWL)
```

```
##       date             individual_ID     region     TEWL_g_m2h        day
##  Min.   :2021-04-19   Min.   : 37.00   dewl:32   Min.   : 7.48   before:163
##  1st Qu.:2021-04-26   1st Qu.: 73.00   dors:33   1st Qu.:20.54
##  Median :2021-05-03   Median : 95.00   head:33   Median :27.43
##  Mean   :2021-05-02   Mean   : 87.46   mite:32   Mean   :29.30
##  3rd Qu.:2021-05-10   3rd Qu.:104.00   vent:33   3rd Qu.:36.91
##  Max.   :2021-05-10   Max.   :122.00             Max.   :62.94
##      n_day   cloacal_temp_C
##  Min.   :0   Min.   :20.00
##  1st Qu.:0   1st Qu.:22.00
##  Median :0   Median :24.00
##  Mean   :0   Mean   :23.84
##  3rd Qu.:0   3rd Qu.:25.00
##  Max.   :0   Max.   :28.00
```

Post-Experiment CEWL

In the future, I could automate this like I did for the HOBO data.

Load in each of the post-rehab datafiles:

```
# trial 1
CEWL_t1 <- read.csv("./data/post_exp_CEWL/4-28-21-CEWL.csv", # filename
                    na.strings=c("","NA")) %>% # fix empty cells
  # rename and select the pertinent variables/cols
  # I have to do this for each one
  # so they all have the same number of columns for joining
  dplyr::select(date = Date,
                Status,
                ID = Comments,
                TEWL_g_m2h = TEWL..g..m2h.. # rename
                )

# trial 2
CEWL_t2 <- read.csv("./data/post_exp_CEWL/5-4-21-CEWL.csv",
                    na.strings=c("","NA")) %>%
  dplyr::select(date = Date,
                Status,
                ID = Comments,
                TEWL_g_m2h = TEWL..g..m2h..
                )
```

```
# trial 3
CEWL_t3 <- read.csv("./data/post_exp_CEWL/5-11-21-CEWL.csv",
                           na.strings=c("","NA")) %>%
  dplyr::select(date = Date,
                Status,
                ID = Comments,
                TEWL_g_m2h = TEWL..g..m2h..
                )

# trial 4
CEWL_t4 <- read.csv("./data/post_exp_CEWL/5-18-21-CEWL.csv",
                           na.strings=c("","NA")) %>%
  dplyr::select(date = Date,
                Status,
                ID = Comments,
                TEWL_g_m2h = TEWL..g..m2h..
                )
```

Load in cloacal temperatures:

```
exp_CT <- read.csv("./data/post_exp_CEWL_cloacal_temps.csv") %>%
  mutate(date = as.Date(date, format = "%Y/%m/%d")) %>%
  dplyr::select(-time)
summary(exp_CT)
```

```
##       date            individual_ID    cloacal_temp_C
##  Min.   :2021-04-28   Min.   : 37.00   Min.   :19.0
##  1st Qu.:2021-05-04   1st Qu.: 69.50   1st Qu.:21.0
##  Median :2021-05-11   Median : 93.00   Median :23.0
##  Mean   :2021-05-09   Mean   : 85.91   Mean   :22.4
##  3rd Qu.:2021-05-18   3rd Qu.:103.50   3rd Qu.:23.0
##  Max.   :2021-05-18   Max.   :122.00   Max.   :26.0
```

Join Dataframes

Merge all post-experiment CEWL, add cloacal temperature, add capture CEWL:

```
# merge all CEWL datafiles & reformat
CEWL <- CEWL_t1 %>% # trial 1
  rbind(., CEWL_t2, # trial 2
        CEWL_t3, # trial 3
        CEWL_t4 # trial 4
        ) %>%
  # remove any unsuccessful measurements
  dplyr::filter(Status == "Normal") %>%
  # extract individual_ID and region separately from the "ID" variable
  separate(ID, c("individual_ID", "region")) %>%
  # reformat data
  dplyr::mutate(# reformat date
                date = as.Date(date, format = "%m/%d/%y"),
                # format individual ID as a factor
                individual_ID = as.integer(individual_ID),
                # set body region as a factor variable after getting only the consistent characters due
                region = as.factor(substring(region, 1, 4)),
                # add when measurement taken
```

```
                day = as.factor("after"), # may rename...
                n_day = 1
                ) %>%
  # remove cols not relevant to stats
  dplyr::select(-Status) %>%
  # remove any rows with missing values
  dplyr::filter(complete.cases(.)) %>%
  # add cloacal temperatures
  left_join(exp_CT, by = c("date", "individual_ID")) %>%
  # now matching dataframes, add capture CEWL data
  rbind(cap_CEWL) %>%
  # add tmt assignments
  left_join(tmts, by = "individual_ID") %>%
  mutate(humidity_tmt_percent = as.factor(humidity_tmt_percent),
         individual_ID = as.factor(individual_ID),
         conclusion = as.factor(conclusion),
         trial_number = as.factor(trial_number)
         )
# every lizard should have 10 measurements
summary(CEWL)
```

```
##       date             individual_ID  region     TEWL_g_m2h         day
##  Min.   :2021-04-19   37     : 10   dewl:67   Min.   :  4.60   after :172
##  1st Qu.:2021-05-03   39     : 10   dors:67   1st Qu.: 19.95   before:163
##  Median :2021-05-10   40     : 10   head:68   Median : 26.82
##  Mean   :2021-05-06   47     : 10   mite:65   Mean   : 30.49
##  3rd Qu.:2021-05-11   52     : 10   vent:68   3rd Qu.: 38.65
##  Max.   :2021-05-18   54     : 10             Max.   :106.38
##                       (Other):275
##      n_day        cloacal_temp_C    temp_tmt_C humidity_tmt_percent
##  Min.   :0.0000   Min.   :19.00   Min.   :25   dry  :162
##  1st Qu.:0.0000   1st Qu.:21.00   1st Qu.:25   humid:173
##  Median :1.0000   Median :23.00   Median :25
##  Mean   :0.5134   Mean   :23.08   Mean   :25
##  3rd Qu.:1.0000   3rd Qu.:24.00   3rd Qu.:25
##  Max.   :1.0000   Max.   :28.00   Max.   :25
##
##  trial_number    conclusion     notes
##  1: 54        complete:335   Length:335
##  2: 53                       Class :character
##  3:110                       Mode  :character
##  4:118
##
##
##
```

Before/after aren't perfectly even because sometimes we were unable to get the AquaFlux to equilibrate and take a measurement.

Finally, make a small edit so the regions are spelled out completely. This requires reordering factor levels:

```
CEWL$region <- factor(CEWL$region,
                      levels = c("head", "dewl", "dors", "vent", "mite"),
                      labels = c("Head", "Dewlap", "Dorsum",
                                 "Ventrum", "Mite Patch")
```

```
                    )
CEWL$humidity_tmt_percent <- factor(CEWL$humidity_tmt_percent,
                                    levels = c("humid", "dry"),
                                    labels = c("Humid", "Dry"))
```

Export Data Frames for Power Analyses

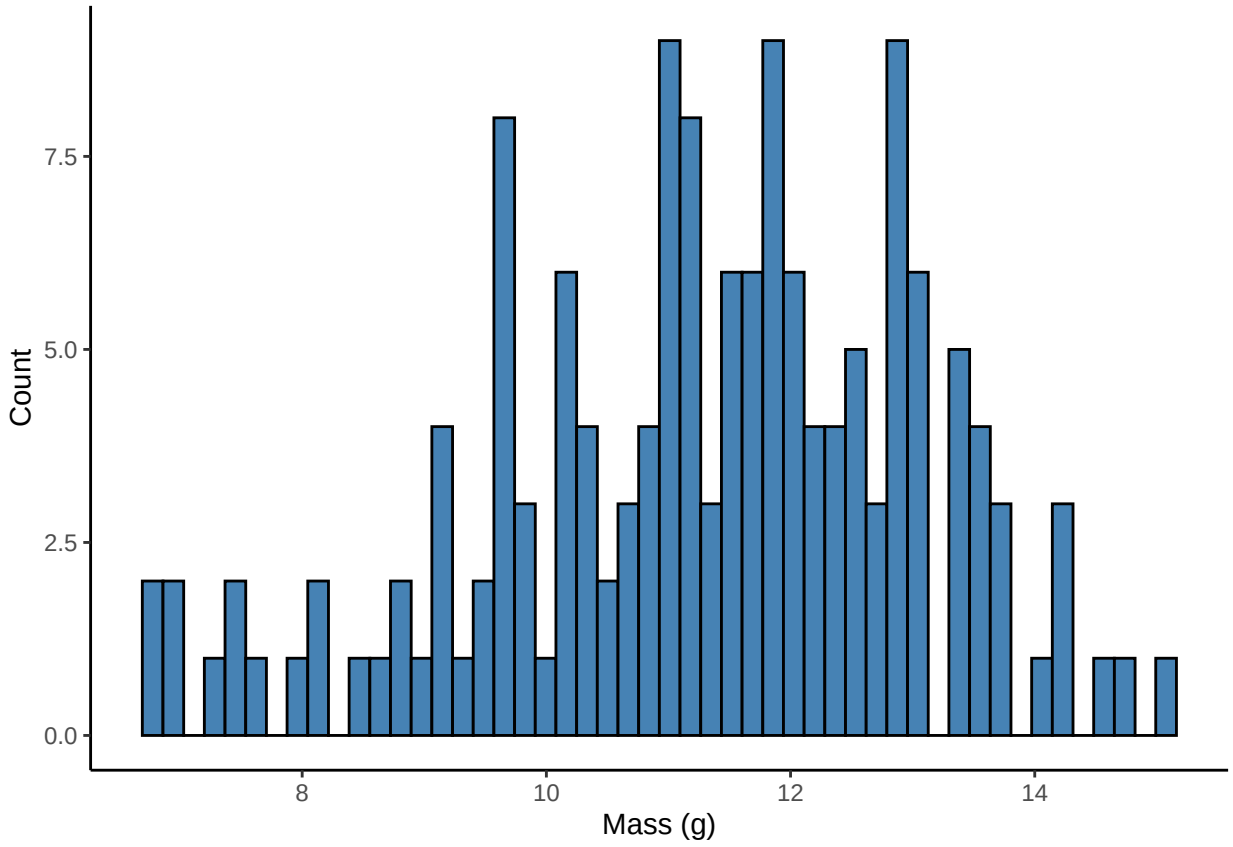
```
write.csv(all_dat, "exported_data/exp_effects_hydration.csv")
write.csv(CEWL, "exported_data/exp_effects_CEWL.csv")
```

Data Distributions

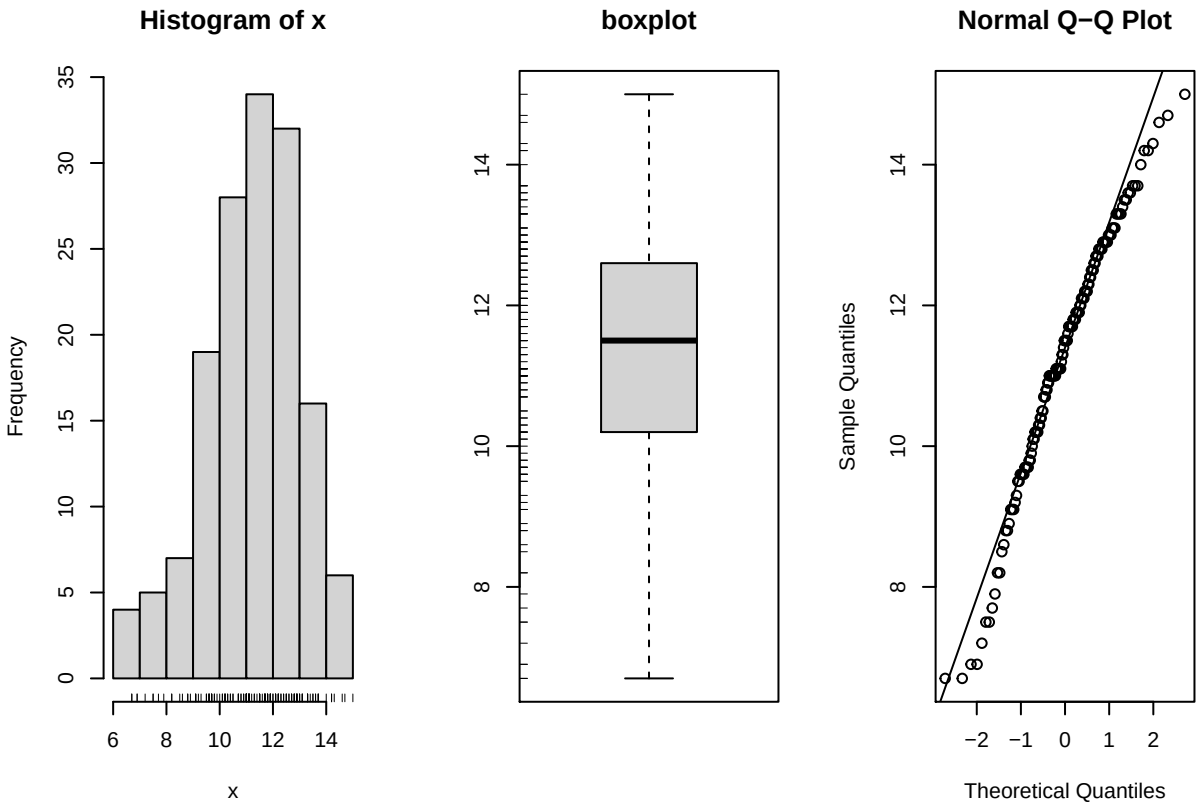
Histograms

Mass

```
all_dat %>%
  ggplot(., aes(x = mass_g)) +
  geom_histogram(color = "black", fill="steelblue", bins=50) +
  theme_classic() +
  xlab("Mass (g)") +
  ylab("Count")
```




```
simple.eda(all_dat$mass_g)
```



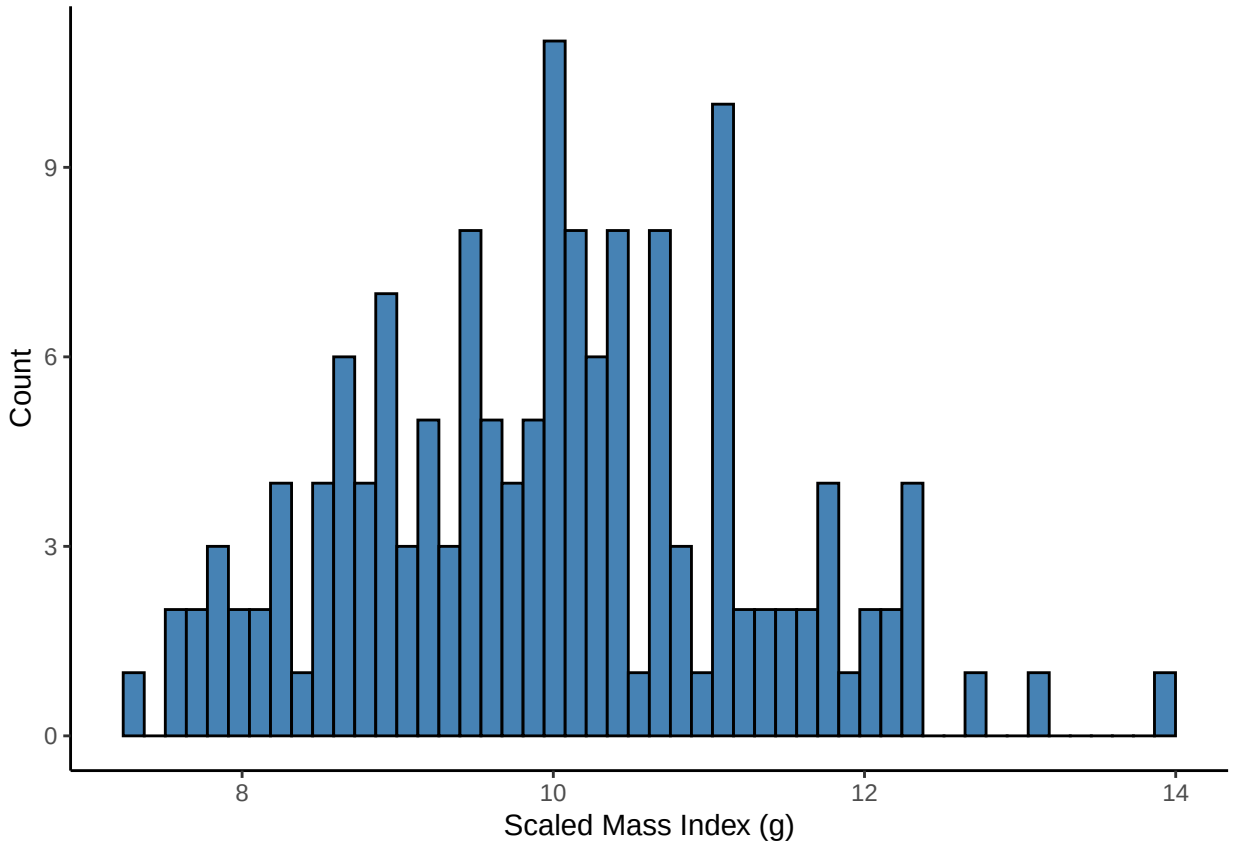

```
shapiro.test(all_dat$mass_g)
```

```
##
##  Shapiro-Wilk normality test
##
## data:  all_dat$mass_g
## W = 0.97747, p-value = 0.01396
```

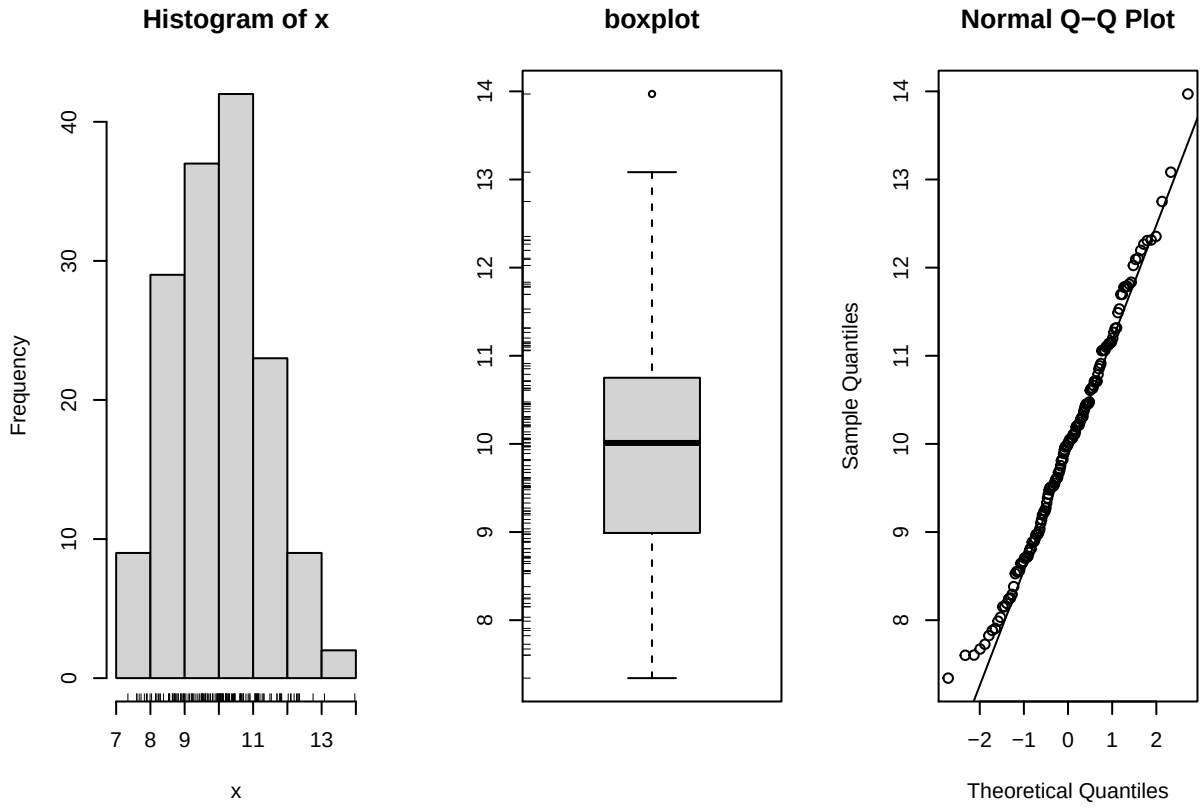
Mass distribution not normal, skewed to the left.

Scaled Mass Index

```
all_dat %>%
  ggplot(., aes(x = SMI)) +
  geom_histogram(color = "black", fill="steelblue", bins=50) +
  theme_classic() +
  xlab("Scaled Mass Index (g)") +
  ylab("Count")
```




```
simple.eda(all_dat$SMI)
```



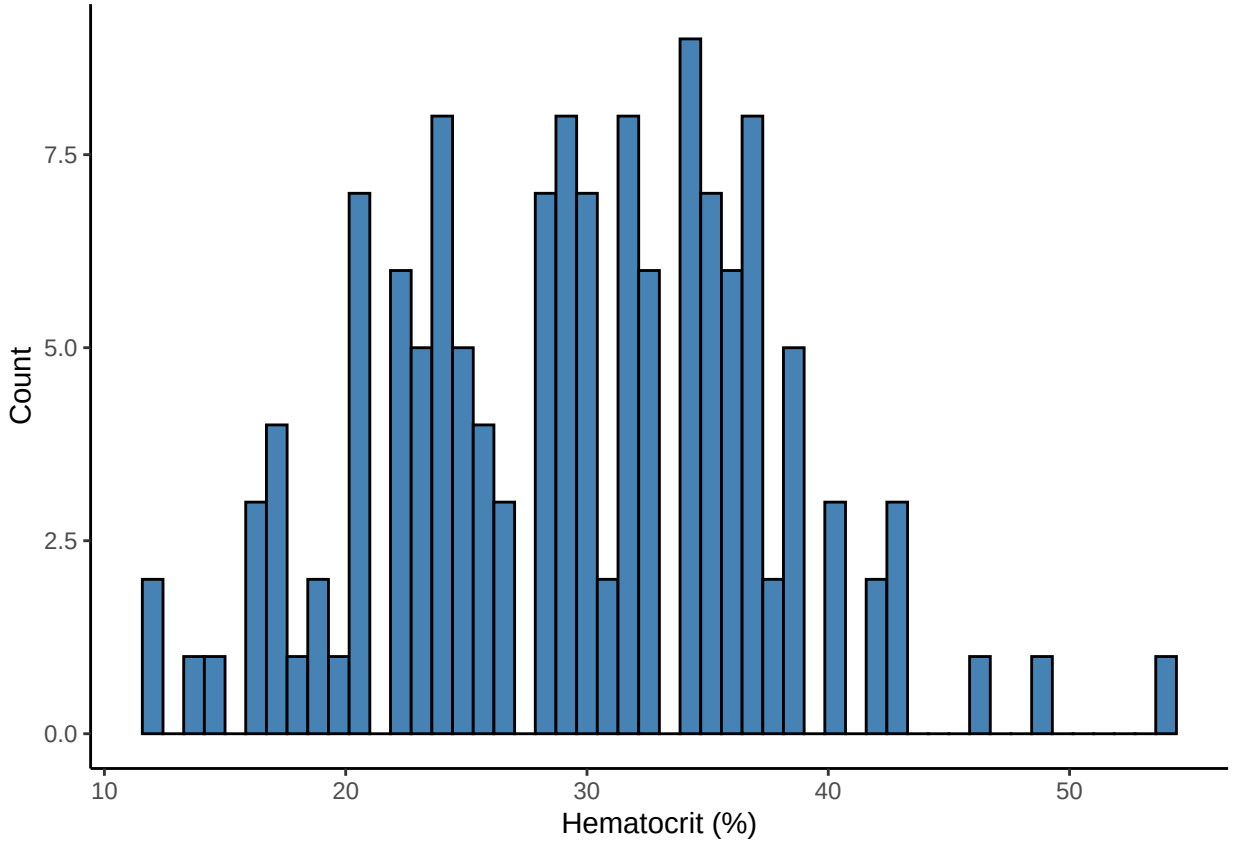
```r
shapiro.test(all_dat$SMI)
```

```
##
##  Shapiro-Wilk normality test
##
## data:  all_dat$SMI
## W = 0.99012, p-value = 0.3712
```

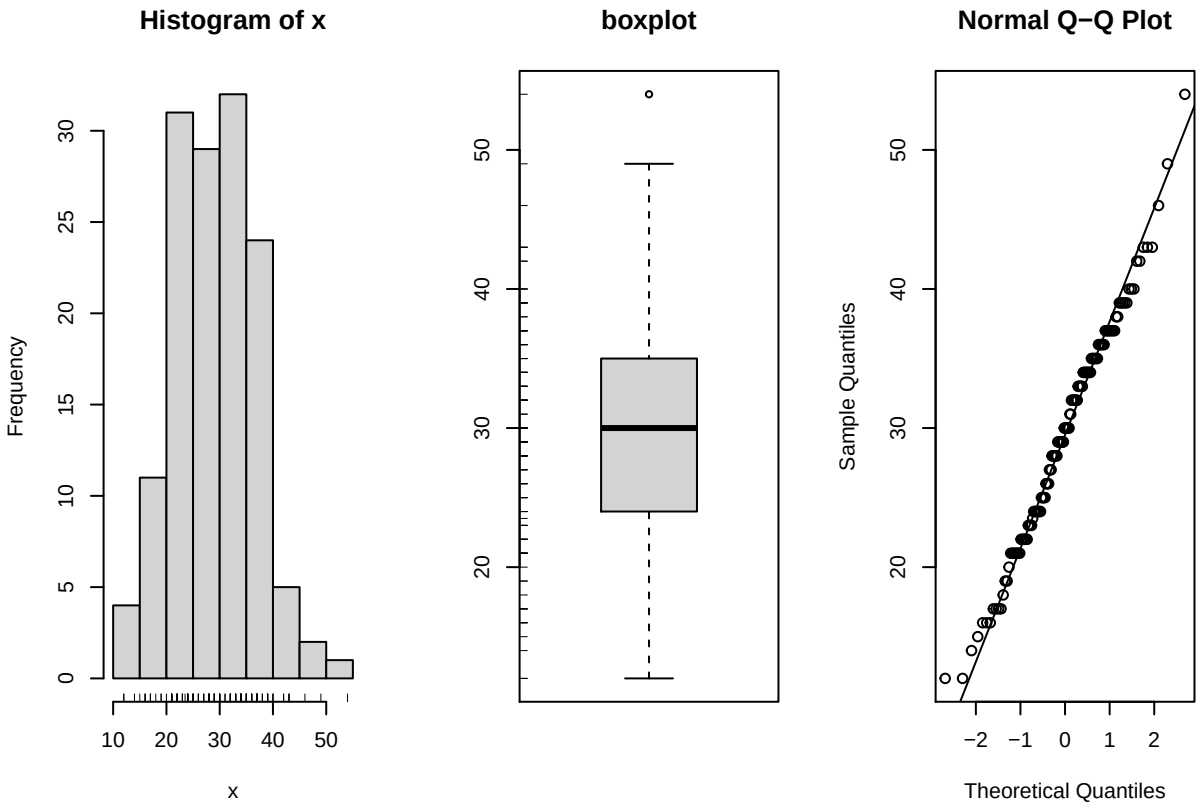
Hematocrit

```r
all_dat %>%
  ggplot(., aes(x = hematocrit_percent)) +
  geom_histogram(color = "black", fill="steelblue", bins=50) +
  theme_classic() +
  xlab("Hematocrit (%)") +
  ylab("Count")
```

```
## Warning: Removed 12 rows containing non-finite values (stat_bin).
```




```r
simple.eda(all_dat$hematocrit_percent)
```




```
shapiro.test(all_dat$hematocrit_percent)
```

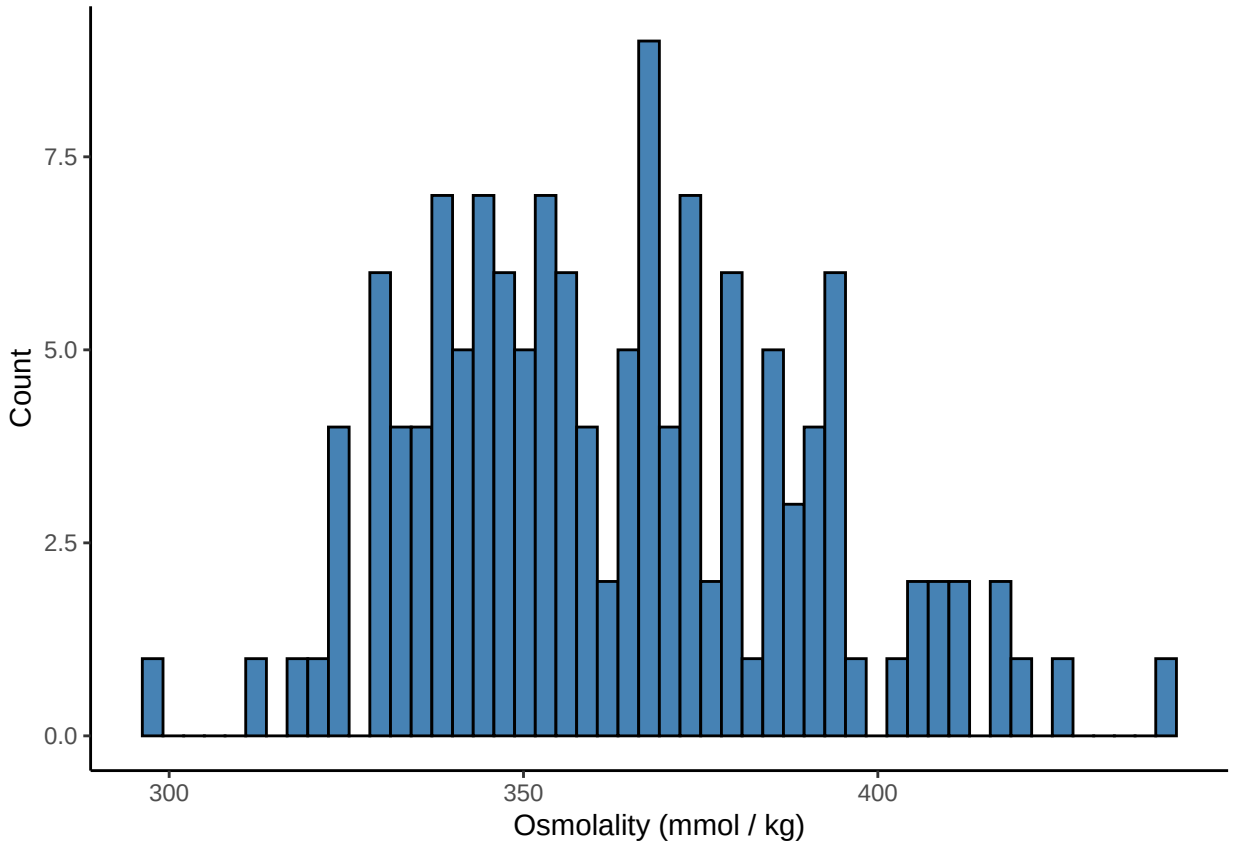
```
##
##  Shapiro-Wilk normality test
##
## data:  all_dat$hematocrit_percent
## W = 0.98984, p-value = 0.4089
```

Visually, looks slightly skewed to the right, but statistically, the distribution of hematocrit is normal.

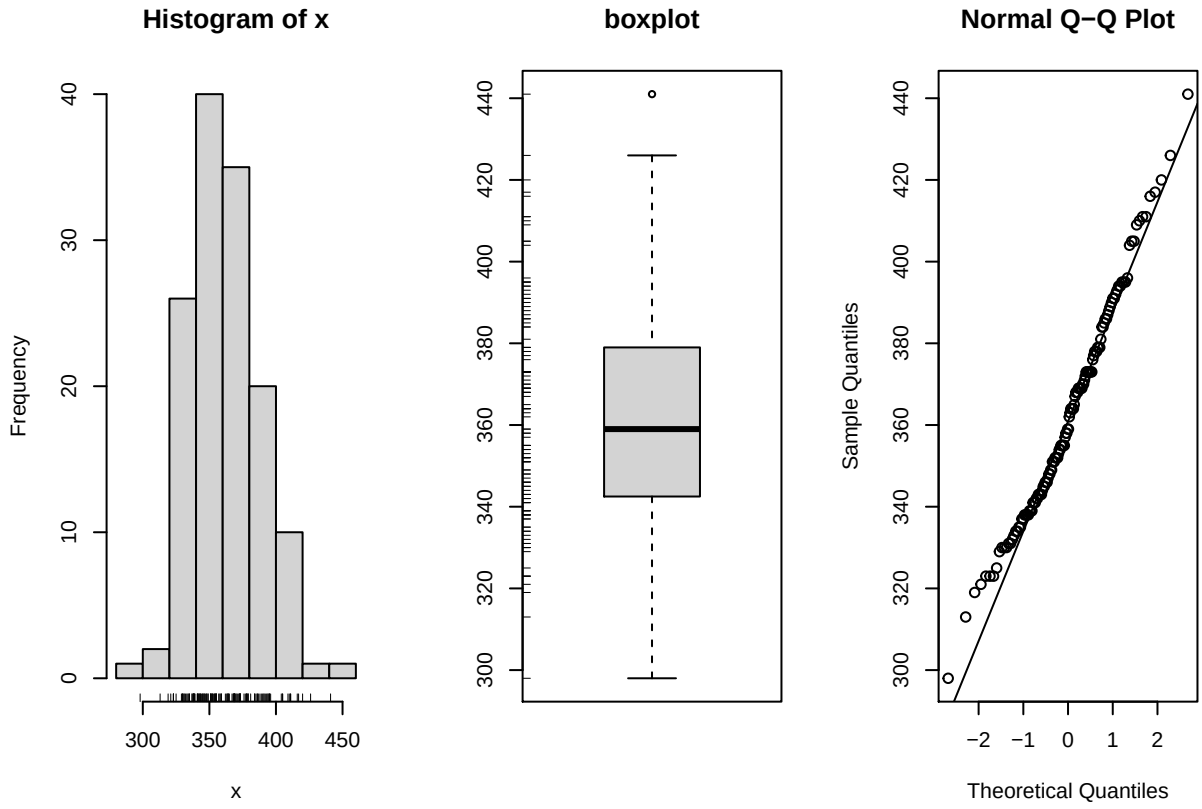
Osmolality

```
all_dat %>%
  ggplot(., aes(x = osmolality_mmol_kg)) +
  geom_histogram(color = "black", fill="steelblue", bins=50) +
  theme_classic() +
  xlab("Osmolality (mmol / kg)") +
  ylab("Count")
```

```
## Warning: Removed 15 rows containing non-finite values (stat_bin).
```




```
simple.eda(all_dat$osmolality_mmol_kg)
```



```
shapiro.test(all_dat$osmolality_mmol_kg)
```

```
##
##  Shapiro-Wilk normality test
##
## data:  all_dat$osmolality_mmol_kg
## W = 0.98331, p-value = 0.09544
```

Visually, looks slightly skewed to the right, but statistically, the distribution of osmolality is normal.

Basic Figures & Models

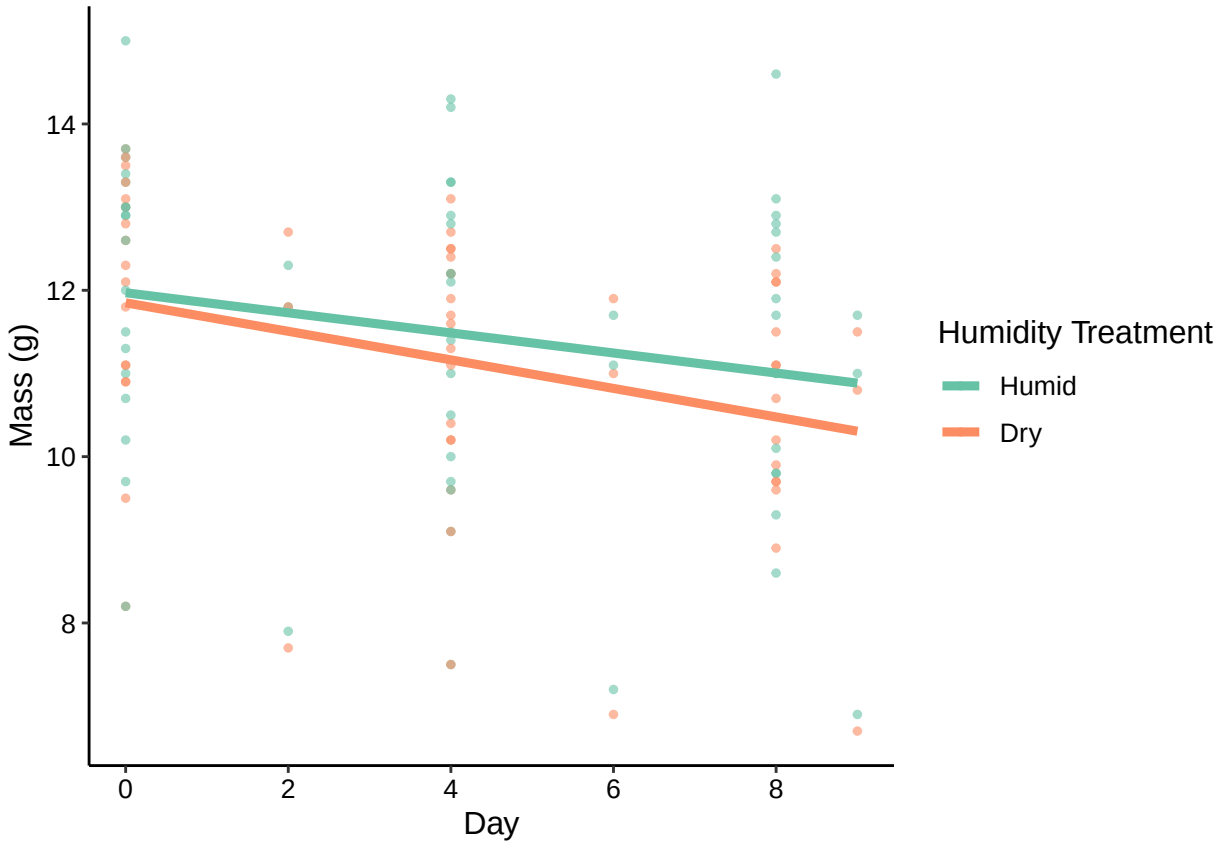
Mass ~ Time

I won't be using this, SMI is more applicable.

Just look at plot:

```
all_dat_no_rehab %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = mass_g,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 0.6) +
  stat_smooth(aes(x = day,
                  y = mass_g,
                  color = humidity_tmt_percent
                  ),
              formula = y ~ x,
              method = "lm",
              se = F,
              size = 1.6,
              alpha = 1 ) +
  theme_classic() +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +

  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  xlab("Day") +
  ylab("Mass (g)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0
)
```

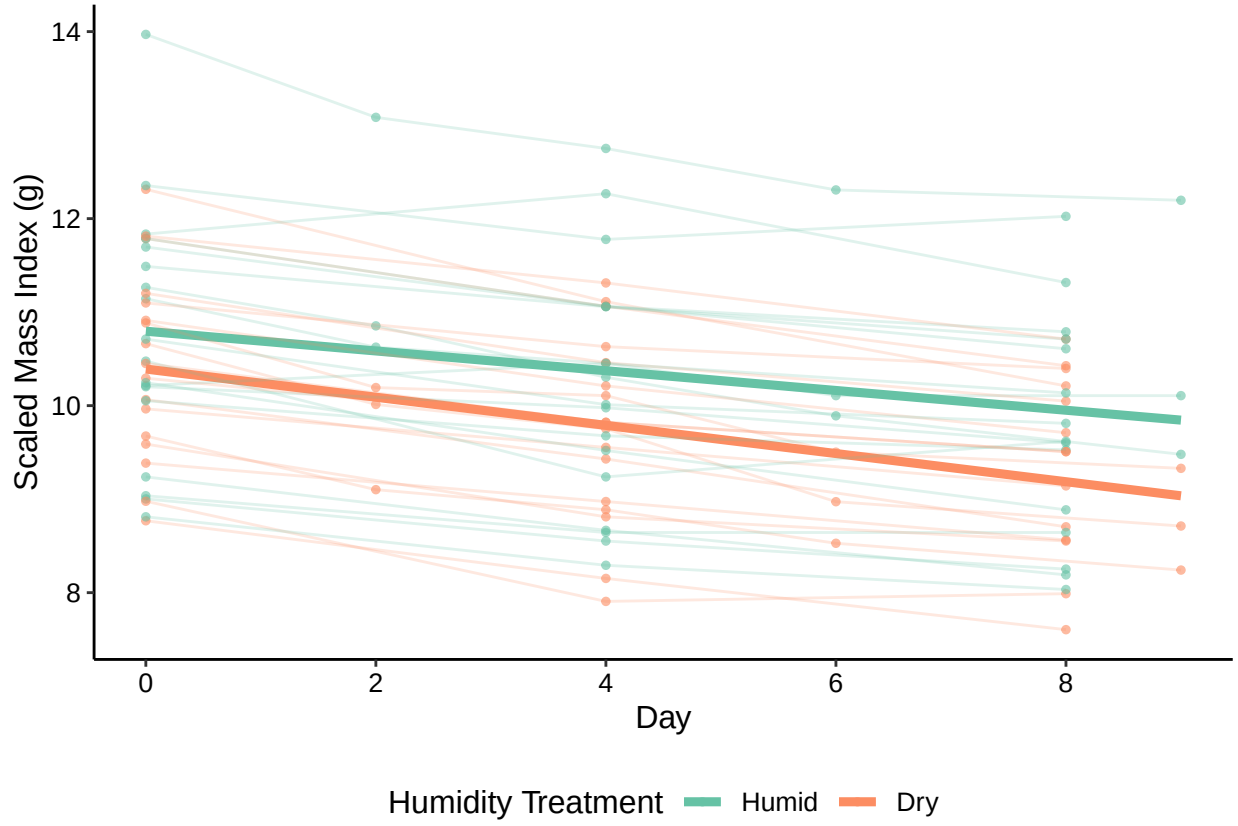


SMI ~ Time

plot over course of experiment:

```
all_dat_no_rehab %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = SMI,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 0.6) +
  stat_smooth(aes(x = day,
                  y = SMI,
                  color = humidity_tmt_percent
                  ),
              formula = y ~ x,
              method = "lm",
              se = F,
              size = 1.6,
              alpha = 1 ) +
  geom_line(aes(x = day,
                y = SMI,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.2) +
  theme_classic() +
```

```
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  xlab("Day") +
  ylab("Scaled Mass Index (g)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
) -> tmt_effects_SMI
tmt_effects_SMI
```




```
# export figure
ggsave(filename = "tmt_effects_SMI.jpeg",
       plot = tmt_effects_SMI,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 5, height = 4)
```

Check whether means started out different:

```
SMI_diff_lm <- all_dat_no_rehab %>%
  dplyr::filter(day == 0) %>%
```

```
  lm(data = ., SMI ~humidity_tmt_percent)
summary(SMI_diff_lm)
```

```
##
## Call:
## lm(formula = SMI ~ humidity_tmt_percent, data = .)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -1.9405 -0.7429 -0.0401  0.7385  3.2183
##
## Coefficients:
##                         Estimate Std. Error t value Pr(>|t|)
## (Intercept)              10.7517     0.2811   38.25   <2e-16 ***
## humidity_tmt_percentDry  -0.2904     0.4033   -0.72    0.476
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.192 on 33 degrees of freedom
## Multiple R-squared:  0.01547,    Adjusted R-squared:  -0.01436
## F-statistic: 0.5187 on 1 and 33 DF,  p-value: 0.4765
```

NOT significantly different, which is good. t-test and p-value have very similar results

model:

```
SMI_mod <- lme4::lmer(data = all_dat_no_rehab,
               SMI ~ day*humidity_tmt_percent +
               (1|trial_number))
summary(SMI_mod)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: SMI ~ day * humidity_tmt_percent + (1 | trial_number)
##    Data: all_dat_no_rehab
##
## REML criterion at convergence: 369.8
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.84407 -0.68219 -0.04622  0.62984  2.74822
##
## Random effects:
##  Groups       Name        Variance Std.Dev.
##  trial_number (Intercept) 0.01241  0.1114
##  Residual                 1.27504  1.1292
## Number of obs: 117, groups:  trial_number, 4
##
## Fixed effects:
##                              Estimate Std. Error t value
## (Intercept)                  10.79499    0.24058  44.871
## day                          -0.10598    0.04510  -2.350
## humidity_tmt_percentDry      -0.40660    0.33497  -1.214
## day:humidity_tmt_percentDry -0.04498    0.06463  -0.696
##
## Correlation of Fixed Effects:
```

```
##             (Intr) day    hmd__D
## day         -0.759
## hmdty_tmt_D -0.678  0.545
## dy:hmdty__D  0.530 -0.698 -0.782
```

```
drop1(SMI_mod)
```

```
## Single term deletions
##
## Model:
## SMI ~ day * humidity_tmt_percent + (1 | trial_number)
##                           npar    AIC
## <none>                         369.24
## day:humidity_tmt_percent    1 367.74
```

```
# drop interaction term
SMI_mod2 <- lme4::lmer(data = all_dat_no_rehab,
               SMI ~ day + humidity_tmt_percent +
               (1|trial_number))
summary(SMI_mod2)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: SMI ~ day + humidity_tmt_percent + (1 | trial_number)
##    Data: all_dat_no_rehab
##
## REML criterion at convergence: 366.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.84939 -0.64254 -0.07319  0.69627  2.67483
##
## Random effects:
##  Groups       Name        Variance Std.Dev.
##  trial_number (Intercept) 0.01262  0.1123
##  Residual                 1.26911  1.1265
## Number of obs: 117, groups:  trial_number, 4
##
## Fixed effects:
##                           Estimate Std. Error t value
## (Intercept)               10.88366    0.20374  53.420
## day                       -0.12788    0.03223  -3.967
## humidity_tmt_percentDry -0.58888    0.20841  -2.826
##
## Correlation of Fixed Effects:
##             (Intr) day
## day         -0.641
## hmdty_tmt_D -0.498  0.000
```

```
drop1(SMI_mod2)
```

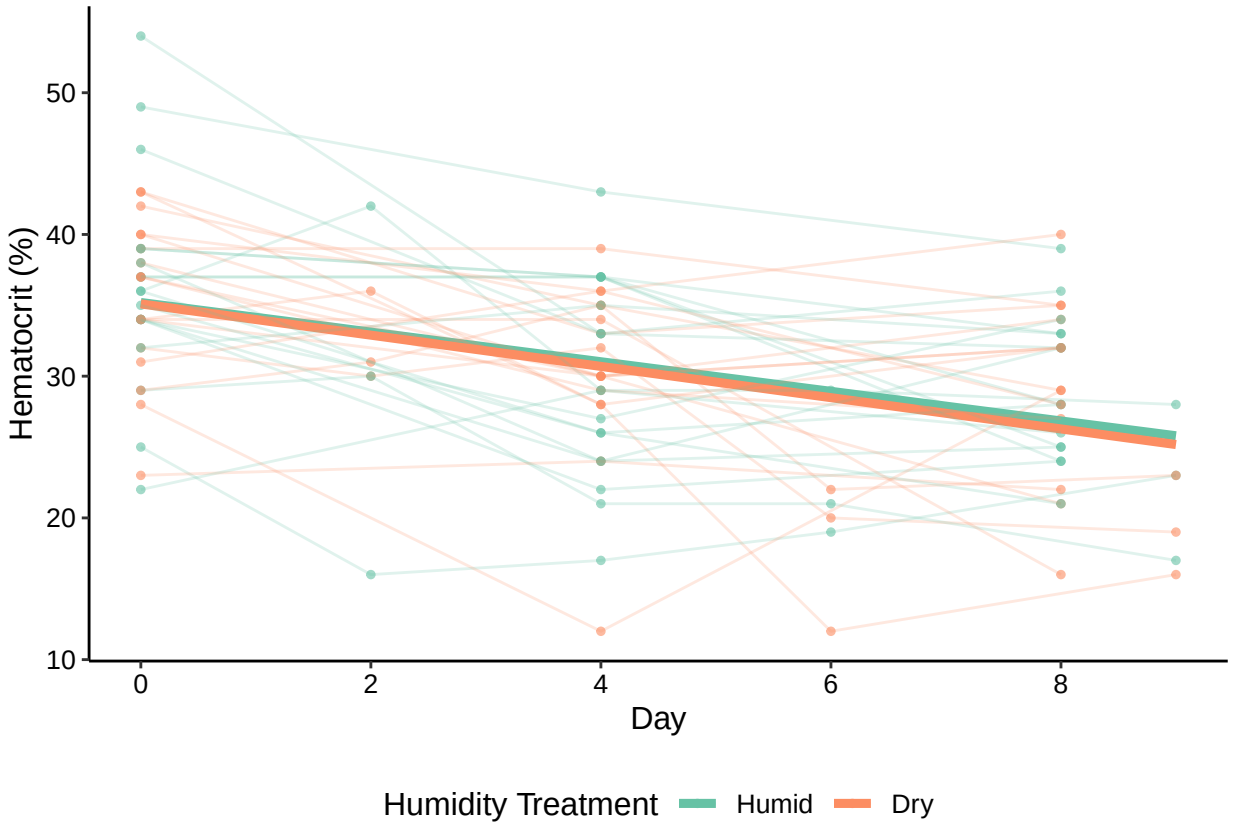
```
## Single term deletions
##
## Model:
## SMI ~ day + humidity_tmt_percent + (1 | trial_number)
##                       npar    AIC
## <none>                     367.74
```

```
## day                      1 380.69
## humidity_tmt_percent     1 373.55
```

SMI is best predicted by day and treatment, but not including their interaction.

Hct ~ Time

```
all_dat_no_rehab %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = hematocrit_percent,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 0.6) +
  stat_smooth(aes(x = day,
                  y = hematocrit_percent,
                  color = humidity_tmt_percent
                  ),
              formula = y ~ x,
              method = "lm",
              se = F,
              size = 1.6,
              alpha = 1 ) +
  geom_line(aes(x = day,
                y = hematocrit_percent,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.2) +
  theme_classic() +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  xlab("Day") +
  ylab("Hematocrit (%)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
        ) -> tmt_effects_hct
tmt_effects_hct
```




```
# export figure
ggsave(filename = "tmt_effects_hct.jpeg",
       plot = tmt_effects_hct,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 5, height = 4)
```

this model seemed to work well with indiv as a random factor, but still excluded because it's probably unnecessary

```
hct_mod <- all_dat_no_rehab %>%
  dplyr::filter(complete.cases(hematocrit_percent)) %>%
  lme4::lmer(data = .,
             hematocrit_percent ~ day + humidity_tmt_percent +
             (1|trial_number))
summary(hct_mod)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: hematocrit_percent ~ day + humidity_tmt_percent + (1 | trial_number)
##    Data: .
##
## REML criterion at convergence: 765.4
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -3.2154 -0.6104  0.0919  0.6453  2.7070
##
## Random effects:
```

```
##  Groups       Name        Variance Std.Dev.
##  trial_number (Intercept)  8.339   2.888
##  Residual                 40.121   6.334
## Number of obs: 117, groups:  trial_number, 4
##
## Fixed effects:
##                          Estimate Std. Error t value
## (Intercept)               35.2145     1.8208  19.340
## day                       -1.0534     0.1813  -5.811
## humidity_tmt_percentDry   -0.2733     1.1724  -0.233
##
## Correlation of Fixed Effects:
##             (Intr) day
## day         -0.403
## hmdty_tmt_D -0.314  0.000
```

```
drop1(hct_mod)
```

```
## Single term deletions
##
## Model:
## hematocrit_percent ~ day + humidity_tmt_percent + (1 | trial_number)
##                       npar    AIC
## <none>                     778.55
## day                      1 806.68
## humidity_tmt_percent     1 776.61
```

```
# drop humidity
hct_mod2 <- all_dat_no_rehab %>%
  dplyr::filter(complete.cases(hematocrit_percent)) %>%
  lme4::lmer(data = .,
             hematocrit_percent ~ day +
             (1|trial_number))
summary(hct_mod2)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: hematocrit_percent ~ day + (1 | trial_number)
##    Data: .
##
## REML criterion at convergence: 767.6
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -3.2511 -0.5977  0.1135  0.6655  2.7398
##
## Random effects:
##  Groups       Name        Variance Std.Dev.
##  trial_number (Intercept)  8.36    2.891
##  Residual                 39.78    6.307
## Number of obs: 117, groups:  trial_number, 4
##
## Fixed effects:
##             Estimate Std. Error t value
## (Intercept)  35.0810     1.7278  20.304
## day          -1.0534     0.1805  -5.836
```

```
##
## Correlation of Fixed Effects:
##     (Intr)
## day -0.423
```

```
drop1(hct_mod2)
```

```
## Single term deletions
##
## Model:
## hematocrit_percent ~ day + (1 | trial_number)
##        npar    AIC
## <none>      776.61
## day       1 804.72
```

The model AIC is slightly better without the interaction effect, so I removed that. The effect of humidity could ALSO be dropped, so humidity treatment was not an important factor affecting hematocrit, but how many days lizards were in treatment was. Both treatment groups lost hematocrit at approximately the same rate.

Osml ~ Time

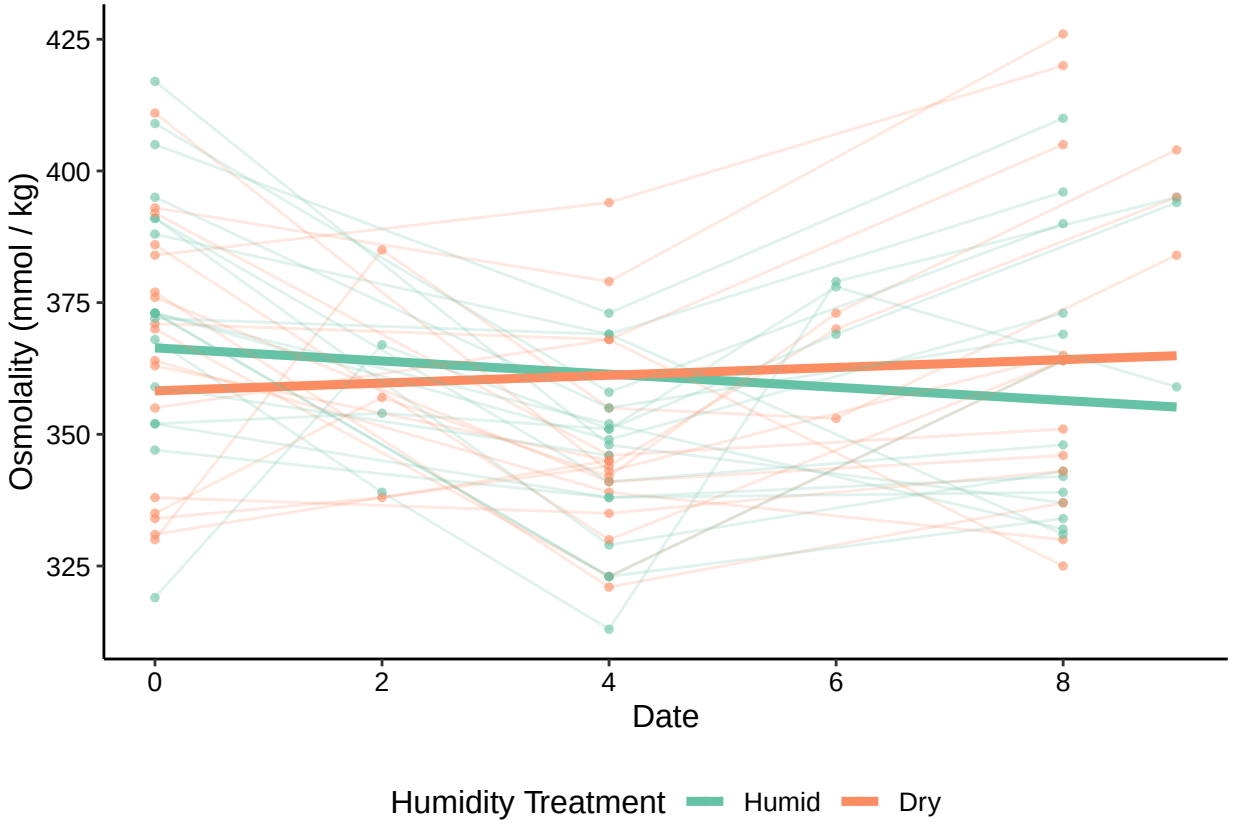
```
all_dat_no_rehab %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = osmolality_mmol_kg,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 0.6) +
  stat_smooth(aes(x = day,
                  y = osmolality_mmol_kg,
                  color = humidity_tmt_percent
                  ),
              formula = y ~ x,
              method = "lm",
              se = F,
              size = 1.6,
              alpha = 1 ) +
  geom_line(aes(x = day,
                y = osmolality_mmol_kg,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.2) +
  theme_classic() +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  xlab("Date") +
  ylab("Osmolality (mmol / kg)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
```

```
                                  size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
        ) -> tmt_effects_osml
tmt_effects_osml
```

```
## Warning: Removed 3 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 3 rows containing missing values (geom_point).
```

```
## Warning: Removed 3 row(s) containing missing values (geom_path).
```




```
# export figure
ggsave(filename = "tmt_effects_osml.jpeg",
       plot = tmt_effects_osml,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 5, height = 4)
```

```
## Warning: Removed 3 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 3 rows containing missing values (geom_point).
```

```
## Warning: Removed 3 row(s) containing missing values (geom_path).
```

singular warning - do NOT include individual ID as a random effect

```
osml_mod <- all_dat_no_rehab %>%
  dplyr::filter(complete.cases(osmolality_mmol_kg)) %>%
  lme4::lmer(data = .,
```

```
              osmolality_mmol_kg ~ day * humidity_tmt_percent +
              (1|trial_number))
summary(osml_mod)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: osmolality_mmol_kg ~ day * humidity_tmt_percent + (1 | trial_number)
##    Data: .
##
## REML criterion at convergence: 1018.4
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -2.1763 -0.6722 -0.1769  0.6925  2.5285
##
## Random effects:
##  Groups       Name        Variance Std.Dev.
##  trial_number (Intercept) 302.6    17.40
##  Residual                 469.3    21.66
## Number of obs: 114, groups:  trial_number, 4
##
## Fixed effects:
##                              Estimate Std. Error t value
## (Intercept)                   369.938      9.804  37.733
## day                            -1.315      0.877  -1.499
## humidity_tmt_percentDry        -8.152      6.439  -1.266
## day:humidity_tmt_percentDry     1.901      1.266   1.502
##
## Correlation of Fixed Effects:
##             (Intr) day    hmd__D
## day         -0.357
## hmdty_tmt_D -0.319  0.542
## dy:hmdty__D  0.246 -0.691 -0.776
```

```
drop1(osml_mod)
```

```
## Single term deletions
##
## Model:
## osmolality_mmol_kg ~ day * humidity_tmt_percent + (1 | trial_number)
##                            npar    AIC
## <none>                          1044.3
## day:humidity_tmt_percent    1 1044.6
```

The model seems good as-is.

Change in Osmolality

```
osml_d0 <- all_dat_no_rehab %>%
  dplyr::filter(day == 0) %>%
  dplyr::select(individual_ID, osml0 = osmolality_mmol_kg,
                humidity_tmt_percent)
osml_d8 <- all_dat_no_rehab %>%
  dplyr::filter(day %in% c(8,9)) %>%
  dplyr::select(individual_ID, osml89 = osmolality_mmol_kg,
                humidity_tmt_percent)
```

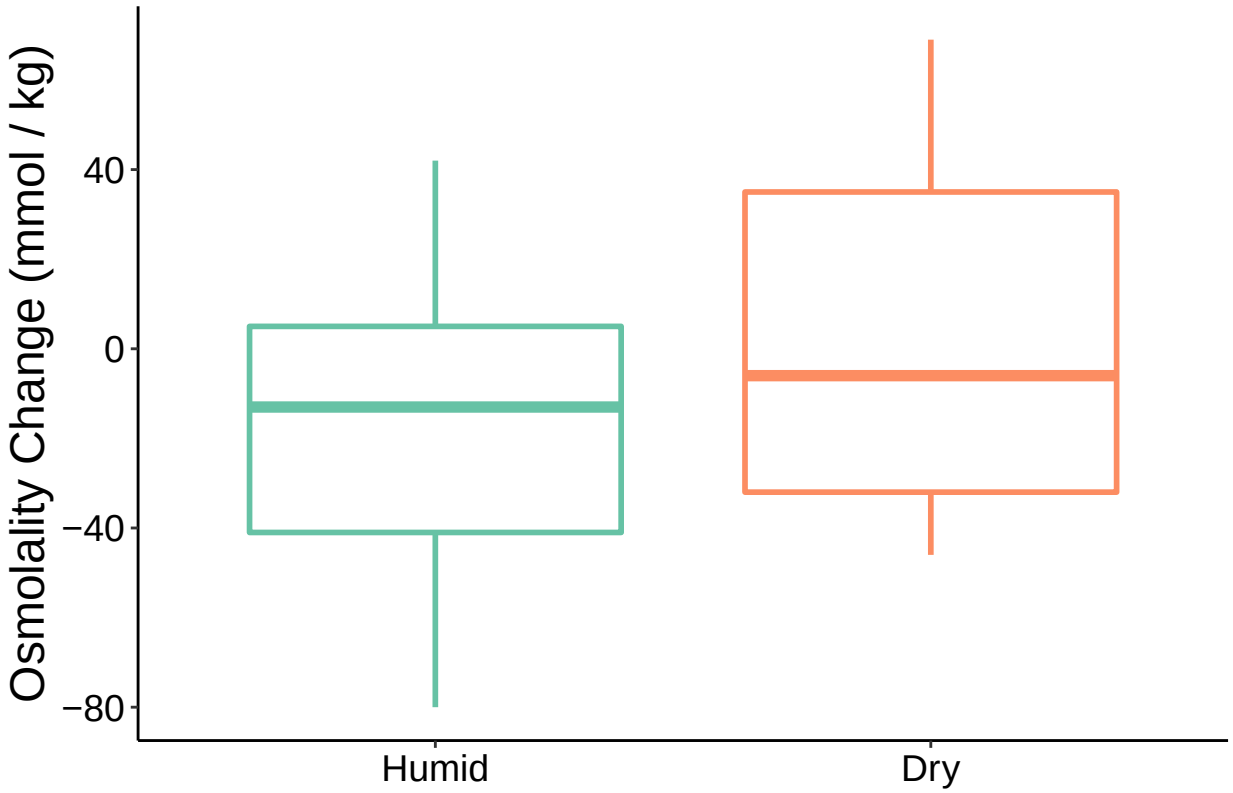
```
osml_diffs <- osml_d0 %>%
  left_join(osml_d8) %>%
  mutate(osml_change = osml89 - osml0)
```

```
## Joining, by = c("individual_ID", "humidity_tmt_percent")
```

boxplot:

```
osml_diffs %>%
  ggplot(data = .) +
  geom_boxplot(aes(x = humidity_tmt_percent,
                   y = osml_change,
                   group = humidity_tmt_percent,
                   color = humidity_tmt_percent
                   ),
               size = 1,
               alpha = 1) +
  theme_classic() +
  xlab("") +
  ylab("Osmolality Change (mmol / kg)") +
  scale_color_brewer(palette = "Set2") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 18),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 14),
        legend.text.align = 0,
        legend.position = "none"
)
```

```
## Warning: Removed 3 rows containing non-finite values (stat_boxplot).
```



model to test statistical significance of difference:

```
osml_diff_mod <- lm(data = osml_diffs,
                    osml_change ~ humidity_tmt_percent)
summary(osml_diff_mod)
```

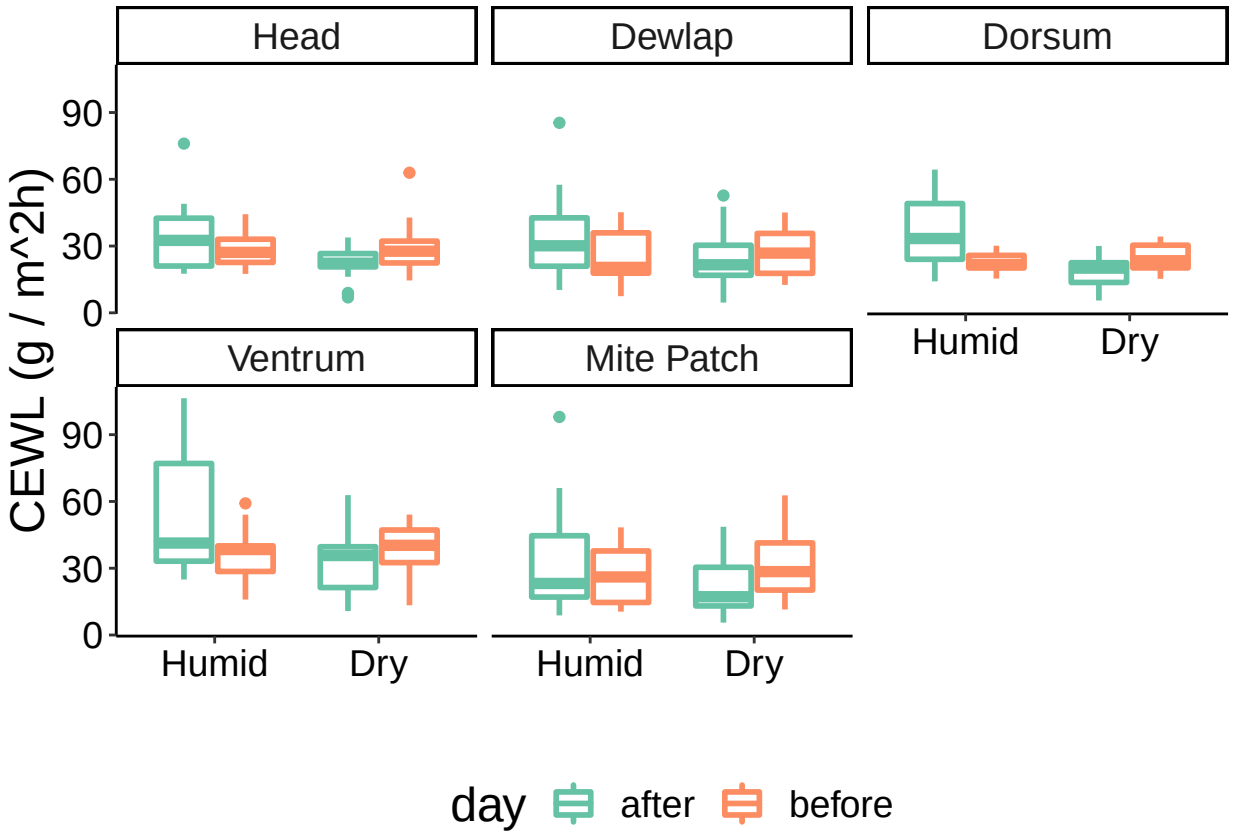
```
##
## Call:
## lm(formula = osml_change ~ humidity_tmt_percent, data = osml_diffs)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -65.765 -29.790  -1.765  30.383  66.133
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## (Intercept)               -14.235      8.988  -1.584    0.124
## humidity_tmt_percentDry    17.102     13.128   1.303    0.203
##
## Residual standard error: 37.06 on 30 degrees of freedom
##   (3 observations deleted due to missingness)
## Multiple R-squared:  0.05354,    Adjusted R-squared:  0.02199
## F-statistic: 1.697 on 1 and 30 DF,  p-value: 0.2026
```

Not difference between the two treatment groups.

CEWL ~ Before/After

try a boxplot:

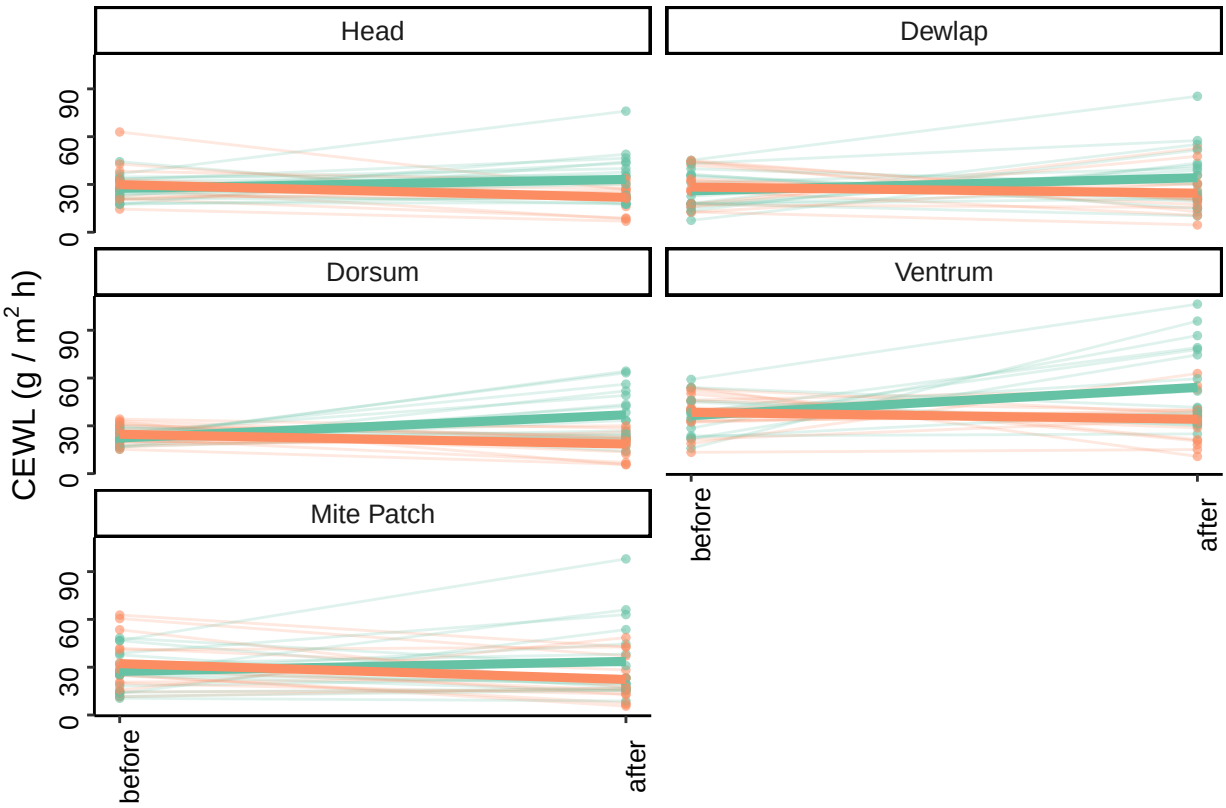
```
CEWL %>%
  ggplot(data = .) +
  geom_boxplot(aes(x = humidity_tmt_percent,
                   y = TEWL_g_m2h,
                   color = day
                   ),
               size = 1,
               alpha = 1) +
  facet_wrap(~region) +
  theme_classic() +
  xlab("") +
  ylab("CEWL (g / m^2h)") +
  scale_color_brewer(palette = "Set2") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 18),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 14),
        legend.text.align = 0,
        legend.position = "bottom"
)
```



this is difficult to see changes, I think a line graph would be better...

```
CEWL %>%
  ggplot(data = .) +
  geom_point(aes(x = n_day,
```

```
                y = TEWL_g_m2h,
                color = humidity_tmt_percent
                ),
            size = 1,
            alpha = 0.6) +
  geom_line(aes(x = n_day,
                y = TEWL_g_m2h,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.2) +
  stat_smooth(aes(x = n_day,
                  y = TEWL_g_m2h,
                  color = humidity_tmt_percent
                  ),
              formula = y ~ x,
              method = "lm",
              se = F,
              size = 1.6,
              alpha = 1) +
  theme_classic() +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  facet_wrap(~region, ncol = 2) +
  scale_x_continuous(breaks = c(0, 1),
                     labels = c("0" = "before", "1" = "after")
                     ) +
  xlab("") +
  ylab(bquote('CEWL (g / '*m^2~h*')')) +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10,
                                 angle = 90),
        legend.text.align = 0,
        legend.position = "none"
) -> CEWL_tmt_fig
CEWL_tmt_fig
```




```
# export figure
ggsave(filename = "tmt_effects_CEWL.jpeg",
       plot = CEWL_tmt_fig,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 4, height = 6)
```

I saved the legend separately to make the figure layout better.

```
CEWL_mod <- CEWL %>%
  dplyr::filter(complete.cases(.)) %>%
  lme4::lmer(data = .,
             TEWL_g_m2h ~ day * humidity_tmt_percent * region +
             (1|trial_number/individual_ID))
summary(CEWL_mod)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula:
## TEWL_g_m2h ~ day * humidity_tmt_percent * region + (1 | trial_number/individual_ID)
##    Data: .
##
## REML criterion at convergence: 2564.8
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -1.9542 -0.5724 -0.0461  0.4516  3.7679
##
## Random effects:
```

```
##  Groups                      Name        Variance Std.Dev.
##  individual_ID:trial_number (Intercept)  31.97    5.655
##  trial_number               (Intercept)  41.24    6.422
##  Residual                                146.69   12.112
## Number of obs: 335, groups:  individual_ID:trial_number, 35; trial_number, 4
##
## Fixed effects:
##                                                  Estimate Std. Error t value
## (Intercept)                                      32.53840    4.51507   7.207
## daybefore                                        -6.17985    4.10281  -1.506
## humidity_tmt_percentDry                         -11.09902    4.52154  -2.455
## regionDewlap                                      1.81474    3.98753   0.455
## regionDorsum                                      2.94941    4.10076   0.719
## regionVentrum                                    20.98333    4.03721   5.197
## regionMite Patch                                  0.51328    4.10058   0.125
## daybefore:humidity_tmt_percentDry                14.12190    5.89079   2.397
## daybefore:regionDewlap                           -4.33097    5.81485  -0.745
## daybefore:regionDorsum                           -8.15118    5.83730  -1.396
## daybefore:regionVentrum                         -12.12039    5.79283  -2.092
## daybefore:regionMite Patch                       -0.69387    5.83717  -0.119
## humidity_tmt_percentDry:regionDewlap              1.06837    5.80840   0.184
## humidity_tmt_percentDry:regionDorsum             -6.26471    5.83730  -1.073
## humidity_tmt_percentDry:regionVentrum            -8.74039    5.79283  -1.509
## humidity_tmt_percentDry:regionMite Patch         -0.14404    5.88920  -0.024
## daybefore:humidity_tmt_percentDry:regionDewlap   -0.04119    8.36570  -0.005
## daybefore:humidity_tmt_percentDry:regionDorsum    6.29866    8.34675   0.755
## daybefore:humidity_tmt_percentDry:regionVentrum   8.42839    8.31571   1.014
## daybefore:humidity_tmt_percentDry:regionMite Patch  2.25453  8.42234   0.268
##
## Correlation matrix not shown by default, as p = 20 > 12.
## Use print(x, correlation=TRUE)  or
##     vcov(x)        if you need it
```

```
drop1(CEWL_mod)
```

```
## Single term deletions
##
## Model:
## TEWL_g_m2h ~ day * humidity_tmt_percent * region + (1 | trial_number/individual_ID)
##                                  npar    AIC
## <none>                                2694.7
## day:humidity_tmt_percent:region    4 2688.5
```

Drop triple interaction. I think the day:region standalone would be weird too.

```
CEWL_mod2 <- CEWL %>%
  dplyr::filter(complete.cases(.)) %>%
  lme4::lmer(data = .,
             TEWL_g_m2h ~
             day*humidity_tmt_percent +
             humidity_tmt_percent*region +
             (1|trial_number/individual_ID))
summary(CEWL_mod2)
```

```
## Linear mixed model fit by REML ['lmerMod']
```

```
## Formula: TEWL_g_m2h ~ day * humidity_tmt_percent + humidity_tmt_percent *
##     region + (1 | trial_number/individual_ID)
##    Data: .
##
## REML criterion at convergence: 2613.1
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -2.2566 -0.6095 -0.0649  0.4249  3.7327
##
## Random effects:
##  Groups                       Name        Variance Std.Dev.
##  individual_ID:trial_number (Intercept)  31.80    5.639
##  trial_number               (Intercept)  41.97    6.478
##  Residual                                146.49   12.103
## Number of obs: 335, groups:  individual_ID:trial_number, 35; trial_number, 4
##
## Fixed effects:
##                                          Estimate Std. Error t value
## (Intercept)                               34.9974     4.1701   8.393
## daybefore                                -11.2541     1.8584  -6.056
## humidity_tmt_percentDry                  -12.7384     3.7313  -3.414
## regionDewlap                              -0.3101     2.8937  -0.107
## regionDorsum                              -1.0593     2.9161  -0.363
## regionVentrum                             15.0963     2.8932   5.218
## regionMite Patch                           0.2436     2.9161   0.084
## daybefore:humidity_tmt_percentDry         17.5119     2.6689   6.561
## humidity_tmt_percentDry:regionDewlap       1.0297     4.1715   0.247
## humidity_tmt_percentDry:regionDorsum      -3.1541     4.1691  -0.757
## humidity_tmt_percentDry:regionVentrum     -4.6434     4.1532  -1.118
## humidity_tmt_percentDry:regionMite Patch   0.8818     4.2069   0.210
##
## Correlation of Fixed Effects:
##                            (Intr) daybfr hmd__D rgnDwl rgnDrs rgnVnt rgnMtP
## daybefore                  -0.210
## hmdty_tmt_D                -0.435  0.238
## regionDewlp                -0.351  0.018  0.392
## regionDorsm                -0.342 -0.008  0.383  0.496
## regionVntrm                -0.347  0.000  0.388  0.500  0.496
## reginMtPtch                -0.342 -0.010  0.382  0.496  0.492  0.496
## dybfr:hm__D                 0.148 -0.696 -0.342 -0.013  0.006  0.000  0.007
## hmdty_tmt_prcntDry:rgnDw    0.243 -0.013 -0.555 -0.694 -0.344 -0.347 -0.344
## hmdty_tmt_prcntDry:rgnDr    0.239  0.006 -0.553 -0.347 -0.699 -0.347 -0.344
## hmdty_t_D:V                 0.242  0.000 -0.557 -0.348 -0.346 -0.697 -0.346
## hmdty__D:MP                 0.238  0.007 -0.548 -0.344 -0.341 -0.344 -0.693
##                            dy:__D hmdty_tmt_prcntDry:rgnDw
## daybefore
## hmdty_tmt_D
## regionDewlp
## regionDorsm
## regionVntrm
## reginMtPtch
## dybfr:hm__D
## hmdty_tmt_prcntDry:rgnDw  0.003
```

```
## hmdty_tmt_prcntDry:rgnDr -0.004  0.496
## hmdty_t_D:V               0.000  0.498
## hmdty__D:MP              -0.002  0.491
##                          hmdty_tmt_prcntDry:rgnDr h__D:V
## daybefore
## hmdty_tmt_D
## regionDewlp
## regionDorsm
## regionVntrm
## reginMtPtch
## dybfr:hm__D
## hmdty_tmt_prcntDry:rgnDw
## hmdty_tmt_prcntDry:rgnDr
## hmdty_t_D:V               0.498
## hmdty__D:MP               0.492                    0.494
```

```
drop1(CEWL_mod2)
```

```
## Single term deletions
##
## Model:
## TEWL_g_m2h ~ day * humidity_tmt_percent + humidity_tmt_percent *
##     region + (1 | trial_number/individual_ID)
##                              npar    AIC
## <none>                            2686.6
## day:humidity_tmt_percent        1 2726.1
## humidity_tmt_percent:region     4 2681.7
```

We can drop the humidity:region interaction.

```
CEWL_mod3 <- CEWL %>%
  dplyr::filter(complete.cases(.)) %>%
  lme4::lmer(data = .,
             TEWL_g_m2h ~
             day*humidity_tmt_percent + region +
             (1|trial_number/individual_ID))
summary(CEWL_mod3)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula:
## TEWL_g_m2h ~ day * humidity_tmt_percent + region + (1 | trial_number/individual_ID)
##    Data: .
##
## REML criterion at convergence: 2633.8
##
## Scaled residuals:
##     Min      1Q  Median      3Q     Max
## -2.1225 -0.6012 -0.0847  0.4543  3.8778
##
## Random effects:
##  Groups                      Name        Variance Std.Dev.
##  individual_ID:trial_number (Intercept)  31.92    5.650
##  trial_number               (Intercept)  42.06    6.486
##  Residual                                145.98   12.082
## Number of obs: 335, groups:  individual_ID:trial_number, 35; trial_number, 4
##
```

```
## Fixed effects:
##                                   Estimate Std. Error t value
## (Intercept)                        35.5635     3.9735   8.950
## daybefore                         -11.2321     1.8548  -6.056
## humidity_tmt_percentDry           -13.9332     2.6578  -5.242
## regionDewlap                        0.1719     2.0806   0.083
## regionDorsum                       -2.6057     2.0804  -1.252
## regionVentrum                      12.8429     2.0721   6.198
## regionMite Patch                    0.6519     2.0980   0.311
## daybefore:humidity_tmt_percentDry  17.4957     2.6643   6.567
##
## Correlation of Fixed Effects:
##             (Intr) daybfr hmd__D rgnDwl rgnDrs rgnVnt rgnMtP
## daybefore   -0.220
## hmdty_tmt_D -0.325  0.333
## regionDewlp -0.264  0.013  0.009
## regionDorsm -0.257 -0.006 -0.005  0.496
## regionVntrm -0.261  0.000  0.000  0.498  0.498
## reginMtPtch -0.256 -0.007  0.000  0.492  0.492  0.494
## dybfr:hm__D  0.155 -0.696 -0.479 -0.014  0.004  0.000  0.008
```

```
drop1(CEWL_mod3)
```

```
## Single term deletions
##
## Model:
## TEWL_g_m2h ~ day * humidity_tmt_percent + region + (1 | trial_number/individual_ID)
##                            npar    AIC
## <none>                          2681.7
## region                        4 2736.6
## day:humidity_tmt_percent      1 2720.8
```

The model is best with all the parameters currently included in model 3.

Change in CEWL

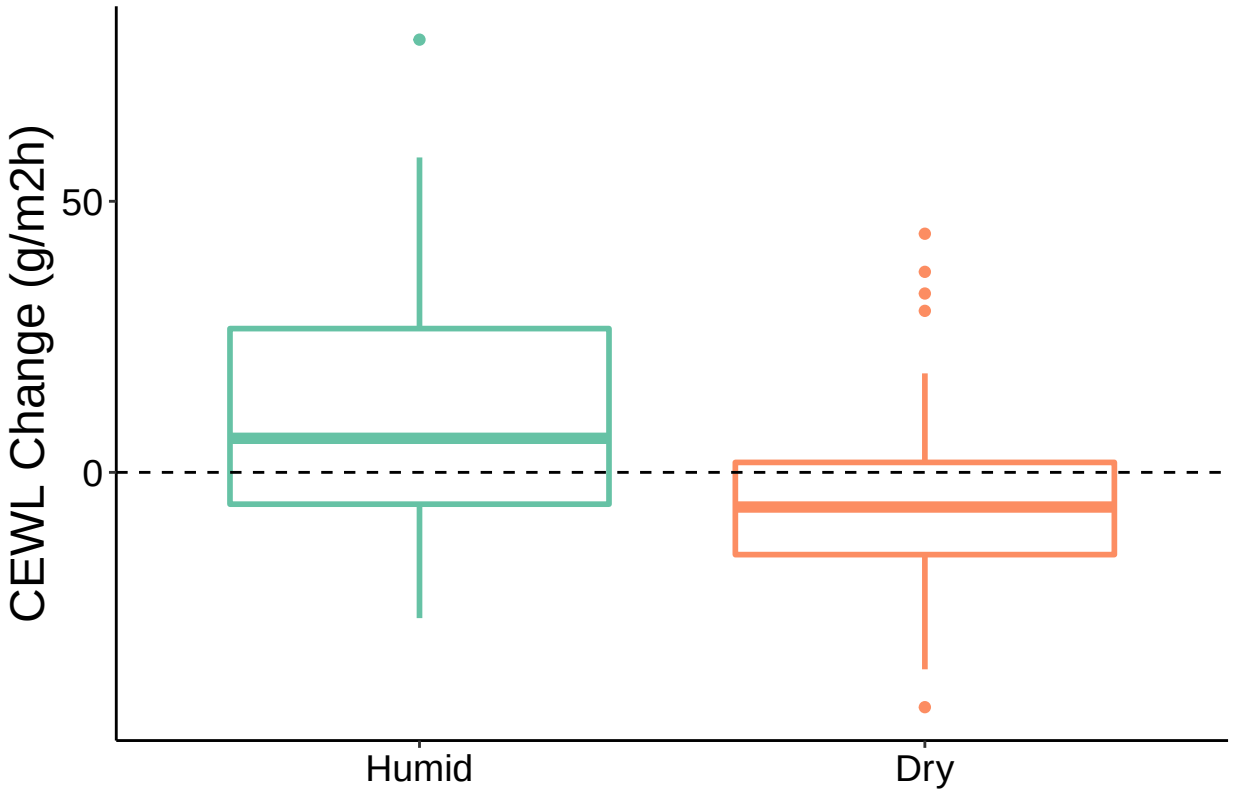
```
CEWL_before <- CEWL %>%
  dplyr::filter(day == "before") %>%
  dplyr::select(CEWL_before = TEWL_g_m2h,
                humidity_tmt_percent, trial_number,
                individual_ID, region)
CEWL_after <- CEWL %>%
  dplyr::filter(day == "after") %>%
  dplyr::select(CEWL_after = TEWL_g_m2h,
                humidity_tmt_percent, trial_number,
                individual_ID, region)

CEWL_diffs <- CEWL_before %>%
  left_join(CEWL_after, by = c('individual_ID', 'region',
                'humidity_tmt_percent', 'trial_number')) %>%
  mutate(CEWL_diff = CEWL_after - CEWL_before)
```

plot:

```
CEWL_diffs %>%
  ggplot(data = .) +
  geom_boxplot(aes(x = humidity_tmt_percent,
                   y = CEWL_diff,
                   group = humidity_tmt_percent,
                   color = humidity_tmt_percent
                   ),
               size = 1,
               alpha = 1) +
  #facet_wrap(~humidity_tmt_percent) +
  theme_classic() +
  geom_hline(yintercept = 0, lty = 2) +
  xlab("") +
  ylab("CEWL Change (g/m2h)") +
  #annotate("text", x = 1.5, y = 45,
   #        label = "paste(italic(p), \" = 0.0152\")",
    #       parse = TRUE,
     #      size = 6) +
  #ylim(10, 50) +
  #scale_x_discrete(labels = c("F" = "Female",
   #                           "M" = "Male")) +
  scale_color_brewer(palette = "Set2") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 18),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 14),
        legend.text.align = 0,
        legend.position = "none"
)
```

```
## Warning: Removed 3 rows containing non-finite values (stat_boxplot).
```



model:

```
CEW_diffs_mod <- lm(data = CEWL_diffs,
                   CEWL_diff ~ humidity_tmt_percent)
summary(CEW_diffs_mod)
```

```
##
## Call:
## lm(formula = CEWL_diff ~ humidity_tmt_percent, data = CEWL_diffs)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -38.213 -13.185  -1.588  10.250  68.497
##
## Coefficients:
##                          Estimate Std. Error t value Pr(>|t|)
## (Intercept)                11.323      2.126   5.326 3.40e-07 ***
## humidity_tmt_percentDry   -17.215      3.045  -5.653 7.18e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 19.25 on 158 degrees of freedom
##   (3 observations deleted due to missingness)
## Multiple R-squared:  0.1682, Adjusted R-squared:  0.163
## F-statistic: 31.96 on 1 and 158 DF,  p-value: 7.18e-08
```

Rehydration. . .

Data

First, get only the data for before experiment, after experiment, and after rehab.

```
summary(all_dat)
```

```
##       date            individual_ID     mass_g      hematocrit_percent
##  Min.   :2021-04-19   37     :  6   Min.   : 6.70   Min.   :12.00
##  1st Qu.:2021-04-30   39     :  6   1st Qu.:10.20   1st Qu.:24.00
##  Median :2021-05-07   40     :  6   Median :11.50   Median :30.00
##  Mean   :2021-05-06   49     :  6   Mean   :11.27   Mean   :29.58
##  3rd Qu.:2021-05-13   52     :  6   3rd Qu.:12.60   3rd Qu.:35.00
##  Max.   :2021-05-20   47     :  5   Max.   :15.00   Max.   :54.00
##                       (Other):116                   NA's   :12
##       type    osmolality_mmol_kg temp_tmt_C humidity_tmt_percent trial_number
##  exp    :82   Min.   :298.0      25:151     Humid:77             1:35
##  rehab  :34   1st Qu.:342.8                 Dry  :74             2:24
##  capture:35   Median :359.0                                      3:44
##               Mean   :362.5                                      4:48
##               3rd Qu.:379.0
##               Max.   :441.0
##               NA's   :15
##     conclusion      SVL_mm       capture_date            day
##  complete:151   Min.   :59.00   Min.   :2021-04-19   Min.   : 0.000
##                 1st Qu.:66.00   1st Qu.:2021-04-26   1st Qu.: 2.000
##                 Median :68.00   Median :2021-05-03   Median : 4.000
##                 Mean   :67.45   Mean   :2021-04-30   Mean   : 5.424
##                 3rd Qu.:70.00   3rd Qu.:2021-05-10   3rd Qu.: 9.000
##                 Max.   :73.00   Max.   :2021-05-10   Max.   :11.000
##
##       SMI
##  Min.   : 7.343
##  1st Qu.: 8.990
##  Median :10.011
##  Mean   : 9.983
##  3rd Qu.:10.751
##  Max.   :13.970
##
```

```
rehydrat_dat <- all_dat %>%
  dplyr::filter(day %in% c(0, 8, 9, 10, 11))
rehydrat_dat$day <- factor(rehydrat_dat$day,
                           levels = c(0, 8, 9, 10, 11),
                           labels = c("Before Experiment",
                                      "After Experiment",
                                      "After Experiment",
                                      "After Rehydration",
                                      "After Rehydration"))
summary(rehydrat_dat)
```

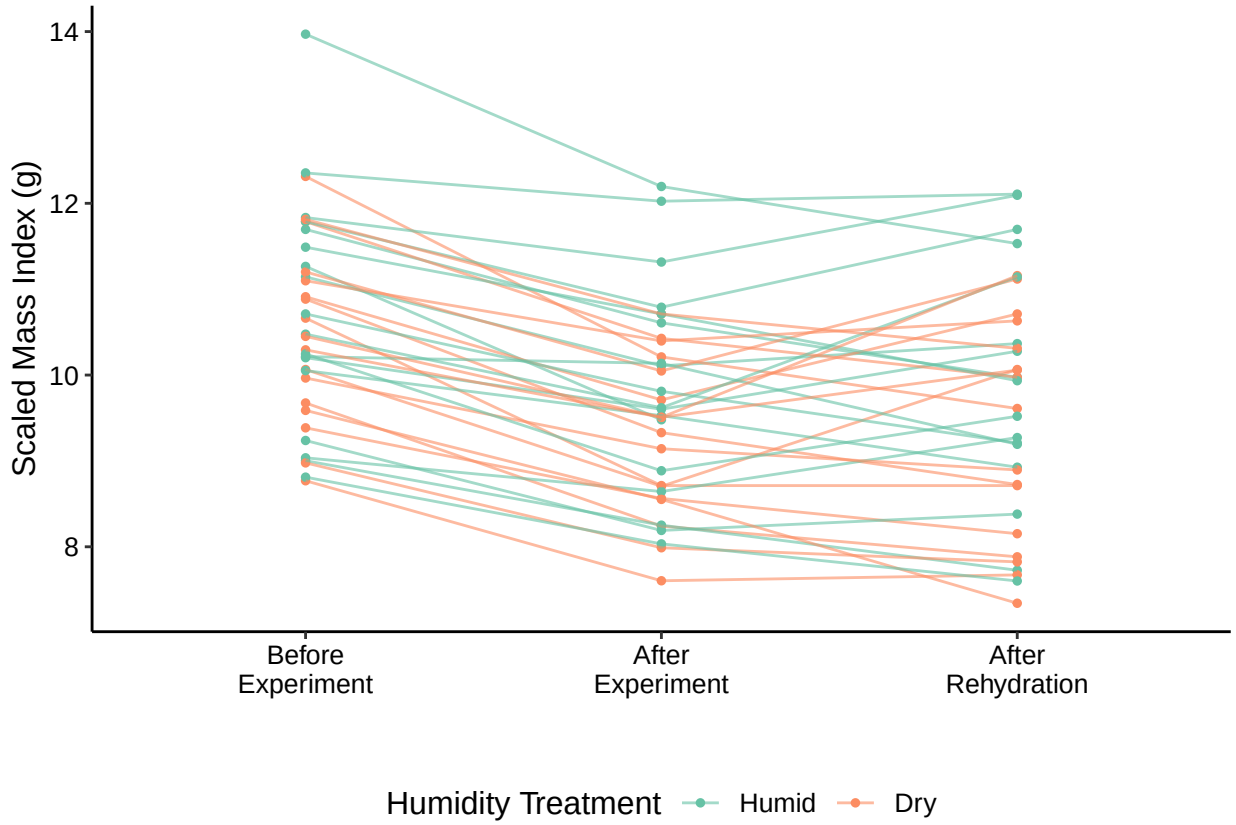
```
##       date            individual_ID     mass_g      hematocrit_percent
##  Min.   :2021-04-19   37     : 3    Min.   : 6.70   Min.   :14.00
##  1st Qu.:2021-05-03   39     : 3    1st Qu.:10.20   1st Qu.:23.88
##  Median :2021-05-10   40     : 3    Median :11.40   Median :29.50
```

```
##  Mean   :2021-05-07   49     : 3    Mean   :11.35   Mean   :29.84
##  3rd Qu.:2021-05-13   52     : 3    3rd Qu.:12.80   3rd Qu.:35.00
##  Max.   :2021-05-20   54     : 3    Max.   :15.00   Max.   :54.00
##                       (Other):86                    NA's   :12
##       type    osmolality_mmol_kg temp_tmt_C humidity_tmt_percent trial_number
##  exp    :35   Min.   :298.0      25:104     Humid:53             1:17
##  rehab  :34   1st Qu.:347.0                 Dry  :51             2:18
##  capture:35   Median :369.0                                      3:33
##               Mean   :368.5                                      4:36
##               3rd Qu.:389.0
##               Max.   :441.0
##               NA's   :15
##     conclusion      SVL_mm       capture_date                         day
##  complete:104   Min.   :59.00   Min.   :2021-04-19   Before Experiment:35
##                 1st Qu.:66.00   1st Qu.:2021-04-26   After Experiment :35
##                 Median :68.00   Median :2021-05-03   After Rehydration:34
##                 Mean   :67.68   Mean   :2021-05-01
##                 3rd Qu.:70.00   3rd Qu.:2021-05-10
##                 Max.   :73.00   Max.   :2021-05-10
##
##       SMI
##  Min.   : 7.343
##  1st Qu.: 8.965
##  Median :10.011
##  Mean   : 9.947
##  3rd Qu.:10.731
##  Max.   :13.970
##
```

SMI

```
rehydrat_dat %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = SMI,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 1) +
  geom_line(aes(x = day,
                y = SMI,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.6) +
  theme_classic() +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  scale_x_discrete(labels = c("Before\nExperiment",
                              "After\nExperiment",
                              "After\nRehydration")) +
  xlab("") +
  xlab("") +
  ylab("Scaled Mass Index (g)") +
```

```
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
        ) -> rehab_SMI_fig
rehab_SMI_fig
```



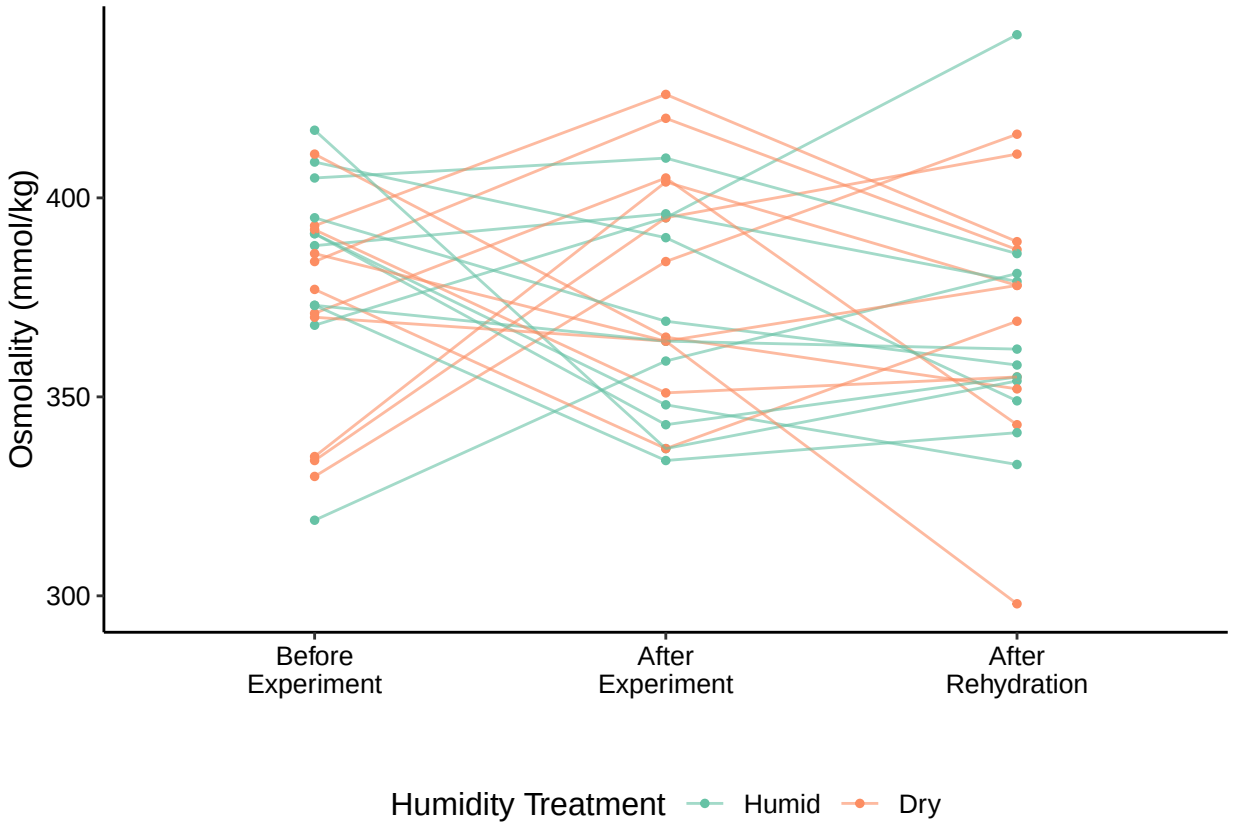

```
# export figure
ggsave(filename = "rehab_SMI_fig.jpeg",
       plot = rehab_SMI_fig,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 5, height = 4)
```

Osmolality

first, make a list of all the IDs that have a post-rehab osmolality measurement, since this has a lot of missing data

```
rehab_osmols <- rehydrat_dat %>%
  dplyr::filter(day == "After Rehydration") %>%
  dplyr::filter(complete.cases(osmolality_mmol_kg))
```

```
rehydrat_dat %>%
  dplyr::filter(individual_ID %in% rehab_osmols$individual_ID) %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = osmolality_mmol_kg,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 1) +
  geom_line(aes(x = day,
                y = osmolality_mmol_kg,
                group = individual_ID,
                color = humidity_tmt_percent),
            alpha = 0.6) +
  theme_classic() +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  scale_x_discrete(labels = c("Before\nExperiment",
                              "After\nExperiment",
                              "After\nRehydration")) +
  xlab("") +
  xlab("") +
  ylab("Osmolality (mmol/kg)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
  ) -> rehab_osml_fig
rehab_osml_fig
```




```
# export figure
ggsave(filename = "rehab_osml_fig.jpeg",
       plot = rehab_osml_fig,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
       width = 5, height = 4)
```

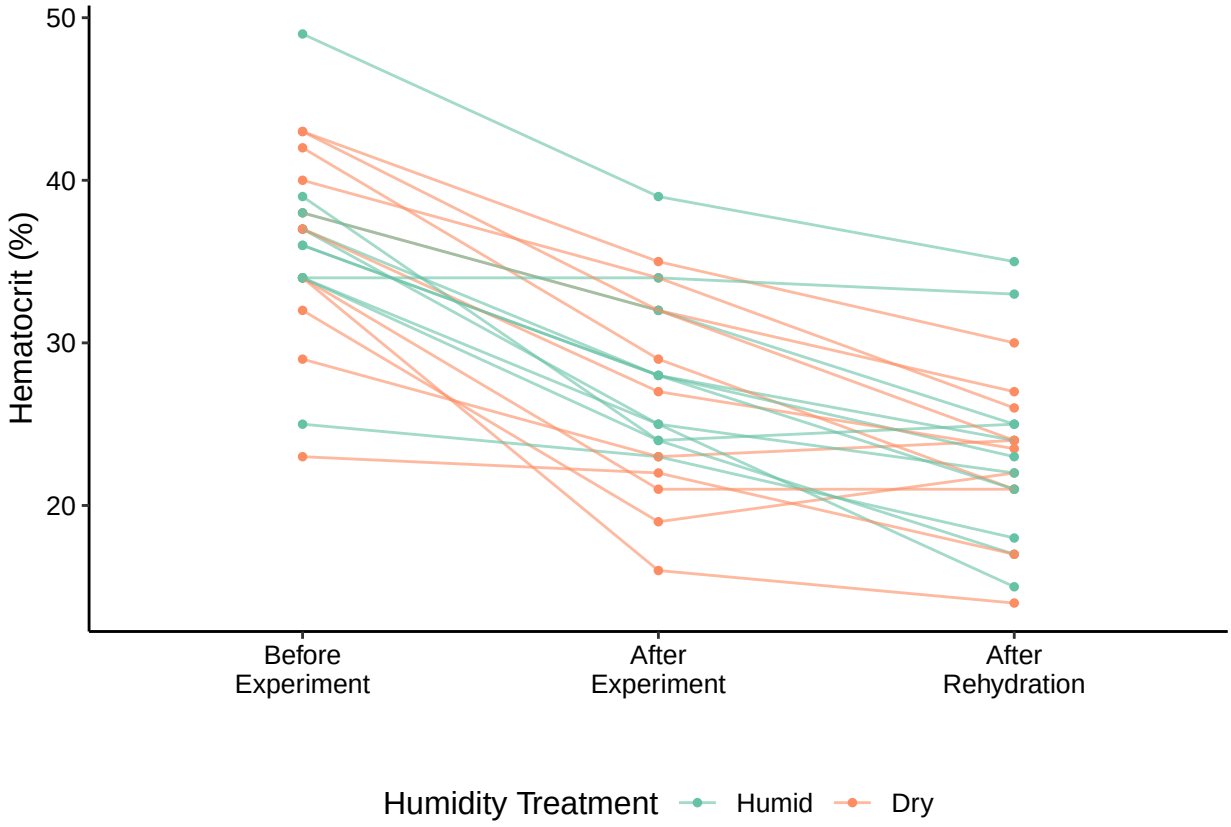
Hematocrit

first, make a list of all the IDs that have all three measurements:

```
rehab_hct <- rehydrat_dat %>%
  dplyr::filter(day == "After Rehydration") %>%
  dplyr::filter(complete.cases(hematocrit_percent))
```

```
rehydrat_dat %>%
  dplyr::filter(individual_ID %in% rehab_osmols$individual_ID) %>%
  ggplot(data = .) +
  geom_point(aes(x = day,
                 y = hematocrit_percent,
                 color = humidity_tmt_percent
                 ),
             size = 1,
             alpha = 1) +
  geom_line(aes(x = day,
                y = hematocrit_percent,
                group = individual_ID,
```

```
                color = humidity_tmt_percent),
            alpha = 0.6) +
  theme_classic() +
  scale_color_brewer(palette = "Set2",
                     name = "Humidity Treatment") +
  scale_x_discrete(labels = c("Before\nExperiment",
                              "After\nExperiment",
                              "After\nRehydration")) +
  xlab("") +
  ylab("Hematocrit (%)") +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                 family = "sans",
                                 size = 10),
        legend.text.align = 0,
        legend.position = "bottom"
  ) -> rehab_hct_fig
rehab_hct_fig
```




```
# export figure
ggsave(filename = "rehab_hct_fig.jpeg",
       plot = rehab_hct_fig,
       path = "./final_figures",
       device = "jpeg",
       dpi = 1200,
```

```
       width = 5, height = 4)
```