

Climate Water Loss Experiment - Treatment Hydration Analysis

Savannah Weaver

2021

Contents

Packages	2
Background and Goals	3
Data	3
Load	3
Split	4
Check	7
Means by Day	10
End Values Only	11
delta CEWL	11
Experiment Models	11
Body Condition	11
Building	11
Assumptions	12
Comparison	17
Hematocrit	18
Building	18
Assumptions	18
Comparison	23
Osmolality	24
Building	24
Assumptions	25
Comparison	30
Body Temp	31
CEWL	37
t-tests	37
Building	38
Assumptions	39
Comparison	44
Group Export	45
Effect Estimates	46
End Value CIs	46
Rate of Change	46
CEWL ~ VPD	47
Recovery Models	55
SMI	55

Hematocrit	56
Osmolality	56
Group Export	57
Figures	57
Colors & Shapes	57
SMI	58
Ind + Means	58
Means Only <i>MS</i>	59
LM + SE	61
Ending Values <i>MS</i>	62
Hct	65
Ind + Means	65
Means Only <i>MS</i>	66
Ending Values <i>MS</i>	68
Osml	71
Ind + Means	71
Means Only <i>MS</i>	72
Stats! Check Pairwise Diffs ~ Time	74
LM + SE	75
Ending Values <i>MS</i>	76
CEWL	79
Ind + Means	79
Means Only <i>MS</i>	80
LM + SE	82
delta CEWL ~ VPD	84
Ending Values <i>MS</i>	86
Exp CEWL ~ Osml	89
Multi-Figures	90

Packages

```

`%nin%` = Negate(`%in%`)
if (!require("tidyverse")) install.packages("tidyverse")
library("tidyverse") # workflow and plots
if (!require("lme4")) install.packages("lme4")
library("lme4") # for LMMs
if (!require("lmerTest")) install.packages("lmerTest")
library("lmerTest") # for p-values
if (!require("ggpubr")) install.packages("ggpubr")
library("ggpubr") # for multi-ggplot figs
if (!require("UsingR")) install.packages("UsingR")
library("UsingR") # simple.eda model assumption checker
if (!require("broom.mixed")) install.packages("broom.mixed")
library("broom.mixed") # lmer model export
if (!require("emmeans")) install.packages('emmeans')
library('emmeans')
if (!require("car")) install.packages("car")
library("car") # VIFs
if (!require("AICcmodavg")) install.packages("AICcmodavg")
library("AICcmodavg") # model selection
if (!require("MuMIn")) install.packages("MuMIn")
library("MuMIn") # model selection

```

```
if (!require("RColorBrewer")) install.packages("RColorBrewer")
library("RColorBrewer") # color
```

Background and Goals

This data was collected June - August by Master's student Savannah Weaver, advisor Dr. Emily Taylor, and research assistants Tess McIntyre and Taylor Van Rossum. Adult male *Sceloporus occidentalis* were caught across the Cal Poly campus then acclimated to 4 different climate treatments. **This R file analyzes the effect of experimental climate treatments on lizard body condition, osmotic balance, and osmoregulation.** Please refer to the published scientific journal article for full details.

Data

Load

Read-in data that was compiled, formatted, and checked for completeness in 'wrangling_general'. See that file for information related to the variables.

```
dat <- read_rds("./data/analysis_data_experiment.RDS") %>%
  # add VPD values by tmt and trial group
  left_join(read_rds("./data/HOBO_tmt_trial_diffs.RDS"),
            by = c('tmt', 'trial_number'='trial')) %>%
  # add the per tmt group calculation of VPD from analysis_HOBO
  mutate(VPD_kPa = case_when(substr(tmt, 1, 6) == "Hot Dr" ~ 3.82,
                              substr(tmt, 1, 6) == "Hot Hu" ~ 1.07,
                              substr(tmt, 1, 6) == "Cool D" ~ 2.50,
                              substr(tmt, 1, 6) == "Cool H" ~ 0.64))

summary(dat)
```

```
## measurement_date      type      individual_ID      mass_g
## Min.   :2021-06-16    exp :804    201      : 7    Min.   : 7.00
## 1st Qu.:2021-07-01    rehab:132  202      : 7    1st Qu.: 9.50
## Median :2021-07-25                                203      : 7    Median :10.60
## Mean   :2021-07-22                                204      : 7    Mean   :10.64
## 3rd Qu.:2021-08-14                                205      : 7    3rd Qu.:11.70
## Max.   :2021-09-01                                206      : 7    Max.   :17.40
##                                     (Other):894
## hematocrit_percent trial_number temp_tmt  humidity_tmt  SVL_mm
## Min.   :13.00      1:175      Hot :467    Humid:468    Min.   :60.00
## 1st Qu.:26.00      2:203      Cool:469    Dry :468     1st Qu.:66.00
## Median :32.00      3:231                                Median :67.00
## Mean   :31.99      4:189                                Mean   :67.74
## 3rd Qu.:37.00      5:138                                3rd Qu.:70.00
## Max.   :52.00                                Max.   :77.00
## NA's      :408
##                tmt                day_n                day_factor osmolality_mmol_kg_mean
## Cool Humid (0.6 kPa):238    Min.   : 0.000    0 :134    Min.   :295.3
## Hot Humid (1.1 kPa) :230    1st Qu.: 4.000    4 :134    1st Qu.:336.1
## Cool Dry (2.5 kPa)  :231    Median : 6.000    5 :134    Median :351.3
## Hot Dry (3.8 kPa)   :237    Mean   : 5.705    6 :134    Mean   :354.3
##                                3rd Qu.: 8.000    7 :134    3rd Qu.:370.0
##                                Max.   :10.000    8 :134    Max.   :471.5
##                                10:132    NA's   :414
```

```
## CEWL_g_m2h_mean    msmt_temp_C    msmt_RH_percent    cloacal_temp_C
## Min.      : 7.152    Min.      :24.80    Min.      :25.52    Min.      :23.00
## 1st Qu.:19.755    1st Qu.:26.20    1st Qu.:46.11    1st Qu.:25.00
## Median :24.152    Median :26.74    Median :47.88    Median :26.00
## Mean      :24.767    Mean      :26.72    Mean      :46.74    Mean      :25.92
## 3rd Qu.:28.505    3rd Qu.:27.11    3rd Qu.:50.50    3rd Qu.:27.00
## Max.      :56.066    Max.      :29.20    Max.      :56.16    Max.      :30.00
## NA's      :669      NA's      :668      NA's      :668      NA's      :668
## msmt_temp_K        e_s_kPa_m        e_a_kPa_m        msmt_VPD_kPa
## Min.      :297.9    Min.      :3.219    Min.      :0.9894    Min.      :1.486
## 1st Qu.:299.4    1st Qu.:3.504    1st Qu.:1.6464    1st Qu.:1.767
## Median :299.9    Median :3.620    Median :1.7411    Median :1.853
## Mean      :299.9    Mean      :3.620    Mean      :1.6833    Mean      :1.937
## 3rd Qu.:300.3    3rd Qu.:3.701    3rd Qu.:1.7992    3rd Qu.:2.012
## Max.      :302.4    Max.      :4.194    Max.      :1.9326    Max.      :3.021
## NA's      :668      NA's      :668      NA's      :668      NA's      :668
## SMI                temp_mean_tmttrial    temp_SD_tmttrial    humidity_mean_tmttrial
## Min.      : 6.747    Min.      :23.30    Min.      :0.5966    Min.      :13.75
## 1st Qu.: 9.714    1st Qu.:24.05    1st Qu.:0.7828    1st Qu.:29.21
## Median :10.594    Median :24.88    Median :1.0461    Median :45.24
## Mean      :10.599    Mean      :29.60    Mean      :1.1513    Mean      :52.94
## 3rd Qu.:11.390    3rd Qu.:35.05    3rd Qu.:1.5191    3rd Qu.:82.84
## Max.      :15.063    Max.      :36.00    Max.      :1.8447    Max.      :93.15
##
## humidity_SD_tmttrial    e_s_kPa        VPD_kPa_tmttrial    VPD_kPa
## Min.      : 4.370    Min.      :2.859    Min.      :0.1958    Min.      :0.640
## 1st Qu.: 6.234    1st Qu.:2.992    1st Qu.:0.7925    1st Qu.:0.640
## Median : 7.382    Median :3.142    Median :2.0310    Median :1.785
## Mean      : 8.765    Mean      :4.330    Mean      :1.9985    Mean      :2.010
## 3rd Qu.:12.297    3rd Qu.:5.639    3rd Qu.:3.1520    3rd Qu.:3.820
## Max.      :19.846    Max.      :5.944    Max.      :4.0640    Max.      :3.820
##
```

Split

Make sub-dataframes without recovery data / with only recovery-related data:

```
dat_no_rehab <- dat %>%
  dplyr::filter(day_n %in% c(seq(0,8)))

recovery_values <- dat %>%
  dplyr::filter(day_n == 10) %>%
  dplyr::select(individual_ID,
    end_hct = hematocrit_percent,
    end_osml = osmolality_mmol_kg_mean,
    end_SMI = SMI)

recovery_v_post_exp <- dat %>%
  dplyr::filter(day_n == 8) %>%
  left_join(recovery_values, by = 'individual_ID') %>%
  mutate(delta_osml_10_8 = end_osml - osmolality_mmol_kg_mean,
    delta_hct_10_8 = end_hct - hematocrit_percent,
    delta_SMI_10_8 = end_SMI - SMI)
summary(recovery_v_post_exp)
```

```

## measurement_date      type      individual_ID      mass_g
## Min.      :2021-06-24    exp :134    201      : 1    Min.      : 7.00
## 1st Qu.:2021-07-04    rehab: 0    202      : 1    1st Qu.: 9.10
## Median :2021-07-28                                203      : 1    Median :10.20
## Mean      :2021-07-25                                204      : 1    Mean      :10.13
## 3rd Qu.:2021-08-16                                205      : 1    3rd Qu.:11.10
## Max.      :2021-08-30                                206      : 1    Max.      :14.20
##
##                                (Other):128
## hematocrit_percent trial_number temp_tmt humidity_tmt      SVL_mm
## Min.      :13.00      1:25      Hot :67      Humid:67      Min.      :60.00
## 1st Qu.:25.00      2:29      Cool:67      Dry :67      1st Qu.:66.00
## Median :28.00      3:33                                Median :67.00
## Mean      :28.53      4:27                                Mean      :67.73
## 3rd Qu.:32.50      5:20                                3rd Qu.:70.00
## Max.      :48.00                                Max.      :77.00
## NA's      :3
##
##                                tmt      day_n      day_factor osmolality_mmol_kg_mean
## Cool Humid (0.6 kPa):34    Min.      :8      0 : 0      Min.      :295.3
## Hot Humid (1.1 kPa) :33    1st Qu.:8      4 : 0      1st Qu.:332.6
## Cool Dry (2.5 kPa) :33    Median :8      5 : 0      Median :345.0
## Hot Dry (3.8 kPa) :34    Mean      :8      6 : 0      Mean      :351.2
##                                3rd Qu.:8      7 : 0      3rd Qu.:362.5
##                                Max.      :8      8 :134     Max.      :445.5
##                                10: 0      NA's      :10
## CEWL_g_m2h_mean msmt_temp_C msmt_RH_percent cloacal_temp_C
## Min.      :12.38    Min.      :24.80    Min.      :43.68    Min.      :23.00
## 1st Qu.:23.80    1st Qu.:25.83    1st Qu.:47.13    1st Qu.:25.00
## Median :26.88    Median :26.29    Median :49.74    Median :26.00
## Mean      :28.79    Mean      :26.25    Mean      :49.35    Mean      :25.39
## 3rd Qu.:33.22    3rd Qu.:26.77    3rd Qu.:51.77    3rd Qu.:26.00
## Max.      :56.07    Max.      :27.23    Max.      :56.16    Max.      :28.00
## NA's      :1
## msmt_temp_K      e_s_kPa_m      e_a_kPa_m      msmt_VPD_kPa
## Min.      :297.9    Min.      :3.219    Min.      :1.595    Min.      :1.486
## 1st Qu.:299.0    1st Qu.:3.427    1st Qu.:1.638    1st Qu.:1.661
## Median :299.4    Median :3.523    Median :1.765    Median :1.802
## Mean      :299.4    Mean      :3.517    Mean      :1.734    Mean      :1.784
## 3rd Qu.:299.9    3rd Qu.:3.627    3rd Qu.:1.809    3rd Qu.:1.855
## Max.      :300.4    Max.      :3.728    Max.      :1.933    Max.      :2.098
##
## SMI      temp_mean_tmttrial temp_SD_tmttrial humidity_mean_tmttrial
## Min.      : 7.317    Min.      :23.30    Min.      :0.5966    Min.      :13.75
## 1st Qu.: 9.254    1st Qu.:24.11    1st Qu.:0.7828    1st Qu.:29.21
## Median :10.141    Median :29.74    Median :1.0461    Median :45.24
## Mean      :10.099    Mean      :29.61    Mean      :1.1502    Mean      :52.95
## 3rd Qu.:10.877    3rd Qu.:35.05    3rd Qu.:1.4894    3rd Qu.:82.84
## Max.      :13.545    Max.      :36.00    Max.      :1.8447    Max.      :93.15
##
## humidity_SD_tmttrial e_s_kPa      VPD_kPa_tmttrial      VPD_kPa
## Min.      : 4.370    Min.      :2.859    Min.      :0.1958    Min.      :0.6400
## 1st Qu.: 6.234    1st Qu.:3.001    1st Qu.:0.7925    1st Qu.:0.7475
## Median : 7.382    Median :4.323    Median :2.0310    Median :1.7850
## Mean      : 8.758    Mean      :4.333    Mean      :1.9993    Mean      :2.0108
## 3rd Qu.:11.490    3rd Qu.:5.639    3rd Qu.:3.0278    3rd Qu.:3.4900

```

```
## Max. :19.846      Max. :5.944      Max. :4.0640      Max. :3.8200
##
##      end_hct      end_osml      end_SMI      delta_osml_10_8
## Min. :13.0      Min. :308.0      Min. : 6.747      Min. : -78.33
## 1st Qu.:23.0      1st Qu.:341.2      1st Qu.: 9.603      1st Qu.: -12.25
## Median :25.5      Median :358.3      Median :10.459      Median : 15.00
## Mean :26.6      Mean :360.1      Mean :10.549      Mean : 10.05
## 3rd Qu.:30.0      3rd Qu.:374.8      3rd Qu.:11.532      3rd Qu.: 33.50
## Max. :47.0      Max. :471.5      Max. :15.063      Max. : 86.67
## NA's :4      NA's :3      NA's :2      NA's :12
## delta_hct_10_8      delta_SMI_10_8
## Min. : -22.000      Min. : -1.13086
## 1st Qu.: -5.000      1st Qu.: -0.09969
## Median : -1.000      Median : 0.38598
## Mean : -2.063      Mean : 0.45063
## 3rd Qu.: 1.000      3rd Qu.: 1.00670
## Max. : 11.000      Max. : 2.46394
## NA's :7      NA's :2
```

```
recovery_v_pre_exp <- dat %>%
  dplyr::filter(day_n == 0) %>%
  left_join(recovery_values, by = 'individual_ID') %>%
  mutate(delta_osml_10_0 = end_osml - osmolality_mmol_kg_mean,
         delta_hct_10_0 = end_hct - hematocrit_percent,
         delta_SMI_10_0 = end_SMI - SMI)
summary(recovery_v_pre_exp)
```

```
## measurement_date      type      individual_ID      mass_g
## Min. :2021-06-16      exp :134      201 : 1      Min. : 8.80
## 1st Qu.:2021-06-26      rehab: 0      202 : 1      1st Qu.:10.60
## Median :2021-07-20      203 : 1      Median :11.70
## Mean :2021-07-17      204 : 1      Mean :11.74
## 3rd Qu.:2021-08-08      205 : 1      3rd Qu.:12.70
## Max. :2021-08-22      206 : 1      Max. :17.40
##                      (Other):128
## hematocrit_percent trial_number temp_tmt humidity_tmt      SVL_mm
## Min. :27.00      1:25      Hot:67      Humid:67      Min. :60.00
## 1st Qu.:34.25      2:29      Cool:67      Dry :67      1st Qu.:66.00
## Median :39.00      3:33      3rd Qu.:70.00
## Mean :38.92      4:27      Max. :77.00
## 3rd Qu.:43.00      5:20
## Max. :52.00
##
##      tmt      day_n      day_factor      osmolality_mmol_kg_mean
## Cool Humid (0.6 kPa):34      Min. :0      0 :134      Min. :305.0
## Hot Humid (1.1 kPa) :33      1st Qu.:0      4 : 0      1st Qu.:334.1
## Cool Dry (2.5 kPa) :33      Median :0      5 : 0      Median :344.6
## Hot Dry (3.8 kPa) :34      Mean :0      6 : 0      Mean :348.2
##                      3rd Qu.:0      7 : 0      3rd Qu.:361.9
##                      Max. :0      8 : 0      Max. :395.0
##                      10: 0
## CEWL_g_m2h_mean      msmt_temp_C      msmt_RH_percent      cloacal_temp_C
## Min. : 7.152      Min. :25.90      Min. :25.52      Min. :25.00
## 1st Qu.:17.348      1st Qu.:26.72      1st Qu.:45.77      1st Qu.:26.00
## Median :21.030      Median :26.96      Median :47.13      Median :26.00
```

```
## Mean :20.779 Mean :27.19 Mean :44.14 Mean :26.46
## 3rd Qu.:24.416 3rd Qu.:27.50 3rd Qu.:48.51 3rd Qu.:27.00
## Max. :34.660 Max. :29.20 Max. :53.15 Max. :30.00
##
## msmt_temp_K e_s_kPa_m e_a_kPa_m msmt_VPD_kPa
## Min. :299.1 Min. :3.441 Min. :0.9894 Min. :1.612
## 1st Qu.:299.9 1st Qu.:3.616 1st Qu.:1.6916 1st Qu.:1.846
## Median :300.1 Median :3.669 Median :1.7367 Median :1.937
## Mean :300.3 Mean :3.724 Mean :1.6331 Mean :2.091
## 3rd Qu.:300.6 3rd Qu.:3.790 3rd Qu.:1.7880 3rd Qu.:2.053
## Max. :302.4 Max. :4.194 Max. :1.8502 Max. :3.021
##
## SMI temp_mean_tmttrial temp_SD_tmttrial humidity_mean_tmttrial
## Min. : 9.122 Min. :23.30 Min. :0.5966 Min. :13.75
## 1st Qu.:10.983 1st Qu.:24.11 1st Qu.:0.7828 1st Qu.:29.21
## Median :11.687 Median :29.74 Median :1.0461 Median :45.24
## Mean :11.693 Mean :29.61 Mean :1.1502 Mean :52.95
## 3rd Qu.:12.347 3rd Qu.:35.05 3rd Qu.:1.4894 3rd Qu.:82.84
## Max. :14.263 Max. :36.00 Max. :1.8447 Max. :93.15
##
## humidity_SD_tmttrial e_s_kPa VPD_kPa_tmttrial VPD_kPa
## Min. : 4.370 Min. :2.859 Min. :0.1958 Min. :0.6400
## 1st Qu.: 6.234 1st Qu.:3.001 1st Qu.:0.7925 1st Qu.:0.7475
## Median : 7.382 Median :4.323 Median :2.0310 Median :1.7850
## Mean : 8.758 Mean :4.333 Mean :1.9993 Mean :2.0108
## 3rd Qu.:11.490 3rd Qu.:5.639 3rd Qu.:3.0278 3rd Qu.:3.4900
## Max. :19.846 Max. :5.944 Max. :4.0640 Max. :3.8200
##
## end_hct end_osml end_SMI delta_osml_10_0
## Min. :13.0 Min. :308.0 Min. : 6.747 Min. : -67.33
## 1st Qu.:23.0 1st Qu.:341.2 1st Qu.: 9.603 1st Qu.: -12.08
## Median :25.5 Median :358.3 Median :10.459 Median : 14.00
## Mean :26.6 Mean :360.1 Mean :10.549 Mean : 12.19
## 3rd Qu.:30.0 3rd Qu.:374.8 3rd Qu.:11.532 3rd Qu.: 38.92
## Max. :47.0 Max. :471.5 Max. :15.063 Max. :130.00
## NA's :4 NA's :3 NA's :2 NA's :3
## delta_hct_10_0 delta_SMI_10_0
## Min. : -32.0 Min. : -4.1183
## 1st Qu.: -15.0 1st Qu.: -1.9496
## Median : -12.0 Median : -1.1513
## Mean : -12.4 Mean : -1.1462
## 3rd Qu.: -8.0 3rd Qu.: -0.2646
## Max. : 2.0 Max. : 1.6871
## NA's :4 NA's :2
```

Check

Dates:

```
unique(dat$measurement_date)
```

```
## [1] "2021-06-16" "2021-06-20" "2021-06-21" "2021-06-22" "2021-06-23"
## [6] "2021-06-24" "2021-06-26" "2021-06-30" "2021-07-01" "2021-07-02"
## [11] "2021-07-03" "2021-07-04" "2021-07-06" "2021-07-20" "2021-07-24"
## [16] "2021-07-25" "2021-07-26" "2021-07-27" "2021-07-28" "2021-07-30"
```

```
## [21] "2021-08-08" "2021-08-12" "2021-08-13" "2021-08-14" "2021-08-15"
## [26] "2021-08-16" "2021-08-18" "2021-08-22" "2021-08-26" "2021-08-27"
## [31] "2021-08-28" "2021-08-29" "2021-08-30" "2021-09-01"
```

Number of measurements for each lizard:

```
dat_no_rehab %>%
  group_by(individual_ID) %>%
  summarise(n = n()) %>%
  arrange(n)
```

```
## # A tibble: 134 x 2
##   individual_ID     n
##   <fct>         <int>
## 1 201             6
## 2 202             6
## 3 203             6
## 4 204             6
## 5 205             6
## 6 206             6
## 7 207             6
## 8 208             6
## 9 209             6
## 10 210            6
## # ... with 124 more rows
```

Every lizard has 6 experimental measurements: pre-tmt, mid-tmt, post-tmt, and mass checks on each of the 3 days between mid and post-tmt.

Did any of the treatment groups inherently start out with large differences in response variables?

```
dat %>%
  dplyr::filter(day_n == 0) %>%
  group_by(tmt) %>%
  summarise(mean(mass_g),
            sd(mass_g),
            mean(SMI),
            mean(hematocrit_percent),
            mean(osmolality_mmol_kg_mean),
            mean(CEWL_g_m2h_mean))
```

```
## # A tibble: 4 x 7
##   tmt                `mean(mass_g)` sd(mass_g) mean(~2 mean(~3 mean(~4 mean(~5
##   <fct>                <dbl>         <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 Cool Humid (0.6 kPa)    11.6         1.35    11.7    39.6    351.    20.9
## 2 Hot Humid (1.1 kPa)    11.6         1.75    11.5    37.9    347.    21.4
## 3 Cool Dry (2.5 kPa)     11.8         1.61    11.8    39.3    346.    20.0
## 4 Hot Dry (3.8 kPa)      12.0         1.68    11.8    38.9    347.    20.9
## # ... with abbreviated variable names 1: `sd(mass_g)`, 2: `mean(SMI)`,
## #   3: `mean(hematocrit_percent)`, 4: `mean(osmolality_mmol_kg_mean)`,
## #   5: `mean(CEWL_g_m2h_mean)`
```

There are slight differences, but overall the starting values across groups are more or less the same.

Temp & RH during (all, before and after exp) CEWL measurements:

```
summary(dat_no_rehab)
```

```
## measurement_date      type      individual_ID      mass_g
```



```

## Min. :2021-06-16 exp :804 201 : 6 Min. : 7.00
## 1st Qu.:2021-06-30 rehab: 0 202 : 6 1st Qu.: 9.50
## Median :2021-07-25 203 : 6 Median :10.60
## Mean :2021-07-22 204 : 6 Mean :10.65
## 3rd Qu.:2021-08-13 205 : 6 3rd Qu.:11.60
## Max. :2021-08-30 206 : 6 Max. :17.40
## (Other):768
## hematocrit_percent trial_number temp_tmt humidity_tmt SVL_mm
## Min. :13.00 1:150 Hot :402 Humid:402 Min. :60.00
## 1st Qu.:28.25 2:174 Cool:402 Dry :402 1st Qu.:66.00
## Median :33.00 3:198 Median :67.00
## Mean :33.75 4:162 Mean :67.73
## 3rd Qu.:39.00 5:120 3rd Qu.:70.00
## Max. :52.00 Max. :77.00
## NA's :406
## tmt day_n day_factor osmolality_mmol_kg_mean
## Cool Humid (0.6 kPa):204 Min. :0.0 0 :134 Min. :295.3
## Hot Humid (1.1 kPa) :198 1st Qu.:4.0 4 :134 1st Qu.:334.7
## Cool Dry (2.5 kPa) :198 Median :5.5 5 :134 Median :348.3
## Hot Dry (3.8 kPa) :204 Mean :5.0 6 :134 Mean :352.3
## 3rd Qu.:7.0 7 :134 3rd Qu.:367.4
## Max. :8.0 8 :134 Max. :445.5
## 10: 0 NA's :413
## CEWL_g_m2h_mean msmt_temp_C msmt_RH_percent cloacal_temp_C
## Min. : 7.152 Min. :24.80 Min. :25.52 Min. :23.00
## 1st Qu.:19.755 1st Qu.:26.20 1st Qu.:46.11 1st Qu.:25.00
## Median :24.152 Median :26.74 Median :47.88 Median :26.00
## Mean :24.767 Mean :26.72 Mean :46.74 Mean :25.92
## 3rd Qu.:28.505 3rd Qu.:27.11 3rd Qu.:50.50 3rd Qu.:27.00
## Max. :56.066 Max. :29.20 Max. :56.16 Max. :30.00
## NA's :537 NA's :536 NA's :536 NA's :536
## msmt_temp_K e_s_kPa_m e_a_kPa_m msmt_VPD_kPa
## Min. :297.9 Min. :3.219 Min. :0.9894 Min. :1.486
## 1st Qu.:299.4 1st Qu.:3.504 1st Qu.:1.6464 1st Qu.:1.767
## Median :299.9 Median :3.620 Median :1.7411 Median :1.853
## Mean :299.9 Mean :3.620 Mean :1.6833 Mean :1.937
## 3rd Qu.:300.3 3rd Qu.:3.701 3rd Qu.:1.7992 3rd Qu.:2.012
## Max. :302.4 Max. :4.194 Max. :1.9326 Max. :3.021
## NA's :536 NA's :536 NA's :536 NA's :536
## SMI temp_mean_tmttrial temp_SD_tmttrial humidity_mean_tmttrial
## Min. : 7.317 Min. :23.30 Min. :0.5966 Min. :13.75
## 1st Qu.: 9.748 1st Qu.:24.05 1st Qu.:0.7828 1st Qu.:29.21
## Median :10.624 Median :29.74 Median :1.0461 Median :45.24
## Mean :10.607 Mean :29.61 Mean :1.1502 Mean :52.95
## 3rd Qu.:11.348 3rd Qu.:35.05 3rd Qu.:1.5191 3rd Qu.:82.84
## Max. :14.263 Max. :36.00 Max. :1.8447 Max. :93.15
## humidity_SD_tmttrial e_s_kPa VPD_kPa_tmttrial VPD_kPa
## Min. : 4.370 Min. :2.859 Min. :0.1958 Min. :0.640
## 1st Qu.: 6.234 1st Qu.:2.992 1st Qu.:0.7925 1st Qu.:0.640
## Median : 7.382 Median :4.323 Median :2.0310 Median :1.785
## Mean : 8.758 Mean :4.333 Mean :1.9993 Mean :2.011
## 3rd Qu.:12.297 3rd Qu.:5.639 3rd Qu.:3.1520 3rd Qu.:3.820
## Max. :19.846 Max. :5.944 Max. :4.0640 Max. :3.820

```

```
##
dat_no_rehab %>%
  group_by(type) %>%
  summarise(mean(msmt_temp_C, na.rm = T),
            sd(msmt_temp_C, na.rm = T),
            mean(msmt_RH_percent, na.rm = T),
            sd(msmt_RH_percent, na.rm = T),
            mean(msmt_VPD_kPa, na.rm = T),
            mean(msmt_VPD_kPa, na.rm = T))

## # A tibble: 1 x 6
##   type `mean(msmt_temp_C, na.rm = T)` sd(msmt_temp_C,~1 mean(~2 sd(ms~3 mean(~4
##   <fct>                <dbl>                <dbl>    <dbl>    <dbl>    <dbl>
## 1 exp                    26.7                    0.799    46.7    6.76    1.94
## # ... with abbreviated variable names 1: `sd(msmt_temp_C, na.rm = T)`,
## #   2: `mean(msmt_RH_percent, na.rm = T)`, 3: `sd(msmt_RH_percent, na.rm = T)`,
## #   4: `mean(msmt_VPD_kPa, na.rm = T)`
```

Means by Day

Calculate mean values per day per tmt group.

```
means <- dat %>% # use whole dat because want for both exp and rehyd
  group_by(day_n, tmt) %>%
  summarise(n_lizards = n(),
            mean_CEWL = mean(CEWL_g_m2h_mean, na.rm = T),
            sd_CEWL = sd(CEWL_g_m2h_mean, na.rm = T),
            mean_osml = mean(osmolality_mmol_kg_mean, na.rm = T),
            sd_osml = sd(osmolality_mmol_kg_mean, na.rm = T),
            mean_hct = mean(hematocrit_percent, na.rm = T),
            sd_hct = sd(hematocrit_percent, na.rm = T),
            mean_SMI = mean(SMI, na.rm = T),
            sd_SMI = sd(SMI, na.rm = T)) %>%
  mutate(se_CEWL = (sd_CEWL/sqrt(n_lizards)),
         se_osml = (sd_osml/sqrt(n_lizards)),
         se_hct = (sd_hct/sqrt(n_lizards)),
         se_SMI = (sd_SMI/sqrt(n_lizards)))
```

```
## `summarise()` has grouped output by 'day_n'. You can override using the
## `.groups` argument.
```

```
# get rid of non-defined points
means$mean_CEWL[is.nan(means$mean_CEWL)] <- NA
means$mean_osml[is.nan(means$mean_osml)] <- NA
means$mean_hct[is.nan(means$mean_hct)] <- NA
means$mean_SMI[is.nan(means$mean_SMI)] <- NA
means
```

```
## # A tibble: 28 x 15
## # Groups:   day_n [7]
##   day_n tmt    n_liz~1 mean_~2 sd_CEWL mean_~3 sd_osml mean_~4 sd_hct mean_~5
##   <dbl> <fct>    <int>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1     0 Cool Hu~    34    20.9     4.78    351.     20.3     39.6     5.30     11.7
## 2     0 Hot Hum~    33    21.4     4.85    347.     18.7     37.9     5.46     11.5
## 3     0 Cool Dr~    33    20.0     6.08    346.     20.6     39.3     5.96     11.8
## 4     0 Hot Dry~    34    20.9     5.93    347.     17.9     38.9     5.05     11.8
```

```
## 5      4 Cool Hu~      34      NA      NA      356.      24.5      34.5      5.34      11.2
## 6      4 Hot Hum~     33      NA      NA      359.      22.5      32.0      5.38      10.5
## 7      4 Cool Dr~     33      NA      NA      355.      27.0      35.0      7.02      11.1
## 8      4 Hot Dry~     34      NA      NA      361.      25.8      33.1      5.00      10.4
## 9      5 Cool Hu~     34      NA      NA      NA      NA      NA      NA      11.0
## 10     5 Hot Hum~     33      NA      NA      NA      NA      NA      NA      10.1
## # ... with 18 more rows, 5 more variables: sd_SMI <dbl>, se_CEWL <dbl>,
## #   se_osml <dbl>, se_hct <dbl>, se_SMI <dbl>, and abbreviated variable names
## #   1: n_lizards, 2: mean_CEWL, 3: mean_osml, 4: mean_hct, 5: mean_SMI

# get only means for the very end
end_means <- means %>%
  dplyr::filter(day_n == 8)
#write.csv(end_means, "./results_statistics/exp_end_means.csv")
```

End Values Only

Select for only day=8 values.

```
end_vals <- dat %>%
  dplyr::filter(day_n == 8)
```

delta CEWL

Get a df that only has complete observations that include CEWL values (only obs from before and after the experiment). Then, calculate the CHANGE (delta) in CEWL from before to after the experiment. Because we only measured CEWL at those two time points, it makes more sense to assess the **amount of change** in CEWL for each lizard, rather than measuring the change over time.

```
start_CEWL <- dat_no_rehab %>%
  dplyr::filter(day_n == 0) %>%
  dplyr::select(individual_ID, start_CEWL = CEWL_g_m2h_mean)
dat_no_rehab_deltaCEWL <- dat_no_rehab %>% # initiate new df
  dplyr::filter(complete.cases(CEWL_g_m2h_mean)) %>% # only use obs incl CEWL
  dplyr::filter(day_n == 8) %>% # get only obs for post-exp
  left_join(start_CEWL, by = 'individual_ID') %>% # add start CEWL to both obs for each lizard
  mutate(delta_CEWL = CEWL_g_m2h_mean - start_CEWL) # calculate deltaCEWL after-before experiment
```

Experiment Models

We predicted that there would be effects of day, humidity treatment, temperature treatment, and treatment VPD. However, we can't use the standard backwards model selection because the three treatment variables are collinear (VIF much higher than acceptable) and it leads to issues with changing the sign of the estimates when all three are included together. So, we will run singular models with each treatment variable alone:

response ~ dayhumidity response ~ daytemperature response ~ day*VPD

Then, we will use AIC, RMSE, and R-sq to assess which treatment effect is most important to that response variable.

Body Condition

Building

Build each treatment effect model.

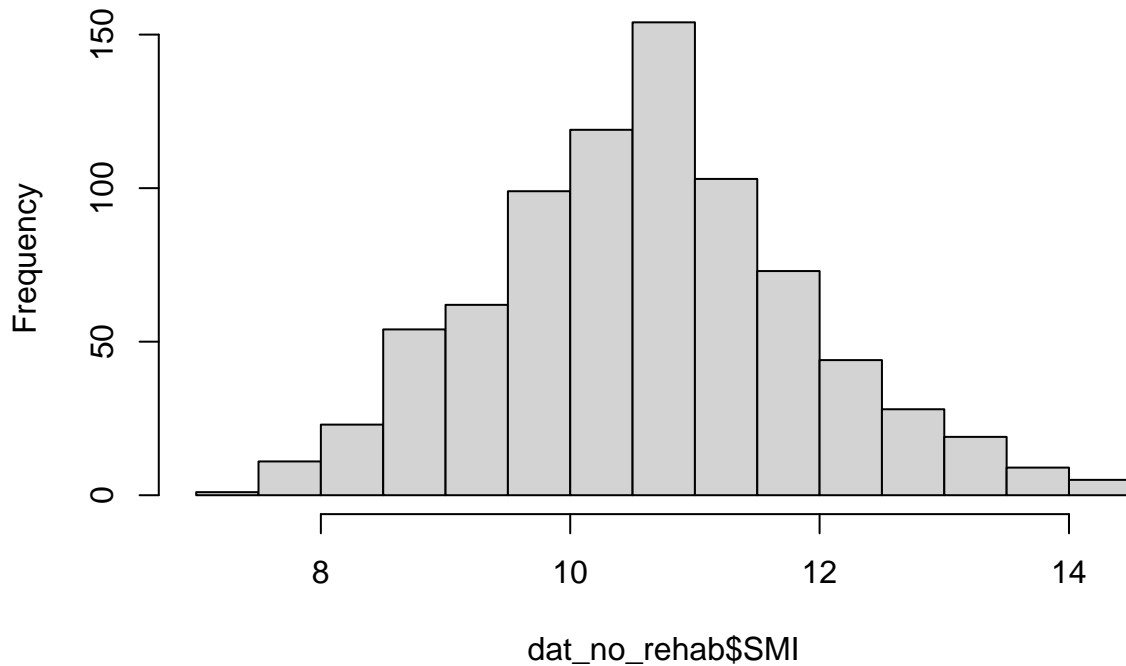
```
SMI_mod_VPD <- lmerTest::lmer(data = dat_no_rehab,
                             SMI ~ day_n * VPD_kPa +
                             (1|trial_number/individual_ID))
SMI_mod_hum <- lmerTest::lmer(data = dat_no_rehab,
                             SMI ~ day_n * humidity_tmt +
                             (1|trial_number/individual_ID))
SMI_mod_temp <- lmerTest::lmer(data = dat_no_rehab,
                              SMI ~ day_n * temp_tmt +
                              (1|trial_number/individual_ID))
```

Assumptions

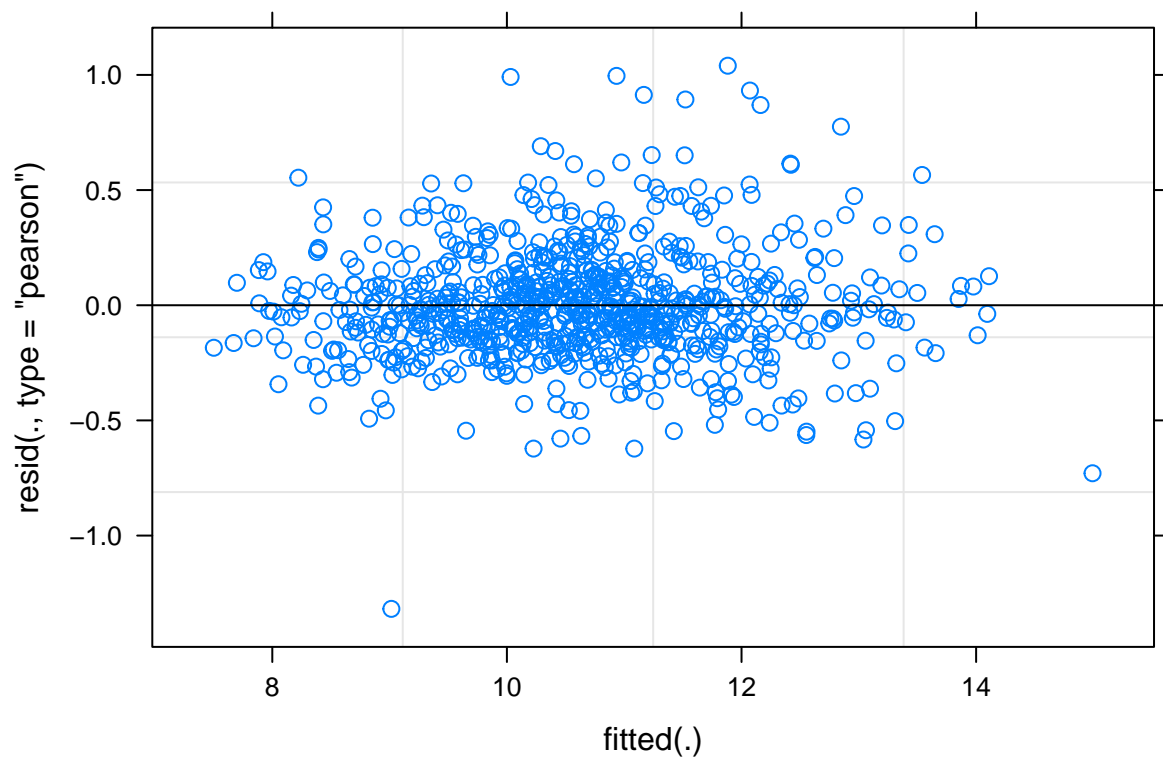
Check linear regression assumptions/conditions.

```
# distribution of SMI
hist(dat_no_rehab$SMI)
```

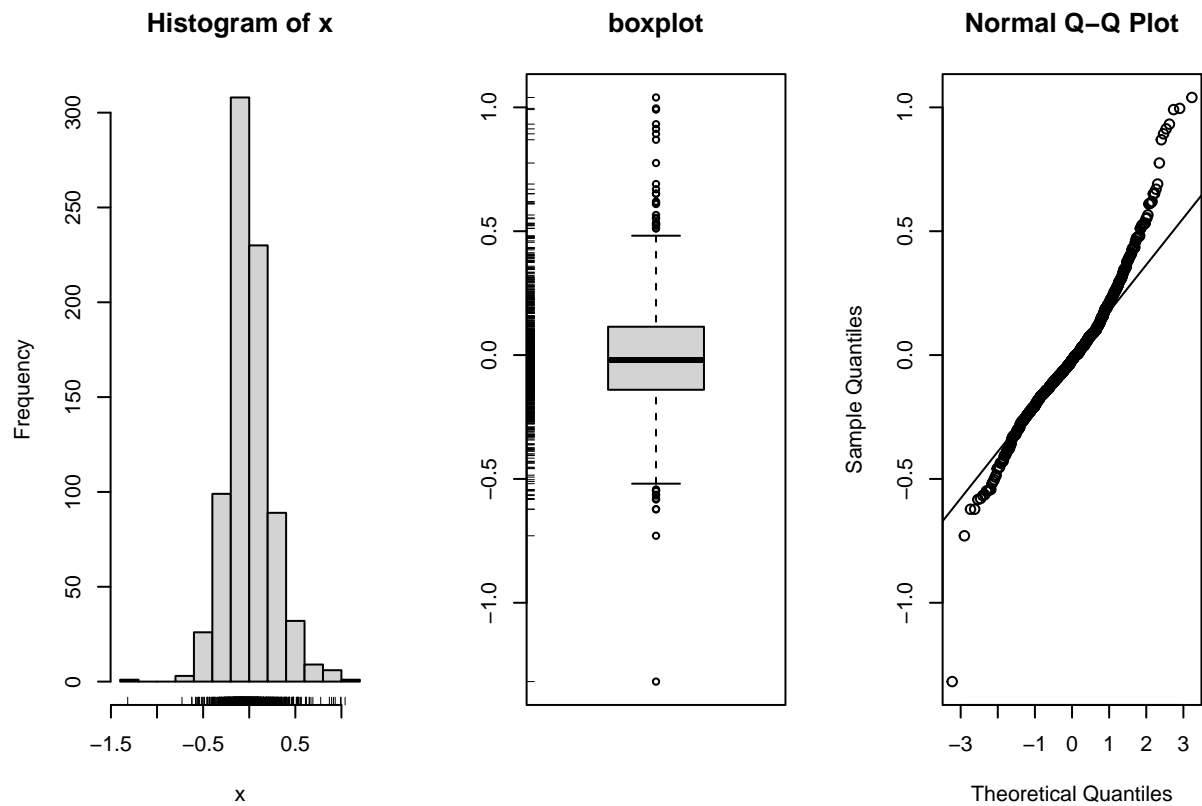
Histogram of dat_no_rehab\$SMI



```
# VPD model
plot(SMI_mod_VPD)
```



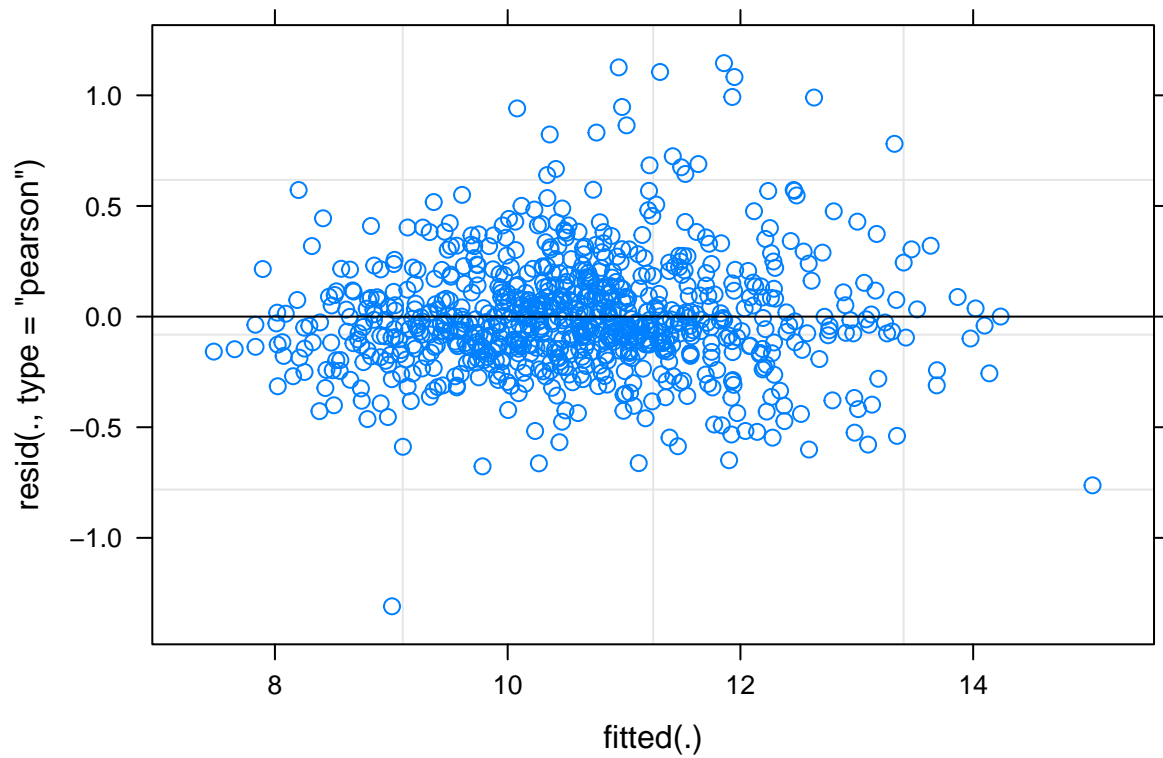
```
simple.eda(residuals(SMI_mod_VPD))
```



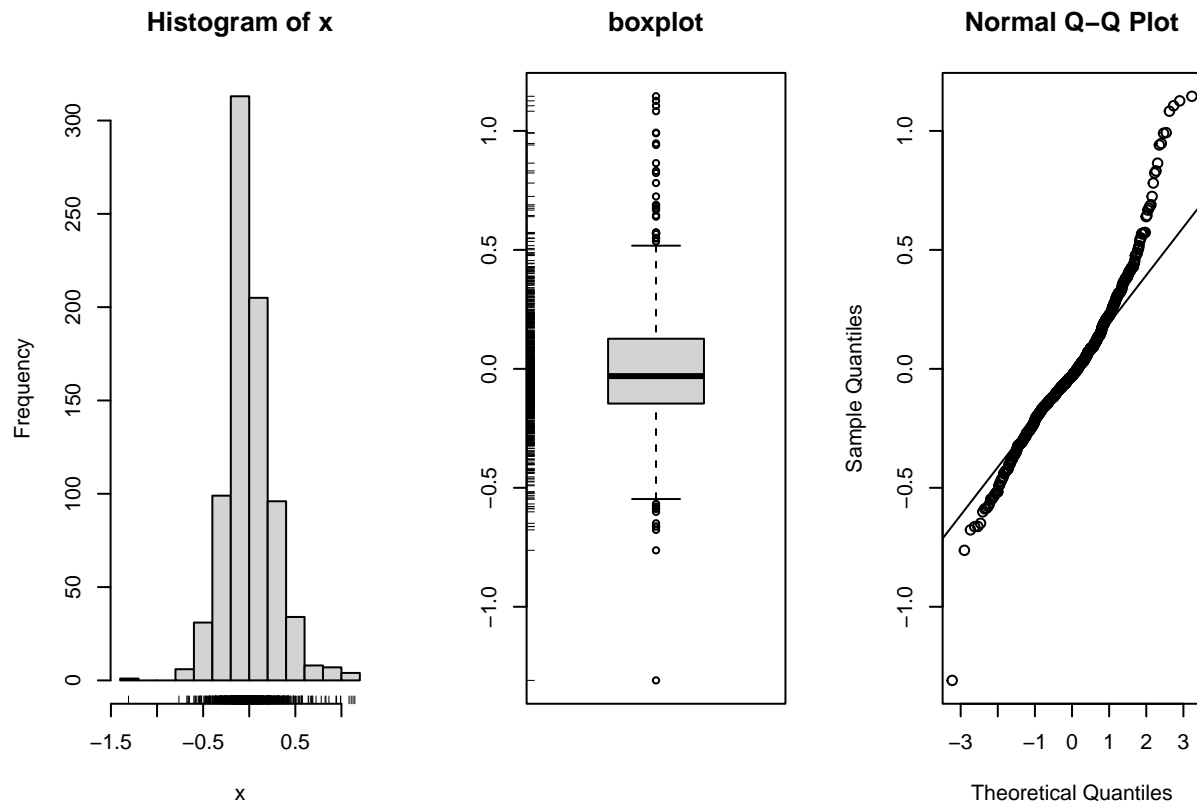
```
shapiro.test(residuals(SMI_mod_VPD))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(SMI_mod_VPD)  
## W = 0.96052, p-value = 6.691e-14
```

```
# humidity model  
plot(SMI_mod_hum)
```



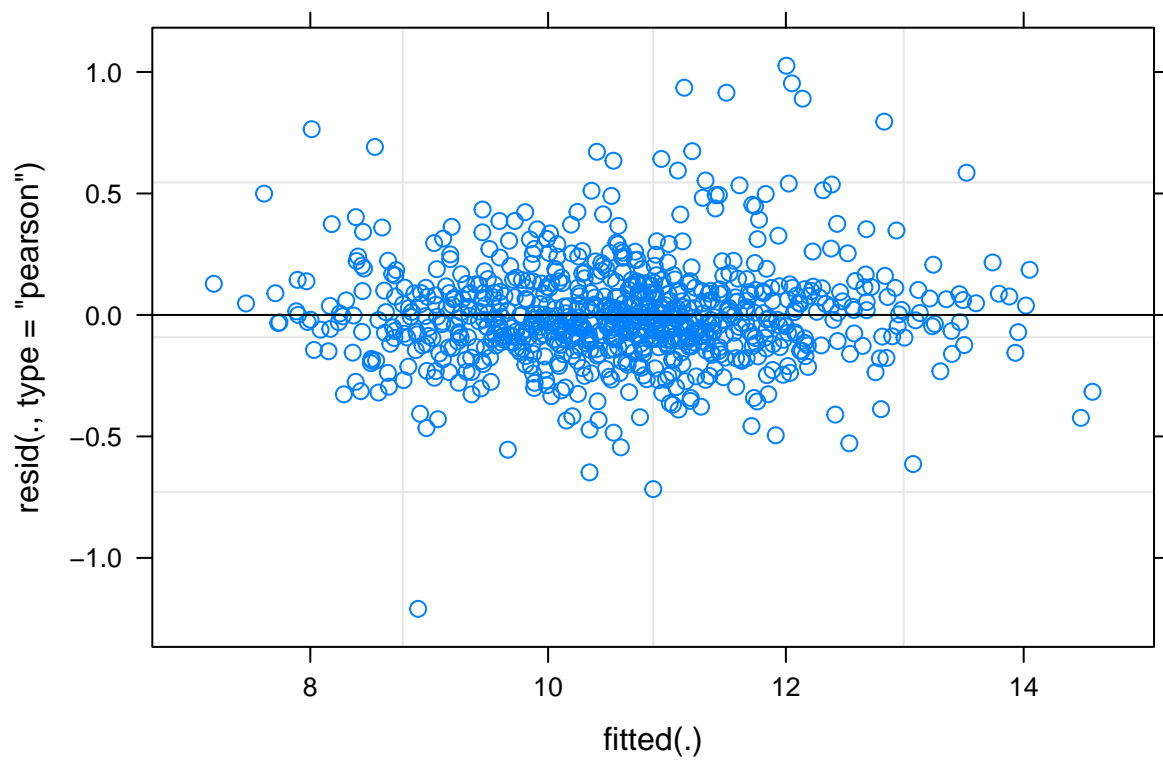
```
simple.eda(residuals(SMI_mod_hum))
```



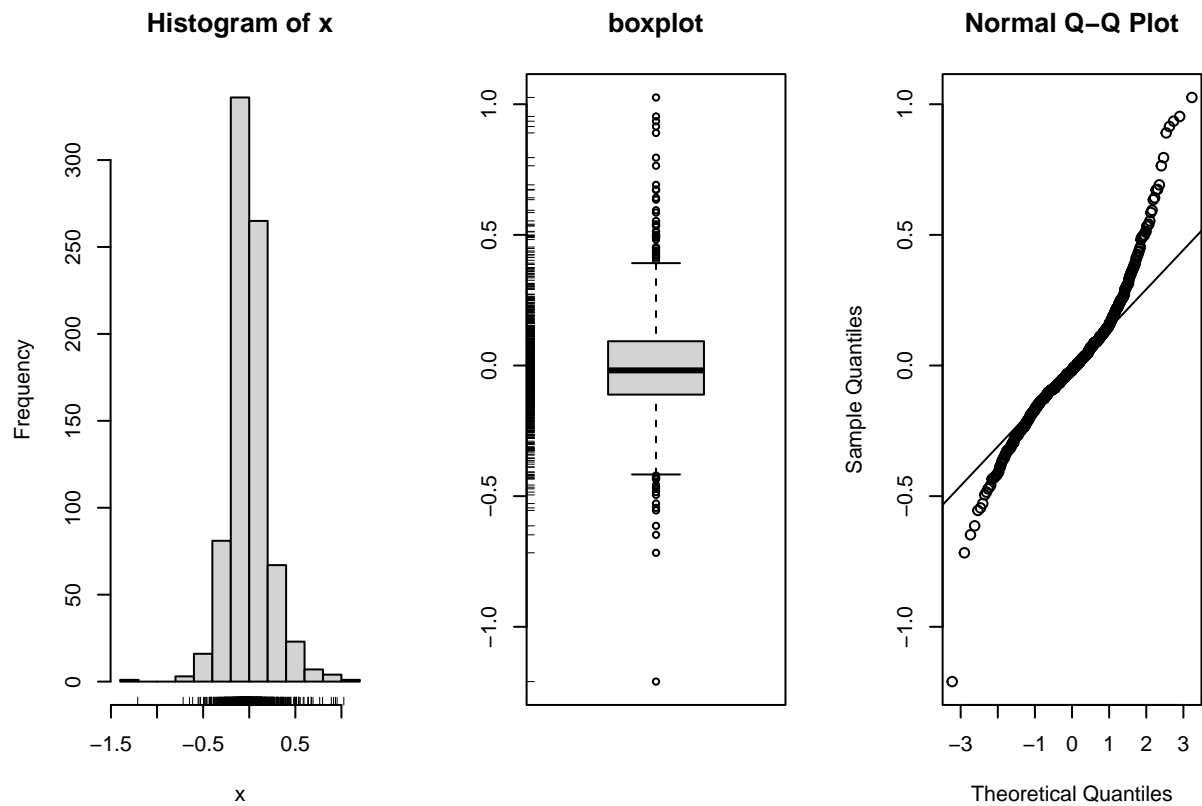
```
shapiro.test(residuals(SMI_mod_hum))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(SMI_mod_hum)
## W = 0.95569, p-value = 7.603e-15
```

```
# temperature model
plot(SMI_mod_temp)
```



```
simple.eda(residuals(SMI_mod_temp))
```




```
shapiro.test(residuals(SMI_mod_temp))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(SMI_mod_temp)  
## W = 0.93877, p-value < 2.2e-16
```

Normality is violated, but linearity, equal error variance, and independence are all good.

Comparison

Now, compare the AIC, RMSE, and R^2 values across models, and the F and p values of the variables for each model.

We... calculate RMSE manually, use the `r.squaredGLMM` function in the `MuMIn` package to get the marginal R^2 , which is how much of the total variance is explained by fixed effects, use the `aictab` function in the `AICmodavg` package to get AIC and deltaAIC values, and get the sum of squares, F, and p-values for each variable from the anova table for each model.

```
# calculate RMSE & R^2  
SMI_RMSE_Rsq <- data.frame(model =  
  c('Day * VPD',  
    'Day * Humidity',  
    'Day * Temp'  
  ),  
  RMSE = c(sqrt(mean((residuals(SMI_mod_VPD))^2)),  
            sqrt(mean((residuals(SMI_mod_hum))^2)),  
            sqrt(mean((residuals(SMI_mod_temp))^2))),  
  # marginal Rsq for the amount of variance  
  # explained by fixed effects only  
  Rsq = c(MuMIn::r.squaredGLMM(SMI_mod_VPD)[,"R2m"],  
          MuMIn::r.squaredGLMM(SMI_mod_hum)[,"R2m"],  
          MuMIn::r.squaredGLMM(SMI_mod_temp)[,"R2m"]))
```

Warning: 'r.squaredGLMM' now calculates a revised statistic. See the help page.

```
# calculate AIC  
SMI_models <- list(SMI_mod_VPD, SMI_mod_hum, SMI_mod_temp)  
EXP_mod_names <- data.frame(model =  
  c('Day * VPD',  
    'Day * Humidity',  
    'Day * Temp'  
  ),  
  )  
SMI_AICc <- data.frame(aictab(cand.set = SMI_models,  
  modnames = EXP_mod_names))
```

Warning in `aictab.AIClmerModLmerTest(cand.set = SMI_models, modnames = EXP_mod_names)`:

Model selection for fixed effects is only appropriate with ML estimation:

REML (default) should only be used to select random effects for a constant set of fixed effects

```
# compare across models  
SMI_across <- SMI_RMSE_Rsq %>%  
  left_join(SMI_AICc, by = 'model') %>%  
  mutate(response = "Body Condition (g)") %>%  
  arrange(Delta_AICc)
```

```

# calculate F & p-values
SMI_VPD_p <- data.frame(anova(SMI_mod_VPD,
                             type = "1",
                             ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * VPD',
         predictor = rownames(.))
SMI_hum_p <- data.frame(anova(SMI_mod_hum,
                             type = "1",
                             ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Humidity',
         predictor = rownames(.))
SMI_temp_p <- data.frame(anova(SMI_mod_temp,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Temp',
         predictor = rownames(.))

# save within model values
SMI_within <- SMI_VPD_p %>%
  rbind(SMI_hum_p) %>%
  rbind(SMI_temp_p) %>%
  mutate(df = paste((NumDF), round(DenDF, 0), sep = ", "),
         response = "Body Condition (g)")

```

Hematocrit

Building

Build each treatment effect model.

```

hct_mod_VPD <- lmerTest::lmer(data = dat_no_rehab,
                             hematocrit_percent ~ day_n * VPD_kPa +
                             (1|trial_number/individual_ID))
hct_mod_hum <- lmerTest::lmer(data = dat_no_rehab,
                             hematocrit_percent ~ day_n * humidity_tmt +
                             (1|trial_number/individual_ID))
hct_mod_temp <- lmerTest::lmer(data = dat_no_rehab,
                              hematocrit_percent ~ day_n * temp_tmt +
                              (1|trial_number/individual_ID))

```

Assumptions

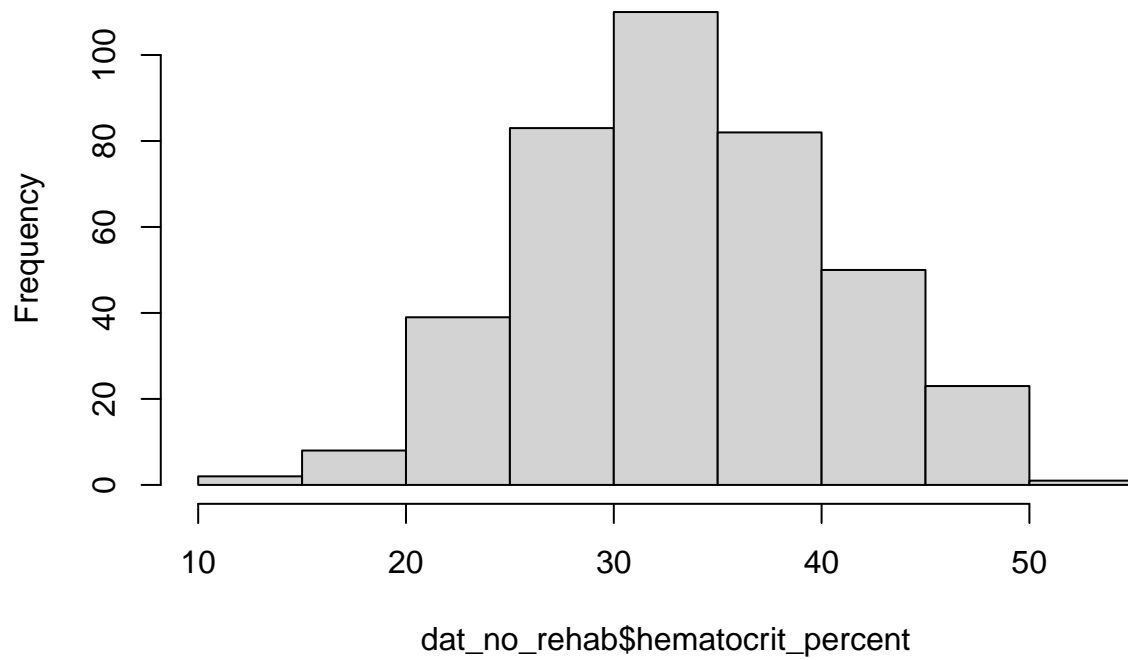
Check linear regression assumptions/conditions.

```

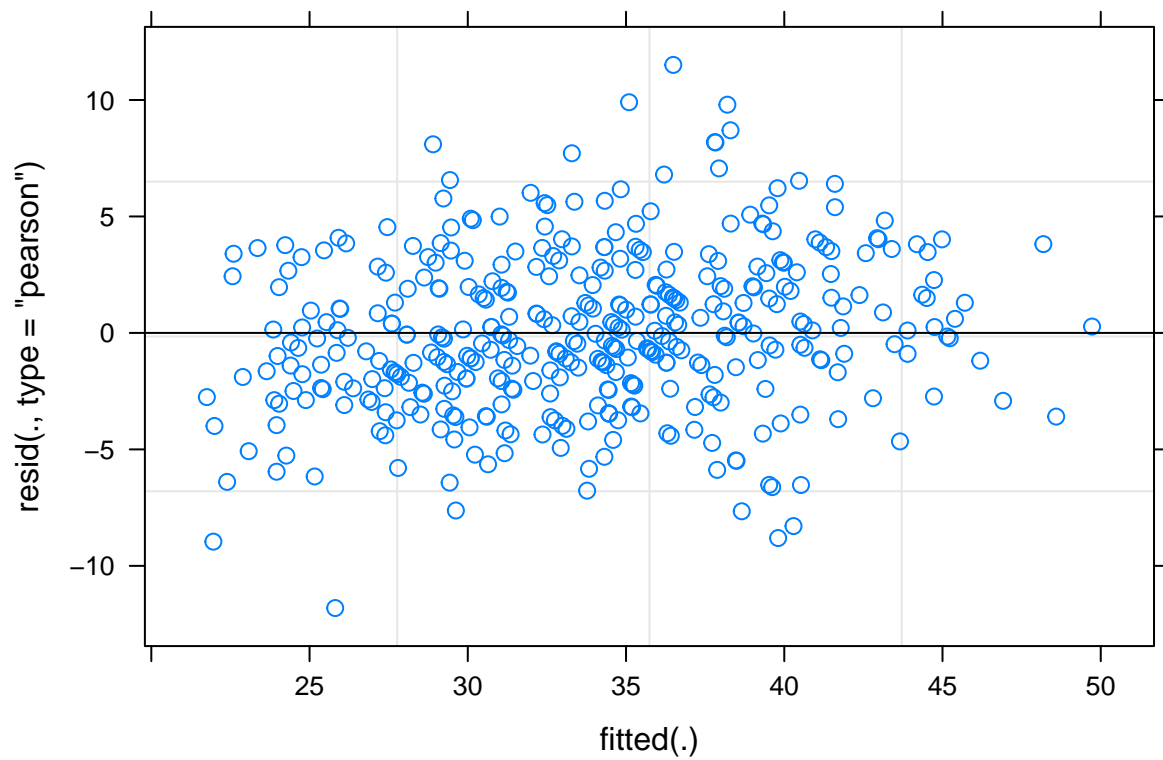
# distribution of
hist(dat_no_rehab$hematocrit_percent)

```

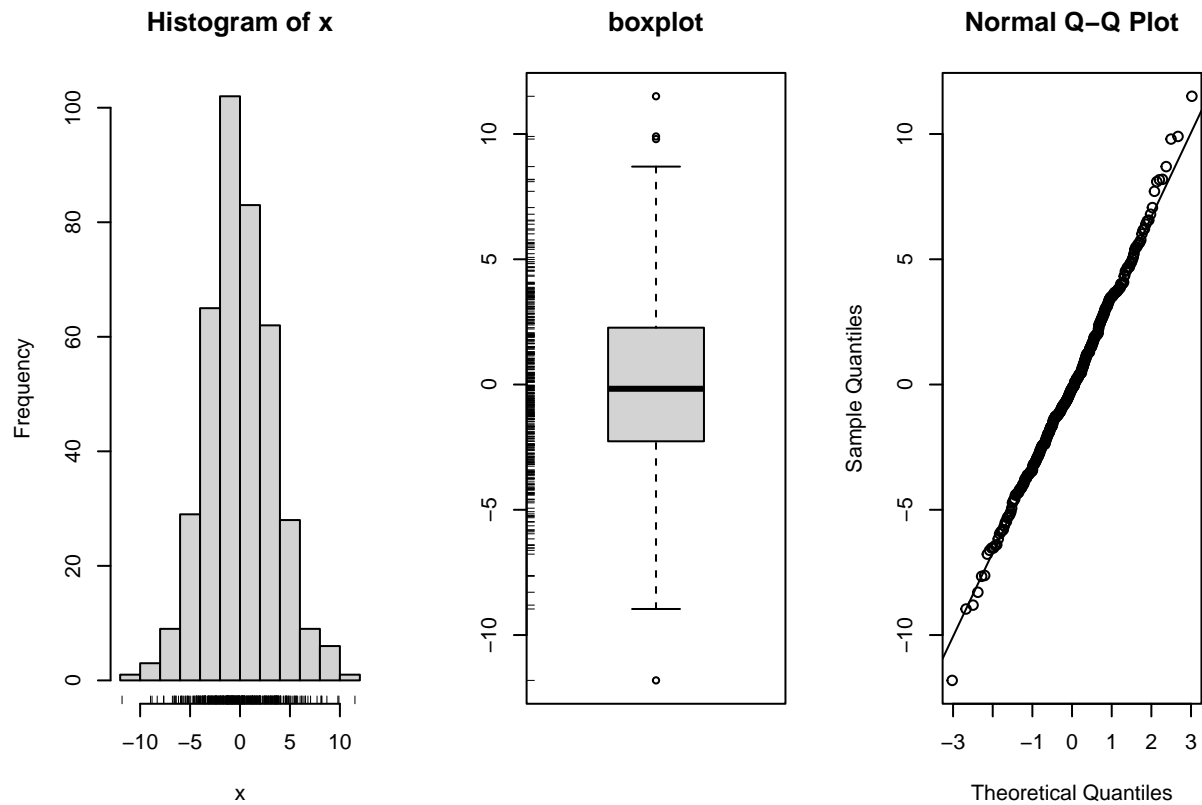
Histogram of dat_no_rehab\$hematocrit_percent



```
# VPD model  
plot(hct_mod_VPD)
```



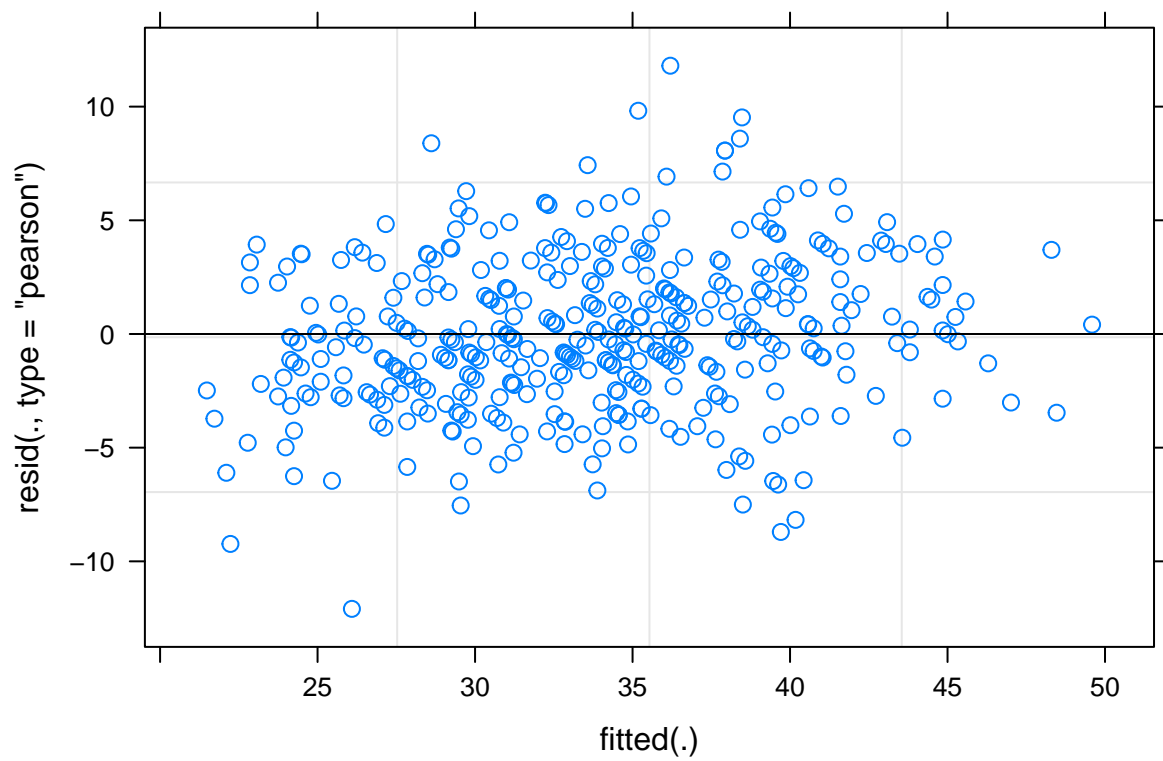
```
simple.eda(residuals(hct_mod_VPD))
```



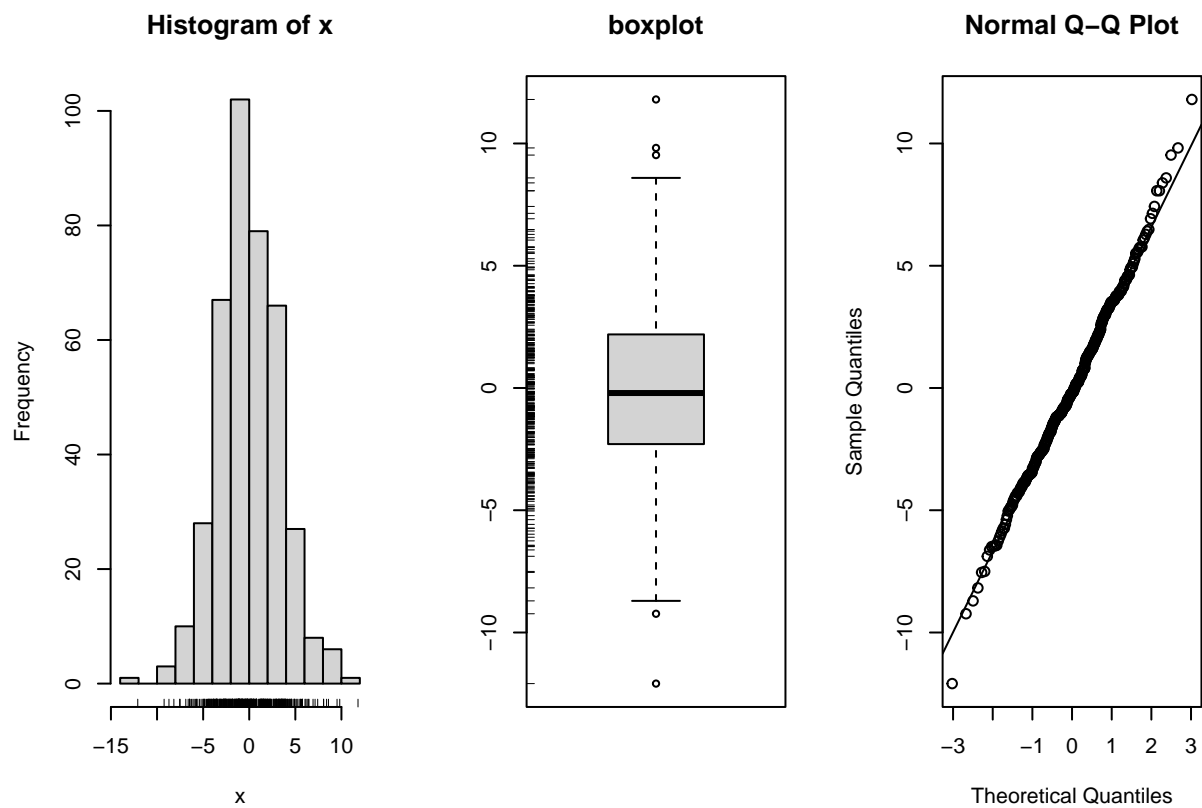
```
shapiro.test(residuals(hct_mod_VPD))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(hct_mod_VPD)
## W = 0.99651, p-value = 0.542
```

```
# humidity model
plot(hct_mod_hum)
```



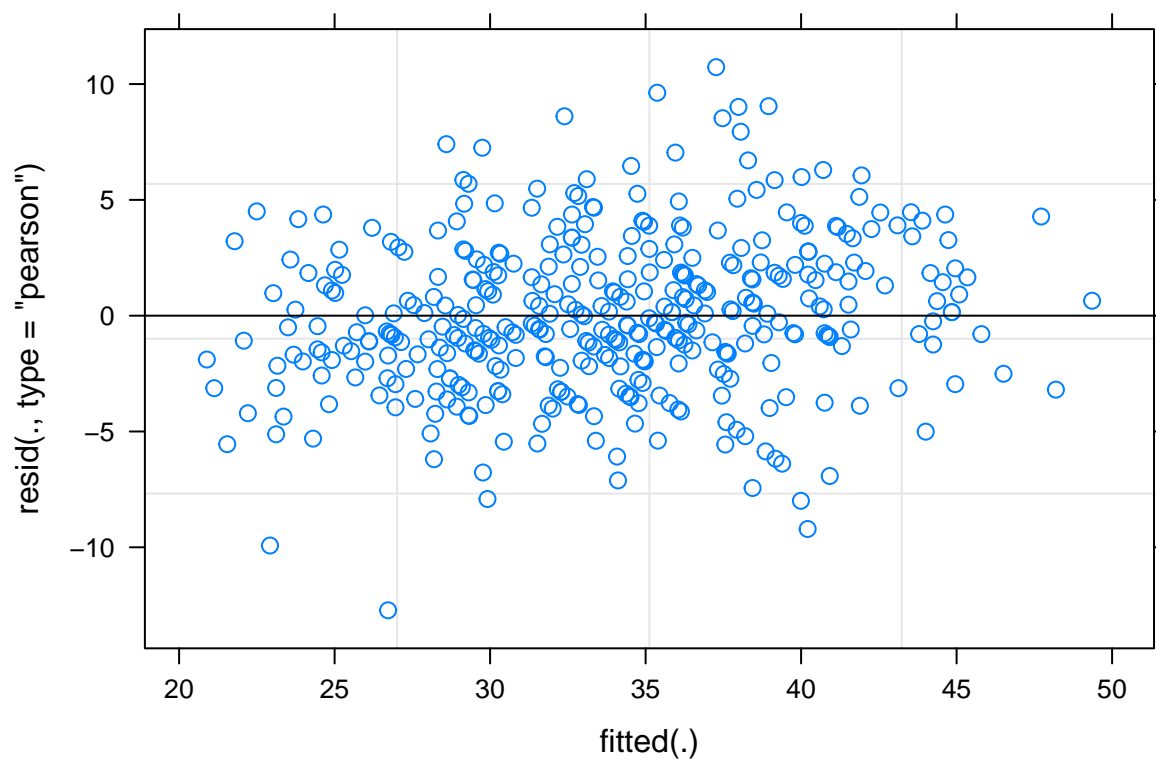
```
simple.eda(residuals(hct_mod_hum))
```



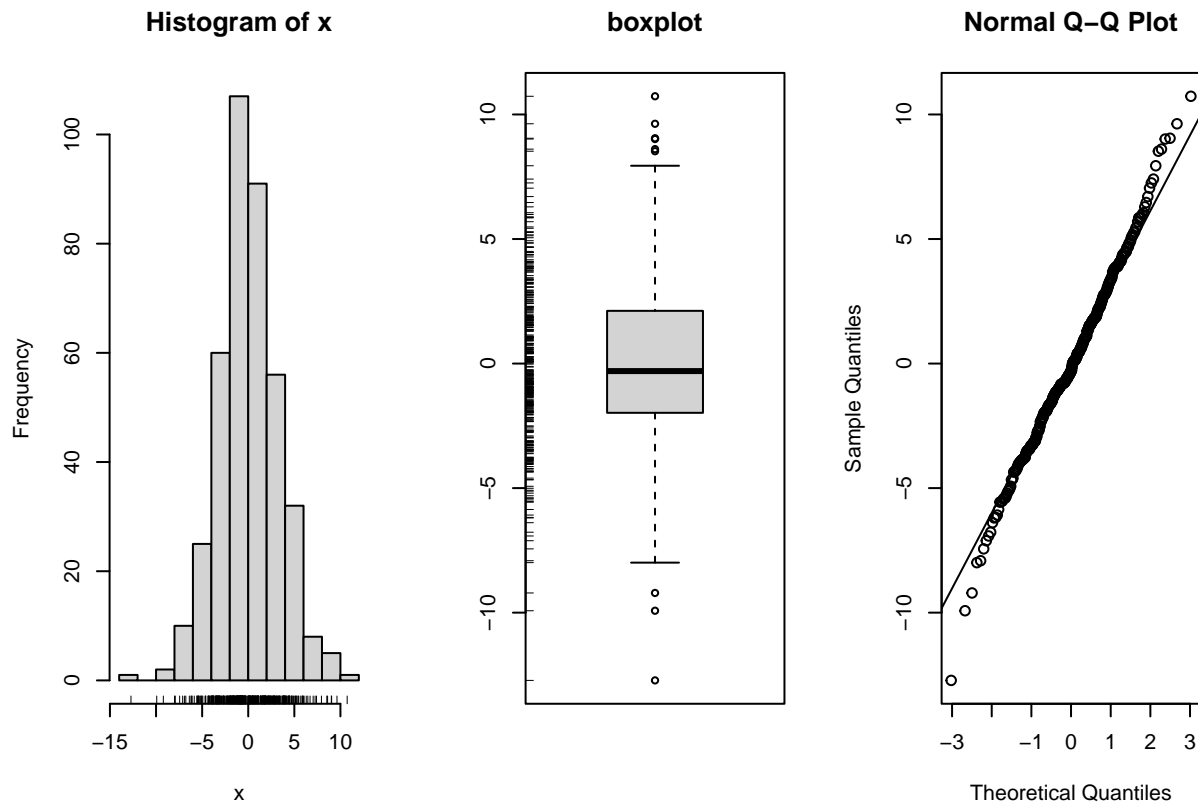
```
shapiro.test(residuals(hct_mod_hum))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(hct_mod_hum)  
## W = 0.99619, p-value = 0.4584
```

```
# temperature model  
plot(hct_mod_temp)
```



```
simple.eda(residuals(hct_mod_temp))
```



```
shapiro.test(residuals(hct_mod_temp))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(hct_mod_temp)
## W = 0.99478, p-value = 0.1975
```

All assumptions, normality, linearity, equal error variance, and independence are all good.

Comparison

Now, compare the AIC, RMSE, and R^2 values across models, and the F and p values of the variables for each model.

We... calculate RMSE manually, use the `r.squaredGLMM` function in the `MuMIn` package to get the marginal R^2 , which is how much of the total variance is explained by fixed effects, use the `aictab` function in the `AICmodavg` package to get AIC and Δ AIC values, and get the sum of squares, F, and p-values for each variable from the anova table for each model.

```
# calculate RMSE & R^2
hct_RMSE_Rsq <- data.frame(model =
  c('Day * VPD',
    'Day * Humidity',
    'Day * Temp'
  ),
  RMSE = c(sqrt(mean((residuals(hct_mod_VPD))^2)),
    sqrt(mean((residuals(hct_mod_hum))^2)),
    sqrt(mean((residuals(hct_mod_temp))^2))),
  # marginal Rsq for the amount of variance
```

```

# explained by fixed effects only
Rsqr = c(MuMIn::r.squaredGLMM(hct_mod_VPD)[, "R2m"],
         MuMIn::r.squaredGLMM(hct_mod_hum)[, "R2m"],
         MuMIn::r.squaredGLMM(hct_mod_temp)[, "R2m"])

# calculate AIC
hct_models <- list(hct_mod_VPD, hct_mod_hum, hct_mod_temp)
hct_AICc <- data.frame(aictab(cand.set = hct_models,
                             modnames = EXP_mod_names))

## Warning in aictab.AIClmerModLmerTest(cand.set = hct_models, modnames = EXP_mod_names):
## Model selection for fixed effects is only appropriate with ML estimation:
## REML (default) should only be used to select random effects for a constant set of fixed effects

# compare across models
hct_across <- hct_RMSE_Rsq %>%
  left_join(hct_AICc, by = 'model') %>%
  mutate(response = "Hematocrit (%)") %>%
  arrange(Delta_AICc)

# calculate F & p-values
hct_VPD_p <- data.frame(anova(hct_mod_VPD,
                             type = "1",
                             ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * VPD',
         predictor = rownames(.))
hct_hum_p <- data.frame(anova(hct_mod_hum,
                             type = "1",
                             ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Humidity',
         predictor = rownames(.))
hct_temp_p <- data.frame(anova(hct_mod_temp,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Temp',
         predictor = rownames(.))

# save within model values
hct_within <- hct_VPD_p %>%
  rbind(hct_hum_p) %>%
  rbind(hct_temp_p) %>%
  mutate(df = paste((NumDF), round(DenDF, 0), sep = ", "),
         response = "Hematocrit (%)")

```

Osmolality

Building

Build each treatment effect model.

```

osml_mod_VPD <- lmerTest::lmer(data = dat_no_rehab,
                             osmolality_mmol_kg_mean ~ day_n * VPD_kPa +
                             (1|trial_number/individual_ID))
osml_mod_hum <- lmerTest::lmer(data = dat_no_rehab,
                             osmolality_mmol_kg_mean ~ day_n * humidity_tmt +

```



```

                                (1|trial_number/individual_ID))
osml_mod_temp <- lmerTest::lmer(data = dat_no_rehab,
                                osmolality_mmol_kg_mean ~ day_n * temp_tmt +
                                (1|trial_number/individual_ID))

```

Assumptions

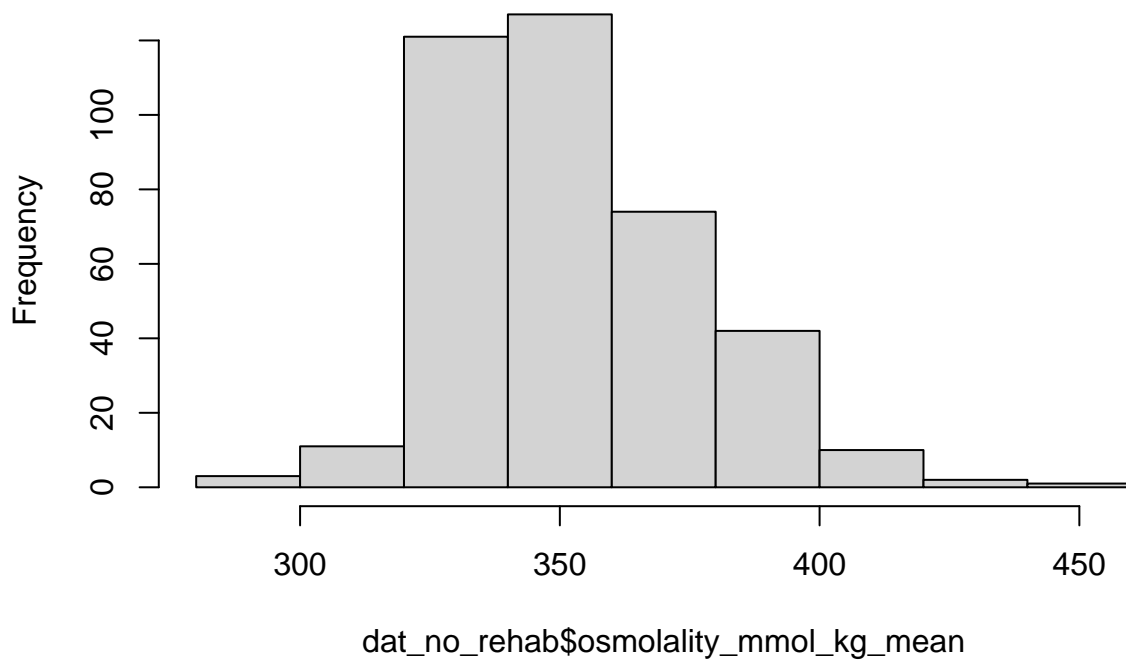
Check linear regression assumptions/conditions.

```

# distribution of
hist(dat_no_rehab$osmolality_mmol_kg_mean)

```

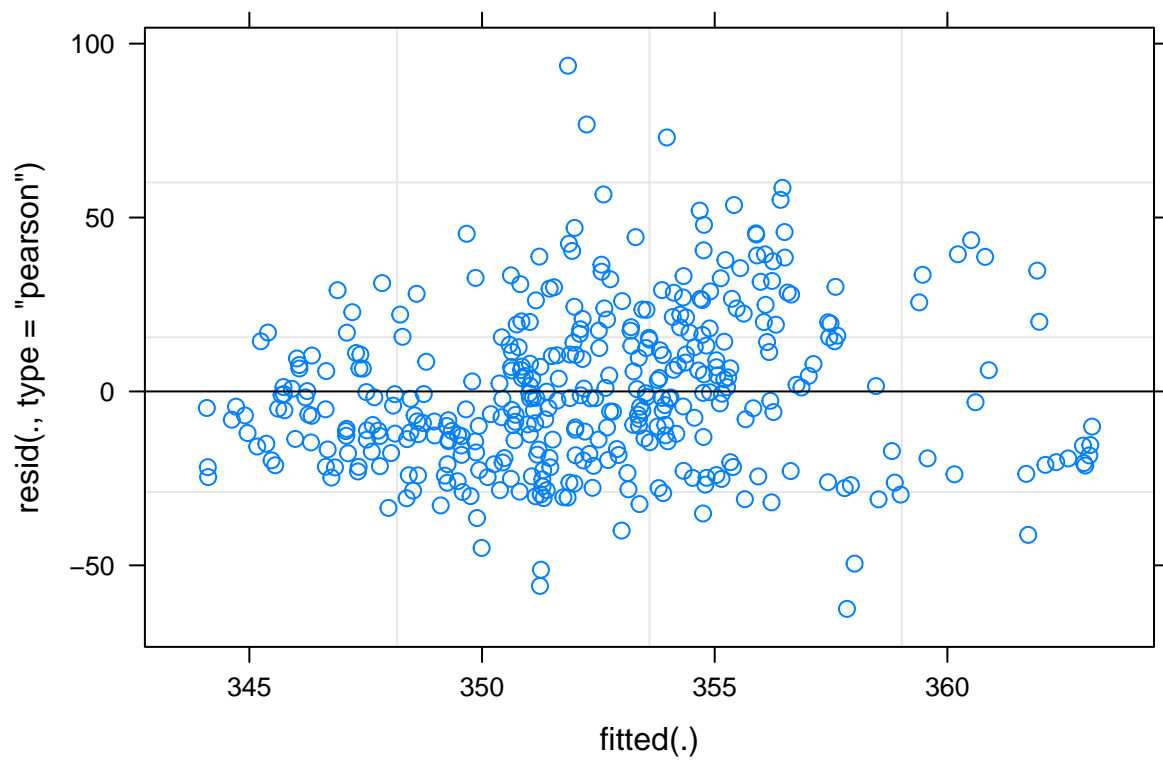
Histogram of dat_no_rehab\$osmolality_mmol_kg_mean



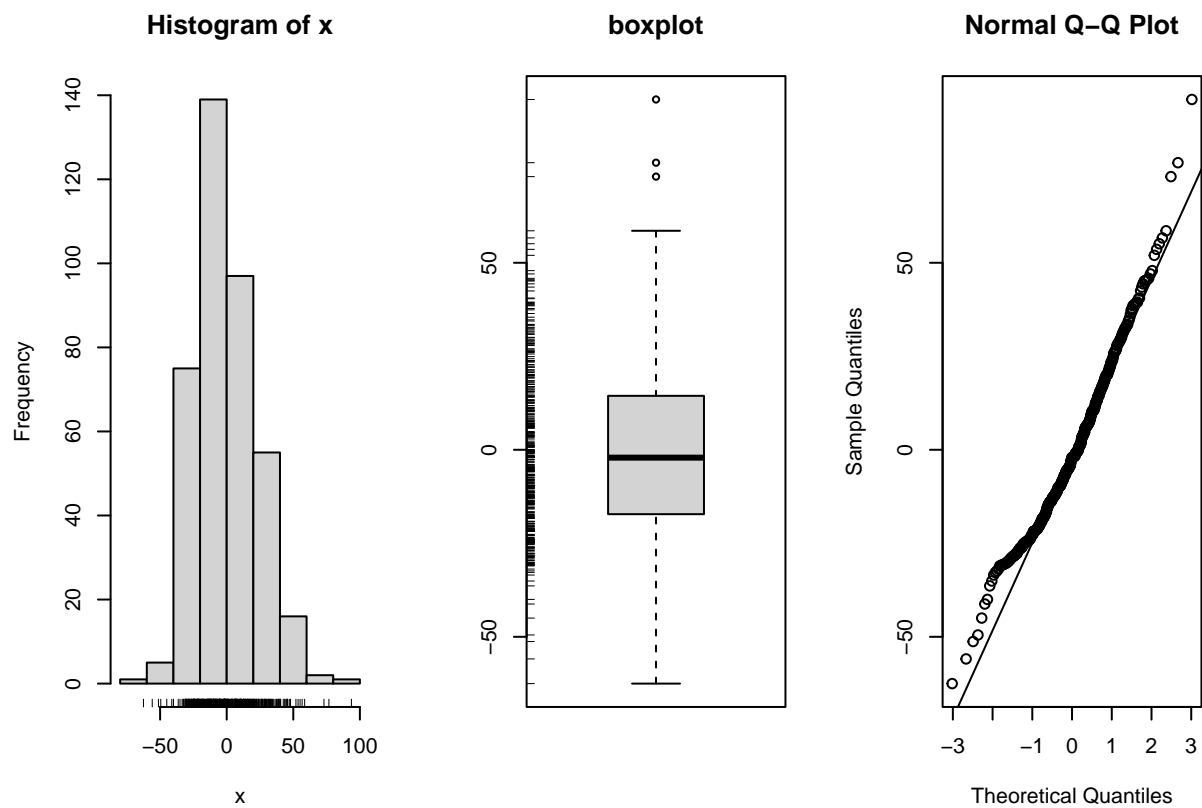
```

# VPD model
plot(osml_mod_VPD)

```



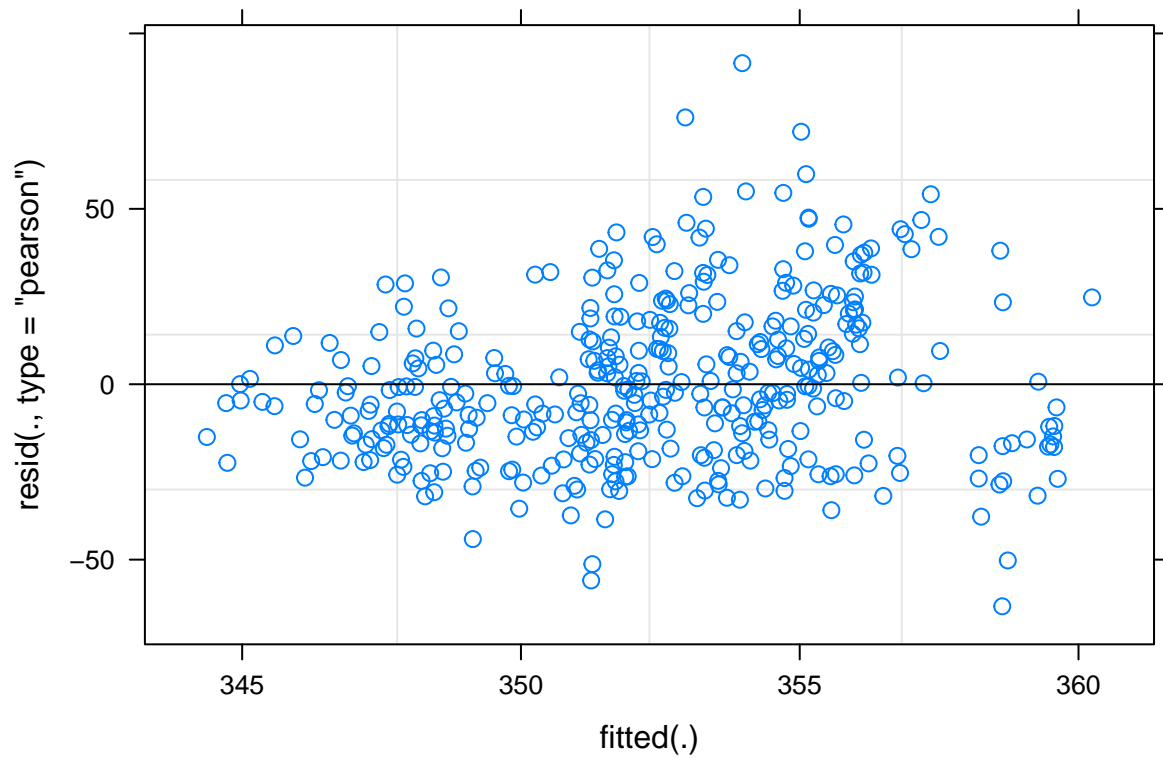
```
simple.eda(residuals(osml_mod_VPD))
```



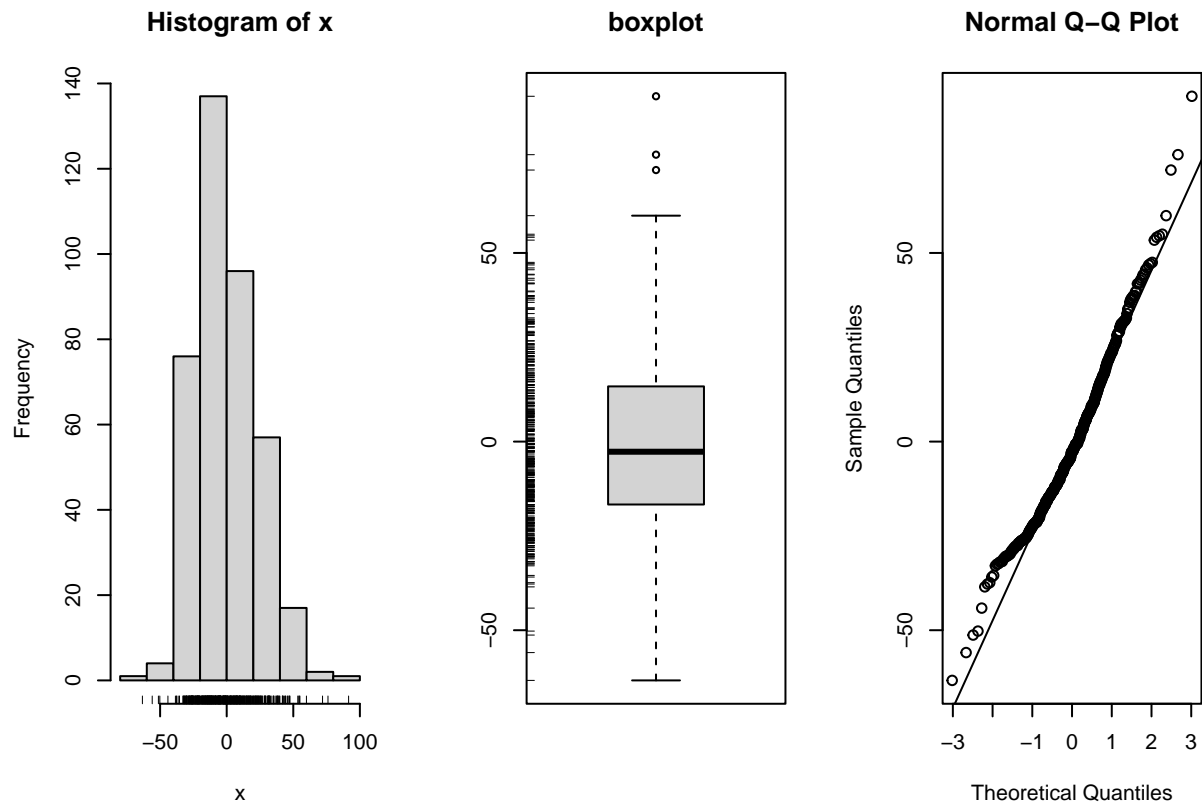
```
shapiro.test(residuals(osml_mod_VPD))
```

```
##  
##  Shapiro-Wilk normality test  
##  
## data:  residuals(osml_mod_VPD)  
## W = 0.9769, p-value = 6.836e-06
```

```
# humidity model  
plot(osml_mod_hum)
```



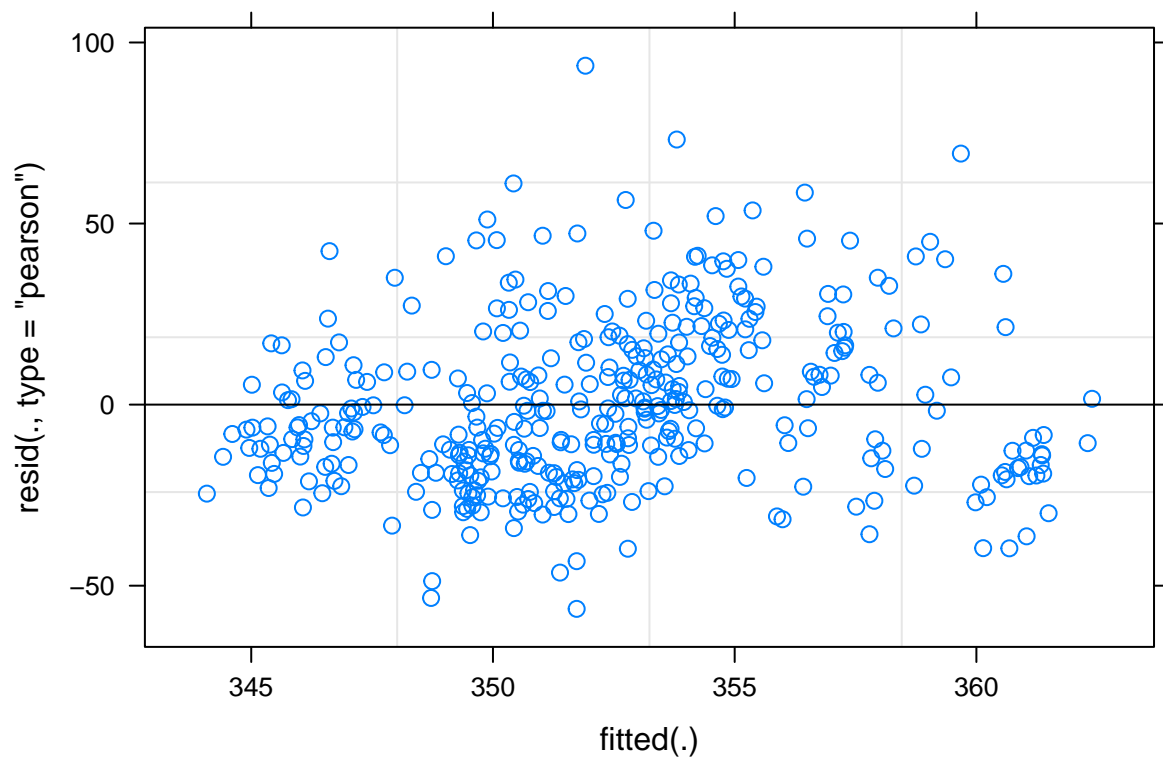
```
simple.eda(residuals(osml_mod_hum))
```



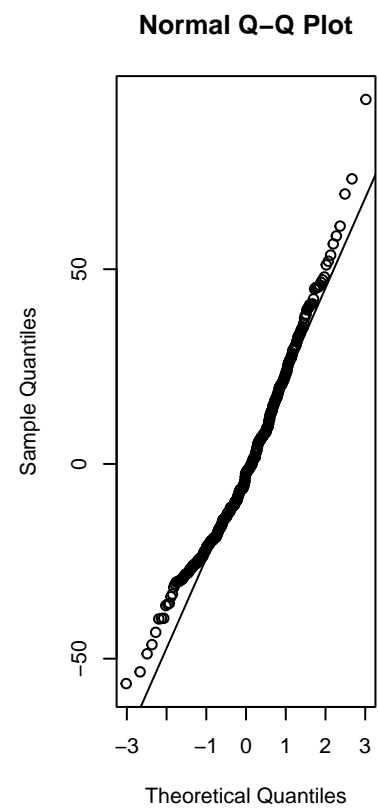
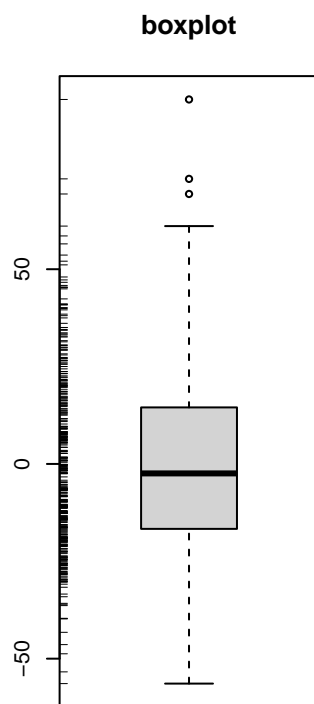
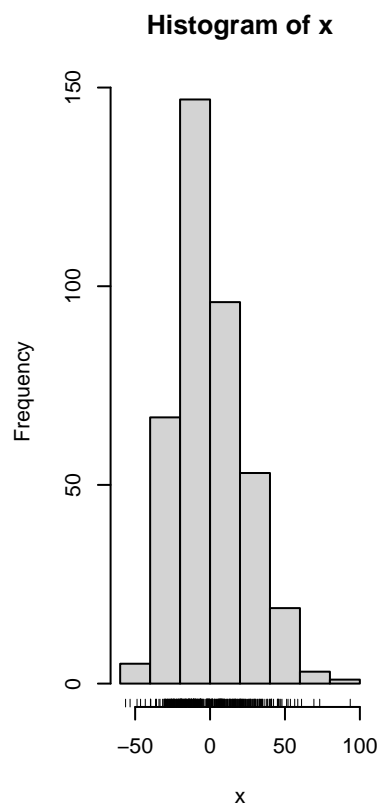
```
shapiro.test(residuals(osml_mod_hum))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(osml_mod_hum)
## W = 0.97746, p-value = 8.914e-06
```

```
# temperature model
plot(osml_mod_temp)
```



```
simple.eda(residuals(osml_mod_temp))
```



```
shapiro.test(residuals(osml_mod_temp))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(osml_mod_temp)  
## W = 0.97346, p-value = 1.44e-06
```

Normality is violated, but linearity, equal error variance, and independence are all okay.

Comparison

Now, compare the AIC, RMSE, and R^2 values across models, and the F and p values of the variables for each model.

We... calculate RMSE manually, use the `r.squaredGLMM` function in the `MuMIn` package to get the marginal R^2 , which is how much of the total variance is explained by fixed effects, use the `aictab` function in the `AICmodavg` package to get AIC and deltaAIC values, and get the sum of squares, F, and p-values for each variable from the anova table for each model.

```
# calculate RMSE & R^2  
osml_RMSE_Rsq <- data.frame(model =  
  c('Day * VPD',  
    'Day * Humidity',  
    'Day * Temp'  
  ),  
  RMSE = c(sqrt(mean((residuals(osml_mod_VPD))^2)),  
            sqrt(mean((residuals(osml_mod_hum))^2)),  
            sqrt(mean((residuals(osml_mod_temp))^2))),  
  # marginal Rsq for the amount of variance  
  # explained by fixed effects only  
  Rsq = c(MuMIn::r.squaredGLMM(osml_mod_VPD)[,"R2m"],  
           MuMIn::r.squaredGLMM(osml_mod_hum)[,"R2m"],  
           MuMIn::r.squaredGLMM(osml_mod_temp)[,"R2m"]))
```

```
# calculate AIC  
osml_models <- list(osml_mod_VPD, osml_mod_hum, osml_mod_temp)  
osml_AICc <- data.frame(aictab(cand.set = osml_models,  
                             modnames = EXP_mod_names))
```

```
## Warning in aictab.AIClmerModLmerTest(cand.set = osml_models, modnames = EXP_mod_names):  
## Model selection for fixed effects is only appropriate with ML estimation:  
## REML (default) should only be used to select random effects for a constant set of fixed effects
```

```
# compare across models  
osml_across <- osml_RMSE_Rsq %>%  
  left_join(osml_AICc, by = 'model') %>%  
  mutate(response = "Plasma Osmolality (mmol/kg)") %>%  
  arrange(Delta_AICc)  
  
# calculate F & p-values  
osml_VPD_p <- data.frame(anova(osml_mod_VPD,  
                               type = "1",  
                               ddf = "Kenward-Roger")) %>%  
  mutate(model = 'Day * VPD',  
         predictor = rownames(.))
```

```

osml_hum_p <- data.frame(anova(osml_mod_hum,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Humidity',
         predictor = rownames(.))
osml_temp_p <- data.frame(anova(osml_mod_temp,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Day * Temp',
         predictor = rownames(.))

# save within model values
osml_within <- osml_VPD_p %>%
  rbind(osml_hum_p) %>%
  rbind(osml_temp_p) %>%
  mutate(df = paste((NumDF), round(DenDF, 0), sep = ", "),
         response = "Plasma Osmolality (mmol/kg)")

```

Body Temp

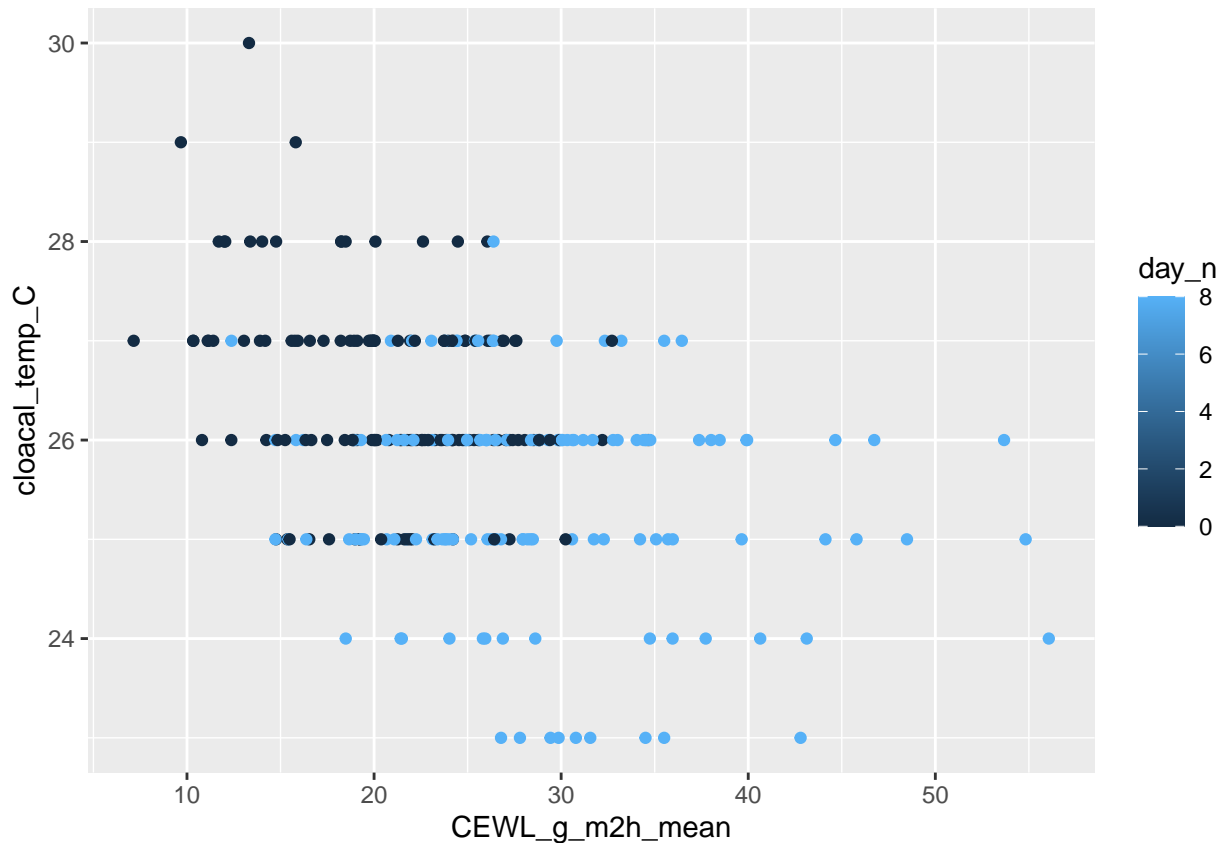
I need to double check whether CEWL has a relationship with body temperature at the point of measurement, to make sure I do not need to account for it.

```

ggplot(dat_no_rehab) +
  geom_point(aes(x = CEWL_g_m2h_mean,
                y = cloacal_temp_C,
                color = day_n))

```

```
## Warning: Removed 537 rows containing missing values (`geom_point()`).
```



Test an lm of raw CEWL ~ body temp for day 8 measurements only.

```
body_temp_test_dat <- dat_no_rehab_deltaCEWL %>%
  dplyr::filter(complete.cases(CEWL_g_m2h_mean)) %>%
  dplyr::filter(day_n == 8)

CEWL_body_temp_mod <- lm(data = body_temp_test_dat,
  CEWL_g_m2h_mean ~ cloacal_temp_C)
summary(CEWL_body_temp_mod)
```

```
##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ cloacal_temp_C, data = body_temp_test_dat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.721  -5.210  -1.534   4.619  25.793
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    55.6681    16.7785   3.318  0.00117 **
## cloacal_temp_C  -1.0581     0.6598  -1.604  0.11120
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.137 on 131 degrees of freedom
## Multiple R-squared:  0.01925,    Adjusted R-squared:  0.01177
## F-statistic: 2.572 on 1 and 131 DF,  p-value: 0.1112
```

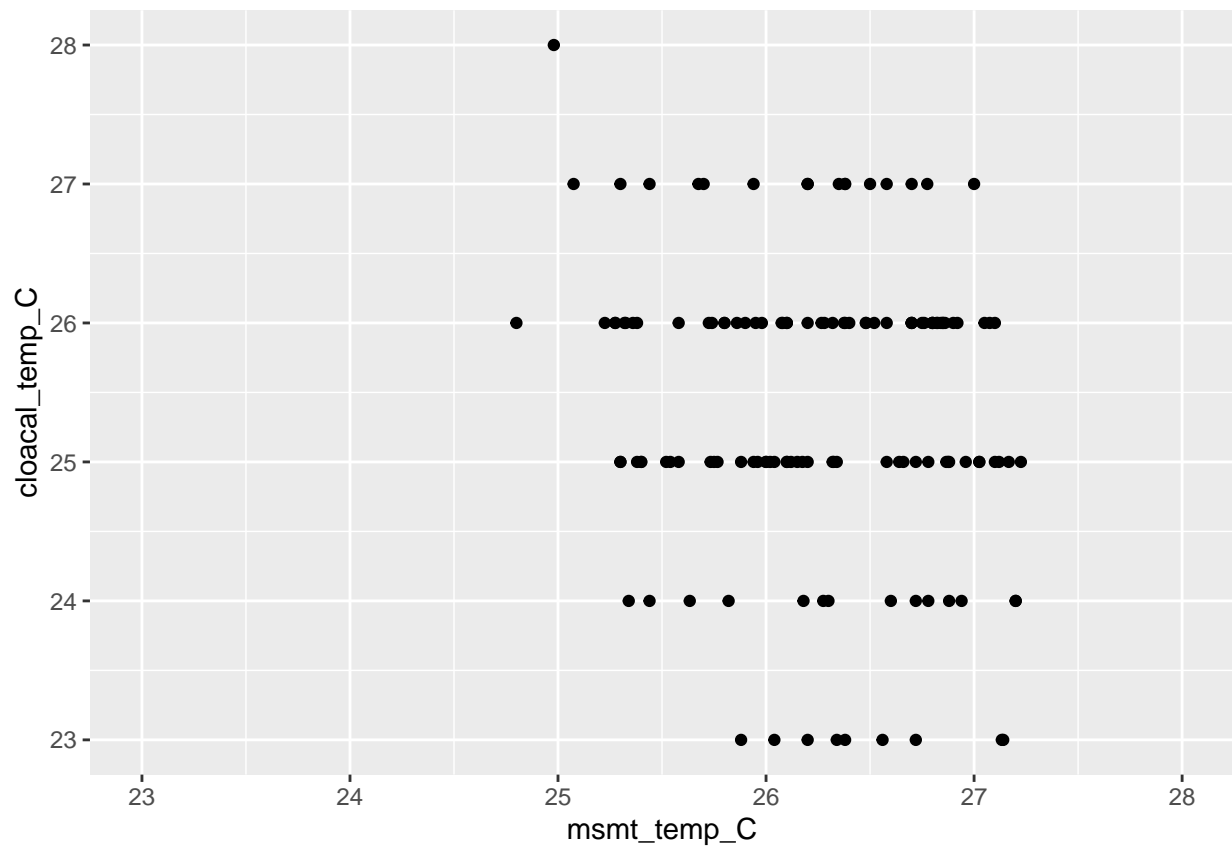

No relationship! So, continue on without.

Also check body temp ~ measurement temp... and CEWL ~ msmt temp

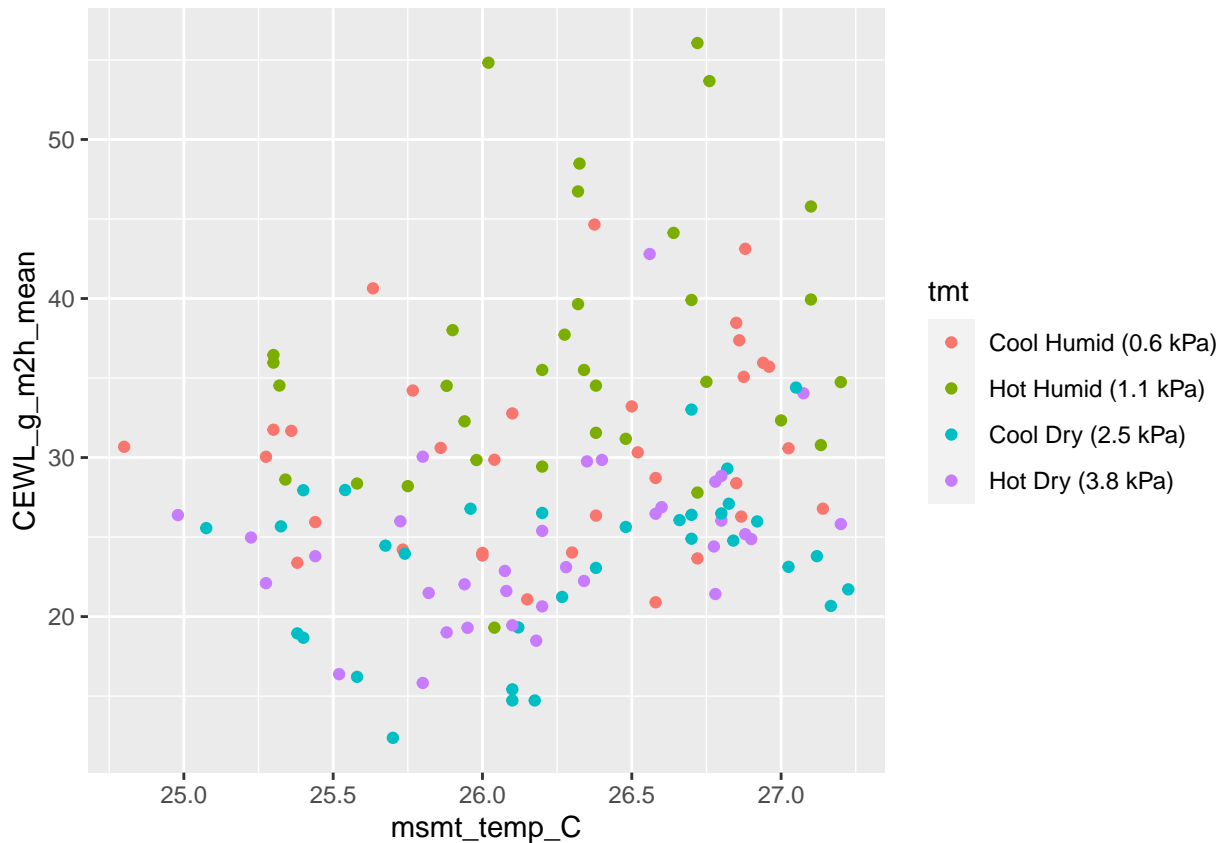
```
temp_body_temp_mod <- lm(data = body_temp_test_dat,
                          cloacal_temp_C ~ msmt_temp_C)
summary(temp_body_temp_mod)

##
## Call:
## lm(formula = cloacal_temp_C ~ msmt_temp_C, data = body_temp_test_dat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5168 -0.5504  0.2894  0.7257  2.2170
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  33.1715     4.1350   8.022 5.13e-13 ***
## msmt_temp_C  -0.2958     0.1575  -1.878  0.0625 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.063 on 131 degrees of freedom
## Multiple R-squared:  0.02623,    Adjusted R-squared:  0.0188
## F-statistic: 3.529 on 1 and 131 DF,  p-value: 0.06254

ggplot(body_temp_test_dat) +
  geom_point(aes(x = msmt_temp_C,
                 y = cloacal_temp_C)) +
  xlim(23, 28) + ylim(23,28)
```



```
ggplot(body_temp_test_dat) +
  geom_point(aes(x = msmt_temp_C,
                 y = CEWL_g_m2h_mean,
                 color = tmt))
```



```
CEWL_msmt_temp_mod <- lm(data = body_temp_test_dat,
                          CEWL_g_m2h_mean ~ msmt_temp_C)
summary(CEWL_msmt_temp_mod)
```

```
##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ msmt_temp_C, data = body_temp_test_dat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.729  -5.780  -1.187   4.445  26.753
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -50.704     31.191  -1.626   0.106
## msmt_temp_C    3.028      1.188   2.549   0.012 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.02 on 131 degrees of freedom
## Multiple R-squared:  0.04726,    Adjusted R-squared:  0.03999
## F-statistic: 6.498 on 1 and 131 DF,  p-value: 0.01195
```

```
HH_temp <- body_temp_test_dat %>%
  dplyr::filter(substr(tmt, 1, 6) == "Hot Hu")
CEWL_msmt_temp_HH <- lm(data = HH_temp,
                        CEWL_g_m2h_mean ~ msmt_temp_C)
summary(CEWL_msmt_temp_HH) # p = 0.12
```

```
##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ msmt_temp_C, data = HH_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.3918  -5.6056  -0.8461   3.3056  19.2123
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -71.115     68.824  -1.033   0.309
## msmt_temp_C    4.102      2.618   1.567   0.127
##
## Residual standard error: 8.193 on 31 degrees of freedom
## Multiple R-squared:  0.07339,    Adjusted R-squared:  0.04349
## F-statistic: 2.455 on 1 and 31 DF,  p-value: 0.1273

HD_temp <- body_temp_test_dat %>%
  dplyr::filter(substr(tmt, 1, 6) == "Hot Dr")
CEWL_msmt_temp_HD <- lm(data = HD_temp,
                        CEWL_g_m2h_mean ~ msmt_temp_C)
summary(CEWL_msmt_temp_HD) # p = 0.02***

##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ msmt_temp_C, data = HD_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
##  -7.300  -2.719  -1.407   2.129  16.991
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -67.804     40.034  -1.694   0.1000
## msmt_temp_C    3.525      1.527   2.308   0.0276 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.889 on 32 degrees of freedom
## Multiple R-squared:  0.1427, Adjusted R-squared:  0.116
## F-statistic: 5.328 on 1 and 32 DF,  p-value: 0.02759

CH_temp <- body_temp_test_dat %>%
  dplyr::filter(substr(tmt, 1, 6) == "Cool H")
CEWL_msmt_temp_CH <- lm(data = CH_temp,
                        CEWL_g_m2h_mean ~ msmt_temp_C)
summary(CEWL_msmt_temp_CH) # p = 0.32

##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ msmt_temp_C, data = CH_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -10.111 -5.218 -0.221 3.556 13.995
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -14.641     45.313  -0.323   0.749
## msmt_temp_C    1.718      1.726   0.995   0.327
##
## Residual standard error: 6.136 on 31 degrees of freedom
## Multiple R-squared: 0.03095, Adjusted R-squared: -0.0003098
## F-statistic: 0.9901 on 1 and 31 DF, p-value: 0.3274

CD_temp <- body_temp_test_dat %>%
  dplyr::filter(substr(tmt, 1, 6) == "Cool D")
CEWL_msmt_temp_CD <- lm(data = CD_temp,
                        CEWL_g_m2h_mean ~ msmt_temp_C)
summary(CEWL_msmt_temp_CD) # p = 0.09

##
## Call:
## lm(formula = CEWL_g_m2h_mean ~ msmt_temp_C, data = CD_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.7712 -2.7535  0.8885  3.1595  8.9908
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -39.970     36.453  -1.096   0.2813
## msmt_temp_C    2.417      1.387   1.743   0.0913 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.962 on 31 degrees of freedom
## Multiple R-squared: 0.08924, Adjusted R-squared: 0.05986
## F-statistic: 3.037 on 1 and 31 DF, p-value: 0.09128
```

CEWL

t-tests

```
CEWL_ttest_HH <- t.test(HH_temp$delta_CEWL,
                        mu = 0, alternative = "two.sided")
CEWL_ttest_HH

##
## One Sample t-test
##
## data: HH_temp$delta_CEWL
## t = 8.7341, df = 32, p-value = 5.573e-10
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##  11.76821 18.92672
## sample estimates:
## mean of x
## 15.34746
```

```
CEWL_ttest_HD <- t.test(HD_temp$delta_CEWL,
                        mu = 0, alternative = "two.sided")
CEWL_ttest_HD
```

```
##
## One Sample t-test
##
## data: HD_temp$delta_CEWL
## t = 2.513, df = 33, p-value = 0.01703
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 0.7042497 6.6929758
## sample estimates:
## mean of x
## 3.698613
```

```
CEWL_ttest_CH <- t.test(CH_temp$delta_CEWL,
                        mu = 0, alternative = "two.sided")
CEWL_ttest_CH
```

```
##
## One Sample t-test
##
## data: CH_temp$delta_CEWL
## t = 8.7488, df = 32, p-value = 5.363e-10
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 7.296566 11.725292
## sample estimates:
## mean of x
## 9.510929
```

```
CEWL_ttest_CD <- t.test(CD_temp$delta_CEWL,
                        mu = 0, alternative = "two.sided")
CEWL_ttest_CD
```

```
##
## One Sample t-test
##
## data: CD_temp$delta_CEWL
## t = 2.75, df = 32, p-value = 0.009721
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 0.9302207 6.2446177
## sample estimates:
## mean of x
## 3.587419
```

Building

Build each treatment effect model.

```
CEWL_mod_VPD <- lmerTest::lmer(data = dat_no_rehab_deltaCEWL,
                              delta_CEWL ~ VPD_kPa +
                              (1|trial_number))
CEWL_mod_hum <- lmerTest::lmer(data = dat_no_rehab_deltaCEWL,
```

```

        delta_CEWL ~ humidity_tmt +
        (1|trial_number))
CEWL_mod_temp <- lmerTest::lmer(data = dat_no_rehab_deltaCEWL,
        delta_CEWL ~ temp_tmt +
        (1|trial_number))

```

Assumptions

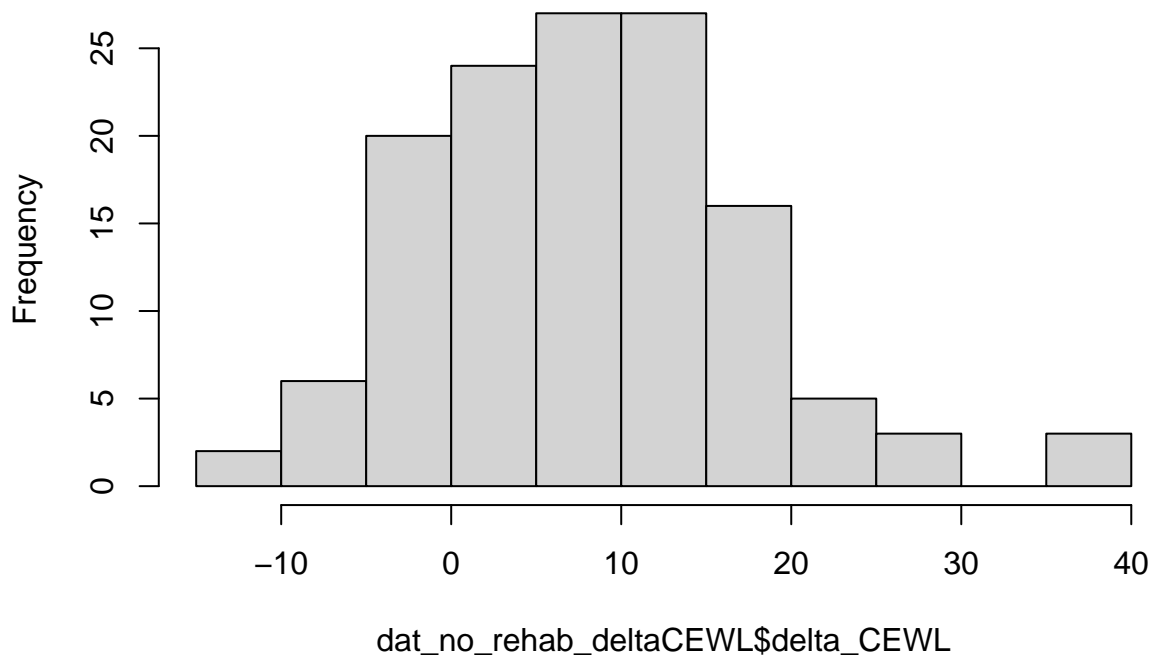
Check linear regression assumptions/conditions.

```

# distribution of
hist(dat_no_rehab_deltaCEWL$delta_CEWL)

```

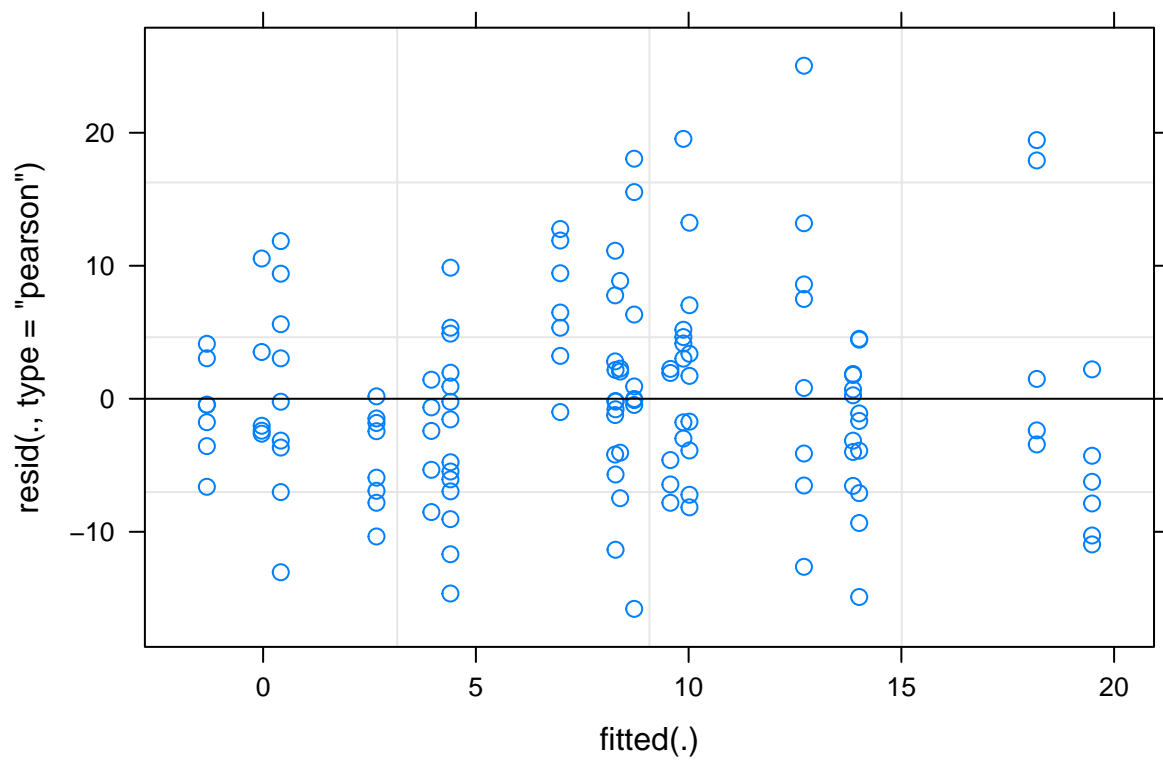
Histogram of dat_no_rehab_deltaCEWL\$delta_CEWL



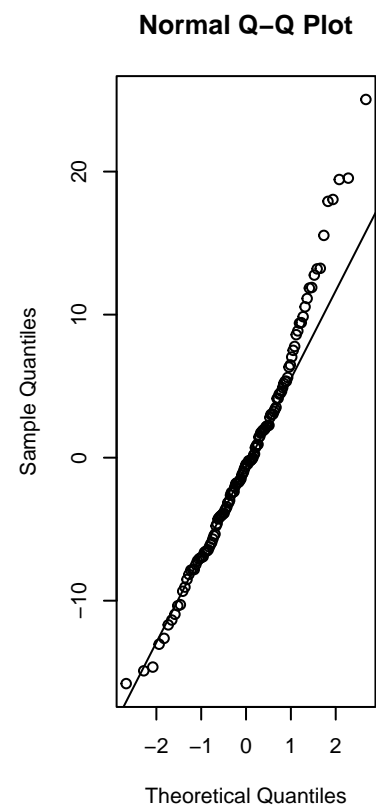
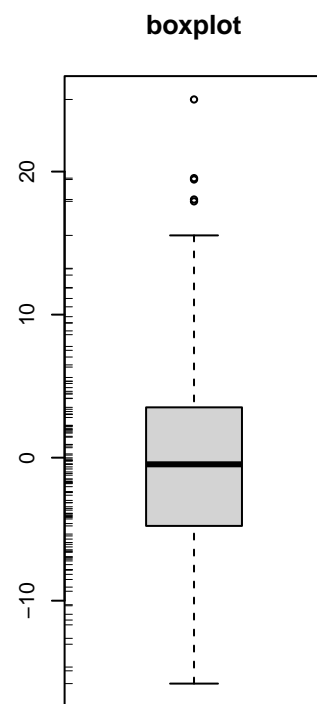
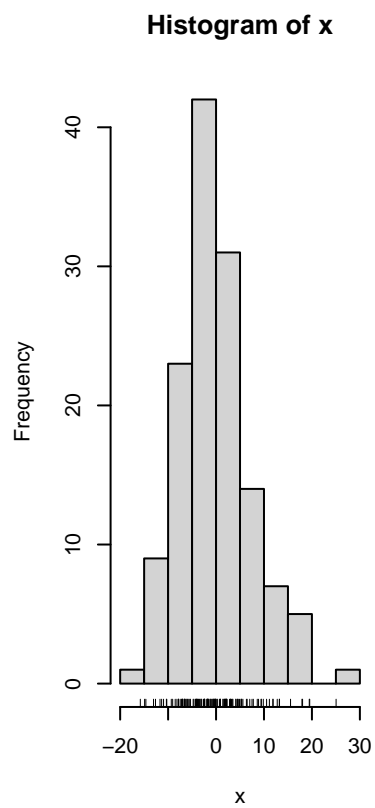
```

# VPD model
plot(CEWL_mod_VPD)

```



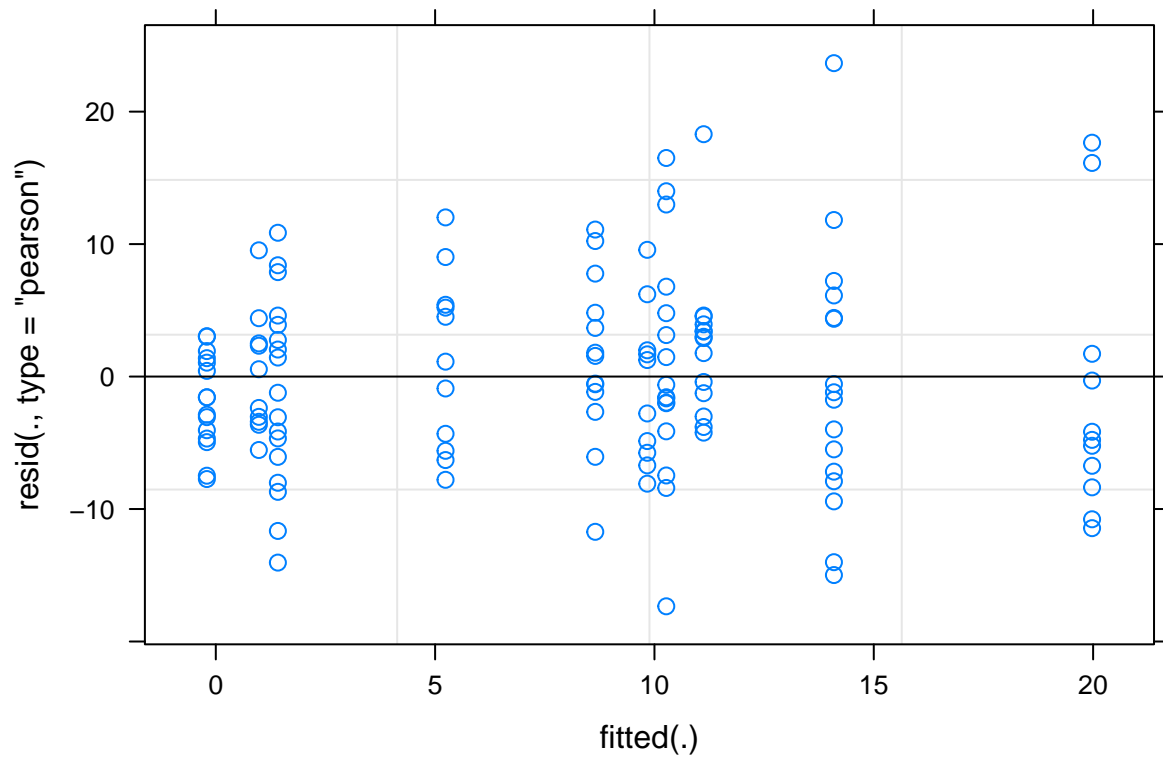
```
simple.eda(residuals(CEWL_mod_VPD))
```



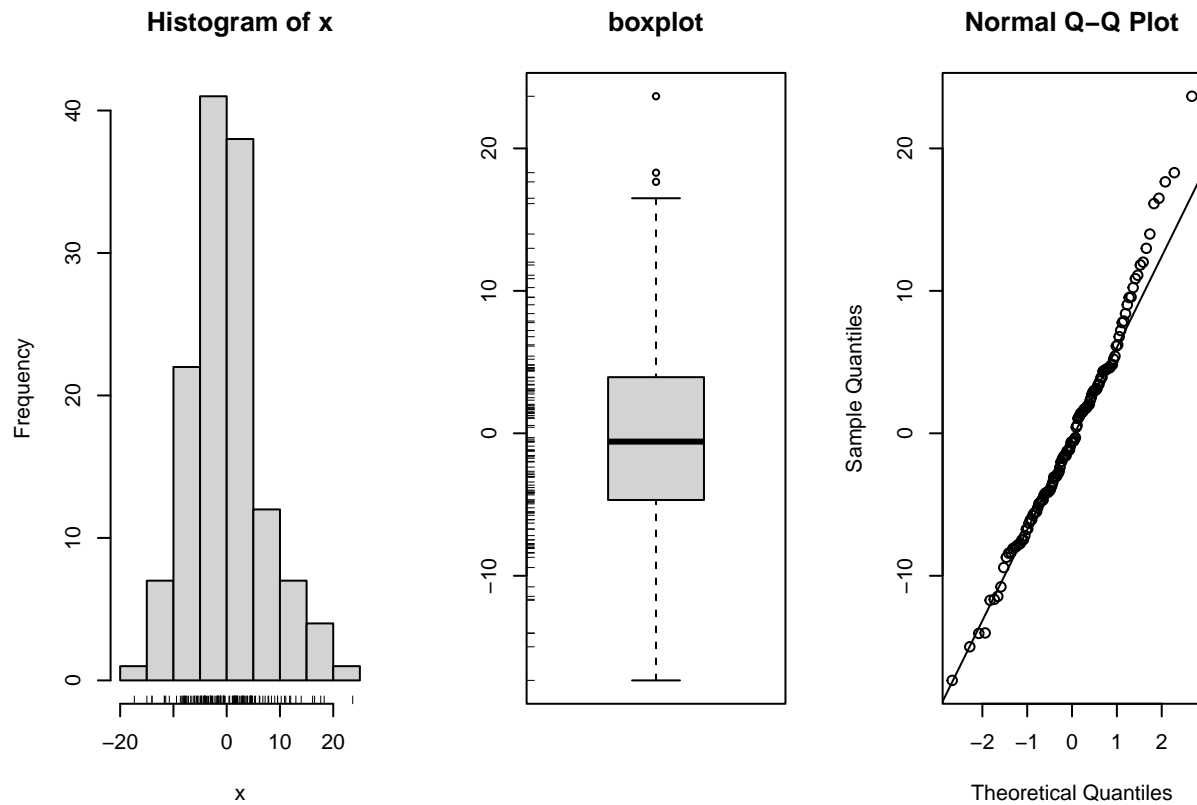

```
shapiro.test(residuals(CEWL_mod_VPD))
```

```
##  
##  Shapiro-Wilk normality test  
##  
## data:  residuals(CEWL_mod_VPD)  
## W = 0.97332, p-value = 0.01016
```

```
# humidity model  
plot(CEWL_mod_hum)
```



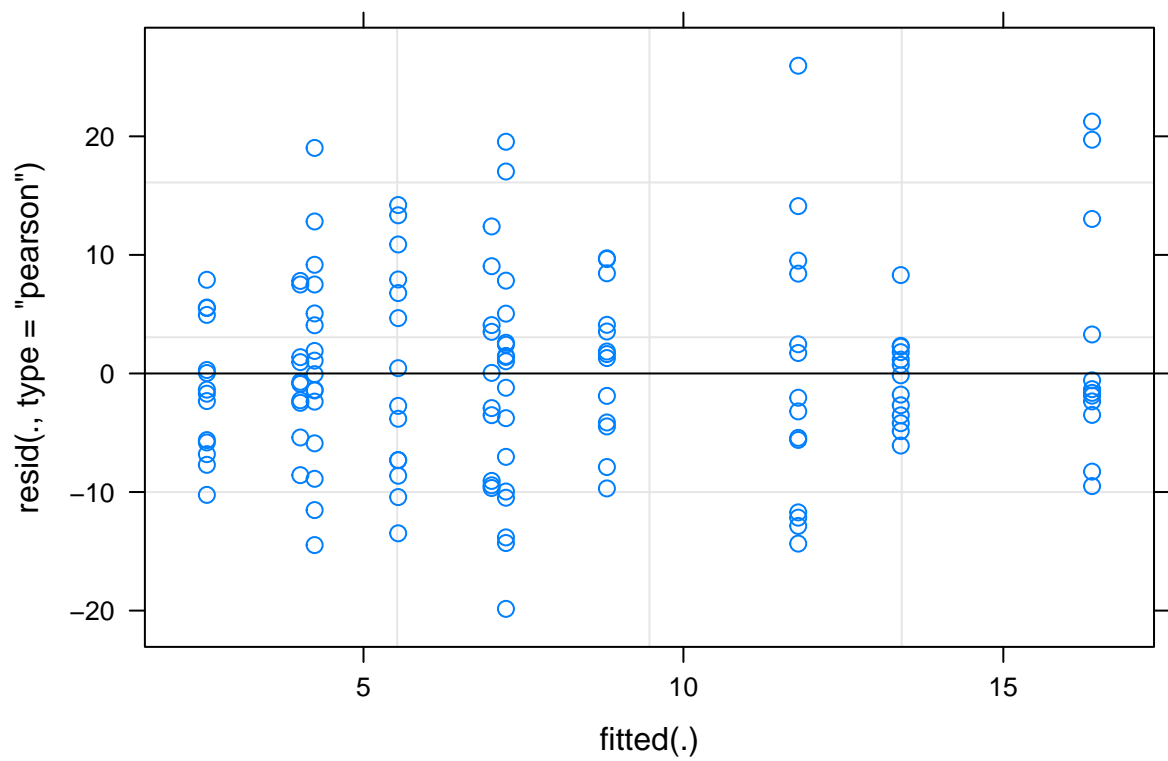
```
simple.eda(residuals(CEWL_mod_hum))
```



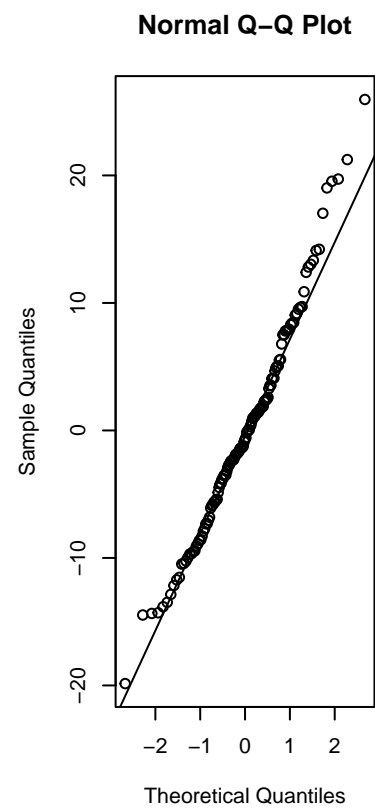
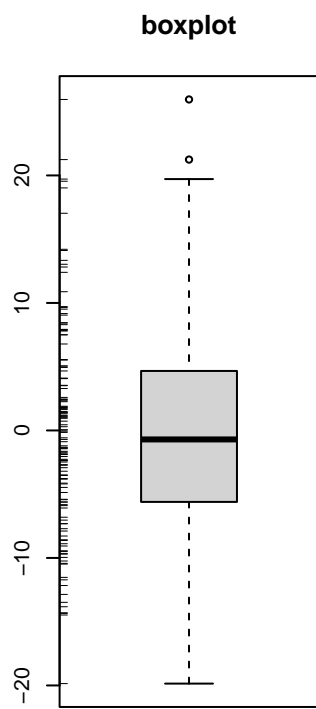
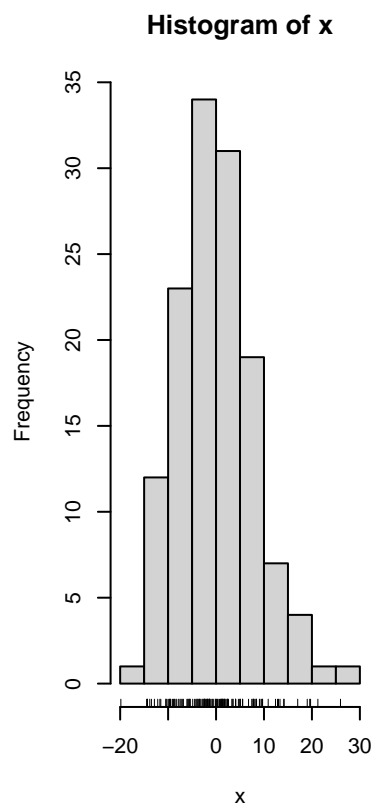
```
shapiro.test(residuals(CEWL_mod_hum))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(CEWL_mod_hum)
## W = 0.98319, p-value = 0.1002
```

```
# temperature model
plot(CEWL_mod_temp)
```



```
simple.eda(residuals(CEWL_mod_temp))
```



```
shapiro.test(residuals(CEWL_mod_temp))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(CEWL_mod_temp)  
## W = 0.98262, p-value = 0.08761
```

All assumptions are fine.

Comparison

Now, compare the AIC, RMSE, and R^2 values across models, and the F and p values of the variables for each model.

We... calculate RMSE manually, use the `r.squaredGLMM` function in the `MuMIn` package to get the marginal R^2 , which is how much of the total variance is explained by fixed effects, use the `aictab` function in the `AICmodavg` package to get AIC and deltaAIC values, and get the sum of squares, F, and p-values for each variable from the anova table for each model.

```
# calculate RMSE & R^2  
CEWL_RMSE_Rsq <- data.frame(model =  
  c('VPD',  
    'Humidity',  
    'Temp'  
  ),  
  RMSE = c(sqrt(mean((residuals(CEWL_mod_VPD))^2)),  
            sqrt(mean((residuals(CEWL_mod_hum))^2)),  
            sqrt(mean((residuals(CEWL_mod_temp))^2))),  
  # marginal Rsq for the amount of variance  
  # explained by fixed effects only  
  Rsq = c(MuMIn::r.squaredGLMM(CEWL_mod_VPD)[,"R2m"],  
           MuMIn::r.squaredGLMM(CEWL_mod_hum)[,"R2m"],  
           MuMIn::r.squaredGLMM(CEWL_mod_temp)[,"R2m"]))  
  
# calculate AIC  
CEWL_models <- list(CEWL_mod_VPD, CEWL_mod_hum, CEWL_mod_temp)  
CEWL_mod_names <- data.frame(model =  
  c('VPD',  
    'Humidity',  
    'Temp'  
  ))  
CEWL_AICc <- data.frame(aictab(cand.set = CEWL_models,  
  modnames = CEWL_mod_names))  
  
## Warning in aictab.AIClmerModLmerTest(cand.set = CEWL_models, modnames = CEWL_mod_names):  
## Model selection for fixed effects is only appropriate with ML estimation:  
## REML (default) should only be used to select random effects for a constant set of fixed effects  
  
# compare across models  
CEWL_across <- CEWL_RMSE_Rsq %>%  
  left_join(CEWL_AICc, by = 'model') %>%  
  mutate(response = "deltaCEWL") %>%  
  arrange(Delta_AICc)  
  
# calculate F & p-values
```

```

CEWL_VPD_p <- data.frame(anova(CEWL_mod_VPD,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'VPD',
         predictor = rownames(.))
CEWL_hum_p <- data.frame(anova(CEWL_mod_hum,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Humidity',
         predictor = rownames(.))
CEWL_temp_p <- data.frame(anova(CEWL_mod_temp,
                              type = "1",
                              ddf = "Kenward-Roger")) %>%
  mutate(model = 'Temp',
         predictor = rownames(.))

# save within model values
CEWL_within <- CEWL_VPD_p %>%
  rbind(CEWL_hum_p) %>%
  rbind(CEWL_temp_p) %>%
  mutate(df = paste((NumDF), round(DenDF, 0), sep = ", "),
         response = "deltaCEWL")

```

Group Export

Put all the model statistics into one df/csv - one for among-model comparisons, and one for within-model stats.

```

experiment_model_compare <- CEWL_across %>%
  rbind(osml_across) %>%
  rbind(hct_across) %>%
  rbind(SMI_across) %>%
  dplyr::select(response, model,
               RMSE, Rsq, AICc, Delta_AICc) %>%
  mutate(RMSE = round(RMSE, 2),
         Rsq = round(Rsq, 2),
         AICc = round(AICc, 2),
         Delta_AICc = round(Delta_AICc, 2))
write.csv(experiment_model_compare,
         "./results_statistics/exp_model_comparisons.csv")

experiment_model_values <- CEWL_within %>%
  rbind(osml_within) %>%
  rbind(hct_within) %>%
  rbind(SMI_within) %>%
  dplyr::select(response, model, predictor,
               seq_sum_of_squares = Sum.Sq,
               df,
               F_statistic = F.value,
               p_value = Pr..F.) %>%
  mutate(seq_sum_of_squares = round(seq_sum_of_squares, 0),
         F_statistic = round(F_statistic, 2))
write.csv(experiment_model_values, "./results_statistics/exp_model_values.csv")

```

Effect Estimates

End Value CIs

Now, we can use the `emmeans` function from the `emmeans` package to estimate marginal means and confidence intervals for the values among treatment groups at the end of the experiment. But, to get for each treatment group, we need to run a new model with day 8 data only and `tmt` as a single, 4-category variable.

```
# Body Condition
SMI_mod_end <- lmerTest::lmer(data = end_vals,
                             SMI ~ tmt +
                             (1|trial_number))
SMI_emmeans <- data.frame(emmeans(SMI_mod_end, "tmt")) %>%
  mutate(response = "Body Condition (g)")
SMI_pairwise <- data.frame(pairs(emmeans(SMI_mod_end, "tmt")) %>%
  mutate(response = "Body Condition (g)")

# Hematocrit
hct_mod_end <- lmerTest::lmer(data = end_vals,
                             hematocrit_percent ~ tmt +
                             (1|trial_number))
hct_emmeans <- data.frame(emmeans(hct_mod_end, "tmt")) %>%
  mutate(response = "Hematocrit (%)")
hct_pairwise <- data.frame(pairs(emmeans(hct_mod_end, "tmt")) %>%
  mutate(response = "Hematocrit (%)")

# Plasma Osmolality
osml_mod_end <- lmerTest::lmer(data = end_vals,
                             osmolality_mmol_kg_mean ~ tmt +
                             (1|trial_number))
osml_emmeans <- data.frame(emmeans(osml_mod_end, "tmt")) %>%
  mutate(response = "Plasma Osmolality (mmol/kg)")
osml_pairwise <- data.frame(pairs(emmeans(osml_mod_end, "tmt")) %>%
  mutate(response = "Plasma Osmolality (mmol/kg)")

# CEWL
CEWL_mod_end <- lmerTest::lmer(data = end_vals,
                              CEWL_g_m2h_mean ~ tmt +
                              (1|trial_number))
CEWL_emmeans <- data.frame(emmeans(CEWL_mod_end, "tmt")) %>%
  mutate(response = "CEWL (g/m2h)")
CEWL_pairwise <- data.frame(pairs(emmeans(CEWL_mod_end, "tmt")) %>%
  mutate(response = "CEWL (g/m2h)")

# put together?
# use indiv for boxplots
```

Rate of Change

```
# Body Condition
SMI_mod_day <- lmerTest::lmer(data = dat_no_rehab,
                             SMI ~ day_n * tmt +
                             (1|trial_number/individual_ID))
SMI_emptrends <- data.frame(emptrends(SMI_mod_day, "tmt", var = "day_n")) %>%
```

```

mutate(response = "Body Condition (g')")
SMI_pairedtrend <- data.frame(pairs(emtrends(SMI_mod_day, "tmt", var = "day_n"))) %>%
  mutate(response = "Body Condition (g')")

# Hematocrit
hct_mod_day <- lmerTest::lmer(data = dat_no_rehab,
                             hematocrit_percent ~ day_n * tmt +
                             (1|trial_number/individual_ID))
hct_emtrends <- data.frame(emtrends(hct_mod_day, "tmt", var = "day_n")) %>%
  mutate(response = "Hematocrit (%)")
hct_pairedtrend <- data.frame(pairs(emtrends(hct_mod_day, "tmt", var = "day_n"))) %>%
  mutate(response = "Hematocrit (%)")

# Plasma Osmolality
osml_mod_day <- lmerTest::lmer(data = dat_no_rehab,
                              osmolality_mmol_kg_mean ~ day_n * tmt +
                              (1|trial_number/individual_ID))
osml_emtrends <- data.frame(emtrends(osml_mod_day, "tmt", var = "day_n")) %>%
  mutate(response = "Plasma Osmolality (mmol/kg)")
osml_pairedtrend <- data.frame(pairs(emtrends(osml_mod_day, "tmt", var = "day_n"))) %>%
  mutate(response = "Plasma Osmolality (mmol/kg)")

# CEWL
CEWL_mod_day <- lmerTest::lmer(data = dat_no_rehab,
                              CEWL_g_m2h_mean ~ day_n * tmt +
                              (1|trial_number))
CEWL_emtrends <- data.frame(emtrends(CEWL_mod_day, "tmt", var = "day_n")) %>%
  mutate(response = "CEWL (g/m2h)")
CEWL_pairedtrend <- data.frame(pairs(emtrends(CEWL_mod_day, "tmt", var = "day_n"))) %>%
  mutate(response = "CEWL (g/m2h)")

# put together
all_emtrends <- CEWL_emtrends %>%
  rbind(osml_emtrends) %>%
  rbind(hct_emtrends) %>%
  rbind(SMI_emtrends) %>%
  mutate(confint95 = paste(round(lower.CL, 2), round(upper.CL, 2), sep = ", ")) %>%
  dplyr::select(response, tmt,
                per_day_change = day_n.trend,
                confint95,
                SE, df)
#write.csv(all_emtrends, "./results_statistics/exp_emtrends_per_day_change.csv")

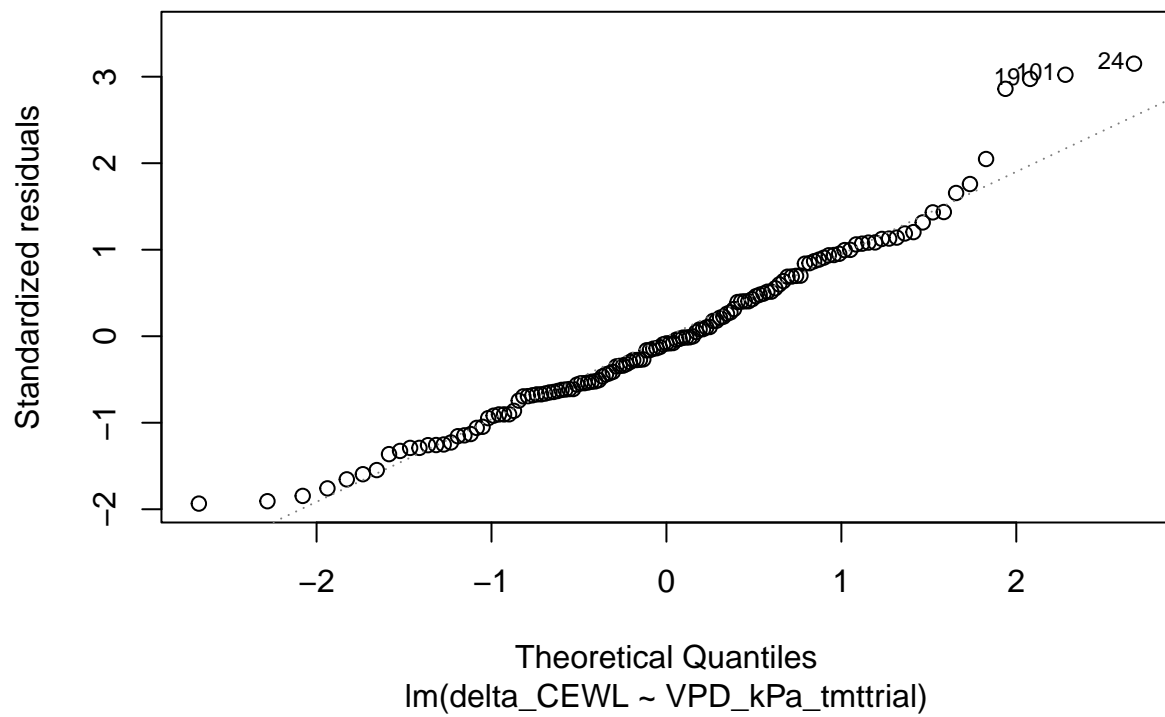
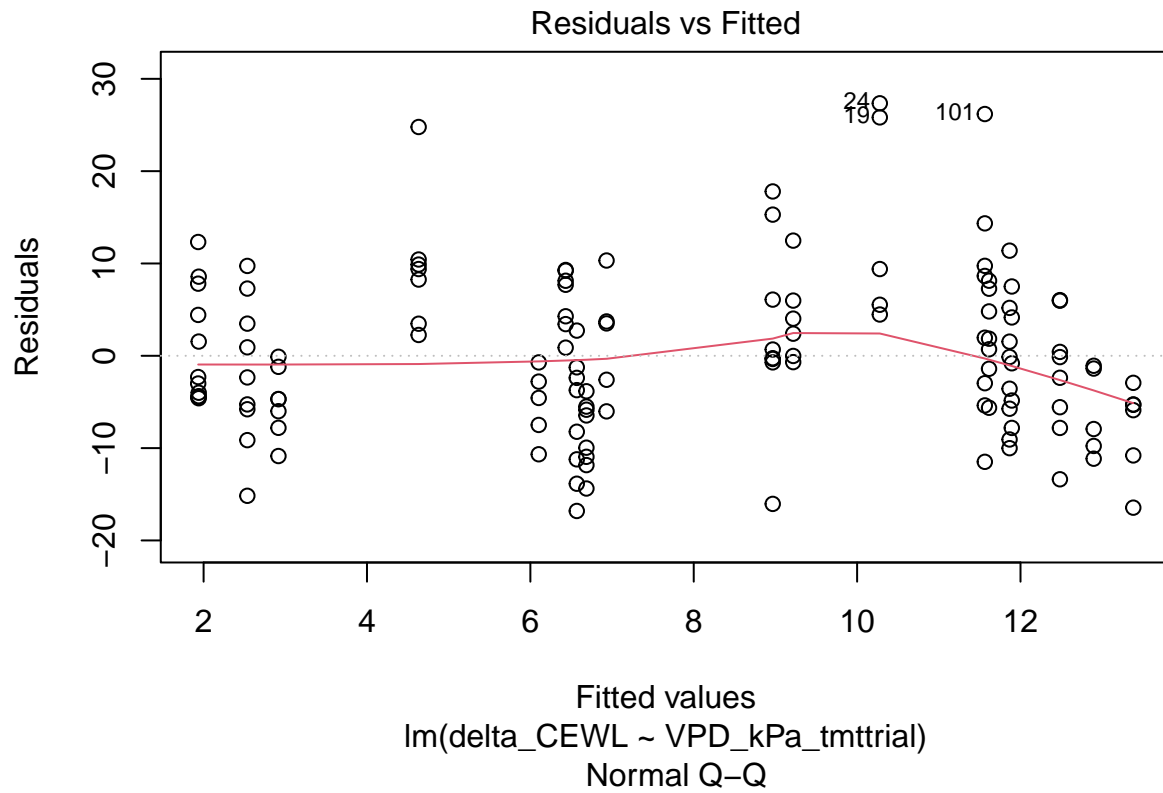
```

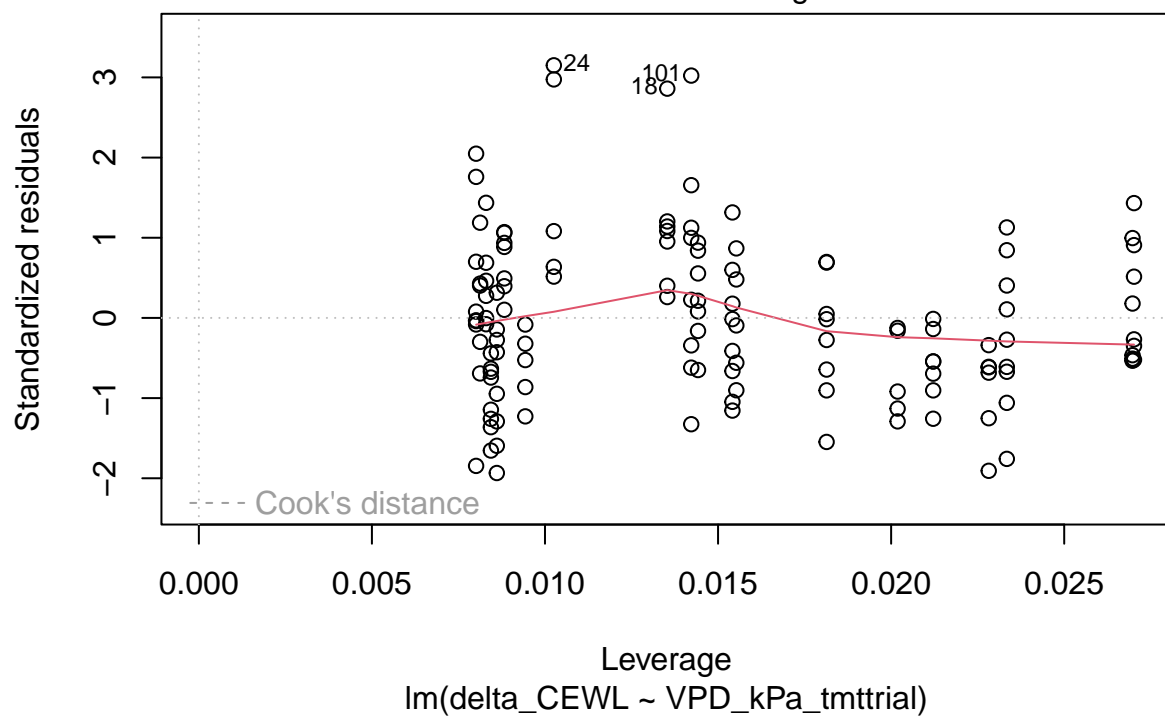
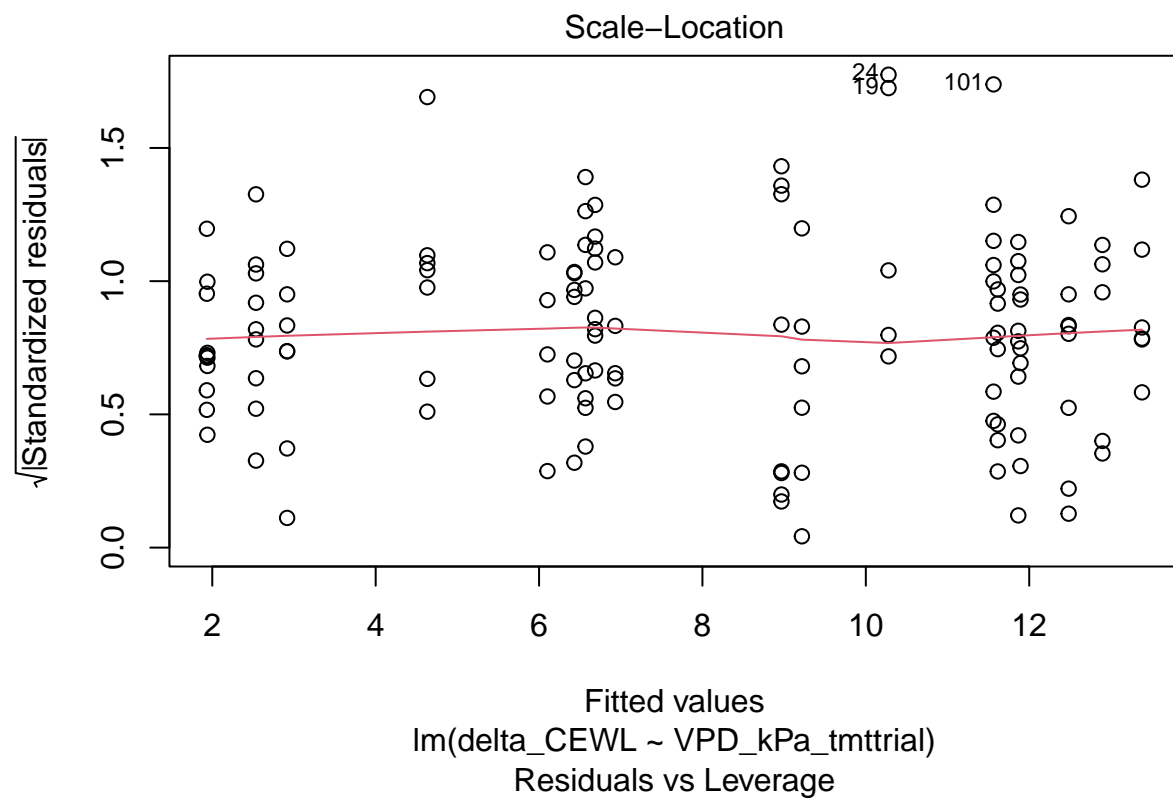
CEWL ~ VPD

```

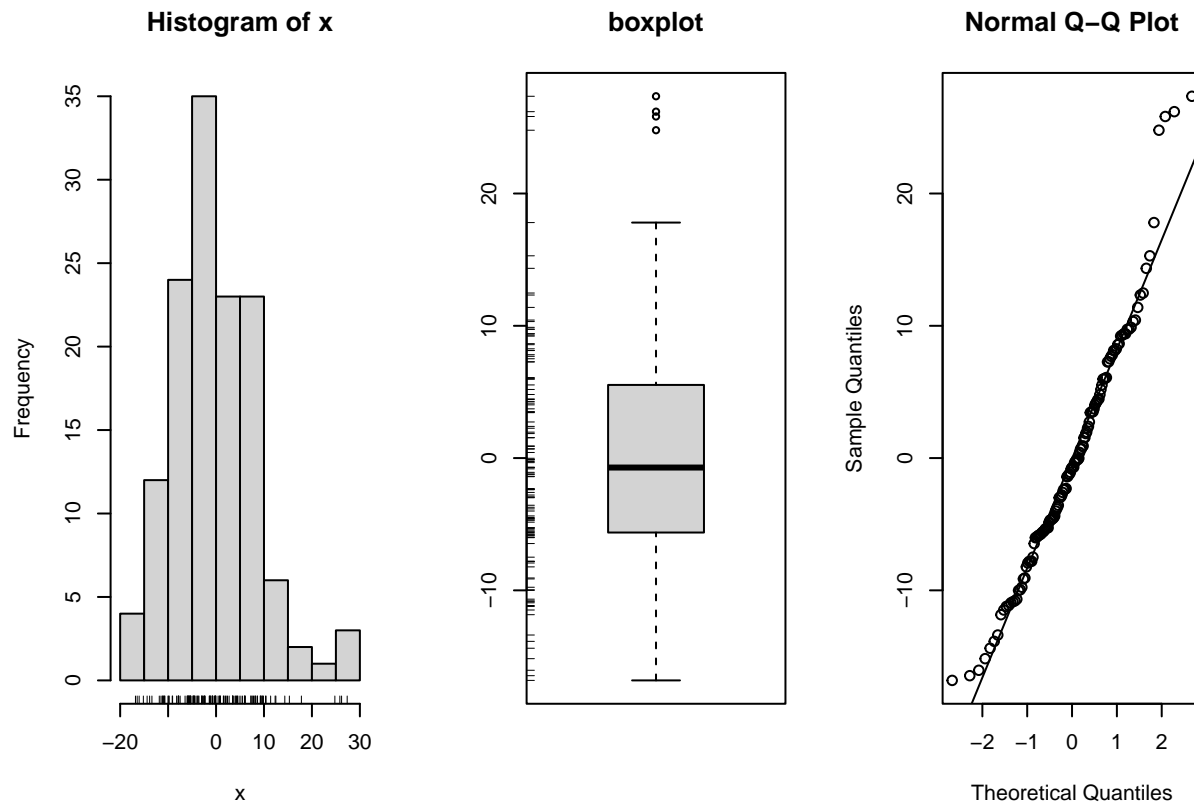
CEWL_VPD_lm <- lm(data = dat_no_rehab_deltaCEWL,
                  delta_CEWL ~ VPD_kPa_tmttrial)
plot(CEWL_VPD_lm)

```





```
simple.eda(residuals(CEWL_VPD_lm))
```



```
shapiro.test(residuals(CEWL_VPD_lm))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(CEWL_VPD_lm)
## W = 0.96694, p-value = 0.002527
```

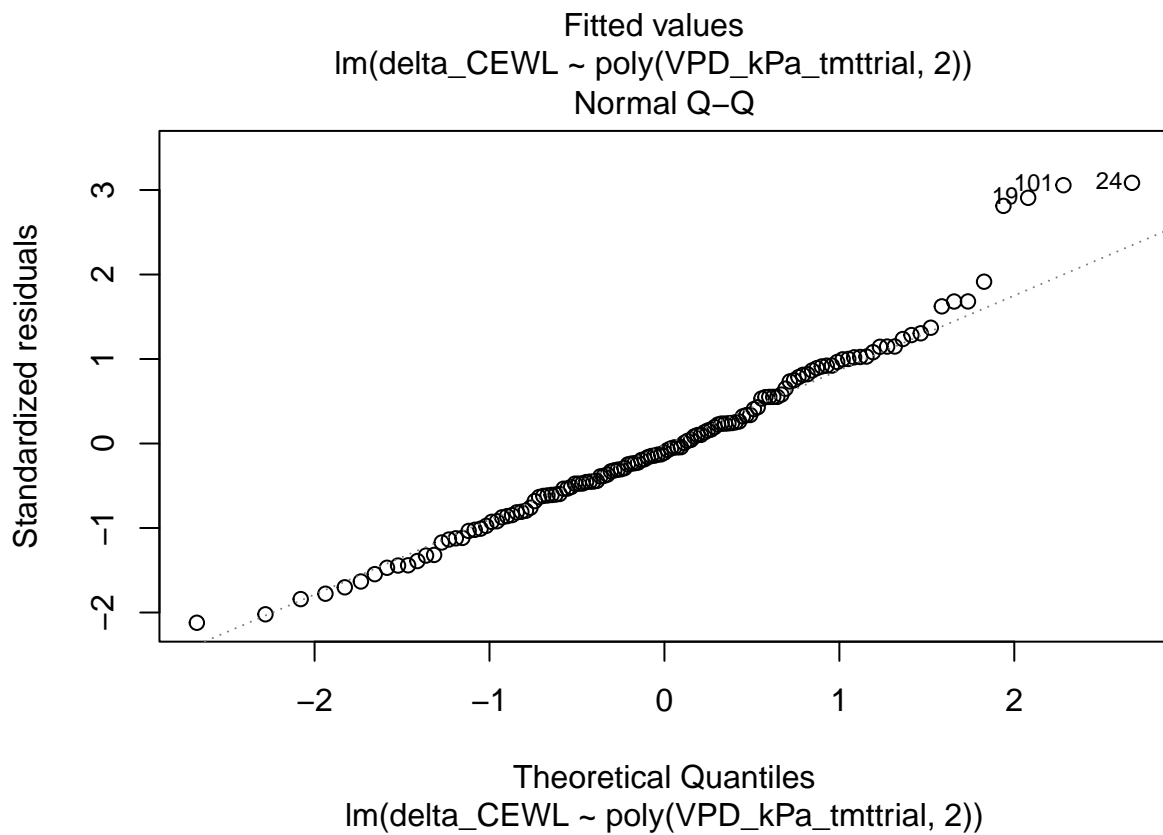
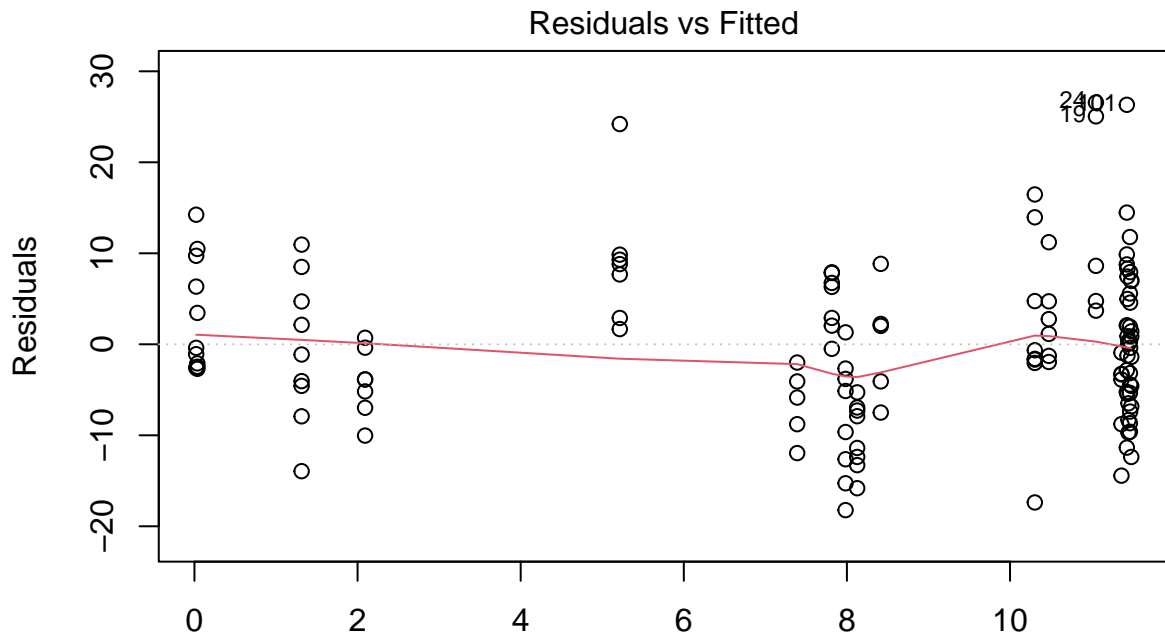
```
summary(CEWL_VPD_lm)
```

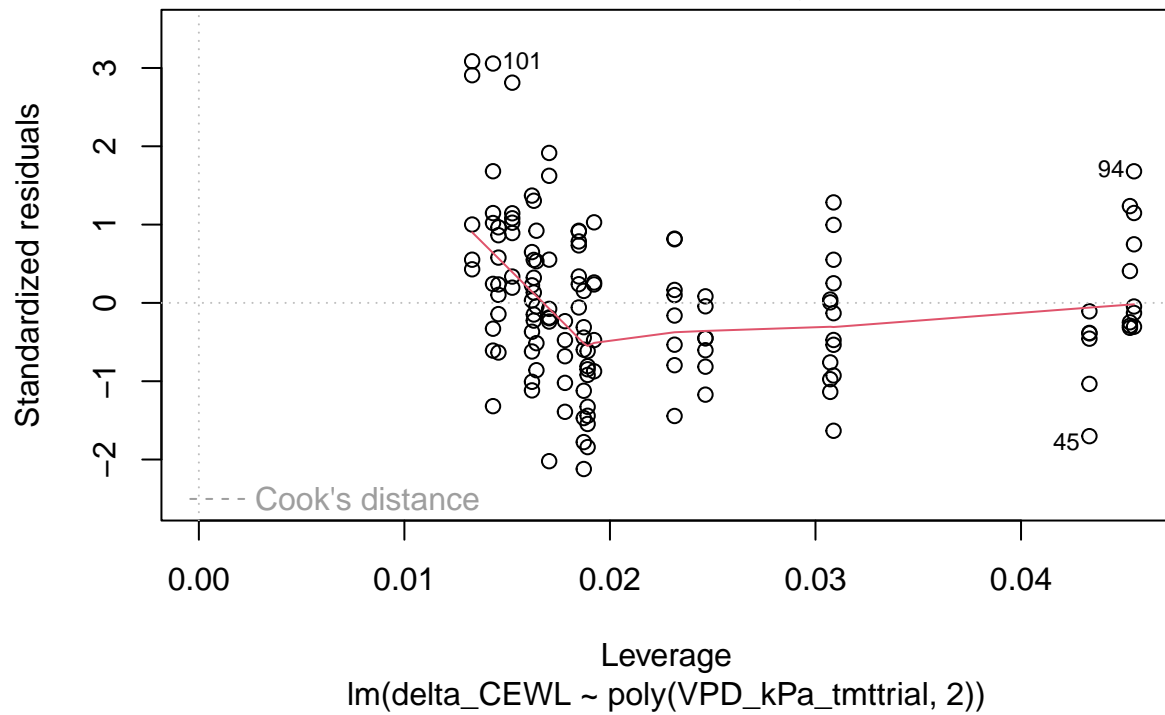
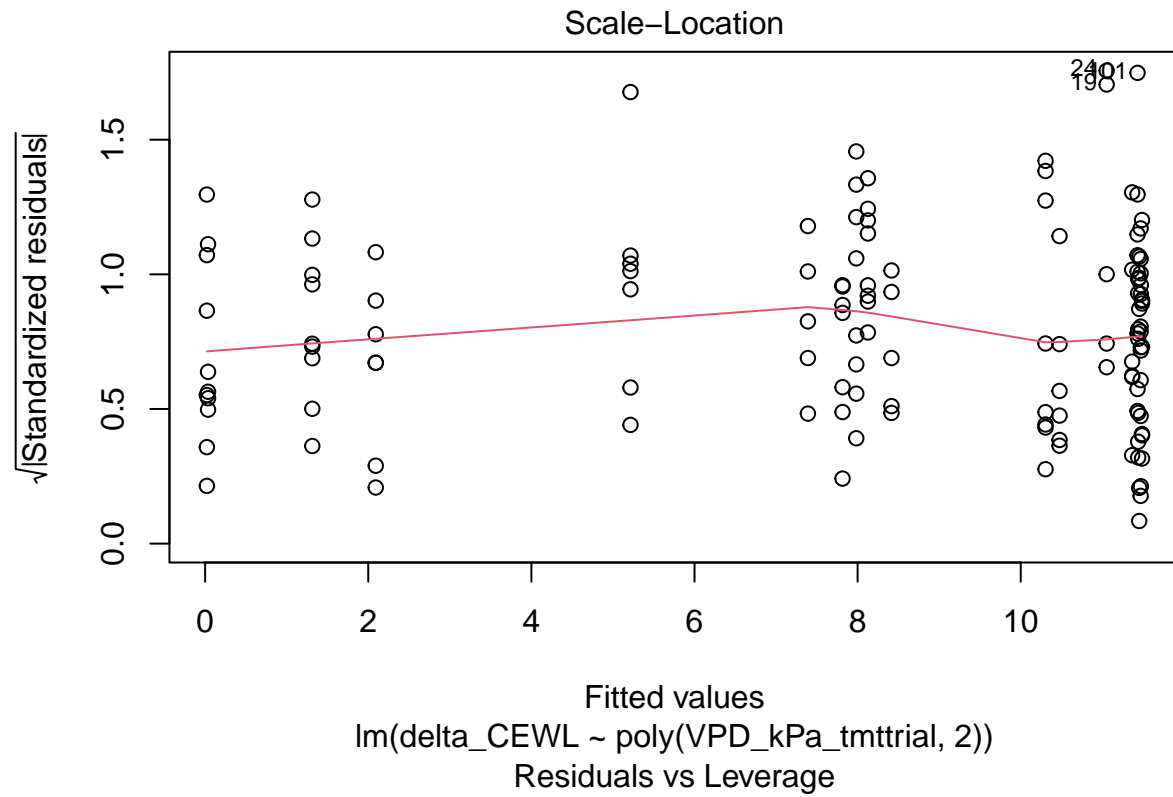
```
##
## Call:
## lm(formula = delta_CEWL ~ VPD_kPa_tmttrial, data = dat_no_rehab_deltaCEWL)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.8069  -5.6289  -0.7159   5.5303  27.3517
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    13.960     1.415   9.866 < 2e-16 ***
## VPD_kPa_tmttrial -2.959     0.594  -4.982 1.95e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.726 on 131 degrees of freedom
## Multiple R-squared:  0.1593, Adjusted R-squared:  0.1529
## F-statistic: 24.82 on 1 and 131 DF, p-value: 1.952e-06
```

Even though the data is slightly nonlinear, a linear model does a fine job explaining the data.

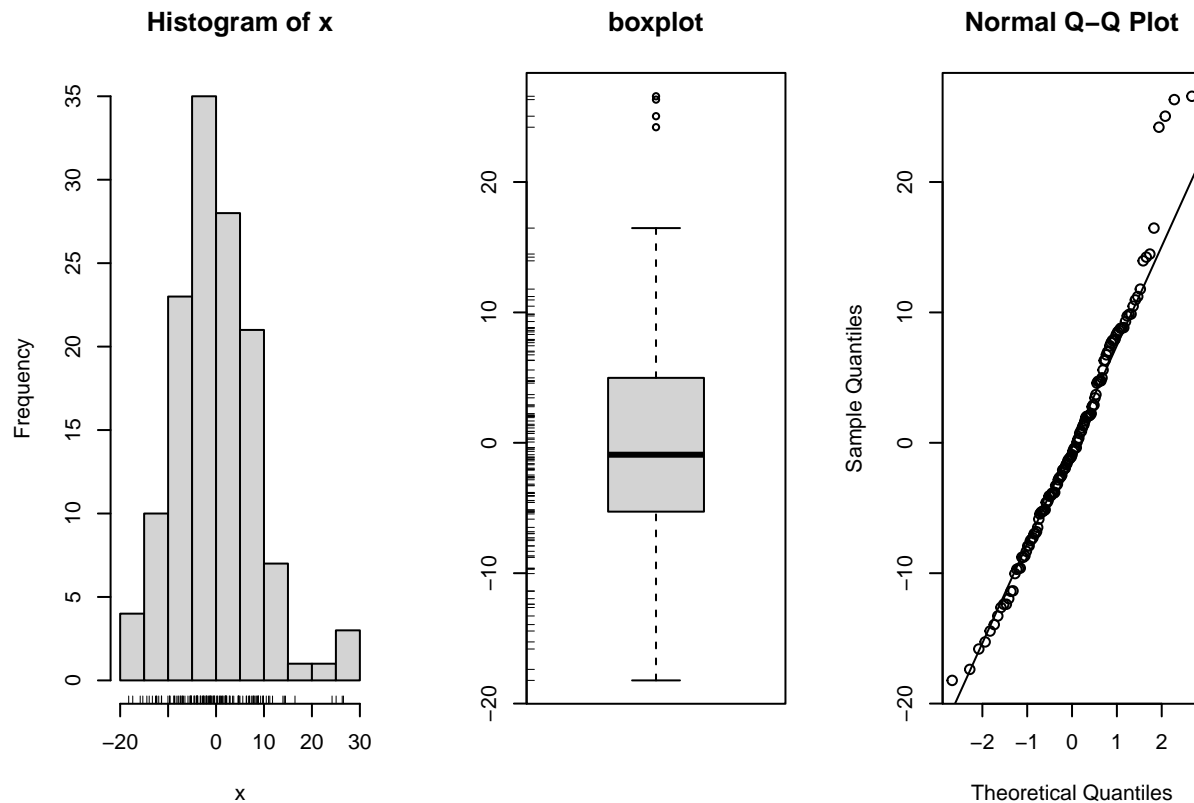
We will double check a comparison of a polynomial model, just to be sure:

```
CEWL_VPD_poly <- lm(data = dat_no_rehab_deltaCEWL,
                     delta_CEWL ~ poly(VPD_kPa_tmtrial, 2))
plot(CEWL_VPD_poly)
```





```
simple.eda(residuals(CEWL_VPD_poly))
```



```
shapiro.test(residuals(CEWL_VPD_poly))
```

```
##
##  Shapiro-Wilk normality test
##
## data:  residuals(CEWL_VPD_poly)
## W = 0.97276, p-value = 0.00896
```

```
summary(CEWL_VPD_poly)
```

```
##
## Call:
## lm(formula = delta_CEWL ~ poly(VPD_kPa_tmttrial, 2), data = dat_no_rehab_deltaCEWL)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -18.222  -5.281  -0.914   4.981  26.576
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      8.004     0.752  10.643 < 2e-16 ***
## poly(VPD_kPa_tmttrial, 2)1 -43.470     8.672  -5.013 1.72e-06 ***
## poly(VPD_kPa_tmttrial, 2)2 -14.069     8.672  -1.622  0.107
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.672 on 130 degrees of freedom
## Multiple R-squared:  0.176, Adjusted R-squared:  0.1633
## F-statistic: 13.88 on 2 and 130 DF, p-value: 3.442e-06
```

LINE assumptions are equally-well-met.

The polynomial factor is not significant, but the R-sq is slightly higher for the poly model.

Compare RMSE and AIC:

```
sqrt(mean((residuals(CEWL_VPD_lm))^2))
```

```
## [1] 8.660173
```

```
sqrt(mean((residuals(CEWL_VPD_poly))^2))
```

```
## [1] 8.573822
```

```
CEWL_VPD_models <- list(CEWL_VPD_lm, CEWL_VPD_poly)
```

```
CEWL_VPD_mod_names <- data.frame(model =  
                                c('linear',  
                                  'polynomial'  
                                ))
```

```
CEWL_VPD_AICc <- data.frame(aictab(cand.set = CEWL_VPD_models,  
                                modnames = CEWL_VPD_mod_names))
```

```
CEWL_VPD_AICc
```

```
##      model K      AICc Delta_AICc Modellik      AICcWt      LL      Cum.Wt  
## 2 polynomial 4 957.3080  0.0000000 1.0000000 0.5669881 -474.4977 0.5669881  
## 1      linear 3 957.8471  0.5391459 0.7637056 0.4330119 -475.8305 1.0000000
```

RMSE is slightly lower for the polynomial model. But, AIC is not meaningfully different between the two versions.

I'll use the lm as the best model/

```
summary(CEWL_VPD_lm)
```

```
##
```

```
## Call:
```

```
## lm(formula = delta_CEWL ~ VPD_kPa_tmttrial, data = dat_no_rehab_deltaCEWL)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max  
## -16.8069  -5.6289  -0.7159   5.5303  27.3517
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept)      13.960      1.415   9.866 < 2e-16 ***  
## VPD_kPa_tmttrial  -2.959      0.594  -4.982 1.95e-06 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 8.726 on 131 degrees of freedom
```

```
## Multiple R-squared:  0.1593, Adjusted R-squared:  0.1529
```

```
## F-statistic: 24.82 on 1 and 131 DF, p-value: 1.952e-06
```

```
anova(CEWL_VPD_lm)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: delta_CEWL
```

```
##              Df Sum Sq Mean Sq F value      Pr(>F)  
## VPD_kPa_tmttrial  1 1889.7 1889.67  24.817 1.952e-06 ***
```

```
## Residuals      131 9974.8   76.14
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(estimate = -2.96, SE = 0.6, df = (1, 131), F = 24.82, p < 0.0001, Rsq = 0.16)
```

Recovery Models

I want to know how the 2-day recovery period affects physiology relative to post- and pre- experiment. To do this, I'll use a two-sided t-test comparing delta to the hypothesis of $\mu=0$.

First, calculate the mean delta and SEM?

```
recovery_v_post_exp_summary <- recovery_v_post_exp %>%
  group_by(type) %>%
  summarise(mean_delta_osml = mean(delta_osml_10_8, na.rm = T),
            mean_delta_hct = mean(delta_hct_10_8, na.rm = T),
            mean_delta_SMI = mean(delta_SMI_10_8, na.rm = T))
```

SMI

```
SMI_rmod_post_exp <- t.test(recovery_v_post_exp$delta_SMI_10_8,
                           mu = 0, alternative = "two.sided")
SMI_rmod_post_exp
```

```
##
## One Sample t-test
##
## data:  recovery_v_post_exp$delta_SMI_10_8
## t = 6.677, df = 131, p-value = 6.292e-10
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##  0.3171205 0.5841444
## sample estimates:
## mean of x
## 0.4506324
```

```
SMI_rmod_pre_exp <- t.test(recovery_v_pre_exp$delta_SMI_10_0,
                           mu = 0, alternative = "two.sided")
SMI_rmod_pre_exp
```

```
##
## One Sample t-test
##
## data:  recovery_v_pre_exp$delta_SMI_10_0
## t = -11.542, df = 131, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -1.3426322 -0.9497445
## sample estimates:
## mean of x
## -1.146188
```

Hematocrit

```
hct_rmod_post_exp <- t.test(recovery_v_post_exp$delta_hct_10_8,  
                             mu = 0, alternative = "two.sided")  
hct_rmod_post_exp
```

```
##  
## One Sample t-test  
##  
## data: recovery_v_post_exp$delta_hct_10_8  
## t = -4.2083, df = 126, p-value = 4.85e-05  
## alternative hypothesis: true mean is not equal to 0  
## 95 percent confidence interval:  
## -3.033119 -1.092866  
## sample estimates:  
## mean of x  
## -2.062992
```

```
hct_rmod_pre_exp <- t.test(recovery_v_pre_exp$delta_hct_10_0,  
                             mu = 0, alternative = "two.sided")  
hct_rmod_pre_exp
```

```
##  
## One Sample t-test  
##  
## data: recovery_v_pre_exp$delta_hct_10_0  
## t = -22.249, df = 129, p-value < 2.2e-16  
## alternative hypothesis: true mean is not equal to 0  
## 95 percent confidence interval:  
## -13.50271 -11.29729  
## sample estimates:  
## mean of x  
## -12.4
```

Osmolality

```
osml_rmod_post_exp <- t.test(recovery_v_post_exp$delta_osml_10_8,  
                              mu = 0, alternative = "two.sided")  
osml_rmod_post_exp
```

```
##  
## One Sample t-test  
##  
## data: recovery_v_post_exp$delta_osml_10_8  
## t = 3.4782, df = 121, p-value = 0.0007021  
## alternative hypothesis: true mean is not equal to 0  
## 95 percent confidence interval:  
## 4.330203 15.772256  
## sample estimates:  
## mean of x  
## 10.05123
```

```
osml_rmod_pre_exp <- t.test(recovery_v_pre_exp$delta_osml_10_0,  
                              mu = 0, alternative = "two.sided")  
osml_rmod_pre_exp
```



```
##
## One Sample t-test
##
## data: recovery_v_pre_exp$delta_osml_10_0
## t = 3.75, df = 130, p-value = 0.0002651
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## 5.758213 18.618378
## sample estimates:
## mean of x
## 12.1883
```

Group Export

```
recovery_stats <- broom.mixed::tidy(osml_rmod_pre_exp) %>%
  rbind(broom.mixed::tidy(osml_rmod_post_exp)) %>%
  rbind(broom.mixed::tidy(hct_rmod_pre_exp)) %>%
  rbind(broom.mixed::tidy(hct_rmod_post_exp)) %>%
  rbind(broom.mixed::tidy(SMI_rmod_pre_exp)) %>%
  rbind(broom.mixed::tidy(SMI_rmod_post_exp)) %>%
  mutate(response = c(rep("Plasma Osmolality (mmol/kg)", 2),
                        rep("Hematocrit (%)", 2),
                        rep("Body Condition (g)", 2)),
         pre_post_exp = c(rep(c("pre", "post"), 3)))
write.csv(recovery_stats,
         "./results_statistics/recovery_stats.csv")
```

Figures

Colors & Shapes

```
CH_color <- brewer.pal(4, "Spectral")[c(4)]
HH_color <- brewer.pal(4, "Spectral")[c(2)]
CD_color <- brewer.pal(4, "Spectral")[c(3)]
HD_color <- brewer.pal(4, "Spectral")[c(1)]
my_colors <- c(CH_color, HH_color, CD_color, HD_color)

CH_shp <- 15
HH_shp <- 19
CD_shp <- 22
HD_shp <- 21
CH_shp_box <- 22
HH_shp_box <- 21
my_shapes <- c(CH_shp, HH_shp, CD_shp, HD_shp)
my_shapes_box <- c(CH_shp_box, HH_shp_box, CD_shp, HD_shp)

my_labels <- c("Cool Humid\n0.6 kPa",
               "Hot Humid\n1.1 kPa",
               "Cool Dry\n2.5 kPa",
               "Hot Dry\n3.8 kPa")
```

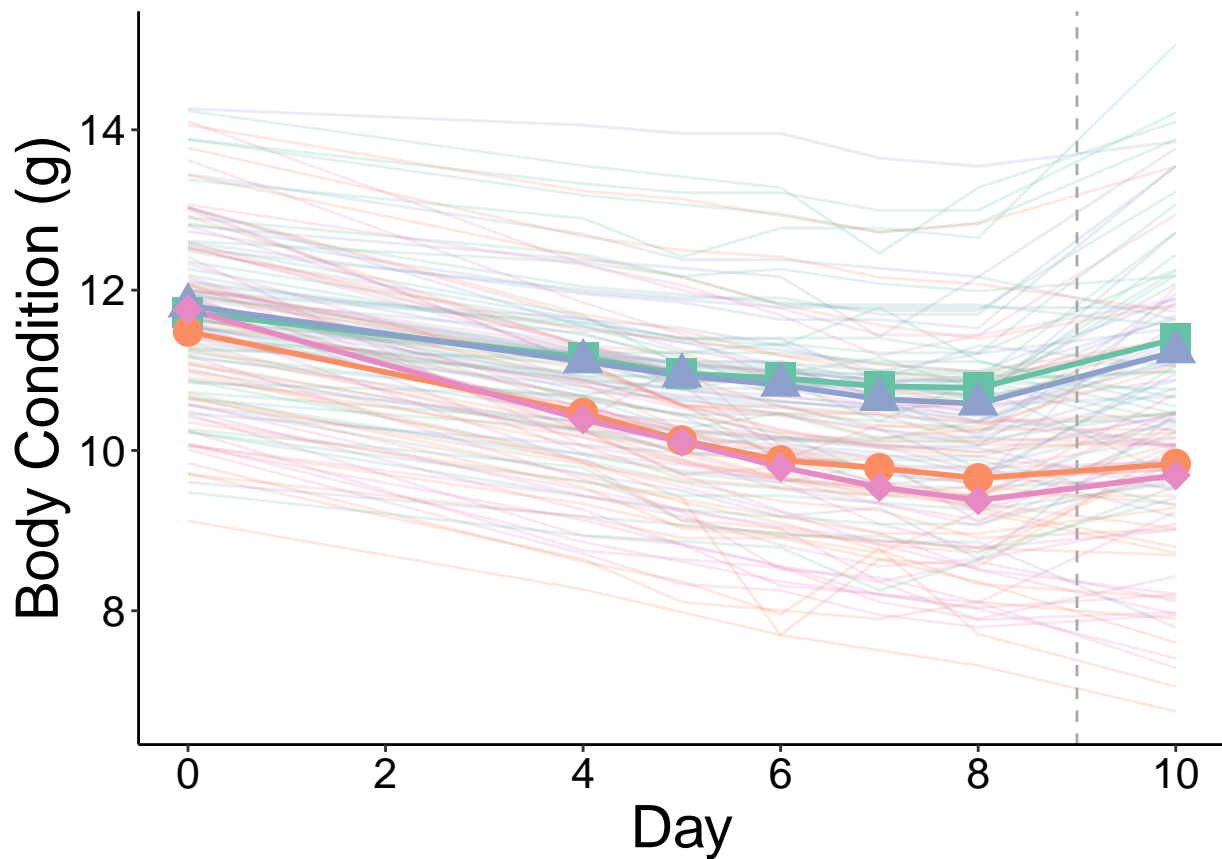
SMI

Ind + Means

```
ggplot() +
  geom_line(data = dat,
    aes(x = day_n,
        y = SMI,
        color = tmt,
        group = individual_ID),
    alpha = 0.2) +
  geom_line(data = means,
    aes(x = day_n,
        y = mean_SMI,
        color = tmt,
        group = tmt),
    alpha = 1,
    size = 1) +
  geom_point(data = means,
    aes(x = day_n,
        y = mean_SMI,
        color = tmt,
        shape = tmt),
    alpha = 1,
    size = 5) +
  geom_vline(xintercept = 9,
    linetype = "dashed",
    color = "darkgrey") +
  theme_classic() +
  scale_shape_manual(values = c(15:18), name = "") +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
  scale_color_brewer(palette = "Set2", name = "") +
  xlab("Day") +
  ylab("Body Condition (g)") +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
    family = "sans",
    size = 22),
    axis.text = element_text(color = "black",
    family = "sans",
    size = 16),
    legend.text = element_text(color = "black",
    family = "sans",
    size = 22),
    legend.text.align = 0,
    legend.position = "none",
    plot.margin = unit(c(0.1, #top
    0.1, #right
    0.1, #bottom
    0.41 #left
    ), "cm")) -> SMI_fig
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
```

SMI_fig



```
#ggsave(filename = "SMI_fig1.jpeg",
#        plot = SMI_fig,
#        path = "./results_figures",
#        device = "jpeg",
#        dpi = 1200,
#        width = 6, height = 6)
```

Means Only *MS*

```
ggplot() +
  #plot these first so they end up on the "bottom"
  geom_smooth(data = dat_no_rehab,
              aes(x = day_n,
                  y = SMI,
                  color = tmt,
                  group = tmt),
              method = "lm",
              se = F,
              size = 0.7) +
  geom_errorbar(data = means,
               aes(x = day_n,
                   y = mean_SMI,
                   color = tmt,
                   group = tmt,
```

```

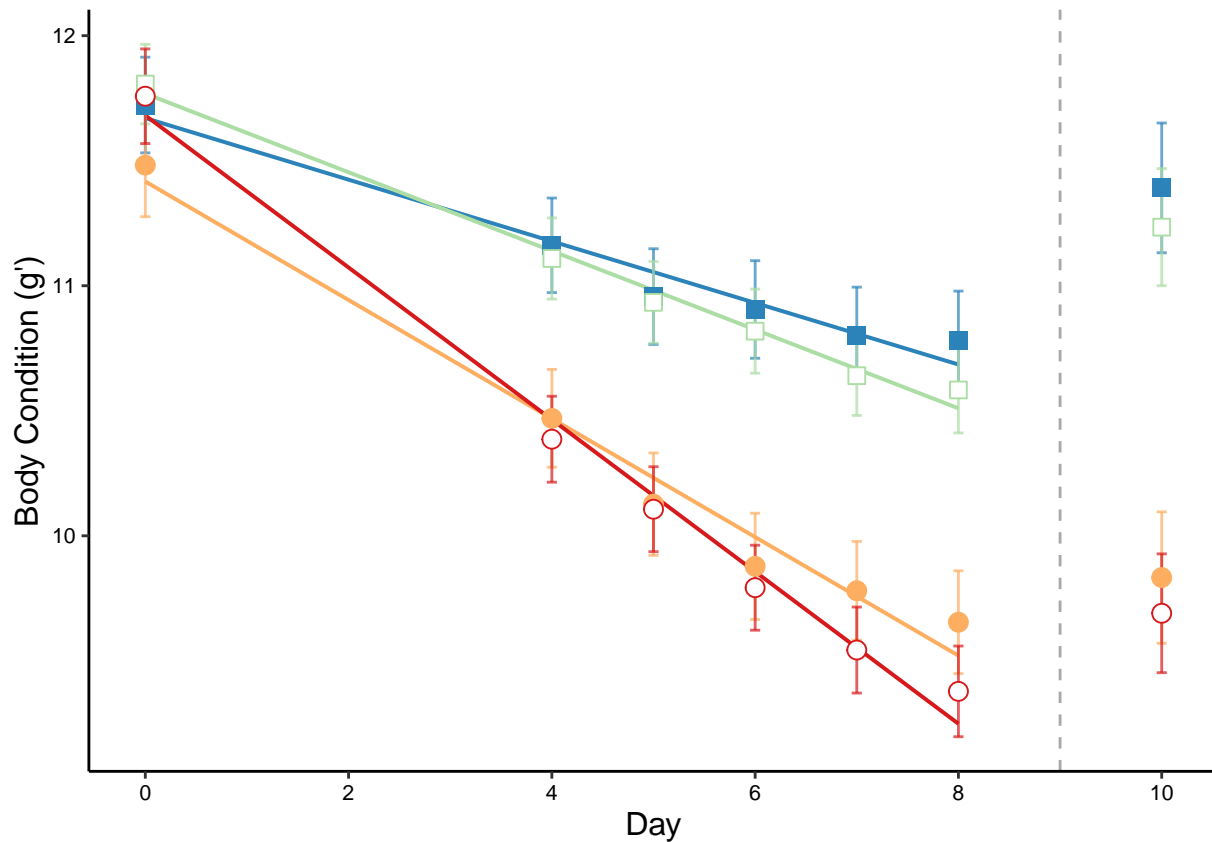
        ymin = mean_SMI-se_SMI,
        ymax = mean_SMI+se_SMI),
        width = .1,
        #position=position_dodge(.01),
        alpha = 0.7) +
#geom_line(data = means,
#          aes(x = day_n,
#              y = mean_SMI,
#              color = tmt,
#              #linetype = tmt,
#              group = tmt),
#          alpha = 1,
#          size = 0.5) +
geom_point(data = means,
           aes(x = day_n,
               y = mean_SMI,
               color = tmt,
               #fill = tmt,
               shape = tmt),
           fill = "white",
           alpha = 1,
           size = 3) +

theme_classic() +
scale_shape_manual(values = my_shapes, name = "",
                   labels = my_labels) +
scale_fill_manual(values = my_colors, name = "",
                  labels = my_labels) +
scale_color_manual(values = my_colors, name = "",
                   labels = my_labels) +
scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
scale_y_continuous(breaks = c(seq(8,12))),
                  labels = c(seq(8,12))) +

xlab("Day") +
ylab("Body Condition (g')") +
guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
geom_vline(xintercept = 9,
           linetype = "dashed",
           color = "darkgrey") +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 12),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 8),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
      legend.text.align = 0,
      legend.position = "none",
      plot.margin = margin(t = 6, r = 6, b = 6, l = 10.8, unit = "pt"))
) -> SMI_fig_min
SMI_fig_min

```

```
## `geom_smooth()` using formula = 'y ~ x'
```



LM + SE

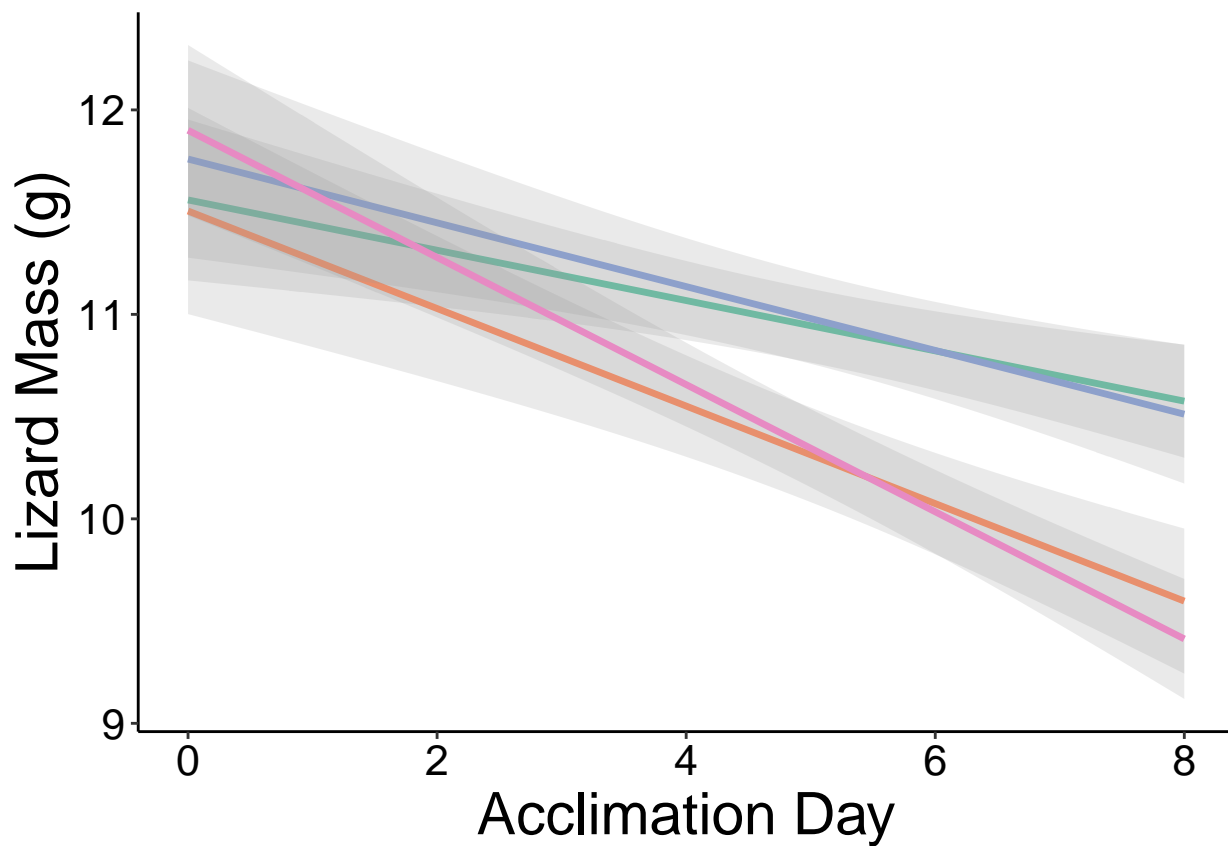
```
ggplot() +
  geom_smooth(data = dat_no_rehab,
             aes(x = day_n,
                 y = mass_g,
                 color = tmt,
                 group = tmt),
             formula = y ~ x,
             method = "lm",
             se = T,
             size = 1.2,
             alpha = 0.2) +
  theme_classic() +
  #scale_shape_manual(values = c(15:18), name = "") +
  scale_x_continuous(limits = c(0,8),
                    breaks = c(0, 2, 4, 6, 8)) +
  scale_color_brewer(palette = "Set2", name = "") +
  xlab("Acclimation Day") +
  ylab("Lizard Mass (g)") +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
                             family = "sans",
                             size = 22),
        axis.text = element_text(color = "black",
```

```

    family = "sans",
    size = 16),
  legend.text = element_text(color = "black",
    family = "sans",
    size = 22),

  legend.text.align = 0,
  legend.position = "none"
) -> SMI_lm_fig
SMI_lm_fig

```



Ending Values *MS*

```

ggplot() +
  geom_jitter(data = end_vals,
    aes(x = tmt,
      y = SMI,
      color = tmt,
      fill = tmt,
      shape = tmt),
    size = 1,
    alpha = 0.4,
    position = position_jitter(height = 0, width = 0.2)) +
  geom_errorbar(data = SMI_emmeans,
    aes(x = tmt,
      y = emmean,
      color = tmt,

```

```

      group = tmt,
      ymin = lower.CL,
      ymax = upper.CL),
    width = .1,
    alpha = 0.9) +
geom_point(data = SMI_emmeans,
  aes(x = tmt,
      y = emmean,
      #color = tmt,
      shape = tmt,
      fill = tmt),
  color = "black",
  size = 4) +
theme_classic() +
scale_shape_manual(values = my_shapes_box, name = "") +
scale_fill_manual(values = my_colors, name = "") +
scale_color_manual(values = my_colors, name = "") +
scale_y_continuous(limits = c(7,15),
  breaks = c(seq(7,15, by = 2)),
  labels = c(seq(7,15, by = 2))) +
scale_x_discrete(labels = c("Cool Humid\n0.6 kPa",
  "Hot Humid\n1.1 kPa",
  "Cool Dry\n2.5 kPa",
  "Hot Dry\n3.8 kPa")) +
xlab("") +

annotate(geom = "text", x = 4, y = 12.7, label = "B",
  size = 3) +
annotate(geom = "text", x = 4, y = 12.2, label = "9.4", #HD
  size = 3) +
annotate(geom = "text", x = 2, y = 14.2, label = "B",
  size = 3) +
annotate(geom = "text", x = 2, y = 13.7, label = "9.6", #HH
  size = 3) +
annotate(geom = "text", x = 3, y = 14.8, label = "A",
  size = 3) +
annotate(geom = "text", x = 3, y = 14.3, label = "10.7", #CD
  size = 3) +
annotate(geom = "text", x = 1, y = 14.6, label = "A",
  size = 3) +
annotate(geom = "text", x = 1, y = 14.1, label = "10.8", #CH
  size = 3) +

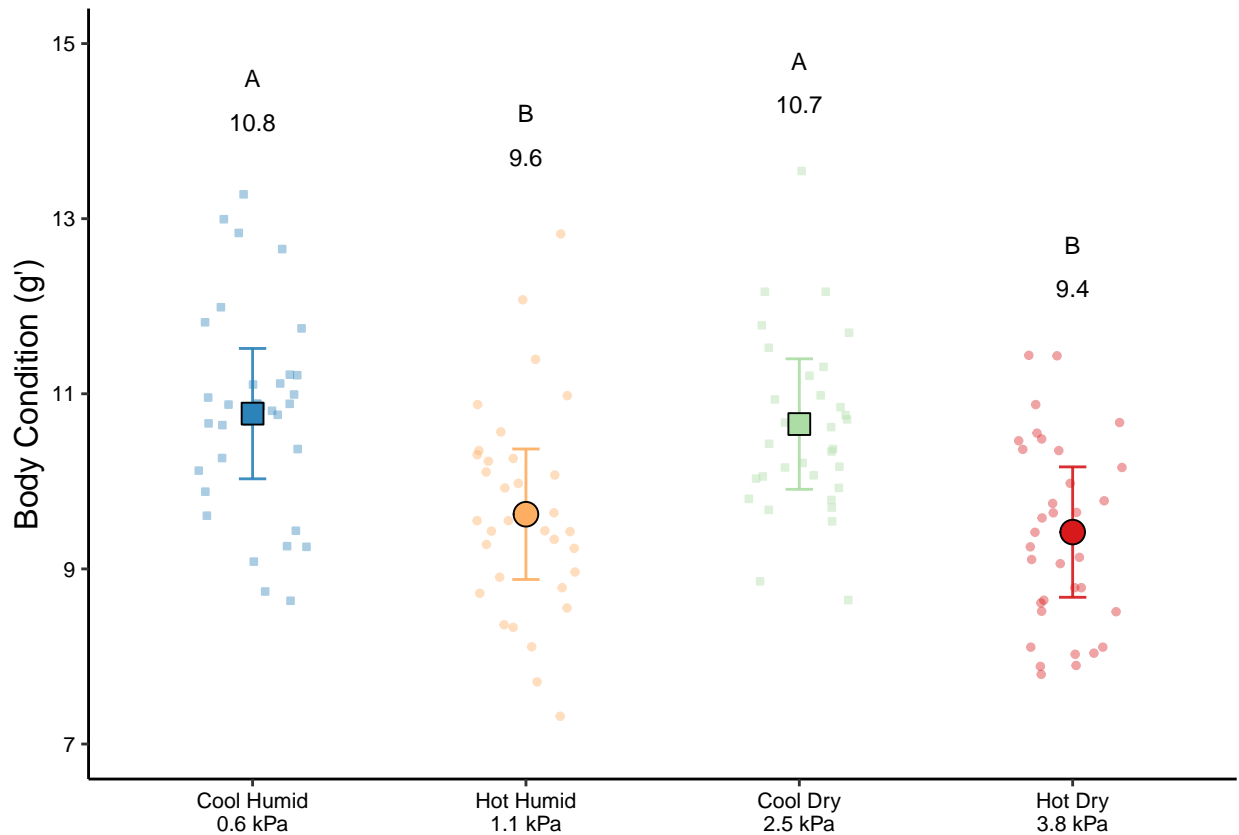
ylab("Body Condition (g')") +
theme(text = element_text(color = "black",
  family = "sans",
  size = 12),
  axis.text = element_text(color = "black",
  family = "sans",
  size = 8),
  #axis.text.x = element_blank(),
  legend.text = element_text(color = "black",
  family = "sans",

```

```

                                size = 8),
  legend.text.align = 0,
  legend.position = "none",
  plot.margin = unit(c(0, #top
                        0, #right
                        0, #bottom
                        3.4), "mm")
) -> SMI_end_boxplot
SMI_end_boxplot

```



SMI_emmeans

```

##           tmt      emmean      SE      df  lower.CL upper.CL
## 1 Cool Humid (0.6 kPa) 10.773306 0.3101046 6.511477 10.028724 11.51789
## 2 Hot Humid (1.1 kPa)  9.624568 0.3116093 6.634876  8.879428 10.36971
## 3 Cool Dry (2.5 kPa) 10.654259 0.3115933 6.634043  9.909138 11.39938
## 4 Hot Dry (3.8 kPa)  9.420686 0.3101705 6.516199  8.676064 10.16531
##
##      response
## 1 Body Condition (g')
## 2 Body Condition (g')
## 3 Body Condition (g')
## 4 Body Condition (g')

```

SMI_pairwise

```

##           contrast      estimate      SE      df
## 1 Cool Humid (0.6 kPa) - Hot Humid (1.1 kPa) 1.1487389 0.2384728 126.0137
## 2 Cool Humid (0.6 kPa) - Cool Dry (2.5 kPa)  0.1190474 0.2391374 126.1449

```



```
## 3   Cool Humid (0.6 kPa) - Hot Dry (3.8 kPa)  1.3526208 0.2370101 126.0772
## 4   Hot Humid (1.1 kPa) - Cool Dry (2.5 kPa) -1.0296915 0.2411556 126.1941
## 5   Hot Humid (1.1 kPa) - Hot Dry (3.8 kPa)  0.2038818 0.2390104 126.1219
## 6   Cool Dry (2.5 kPa) - Hot Dry (3.8 kPa)  1.2335733 0.2385858 126.0337
##      t.ratio      p.value      response
## 1  4.8170640 2.420102e-05 Body Condition (g')
## 2  0.4978201 9.594278e-01 Body Condition (g')
## 3  5.7070166 4.643724e-07 Body Condition (g')
## 4 -4.2698224 2.211875e-04 Body Condition (g')
## 5  0.8530250 8.288557e-01 Body Condition (g')
## 6  5.1703556 5.287879e-06 Body Condition (g')
```

Hct

Ind + Means

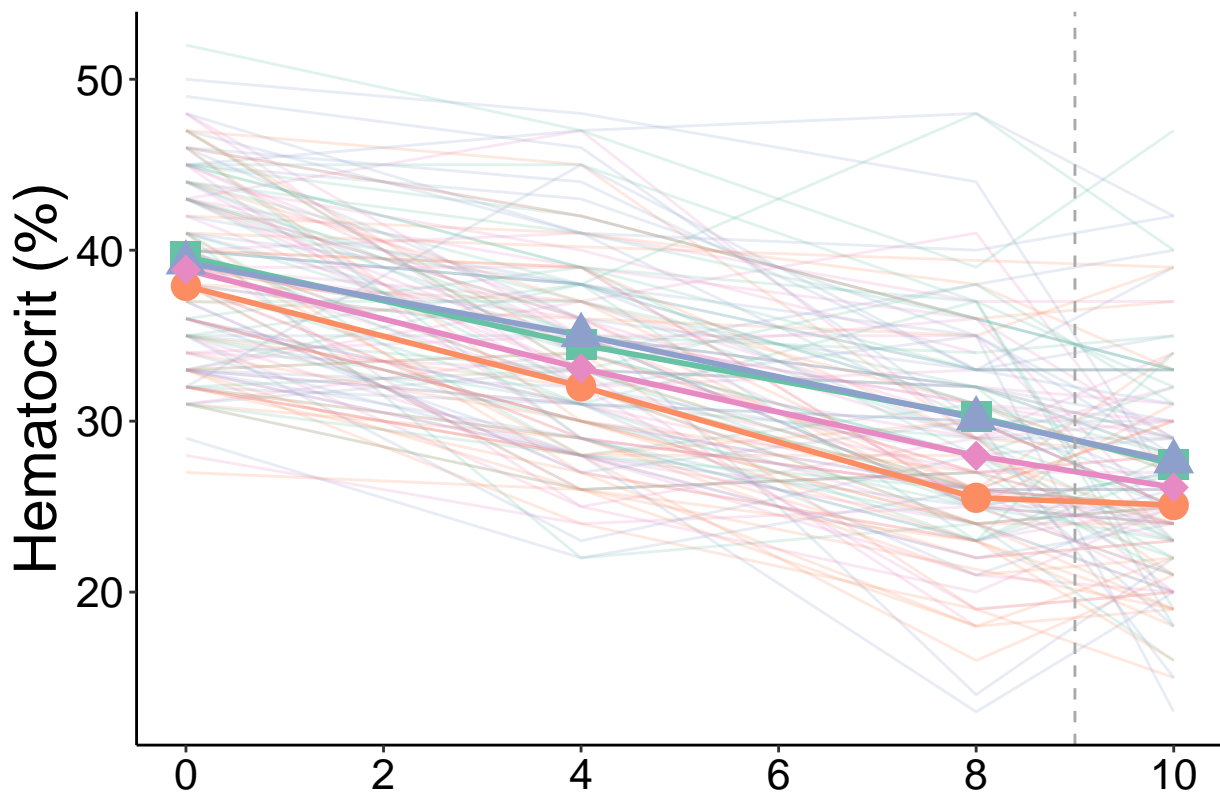
```
ggplot() +
  geom_line(data = dat[complete.cases(dat$hematocrit_percent),],
    aes(x = day_n,
        y = hematocrit_percent,
        color = tmt,
        group = individual_ID),
    alpha = 0.2) +
  geom_line(data = means[complete.cases(means$mean_hct),],
    aes(x = day_n,
        y = mean_hct,
        color = tmt,
        group = tmt),
    alpha = 1,
    size = 1) +
  geom_point(data = means,
    aes(x = day_n,
        y = mean_hct,
        color = tmt,
        shape = tmt),
    alpha = 1,
    size = 5) +
  geom_vline(xintercept = 9,
    linetype = "dashed",
    color = "darkgrey") +
  theme_classic() +
  scale_shape_manual(values = c(15:18), name = "") +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
  scale_color_brewer(palette = "Set2", name = "") +
  xlab("") +
  ylab("Hematocrit (%)") +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
    family = "sans",
    size = 22),
    axis.text = element_text(color = "black",
    family = "sans",
    size = 16),
    legend.text = element_text(color = "black",
```

```

family = "sans",
size = 22),
legend.text.align = 0,
legend.position = "none",
plot.margin = unit(c(0.1, #top
0.1, #right
0.1, #bottom
0.41 #left
), "cm")
) -> hct_fig
hct_fig

```

Warning: Removed 12 rows containing missing values (`geom_point()`).



Means Only *MS*

```

ggplot() +
  geom_smooth(data = dat_no_rehab[complete.cases(dat_no_rehab$hematocrit_percent),],
    aes(x = day_n,
        y = hematocrit_percent,
        color = tmt,
        group = tmt),
    method = "lm",
    se = F,
    size = 0.7) +
  geom_errorbar(data = means[complete.cases(means$mean_hct),],
    aes(x = day_n,
        y = mean_hct,

```

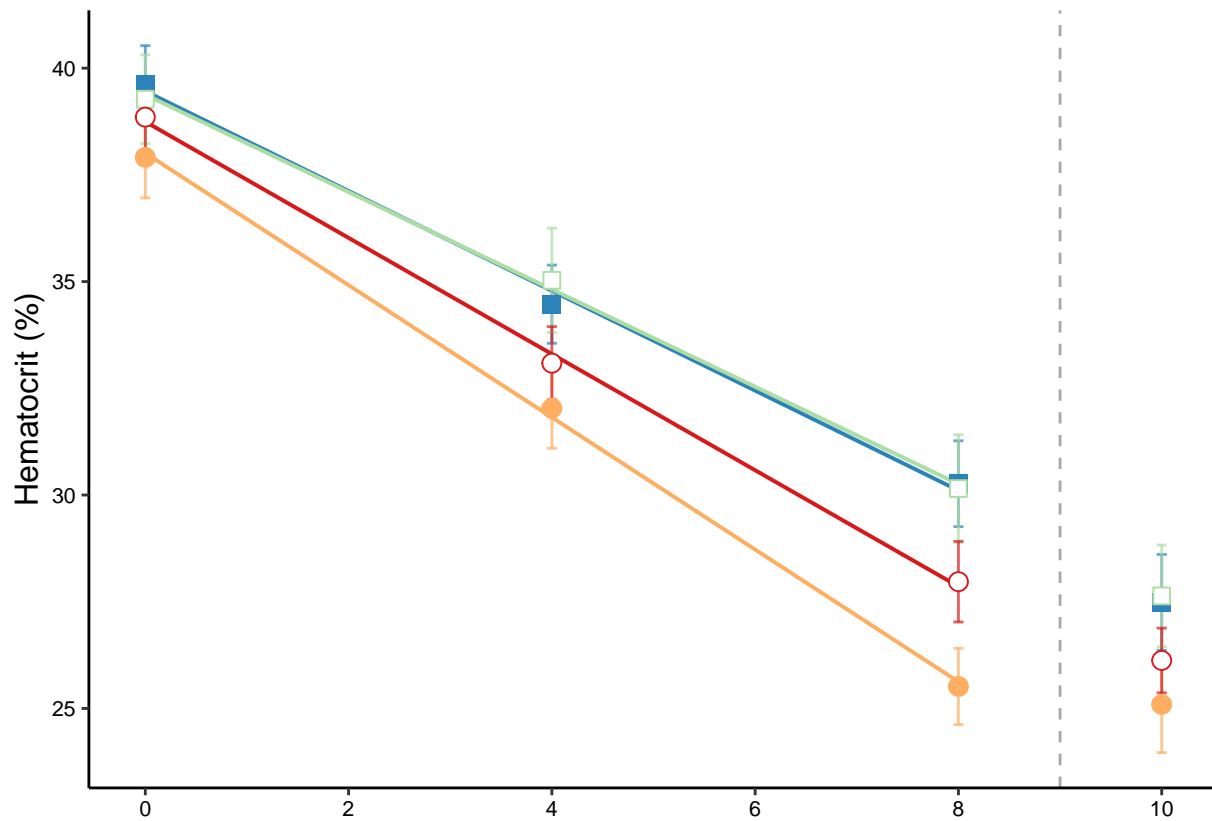
```

        color = tmt,
        group = tmt,
        ymin = mean_hct-se_hct,
        ymax = mean_hct+se_hct),
        width = .1,
        #position=position_dodge(.1),
        alpha = 0.7) +
#geom_line(data = means[complete.cases(means$mean_hct),],
#          aes(x = day_n,
#              y = mean_hct,
#              color = tmt,
#              #linetype = tmt,
#              group = tmt),
#          # alpha = 1,
#          # size = 0.5) +
geom_point(data = means[complete.cases(means$mean_hct),],
           aes(x = day_n,
               y = mean_hct,
               color = tmt,
               #fill = tmt,
               shape = tmt),
           alpha = 1,
           fill = "white",
           size = 3) +
theme_classic() +
scale_shape_manual(values = my_shapes, name = "",
                   labels = my_labels) +
scale_fill_manual(values = my_colors, name = "",
                  labels = my_labels) +
scale_color_manual(values = my_colors, name = "",
                   labels = my_labels) +
scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
scale_y_continuous(breaks = c(25, 30, 35, 40),
                   labels = c(25, 30, 35, 40),) +
xlab("") +
ylab("Hematocrit (%)") +
guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
geom_vline(xintercept = 9,
           linetype = "dashed",
           color = "darkgrey") +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 12),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 8),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
      legend.text.align = 0,
      legend.position = "none",
      plot.margin = margin(t = 6, r = 6, b = 0, l = 10.8, unit = "pt")
) -> hct_fig_min

```

```
hct_fig_min
```

```
## `geom_smooth()` using formula = 'y ~ x'
```



Ending Values *MS*

```
ggplot() +  
  geom_jitter(data = end_vals,  
    aes(x = tmt,  
        y = hematocrit_percent,  
        color = tmt,  
        fill = tmt,  
        shape = tmt),  
    size = 1,  
    alpha = 0.4,  
    position = position_jitter(height = 0, width = 0.2)) +  
  geom_errorbar(data = hct_emmeans,  
    aes(x = tmt,  
        y = emmean,  
        color = tmt,  
        group = tmt,  
        ymin = lower.CL,  
        ymax = upper.CL),  
    width = .1,  
    alpha = 0.9) +  
  geom_point(data = hct_emmeans,  
    aes(x = tmt,
```

```

        y = emmean,
        #color = tmt,
        shape = tmt,
        fill = tmt),
        color = "black",
        size = 4) +
theme_classic() +
scale_shape_manual(values = my_shapes_box, name = "") +
scale_fill_manual(values = my_colors, name = "") +
scale_color_manual(values = my_colors, name = "") +
xlab("") +
scale_y_continuous(limits = c(10,55),
                    breaks = c(seq(10,50, by = 10)),
                    labels = c(seq(10,50, by = 10))) +
scale_x_discrete(labels = c("Cool Humid\n0.6 kPa",
                             "Hot Humid\n1.1 kPa",
                             "Cool Dry\n2.5 kPa",
                             "Hot Dry\n3.8 kPa")) +

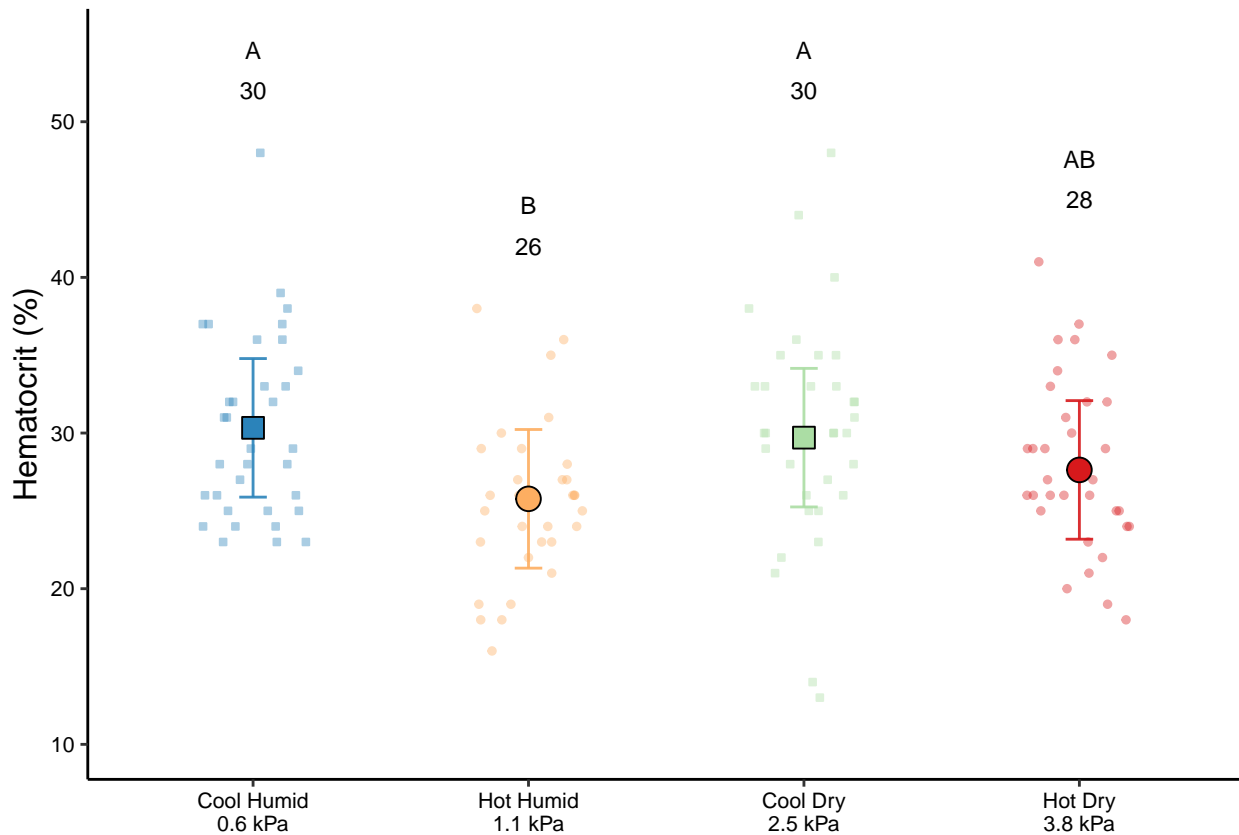
annotate(geom = "text", x = 4, y = 47.6, label = "AB",
         size = 3) +
annotate(geom = "text", x = 4, y = 45, label = "28", #HD
         size = 3) +
annotate(geom = "text", x = 2, y = 44.6, label = "B",
         size = 3) +
annotate(geom = "text", x = 2, y = 42, label = "26", #HH
         size = 3) +
annotate(geom = "text", x = 3, y = 54.6, label = "A",
         size = 3) +
annotate(geom = "text", x = 3, y = 52, label = "30", #CD
         size = 3) +
annotate(geom = "text", x = 1, y = 54.6, label = "A",
         size = 3) +
annotate(geom = "text", x = 1, y = 52, label = "30", #CH
         size = 3) +

ylab("Hematocrit (%)") +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 12),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 8),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
      legend.text.align = 0,
      legend.position = "none",
      plot.margin = unit(c(0, #top
                           0, #right
                           0, #bottom
                           2.24), "mm"))

) -> hct_end_boxplot
hct_end_boxplot

```

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



```
hct_emmeans
```

```
##           tmt      emmean      SE      df lower.CL upper.CL
## 1 Cool Humid (0.6 kPa) 30.33352 1.808392 5.860156 25.88283 34.78422
## 2 Hot Humid (1.1 kPa) 25.77326 1.830146 6.142261 21.32008 30.22645
## 3 Cool Dry (2.5 kPa) 29.70335 1.815476 5.950456 25.25206 34.15464
## 4 Hot Dry (3.8 kPa) 27.63092 1.815279 5.948653 23.17978 32.08205
##      response
## 1 Hematocrit (%)
## 2 Hematocrit (%)
## 3 Hematocrit (%)
## 4 Hematocrit (%)
```

```
hct_pairwise
```

```
##           contrast      estimate      SE      df
## 1 Cool Humid (0.6 kPa) - Hot Humid (1.1 kPa) 4.5602630 1.272328 123.0398
## 2 Cool Humid (0.6 kPa) - Cool Dry (2.5 kPa) 0.6301769 1.254531 123.1123
## 3 Cool Humid (0.6 kPa) - Hot Dry (3.8 kPa) 2.7026074 1.253006 123.0665
## 4 Hot Humid (1.1 kPa) - Cool Dry (2.5 kPa) -3.9300861 1.288296 123.2298
## 5 Hot Humid (1.1 kPa) - Hot Dry (3.8 kPa) -1.8576556 1.283706 123.1026
## 6 Cool Dry (2.5 kPa) - Hot Dry (3.8 kPa) 2.0724305 1.262089 123.0607
##      t.ratio      p.value      response
## 1 3.5841894 0.002703439 Hematocrit (%)
## 2 0.5023206 0.958383983 Hematocrit (%)
```

```
## 3  2.1568990 0.141276348 Hematocrit (%)
## 4 -3.0506087 0.014665591 Hematocrit (%)
## 5 -1.4471033 0.472669199 Hematocrit (%)
## 6  1.6420637 0.359065321 Hematocrit (%)
```

Osmol

Ind + Means

```
ggplot() +
  geom_line(data = dat[complete.cases(dat$osmolality_mmol_kg_mean),],
    aes(x = day_n,
        y = osmolality_mmol_kg_mean,
        color = tmt,
        group = individual_ID),
    alpha = 0.2) +
  geom_line(data = means[complete.cases(means$mean_osml),],
    aes(x = day_n,
        y = mean_osml,
        color = tmt,
        group = tmt),
    alpha = 1,
    size = 1) +
  geom_point(data = means,
    aes(x = day_n,
        y = mean_osml,
        color = tmt,
        shape = tmt),
    alpha = 1,
    size = 5) +
  geom_vline(xintercept = 9,
    linetype = "dashed",
    color = "darkgrey") +
  scale_shape_manual(values = c(15:18), name = "") +
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
  ylim(300,450) +
  scale_color_brewer(palette = "Set2", name = "") +
  xlab("") +
  ylab("Plasma Osmolality (mmol/kg)") +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme_classic() +
  theme(text = element_text(color = "black",
    family = "sans",
    size = 22),
    axis.text = element_text(color = "black",
    family = "sans",
    size = 16),
    legend.text = element_text(color = "black",
    family = "sans",
    size = 22),
    legend.text.align = 0,
    legend.position = "none",
    plot.margin = unit(c(0.6, #top
    0.1, #right
```

```

0.1, #bottom
0.1 #left
), "cm")

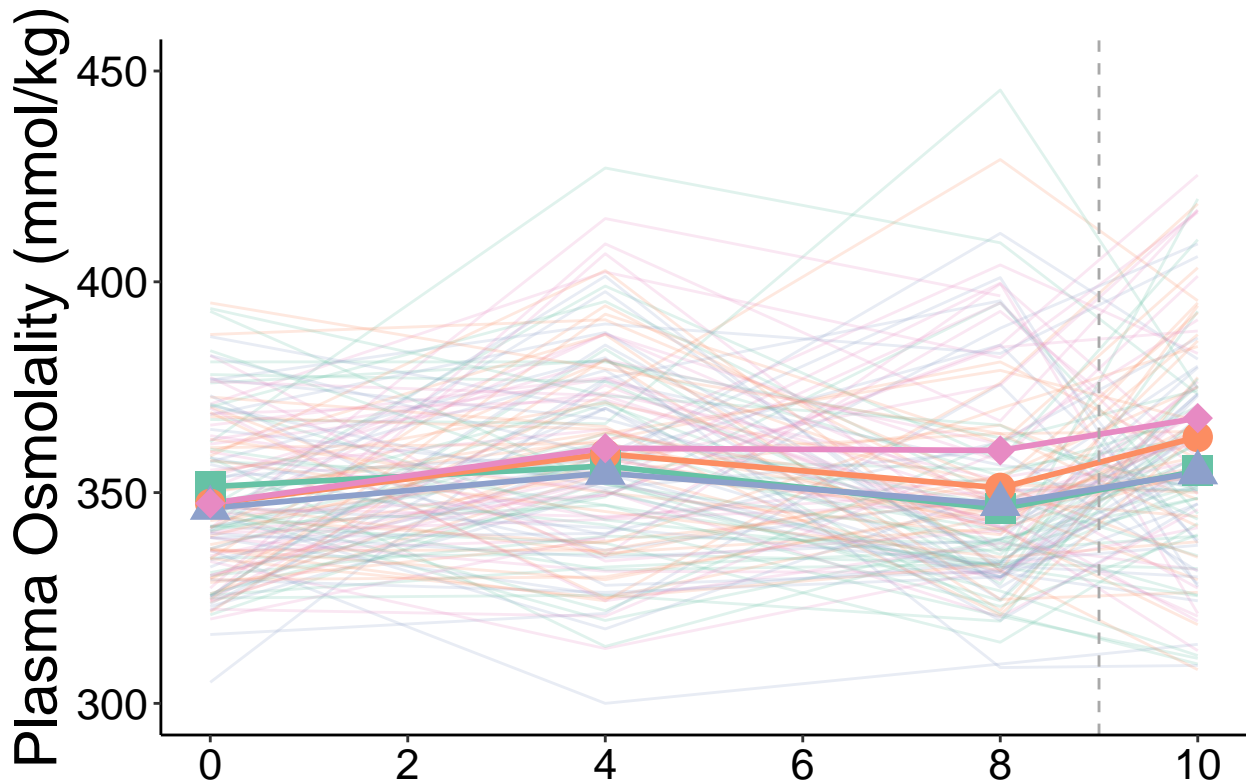
) -> osml_fig
osml_fig

```

```

## Warning: Removed 1 row containing missing values (`geom_line()`).
## Warning: Removed 12 rows containing missing values (`geom_point()`).

```



Means Only *MS*

```

ggplot() +
  geom_smooth(data = dat_no_rehab[complete.cases(dat_no_rehab$osmolality_mmol_kg_mean),],
    aes(x = day_n,
        y = osmolality_mmol_kg_mean,
        color = tmt,
        group = tmt),
    method = "lm",
    se = F,
    size = 0.7) +
  geom_errorbar(data = means,
    aes(x = day_n,
        y = mean_osml,
        color = tmt,
        group = tmt,
        ymin = mean_osml-se_osml,
        ymax = mean_osml+se_osml),
    width = .1,

```

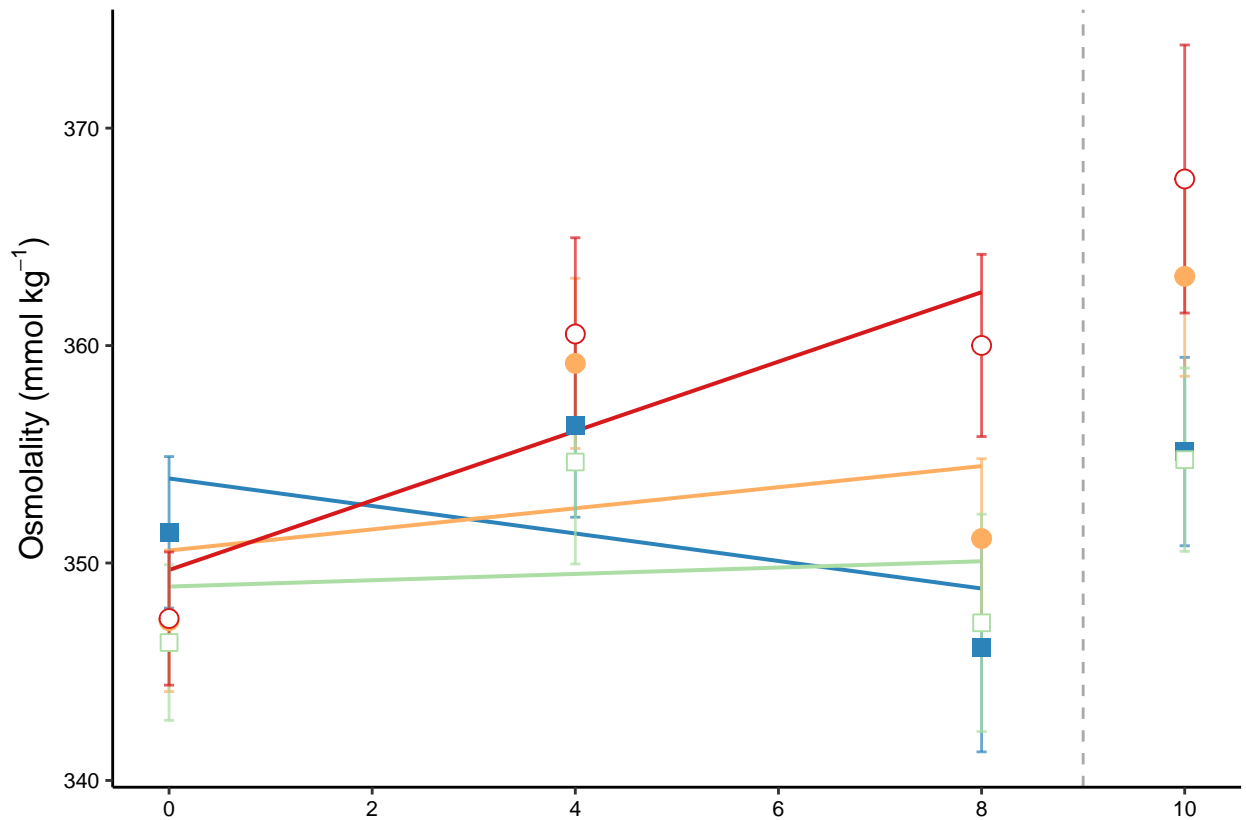


```

        #position=position_dodge(.1),
        alpha = 0.7) +
#geom_line(data = means[complete.cases(means$mean_osml),],
#          aes(x = day_n,
#              y = mean_osml,
#              color = tmt,
#              #linetype = tmt,
#              group = tmt),
#          alpha = 1,
#          size = 0.5) +
geom_point(data = means,
           aes(x = day_n,
               y = mean_osml,
               color = tmt,
               #fill = tmt,
               shape = tmt),
           fill = "white",
           alpha = 1,
           size = 3) +
theme_classic() +
scale_shape_manual(values = my_shapes, name = "",
                   labels = my_labels) +
scale_fill_manual(values = my_colors, name = "",
                  labels = my_labels) +
scale_color_manual(values = my_colors, name = "",
                   labels = my_labels) +
scale_x_continuous(breaks = c(0, 2, 4, 6, 8, 10)) +
scale_y_continuous(breaks = c(seq(320,400, by = 10)),
                   labels = c(seq(320,400, by = 10))) +
xlab("") +
ylab(bquote('Osmolality (mmol '*kg-1')) +
guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
geom_vline(xintercept = 9,
           linetype = "dashed",
           color = "darkgrey") +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 12),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 8),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
      legend.text.align = 0,
      legend.position = "none",
      plot.margin = margin(t = 6, r = 6, b = 0, l = 1, unit = "pt")
) -> osml_fig_min
osml_fig_min

## `geom_smooth()` using formula = 'y ~ x'
## Warning: Removed 12 rows containing missing values (`geom_point()`).

```



Stats! Check Pairwise Diffs ~ Time

Since Plasma osmolality has such a nonlinear trend, we need to test whether the elevated values in the middle of the experiment are significantly different than the values taken before and/or after.

```
# first make sub-dfs for each tmt group
HH <- dat_no_rehab %>%
  dplyr::filter(substr(tmt, 1, 6) == "Hot Hu")
HD <- dat_no_rehab %>%
  dplyr::filter(substr(tmt, 1, 6) == "Hot Dr")
CH <- dat_no_rehab %>%
  dplyr::filter(substr(tmt, 1, 6) == "Cool H")
CD <- dat_no_rehab %>%
  dplyr::filter(substr(tmt, 1, 6) == "Cool D")

# next do pairwise tests for osml on the diff exp days, for each tmt group
pair_HH <- TukeyHSD(aov(data = HH, osmolality_mmol_kg_mean ~ day_factor)) #nonsig
pair_HD <- TukeyHSD(aov(data = HD, osmolality_mmol_kg_mean ~ day_factor)) #nonsig
pair_CH <- TukeyHSD(aov(data = CH, osmolality_mmol_kg_mean ~ day_factor)) #nonsig
pair_CD <- TukeyHSD(aov(data = CD, osmolality_mmol_kg_mean ~ day_factor)) #nonsig

# put into a df and export
osml_pairwise_df <- as.data.frame(pair_HD[[1]]) %>%
  rbind(as.data.frame(pair_HH[[1]])) %>%
  rbind(as.data.frame(pair_CD[[1]])) %>%
  rbind(as.data.frame(pair_CH[[1]])) %>%
  mutate(day_diff = paste("day", substr(rownames(.), 1, 3)),
         tmt = c(rep("Hot Dry", 3),
```

```

      rep("Hot Humid",3),
      rep("Cool Dry",3),
      rep("Cool Humid",3)),
    CI_95 = paste(round(lwr, digits = 2), round(upr, digits = 2), sep = ", "),
    diff = round(diff, digits = 2)) %>%
  dplyr::select(tmt, day_diff, diff, CI_95, p_adj = "p adj")
write.csv(osml_pairwise_df, "./results_statistics/osmolality_pairwise_diffs.csv")

```

nope, none of the differences between days within tmt groups are significantly different

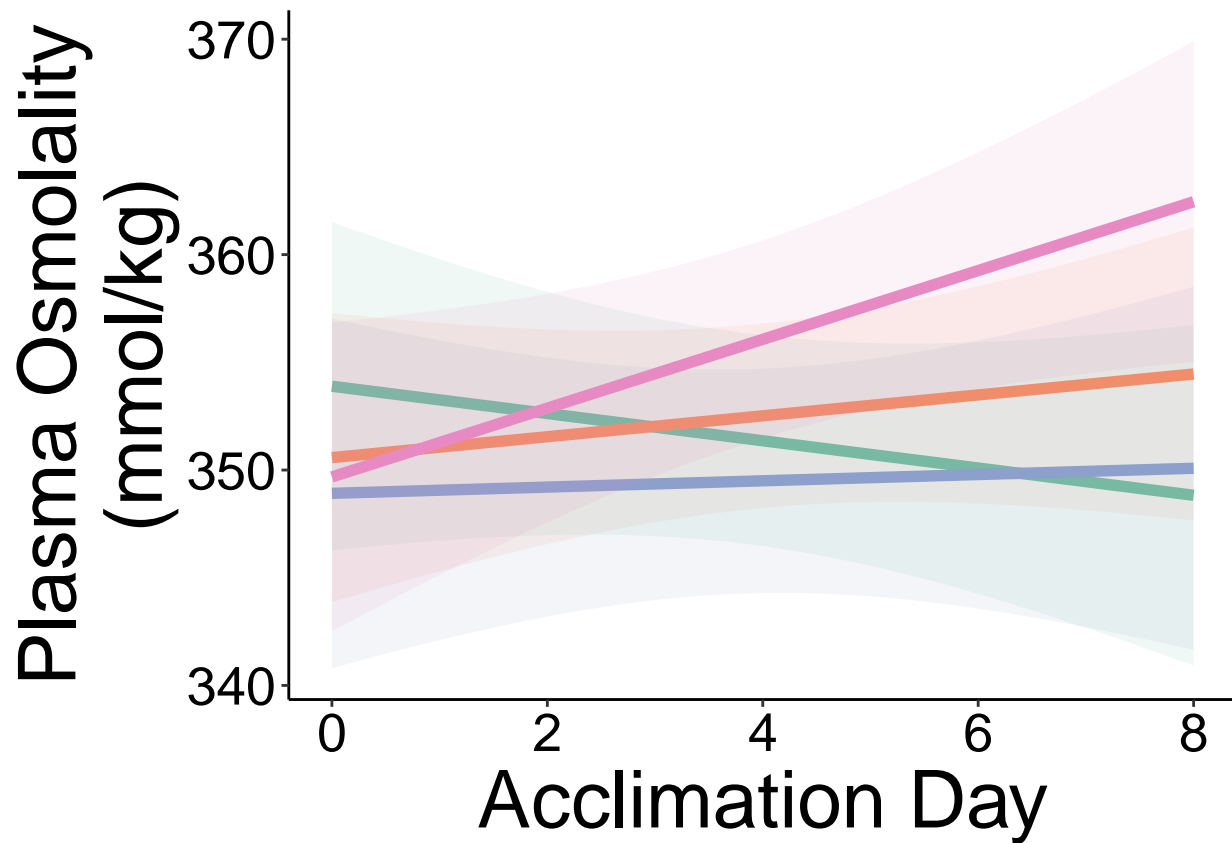
LM + SE

```

ggplot() +
  stat_smooth(data = dat_no_rehab,
    aes(x = day_n,
      y = osmolality_mmol_kg_mean,
      color = tmt,
      fill = tmt,
      group = tmt),
    formula = y ~ x,
    method = "lm",
    se = T,
    size = 2,
    alpha = 0.1) +
  theme_classic() +
  #scale_shape_manual(values = c(15:18), name = "") +
  scale_x_continuous(limits = c(0,8),
    breaks = c(0, 2, 4, 6, 8)) +
  scale_color_brewer(palette = "Set2", name = "") +
  scale_fill_brewer(palette = "Set2", name = "") +
  xlab("Acclimation Day") +
  #ylim(340,370) + #CANNOT put ylims in bc gets rid of pts that are used to make lines
  ylab("Plasma Osmolality\n(mmol/kg)") +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
    family = "sans",
    size = 30),
    axis.text = element_text(color = "black",
      family = "sans",
      size = 20),
    legend.text = element_text(color = "black",
      family = "sans",
      size = 22),
    legend.text.align = 0,
    legend.position = "none"
  ) -> osml_lm_fig
osml_lm_fig

```

```
## Warning: Removed 413 rows containing non-finite values (`stat_smooth()`).
```



Ending Values *MS*

```
ggplot() +
  geom_jitter(data = end_vals,
    aes(x = tmt,
      y = osmolality_mmol_kg_mean,
      color = tmt,
      fill = tmt,
      shape = tmt),
    size = 1,
    alpha = 0.4,
    position = position_jitter(height = 0, width = 0.2)) +
  geom_errorbar(data = osml_emmeans,
    aes(x = tmt,
      y = emmean,
      color = tmt,
      group = tmt,
      ymin = lower.CL,
      ymax = upper.CL),
    width = .1,
    alpha = 0.9) +
  geom_point(data = osml_emmeans,
    aes(x = tmt,
      y = emmean,
      #color = tmt,
      shape = tmt,
```

```

        fill = tmt),
        color = "black",
        size = 4) +
theme_classic() +
scale_shape_manual(values = my_shapes_box, name = "") +
scale_fill_manual(values = my_colors, name = "") +
scale_color_manual(values = my_colors, name = "") +
scale_y_continuous(limits = c(290,470),
                    breaks = c(seq(300,450, by = 50)),
                    labels = c(seq(300,450, by = 50))) +
scale_x_discrete(labels = c("Cool Humid\n0.6 kPa",
                            "Hot Humid\n1.1 kPa",
                            "Cool Dry\n2.5 kPa",
                            "Hot Dry\n3.8 kPa")) +

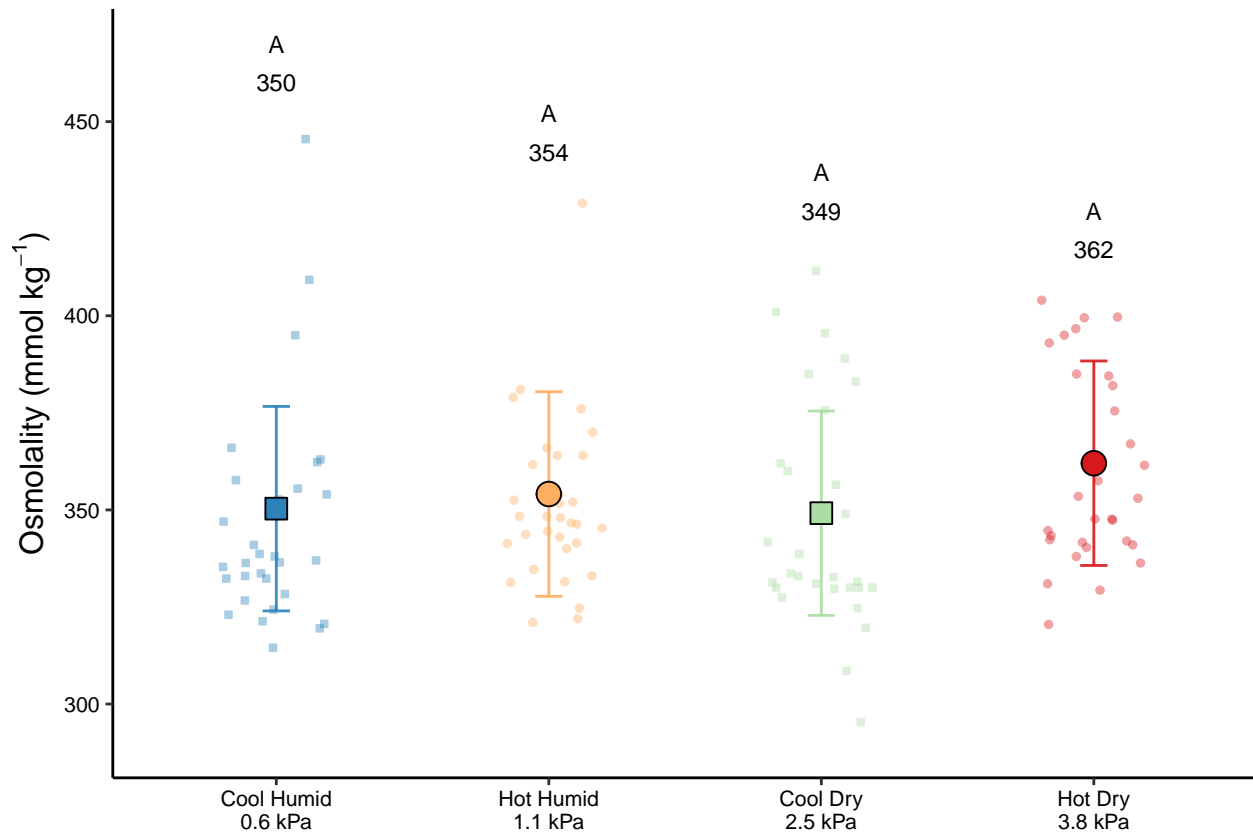
xlab("") +

annotate(geom = "text", x = 4, y = 427, label = "A",
         size = 3) +
annotate(geom = "text", x = 4, y = 417, label = "362", #HD
         size = 3) +
annotate(geom = "text", x = 2, y = 452, label = "A",
         size = 3) +
annotate(geom = "text", x = 2, y = 442, label = "354", #HH
         size = 3) +
annotate(geom = "text", x = 3, y = 437, label = "A",
         size = 3) +
annotate(geom = "text", x = 3, y = 427, label = "349", #CD
         size = 3) +
annotate(geom = "text", x = 1, y = 470, label = "A",
         size = 3) +
annotate(geom = "text", x = 1, y = 460, label = "350", #CH
         size = 3) +

ylab(bquote('Osmolality (mmol '*kg-1')) +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 12),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 8),
      #axis.text.x = element_blank(),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
      legend.text.align = 0,
      legend.position = "none",
      plot.margin = unit(c(0, #top
                          0, #right
                          0, #bottom
                          0), "mm"))
) -> osml_end_boxplot
osml_end_boxplot

```

```
## Warning: Removed 10 rows containing missing values (`geom_point()`).
```



osml_emmeans

```
##               tmt      emmean      SE      df lower.CL upper.CL
## 1 Cool Humid (0.6 kPa) 350.3202 10.16423 4.874923 323.9893 376.6511
## 2 Hot Humid (1.1 kPa) 354.0963 10.13690 4.824373 327.7506 380.4420
## 3 Cool Dry (2.5 kPa) 349.1611 10.17830 4.903510 322.8413 375.4808
## 4 Hot Dry (3.8 kPa) 362.0244 10.15727 4.863220 335.6918 388.3569
##               response
## 1 Plasma Osmolality (mmol/kg)
## 2 Plasma Osmolality (mmol/kg)
## 3 Plasma Osmolality (mmol/kg)
## 4 Plasma Osmolality (mmol/kg)
```

osml_pairwise

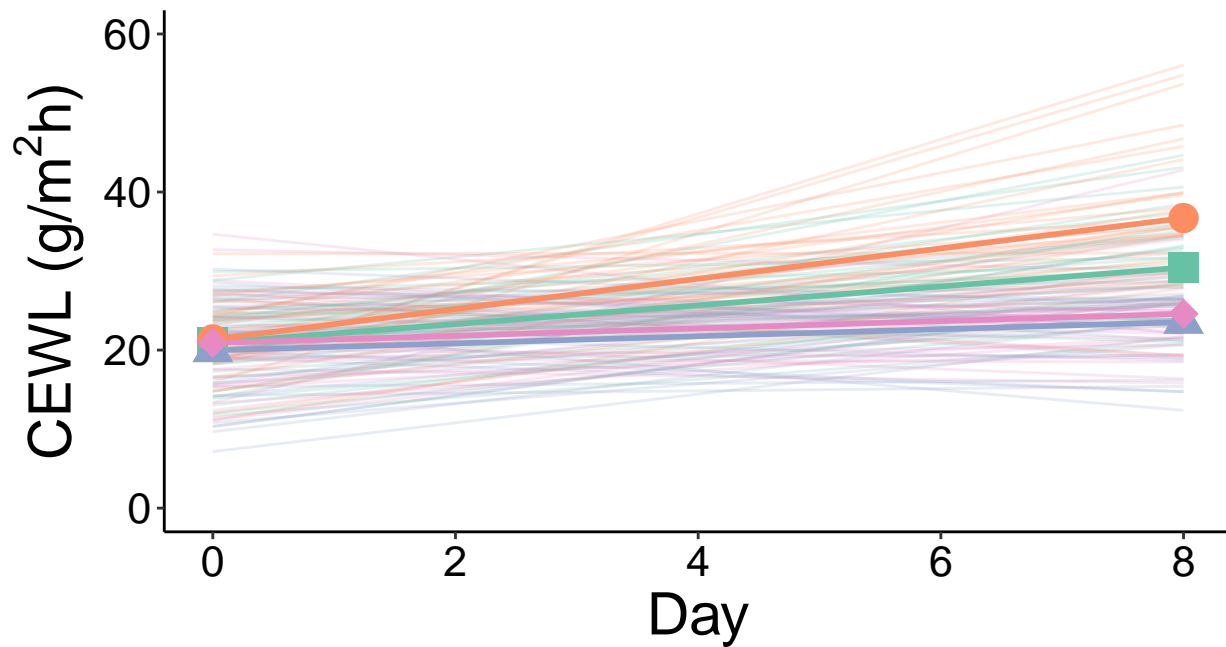
```
##               contrast      estimate      SE      df
## 1 Cool Humid (0.6 kPa) - Hot Humid (1.1 kPa) -3.776107 5.031118 116.0115
## 2 Cool Humid (0.6 kPa) - Cool Dry (2.5 kPa) 1.159115 5.128594 116.0589
## 3 Cool Humid (0.6 kPa) - Hot Dry (3.8 kPa) -11.704175 5.078569 116.0347
## 4 Hot Humid (1.1 kPa) - Cool Dry (2.5 kPa) 4.935222 5.085161 116.0412
## 5 Hot Humid (1.1 kPa) - Hot Dry (3.8 kPa) -7.928068 5.035558 116.0197
## 6 Cool Dry (2.5 kPa) - Hot Dry (3.8 kPa) -12.863291 5.114607 116.0115
##      t.ratio    p.value      response
## 1 -0.7505502 0.87626200 Plasma Osmolality (mmol/kg)
## 2 0.2260104 0.99590771 Plasma Osmolality (mmol/kg)
## 3 -2.3046207 0.10288296 Plasma Osmolality (mmol/kg)
## 4 0.9705144 0.76645671 Plasma Osmolality (mmol/kg)
## 5 -1.5744172 0.39720576 Plasma Osmolality (mmol/kg)
```

```
## 6 -2.5150105 0.06282759 Plasma Osmolality (mmol/kg)
```

CEWL

Ind + Means

```
ggplot() +  
  geom_line(data = dat[complete.cases(dat$CEWL_g_m2h_mean),],  
    aes(x = day_n,  
      y = CEWL_g_m2h_mean,  
      color = tmt,  
      group = individual_ID),  
    alpha = 0.2) +  
  geom_line(data = means[complete.cases(means$mean_CEWL),],  
    aes(x = day_n,  
      y = mean_CEWL,  
      color = tmt,  
      group = tmt),  
    alpha = 1,  
    size = 1) +  
  geom_point(data = means[complete.cases(means$mean_CEWL),],  
    aes(x = day_n,  
      y = mean_CEWL,  
      color = tmt,  
      shape = tmt),  
    alpha = 1,  
    size = 5) +  
  theme_classic() +  
  scale_shape_manual(values = c(15:18), name = "") +  
  scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +  
  scale_color_brewer(palette = "Set2", name = "") +  
  xlab("Day") +  
  ylim(0,60) +  
  ylab(bquote('CEWL (g/'*m^2*'h)')) +  
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +  
  theme(text = element_text(color = "black",  
    family = "sans",  
    size = 22),  
    axis.text = element_text(color = "black",  
      family = "sans",  
      size = 16),  
    legend.text = element_text(color = "black",  
      family = "sans",  
      size = 22),  
    legend.text.align = 0,  
    legend.position = "bottom"  
    #legend.position = c(0.25,0.85)  
  ) -> CEWL_fig  
CEWL_fig
```



■ Cool Humid (0.6 kPa) ● Hot Humid (1.1 kPa)
▲ Cool Dry (2.5 kPa) ◆ Hot Dry (3.8 kPa)

```
#ggsave(filename = "CEWL_fig1.jpeg",
#        plot = CEWL_fig,
#        path = "./results_figures",
#        device = "jpeg",
#        dpi = 1200,
#        width = 6, height = 6)
```

Means Only *MS*

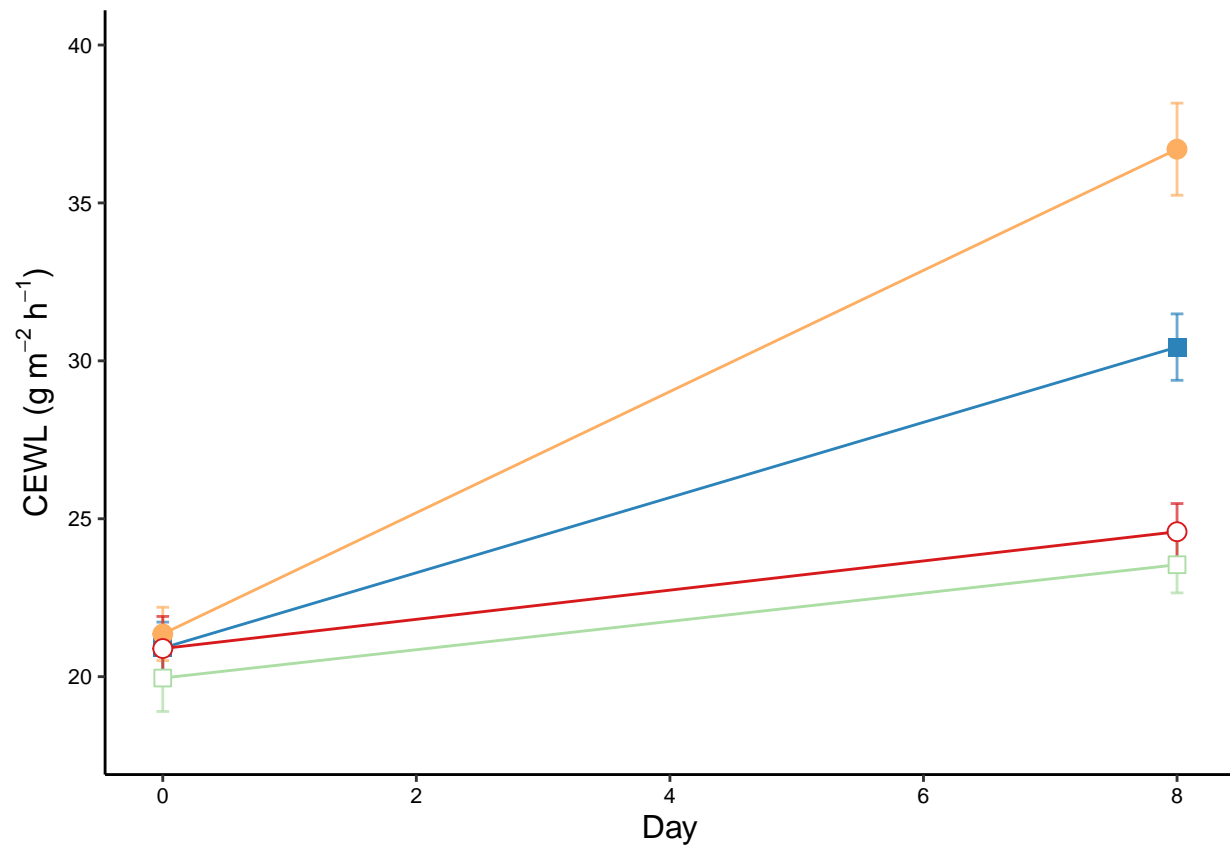
```
ggplot() +
  geom_errorbar(data = means[complete.cases(means$mean_CEWL),],
               aes(x = day_n,
                   y = mean_CEWL,
                   color = tmt,
                   group = tmt,
                   ymin = mean_CEWL-se_CEWL,
                   ymax = mean_CEWL+se_CEWL),
               width = .1,
               #position=position_dodge(.1),
               alpha = 0.7) +
  geom_line(data = means[complete.cases(means$mean_CEWL),],
            aes(x = day_n,
                y = mean_CEWL,
                color = tmt,
                #linetype = tmt,
                group = tmt),
            alpha = 1,
```



```

      size = .5) +
geom_point(data = means[complete.cases(means$mean_CEWL),],
      aes(x = day_n,
          y = mean_CEWL,
          color = tmt,
          shape = tmt),
      fill = "white",
      alpha = 1,
      size = 3) +
theme_classic() +
scale_shape_manual(values = my_shapes, name = "",
      labels = my_labels) +
scale_fill_manual(values = my_colors, name = "",
      labels = my_labels) +
scale_color_manual(values = my_colors, name = "",
      labels = my_labels) +
scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +
scale_y_continuous(breaks = c(20, 25, 30, 35, 40),
      limits = c(18,40)) +
xlab("Day") +
ylab(bquote('CEWL (g *m^-2* ' *h^-1*')')) +
guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
theme(text = element_text(color = "black",
      family = "sans",
      size = 12),
      axis.text = element_text(color = "black",
      family = "sans",
      size = 8),
      legend.text = element_text(color = "black",
      family = "sans",
      size = 8),
      legend.text.align = 0,
      legend.position = "none"
      #legend.position = c(0.25,0.85)
      ) -> CEWL_fig_min
CEWL_fig_min

```



```
# use ggarrange so legend is centered
CEWL_fig_formatted <- ggarrange(CEWL_fig_min,
                                ncol = 1, nrow = 1,
                                common.legend = TRUE,
                                legend = "bottom")

# save
ggsave(filename = "experiment_CEWL_fig.pdf",
        plot = CEWL_fig_formatted,
        path = "./results_figures",
        device = "pdf",
        dpi = 600,
        units = "mm",
        width = 80, height = 90)
```

LM + SE

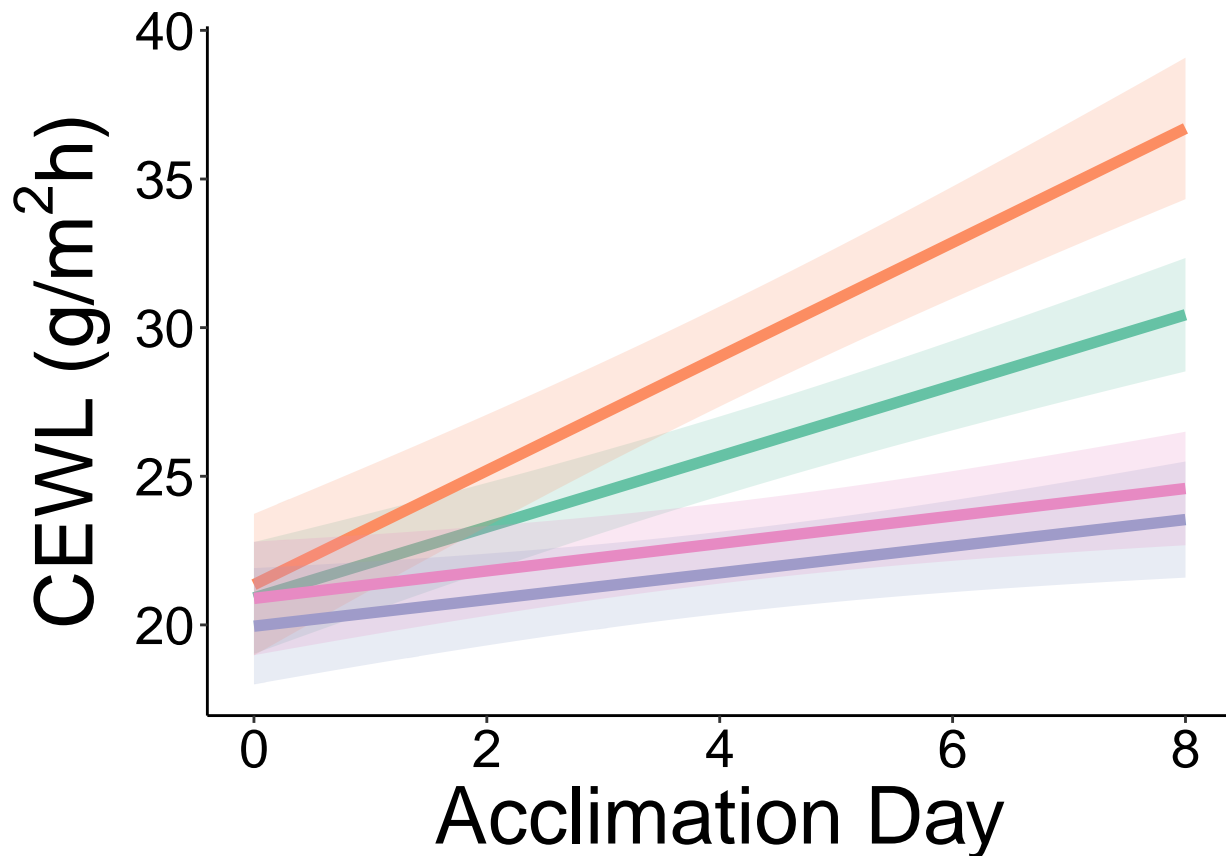
```
ggplot() +
  stat_smooth(data = dat,
             aes(x = day_n,
                 y = CEWL_g_m2h_mean,
                 color = tmt,
                 fill = tmt,
                 group = tmt),
             formula = y ~ x,
             method = "lm",
             se = T,
             size = 2,
```

```

      alpha = 0.2) +
theme_classic() +
scale_x_continuous(breaks = c(0, 2, 4, 6, 8)) +
scale_color_brewer(palette = "Set2", name = "") +
scale_fill_brewer(palette = "Set2", name = "") +
xlab("Acclimation Day") +
ylab(bquote('CEWL (g/*m^2*h)')) +
guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
theme(text = element_text(color = "black",
                           family = "sans",
                           size = 30),
      axis.text = element_text(color = "black",
                                family = "sans",
                                size = 20),
      legend.text = element_text(color = "black",
                                  family = "sans",
                                  size = 22),
      legend.text.align = 0,
      legend.position = "none"
      #legend.position = c(0.25,0.85)
      ) -> CEWL_lm_fig
CEWL_lm_fig

```

```
## Warning: Removed 669 rows containing non-finite values (`stat_smooth()`).
```



delta CEWL ~ VPD

```
ggplot(data = dat_no_rehab_deltaCEWL) +
  geom_point(aes(x = VPD_kPa_tmttrial,
                 y = delta_CEWL,
                 color = tmt,
                 fill = tmt,
                 shape = tmt),
            size = 2,
            alpha = 0.4
  ) +
  geom_smooth(aes(x = VPD_kPa_tmttrial,
                  y = delta_CEWL),
              se = F,
              formula = y ~ x,
              method = "lm",
              color = "black") +
  #geom_smooth(aes(x = VPD_kPa_tmttrial,
  #                y = delta_CEWL),
  #            se = F,
  #            formula = y ~ poly(x, 2),
  #            method = "lm",
  #            color = "red") +
  theme_classic() +
  scale_shape_manual(values = my_shapes_box, name = "",
                    labels = c("Cool Humid (CH)",
                               "Hot Humid (HH)",
                               "Cool Dry (CD)",
                               "Hot Dry (HD)")) +
  scale_fill_manual(values = my_colors, name = "",
                   labels = c("Cool Humid (CH)",
                              "Hot Humid (HH)",
                              "Cool Dry (CD)",
                              "Hot Dry (HD)")) +
  scale_color_manual(values = my_colors, name = "",
                    labels = c("Cool Humid (CH)",
                               "Hot Humid (HH)",
                               "Cool Dry (CD)",
                               "Hot Dry (HD)")) +
  scale_y_continuous(limits = c(-13,40),
                    breaks = c(seq(-10,40, by = 10)),
                    labels = c(seq(-10,40, by = 10))
  ) +
  scale_x_continuous(limits = c(0,4),
                    breaks = c(0.6, 1.1, 2.5, 3.8),
                    labels = c("0.6\nCH",
                               "1.1\nHH",
                               "2.5\nCD",
                               "3.8\nHD")) +
  xlab("Vapor Pressure Deficit (kPa)") +
  ylab(expression(Delta ~ 'CEWL')) +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
                             family = "sans",
```

```

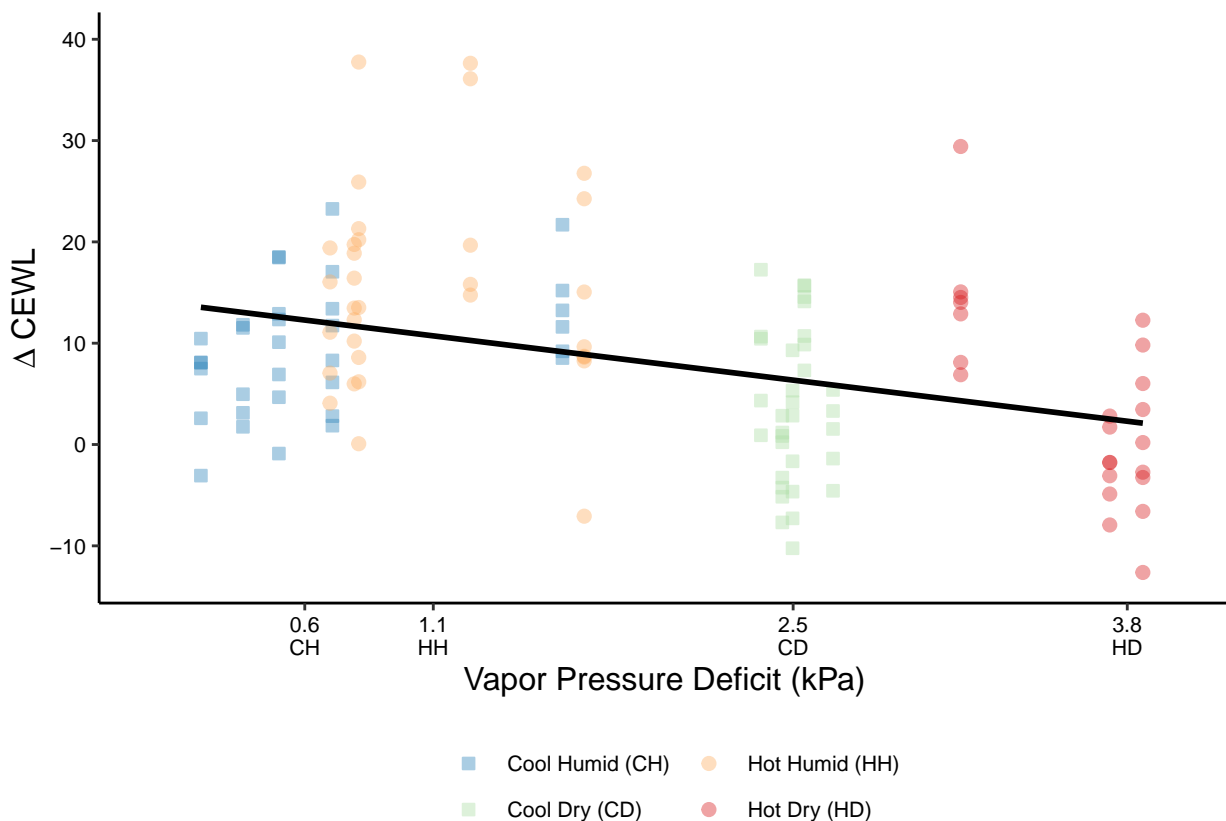
        size = 12),
axis.text = element_text(color = "black",
                          family = "sans",
                          size = 8),
#axis.text.x = element_blank(),
legend.text = element_text(color = "black",
                          family = "sans",
                          size = 8),

legend.text.align = 0,
legend.position = "bottom",
legend.spacing.y = unit(0, "mm")
#plot.margin = unit(c(0, #top
#                    0, #right
#                    0, #bottom
#                    0), "mm")
) -> CEWL_VPD_fig
CEWL_VPD_fig

```

```
## Warning: Removed 11 rows containing non-finite values (`stat_smooth()`).
```

```
## Warning: Removed 11 rows containing missing values (`geom_point()`).
```



```

# use ggarrange so legend is centered
CEWL_VPD_fig_formatted <- ggarrange(CEWL_VPD_fig,
                                   ncol = 1, nrow = 1,
                                   common.legend = TRUE,
                                   legend = "bottom")

```

```
## Warning: Removed 11 rows containing non-finite values (`stat_smooth()`).
```

```
## Removed 11 rows containing missing values (`geom_point()`).
## Warning: Removed 11 rows containing non-finite values (`stat_smooth()`).
## Warning: Removed 11 rows containing missing values (`geom_point()`).

# save
ggsave(filename = "exp_CEWL_delta_VPD.pdf",
        plot = CEWL_VPD_fig_formatted,
        path = "./results_figures",
        device = "pdf",
        dpi = 600,
        units = "mm",
        width = 80, height = 90)
```

Ending Values *MS*

```
ggplot() +
  geom_jitter(data = end_vals,
             aes(x = tmt,
                 y = CEWL_g_m2h_mean,
                 color = tmt,
                 fill = tmt,
                 shape = tmt),
             size = 1,
             alpha = 0.4,
             position = position_jitter(height = 0, width = 0.2)) +
  geom_errorbar(data = CEWL_emmeans,
               aes(x = tmt,
                   y = emmean,
                   color = tmt,
                   group = tmt,
                   ymin = lower.CL,
                   ymax = upper.CL),
               width = .1,
               alpha = 0.9) +
  geom_point(data = CEWL_emmeans,
            aes(x = tmt,
                y = emmean,
                #color = tmt,
                shape = tmt,
                fill = tmt),
            color = "black",
            size = 4) +
  theme_classic() +
  scale_shape_manual(values = my_shapes_box, name = "") +
  scale_fill_manual(values = my_colors, name = "") +
  scale_color_manual(values = my_colors, name = "") +
  scale_x_discrete(labels = c("Cool Humid\n0.6 kPa",
                              "Hot Humid\n1.1 kPa",
                              "Cool Dry\n2.5 kPa",
                              "Hot Dry\n3.8 kPa")) +
  scale_y_continuous(limits = c(9,63),
                     breaks = c(seq(10,60, by = 10)),
                     labels = c(seq(10,60, by = 10))) +
```

```

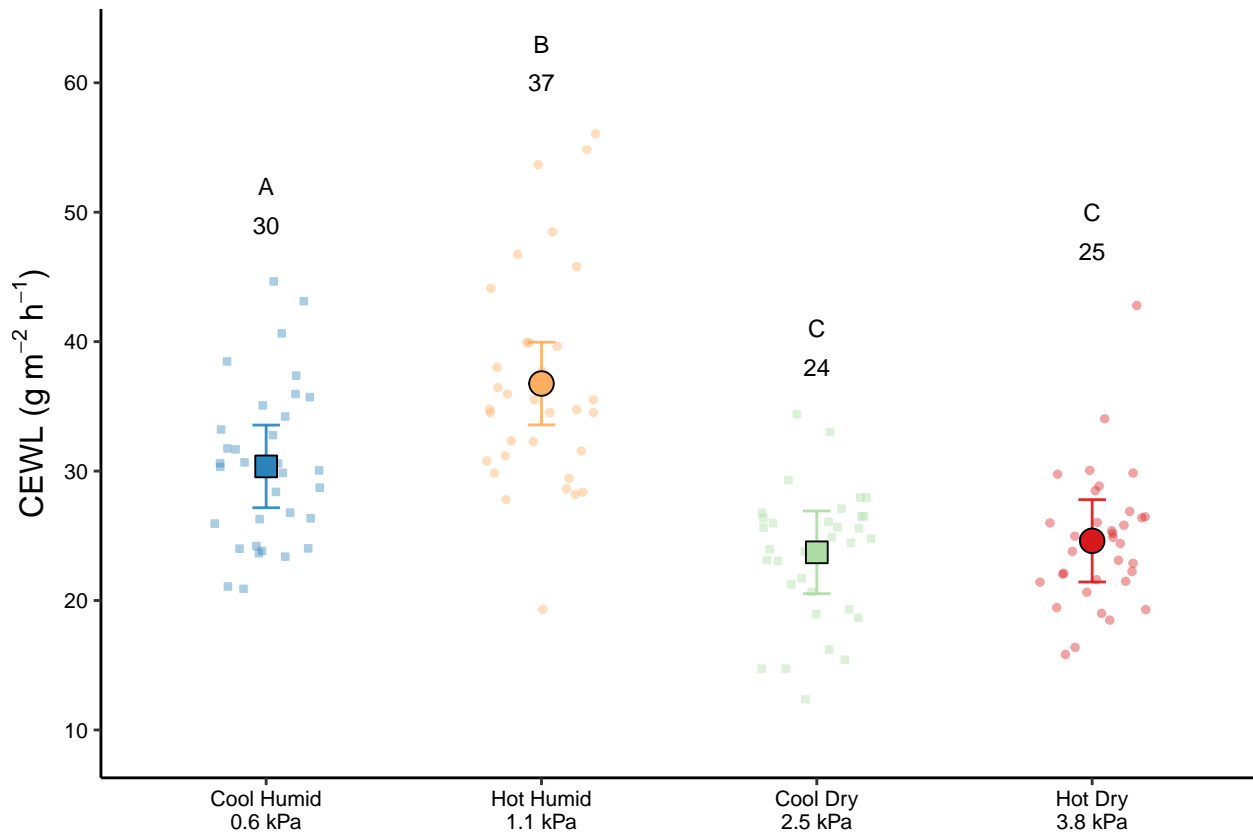
xlab("") +
ylab(bquote('CEWL (g '*m^-2*' '*h^-1*')')) +

annotate(geom = "text", x = 4, y = 50, label = "C",
  size = 3) +
annotate(geom = "text", x = 4, y = 47, label = "25", #HD
  size = 3) +
annotate(geom = "text", x = 2, y = 63, label = "B",
  size = 3) +
annotate(geom = "text", x = 2, y = 60, label = "37", #HH
  size = 3) +
annotate(geom = "text", x = 3, y = 41, label = "C",
  size = 3) +
annotate(geom = "text", x = 3, y = 38, label = "24", #CD
  size = 3) +
annotate(geom = "text", x = 1, y = 52, label = "A",
  size = 3) +
annotate(geom = "text", x = 1, y = 49, label = "30", #CH
  size = 3) +

theme(text = element_text(color = "black",
  family = "sans",
  size = 12),
  axis.text = element_text(color = "black",
    family = "sans",
    size = 8),
  #axis.text.x = element_blank(),
  legend.text = element_text(color = "black",
    family = "sans",
    size = 8),
  legend.text.align = 0,
  legend.position = "none",
  plot.margin = unit(c(0, #top
    0, #right
    0, #bottom
    0), "mm")
) -> CEWL_end_boxplot
CEWL_end_boxplot

```

```
## Warning: Removed 1 rows containing missing values (`geom_point()`).
```



CEWL_emmeans

```
##
##          tmt      emmean      SE      df lower.CL upper.CL
## 1 Cool Humid (0.6 kPa) 30.36157 1.445712 10.68500 27.16810 33.55505
## 2 Hot Humid (1.1 kPa) 36.76428 1.446643 10.69577 33.56916 39.95941
## 3 Cool Dry (2.5 kPa) 23.72439 1.446548 10.69551 20.52946 26.91931
## 4 Hot Dry (3.8 kPa) 24.61233 1.434910 10.37491 21.43074 27.79392
##
##      response
## 1 CEWL (g/m2h)
## 2 CEWL (g/m2h)
## 3 CEWL (g/m2h)
## 4 CEWL (g/m2h)
```

CEWL_pairwise

```
##
##          contrast      estimate      SE      df
## 1 Cool Humid (0.6 kPa) - Hot Humid (1.1 kPa) -6.4027092 1.477747 125.0660
## 2 Cool Humid (0.6 kPa) - Cool Dry (2.5 kPa) 6.6371867 1.482776 125.4155
## 3 Cool Humid (0.6 kPa) - Hot Dry (3.8 kPa) 5.7492451 1.468760 125.2015
## 4 Hot Humid (1.1 kPa) - Cool Dry (2.5 kPa) 13.0398959 1.482912 125.4316
## 5 Hot Humid (1.1 kPa) - Hot Dry (3.8 kPa) 12.1519543 1.469693 125.2717
## 6 Cool Dry (2.5 kPa) - Hot Dry (3.8 kPa) -0.8879416 1.467098 125.0811
##
##      t.ratio      p.value      response
## 1 -4.3327496 1.740823e-04 CEWL (g/m2h)
## 2 4.4761900 9.849699e-05 CEWL (g/m2h)
## 3 3.9143543 8.436015e-04 CEWL (g/m2h)
## 4 8.7934385 9.026113e-14 CEWL (g/m2h)
## 5 8.2683648 1.057265e-12 CEWL (g/m2h)
```

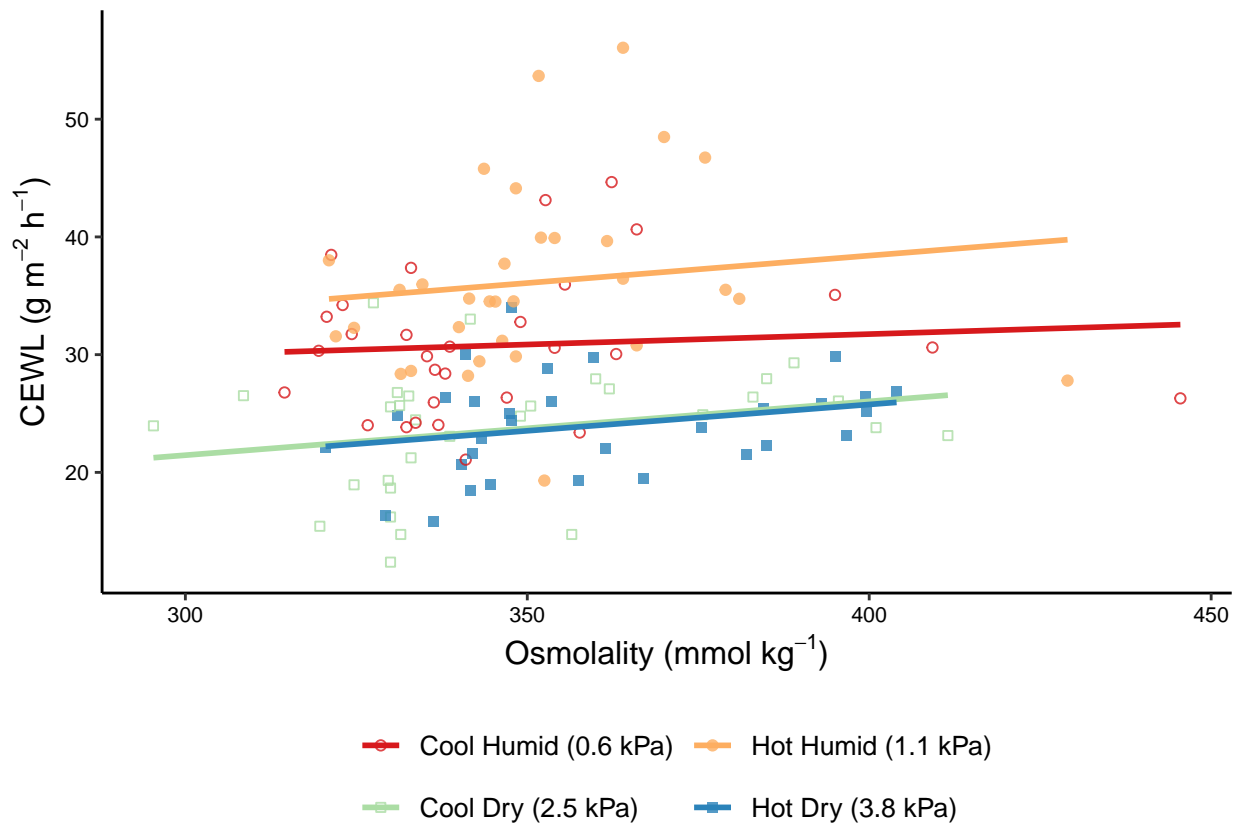


```
## 6 -0.6052366 9.302438e-01 CEWL (g/m2h)
```

Exp CEWL ~ Osmol

```
end_vals_CEWL_osml <- dat %>%
  dplyr::filter(day_n == 8) %>%
  dplyr::filter(complete.cases(CEWL_g_m2h_mean, osmolality_mmol_kg_mean))

ggplot(end_vals_CEWL_osml) +
  aes(x = osmolality_mmol_kg_mean,
      y = CEWL_g_m2h_mean,
      color = tmt,
      shape = tmt) +
  geom_point(size = 1.5,
             alpha = 0.8) +
  stat_smooth(formula = y ~ x,
             method = "lm",
             se = F,
             size = 1,
             alpha = 0.9) +
  theme_classic() +
  xlab(bquote('Osmolality (mmol '*kg-1')) +
  ylab(bquote('CEWL (g '*m-2' '*h-1')) +
  #xlim(300, 400) +
  #ylim(0, 40) +
  scale_shape_manual(values = c(21,19, 22,15), name = "") +
  scale_fill_brewer(palette = "Spectral", name = "") +
  scale_color_brewer(palette = "Spectral", name = "") +
  scale_x_continuous(breaks = c(300, 350, 400, 450)) +
  scale_y_continuous(breaks = c(20, 30, 40, 50),
                    limits = c(12,57)) +
  guides(shape = guide_legend(nrow = 2, byrow = TRUE)) +
  theme(text = element_text(color = "black",
                            family = "sans",
                            size = 12),
        axis.text = element_text(color = "black",
                                  family = "sans",
                                  size = 8),
        legend.position = "bottom"
        #axis.text.y = element_blank(),
        #plot.margin = unit(c(0.1,0,0.1,0.45), "cm")
  ) -> exp_end_CEWL_osml_fig
exp_end_CEWL_osml_fig
```



```
# will need to save using figure arrange to make legend centered if I use that fig
ggsave(filename = "exp_CEWL_osml_fig.pdf",
  plot = exp_end_CEWL_osml_fig,
  path = "./results_figures",
  device = "pdf",
  dpi = 600,
  units = "mm",
  width = 80, height = 90)
```

Multi-Figures

over time:

```
ggarrange(osml_fig_min,
  hct_fig_min,
  SMI_fig_min,
  ncol = 1, nrow = 3,
  labels = c("A", "B", "C"),
  common.legend = TRUE,
  legend = "bottom"
) -> experiment_multi_fig
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

```
## Warning: Removed 12 rows containing missing values (`geom_point()`).
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

```
## Warning: Removed 12 rows containing missing values (`geom_point()`).
```

```
## `geom_smooth()` using formula = 'y ~ x'
## `geom_smooth()` using formula = 'y ~ x'

#experiment_multi_fig
# export figure
ggsave(filename = "experiment_multi_fig.pdf",
        plot = experiment_multi_fig,
        path = "./results_figures",
        device = "pdf",
        dpi = 600,
        units = "mm",
        width = 80, height = 210)
```

end values:

```
ggarrange(CEWL_end_boxplot,
          osml_end_boxplot,
          hct_end_boxplot,
          SMI_end_boxplot,
          ncol = 2, nrow = 2,
          labels = c("A", "B", "C", "D"),
          widths = c(2, 2.045), heights = c(2, 2),
          common.legend = FALSE
          ) -> ending_values_multi_fig
```

```
## Warning: Removed 1 rows containing missing values (`geom_point()`).
## Warning: Removed 10 rows containing missing values (`geom_point()`).
## Warning: Removed 3 rows containing missing values (`geom_point()`).
```

```
#ending_values_multi_fig

ggsave(filename = "exp_end_val_multi_fig.pdf",
        plot = ending_values_multi_fig,
        path = "./results_figures",
        device = "pdf",
        dpi = 600,
        units = "mm",
        width = 180, height = 150)
```