

data wrangling and plots

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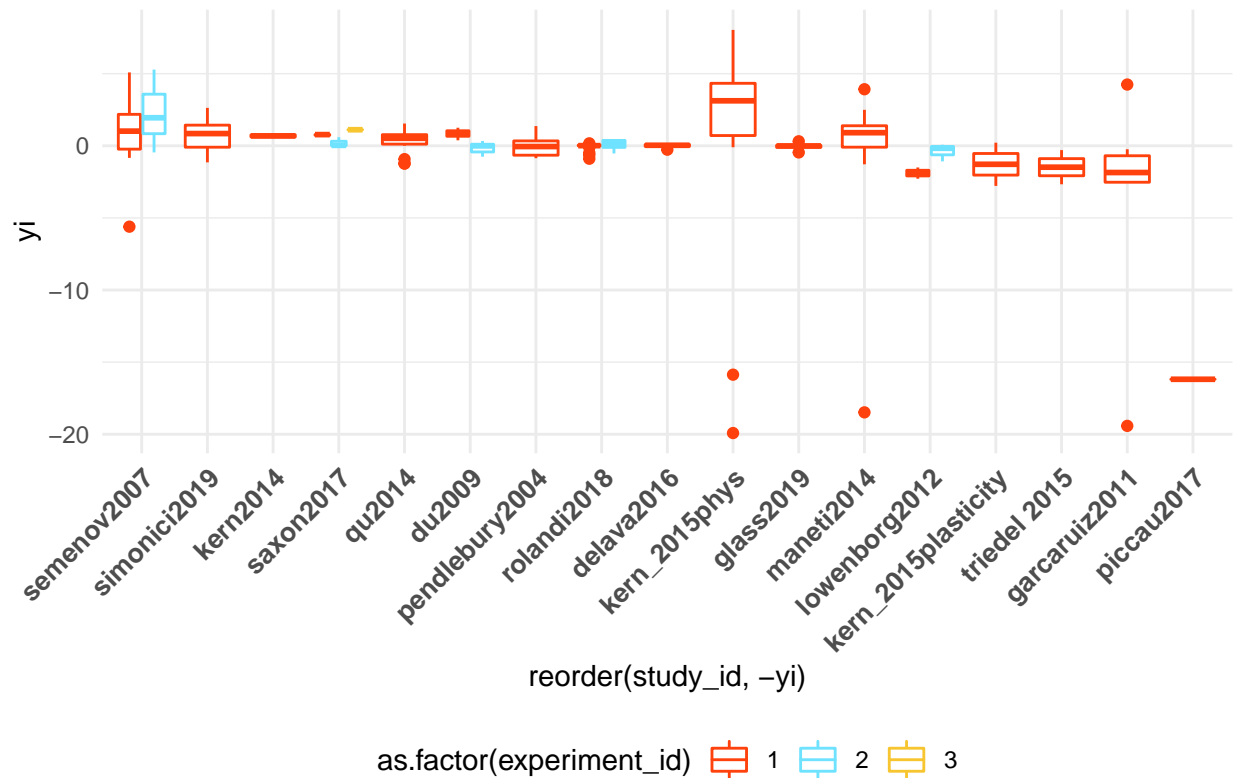
Plots

Question: How does response compare across studies and experiments?

Figure 1.

```
#SMD across all studies
ggplot(normalized, aes(x=reorder(study_id, -yi), y=yi, color = as.factor(experiment_id)))+
  geom_boxplot()+
  scale_color_tron()+
  theme_minimal()+
  theme(axis.text.x = element_text(face = "bold",
                                    size = 10, angle = 45, hjust = 1),
        legend.position = "bottom")+
  ggtitle("SMD across all studies")
```

SMD across all studies



```
#corresponding random effects model
fig1 <- rma.mv(yi, vi, data=dat_MA_ES,
              random = ~1 | experiment_id/ study_id,
              method="REML")
fig1
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0007     3    no      experiment_id
## sigma^2.2 11.4486  3.3836    23    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 202) = 5614.3312, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4951  0.7070  -0.7003  0.4837  -1.8807  0.8905
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#trying out mixed effects model
figlme <- rma(yi, vi, data=dat_MA_ES, mods = ~study_id,
              method="FE")
figlme
```

```
##
## Fixed-Effects with Moderators Model (k = 203)
##
## I2 (residual heterogeneity / unaccounted variability): 95.09%
## H2 (unaccounted variability / sampling variability): 20.37
##
## Test for Residual Heterogeneity:
## QE(df = 186) = 3789.5783, p-val < .0001
##
## Test of Moderators (coefficients 2:17):
## QM(df = 16) = 1824.7529, p-val < .0001
##
## Model Results:
##
##               estimate      se      zval      pval      ci.lb
## intrcpt          0.0006  0.0448   0.0127  0.9899  -0.0872
## study_iddu2009    0.1754  0.0782   2.2433  0.0249   0.0222
## study_idgarcaruiz2011 -1.0855  0.0792 -13.7139 <.0001  -1.2407
## study_idglass2019  -0.0359  0.0636  -0.5638  0.5729  -0.1606
## study_idkern_2015phys  2.0706  0.2485   8.3316 <.0001   1.5835
## study_idkern_2015plasticity -0.7988  0.3360  -2.3774  0.0174  -1.4574
## study_idkern2014     0.6921  0.1156   5.9871 <.0001   0.4655
## study_idlowenborg2012 -0.8458  0.1396  -6.0565 <.0001  -1.1195
## study_idmaneti2014    0.7240  0.1146   6.3198 <.0001   0.4995
## study_idpendlebury2004 -0.0100  0.1239  -0.0807  0.9357  -0.2529
## study_idpiccau2017   -16.1904  0.7515 -21.5444 <.0001 -17.6633
## study_idqu2014       0.3173  0.0858   3.6992  0.0002   0.1492
## study_idrolandi2018   0.0200  0.0467   0.4283  0.6684  -0.0716
## study_idsaxon2017     0.4966  0.0470  10.5578 <.0001   0.4044
## study_idsemenov2007   0.9931  0.2096   4.7384 <.0001   0.5823
## study_idsimonici2019  0.5785  0.0932   6.2079 <.0001   0.3958
## study_idtriedel 2015  -1.1246  0.1483  -7.5818 <.0001  -1.4153
##
##               ci.ub
## intrcpt          0.0883
## study_iddu2009    0.3287 *
## study_idgarcaruiz2011 -0.9304 ***
## study_idglass2019  0.0888
## study_idkern_2015phys  2.5577 ***
## study_idkern_2015plasticity -0.1403 *
## study_idkern2014     0.9187 ***
## study_idlowenborg2012 -0.5721 ***
## study_idmaneti2014    0.9485 ***
## study_idpendlebury2004  0.2329
## study_idpiccau2017   -14.7175 ***
## study_idqu2014       0.4854 ***
## study_idrolandi2018   0.1116
## study_idsaxon2017     0.5888 ***
## study_idsemenov2007   1.4039 ***
```

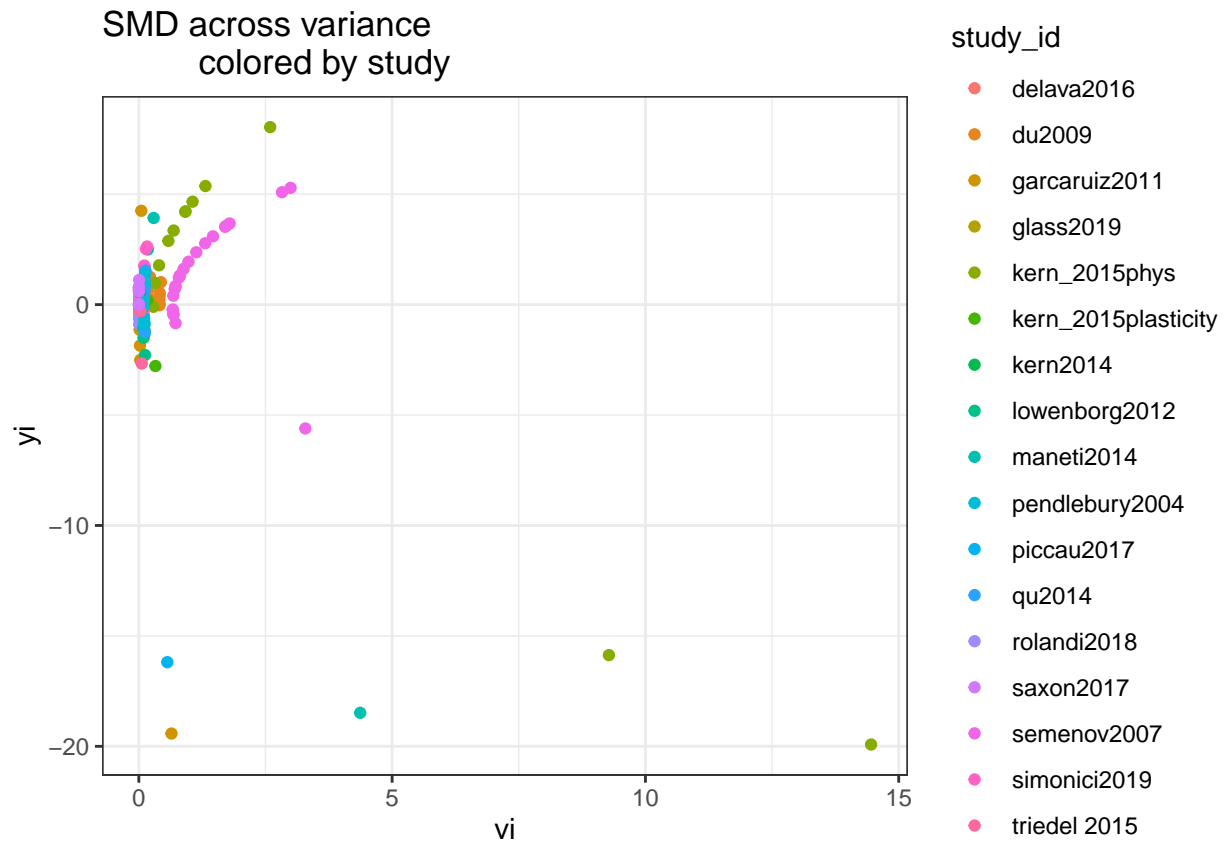
```
## study_idsimonici2019          0.7611 ***
## study_idtriedel 2015         -0.8339 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#basic linear model
```

```
simple1<-lm(yi ~ study_id, data = dat_MA_ES)
summary(fig1)
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
##      logLik      Deviance      AIC      BIC      AICc
## -1624.4609   3248.9218   3254.9218   3264.8466   3255.0430
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0007     3    no      experiment_id
## sigma^2.2  11.4486  3.3836    23    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 202) = 5614.3312, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.4951  0.7070  -0.7003  0.4837  -1.8807  0.8905
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Figure 2.

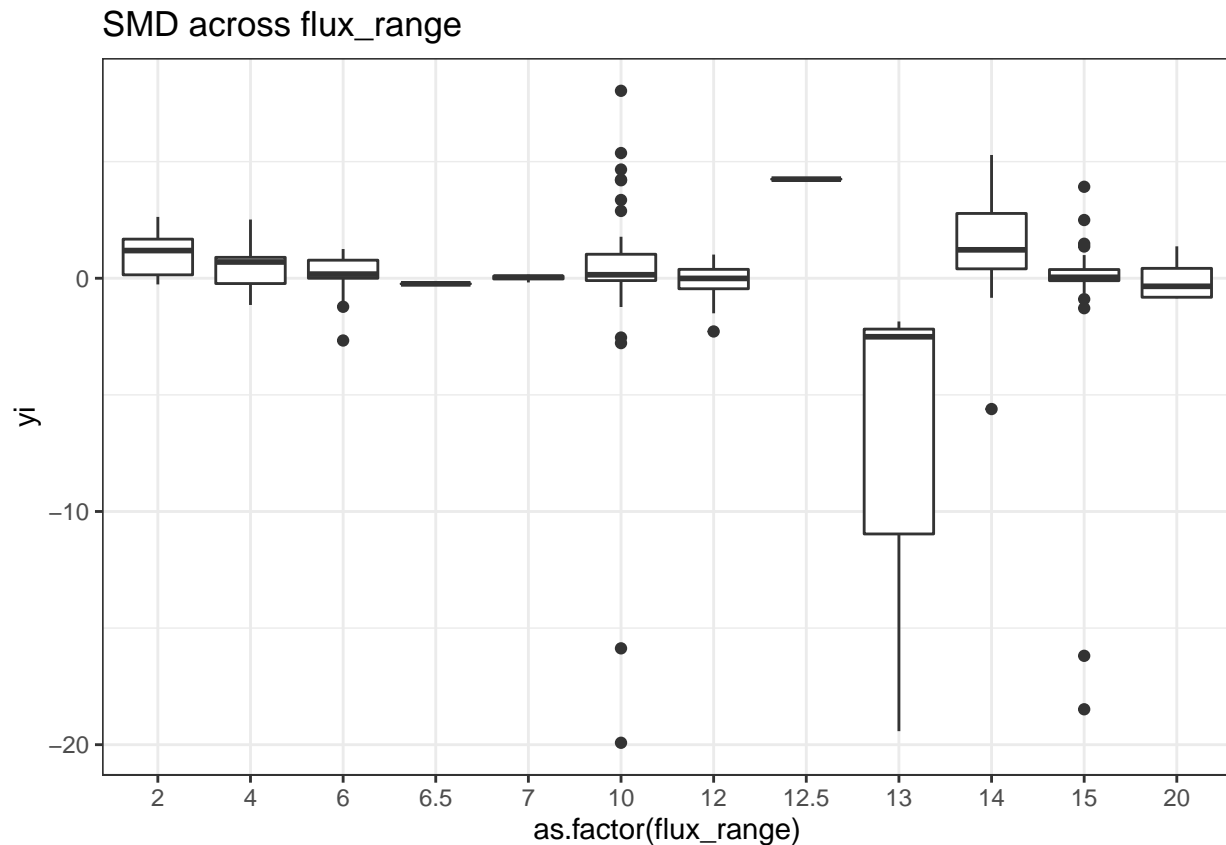


```
##
## Call:
## lm(formula = yi ~ vi, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -45.524  -0.962  -0.529   0.421  21.478
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.9781     0.3499   2.795  0.00569 **
## vi           -2.9315     0.2123 -13.805 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.783 on 201 degrees of freedom
## Multiple R-squared:  0.4867, Adjusted R-squared:  0.4842
## F-statistic: 190.6 on 1 and 201 DF, p-value: < 2.2e-16
```

Question: How does fluctuation amplitude affect response variables?

Figure 3.

```
# boxplots of how fluctuation range influences SMD
ggplot(normalized, aes(x=as.factor(flux_range), y=yi))+
  geom_boxplot()+
  theme_bw()+
  ggtitle("SMD across flux_range")
```



```
#random effects model including org_level as a random variable
fig7 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~flux_range,
  random = ~1 | experiment_id/ study_id,
  method="REML")
fig7
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim    sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0006    3    no      experiment_id
## sigma^2.2 11.1767  3.3432   23    no  experiment_id/study_id
```

```
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 5122.0123, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 18.2850, p-val < .0001
##
## Model Results:
##
##           estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      -0.0057  0.7078  -0.0081  0.9936  -1.3930   1.3816
## flux_range   -0.0466  0.0109  -4.2761  <.0001  -0.0679  -0.0252 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

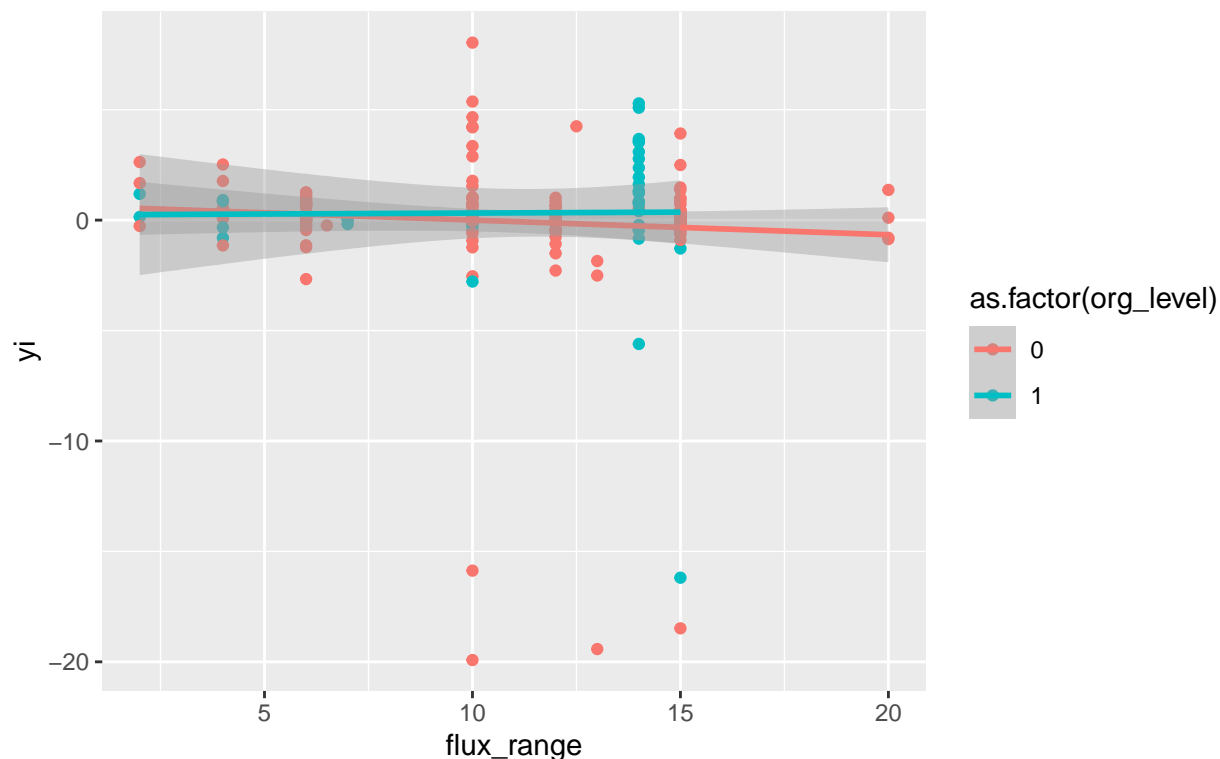
```
#simple linear model looking at how flux_range affects yi
simple7<-lm(yi~flux_range, data =dat_MA_ES)
summary(simple7)
```

```
##
## Call:
## lm(formula = yi ~ flux_range, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.315   0.044   0.662   1.167   8.328
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.7109     1.3490   0.527   0.599
## flux_range   -0.1001     0.1156  -0.866   0.388
##
## Residual standard error: 6.664 on 201 degrees of freedom
## Multiple R-squared:  0.003714,    Adjusted R-squared:  -0.001243
## F-statistic: 0.7493 on 1 and 201 DF,  p-value: 0.3877
```

Figure 4.

```
# scatterplot of standardized mean response vs flux range colored and lm fit by org level
ggplot(normalized, aes(x=flux_range, y=yi, color = as.factor(org_level)))+
  geom_point()+
  geom_smooth(method="lm", formula = y~x)+
  ggtitle("SMD across fluctuation ranges colored by organization level
          and fit with linear model")
```

SMD across fluctuation ranges colored by organization level
and fit with linear model



```
#random effects model including org_level as a random variable
fig8 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~flux_range + org_level,
              random = ~1 | experiment_id/ study_id,
              method="REML")
fig8
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim      sqrt nlvls  fixed      factor
## sigma^2.1  0.0000  0.0007     3    no      experiment_id
## sigma^2.2 10.7755  3.2826    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 200) = 5061.4557, p-val < .0001
##
## Test of Moderators (coefficients 2:3):
## QM(df = 2) = 45.3027, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      0.0960  0.6957   0.1380  0.8903   -1.2675    1.4595
## flux_range  -0.0466  0.0109  -4.2815 <.0001   -0.0680   -0.0253 ***
```



```
## org_level    -0.4312  0.0830  -5.1961  <.0001  -0.5939  -0.2686  ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

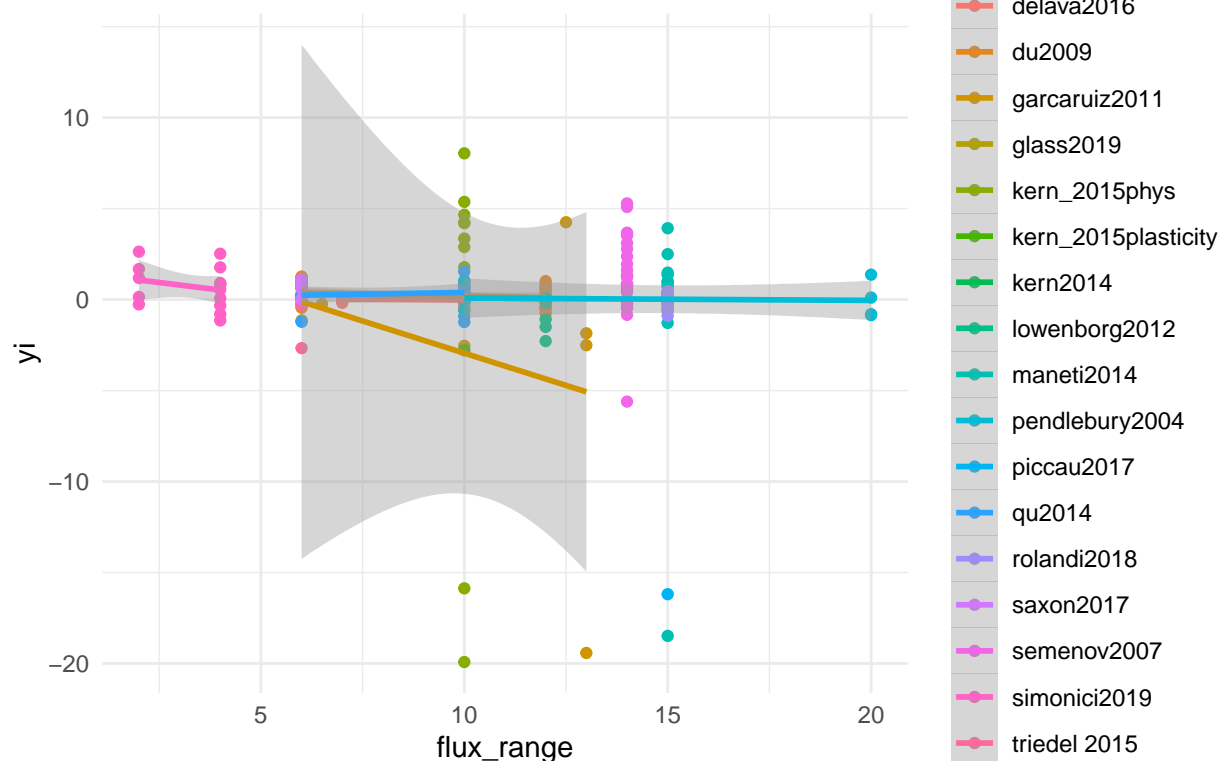
```
#simple linear model looking at how flux_range and org_level affect yi
simple8<-lm(yi~flux_range*org_level, data =dat_MA_ES)
summary(simple8)
```

```
##
## Call:
## lm(formula = yi ~ flux_range * org_level, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.053  -0.047   0.758   1.297   8.488
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.8927     1.4984   0.596   0.552
## flux_range      -0.1343     0.1297  -1.035   0.302
## org_level       -0.6582     3.5015  -0.188   0.851
## flux_range:org_level  0.1428     0.2910   0.491   0.624
##
## Residual standard error: 6.683 on 199 degrees of freedom
## Multiple R-squared:  0.008198,    Adjusted R-squared:  -0.006754
## F-statistic: 0.5483 on 3 and 199 DF,  p-value: 0.6498
```

Figure 5.

```
#scatterplot of standardized mean difference across flux range colored by study fit with lm
ggplot(normalized, aes(x=flux_range, y=yi, color = study_id))+
  geom_point()+
  geom_smooth(method="lm", formula = y~x)+
  theme_minimal()+
  ggtitle("SMD across fluctuation ranges colored by
          studies and fit with linear model by study")
```

SMD across fluctuation ranges colored by studies and fit with linear model by study



```
#random effects model including org_level as a random variable
fig9 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~flux_range + study_id,
              random = ~1 | experiment_id/ study_id,
              method="REML")
fig9
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0002     3    no      experiment_id
## sigma^2.2  0.3737  0.6113    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 185) = 3772.5680, p-val < .0001
##
## Test of Moderators (coefficients 2:18):
## QM(df = 17) = 333.0898, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb
## intrcpt          0.3911  0.6199   0.6309  0.5281  -0.8239
## flux_range       -0.0459  0.0109  -4.2133 <.0001  -0.0673
```

```
## study_iddu2009      0.3311  0.7530   0.4397  0.6602  -1.1447
## study_idgarcaruiz2011 -1.0411  0.8682  -1.1992  0.2305  -2.7427
## study_idglass2019    0.0331  0.8670   0.0382  0.9696  -1.6662
## study_idkern_2015phys 2.1396  0.8997   2.3782  0.0174   0.3763
## study_idkern_2015plasticity -0.7299  0.9276  -0.7868  0.4314  -2.5480
## study_idkern2014     0.9448  0.8742   1.0807  0.2798  -0.7687
## study_idlowenborg2012 -0.9177  0.7640  -1.2013  0.2296  -2.4151
## study_idmaneti2014    1.0227  0.8749   1.1689  0.2424  -0.6921
## study_idpendlebury2004 0.2829  0.8761   0.3229  0.7467  -1.4342
## study_idpiccau2017   -15.8917  1.1477 -13.8471 <.0001 -18.1410
## study_idqu2014       0.2923  0.8688   0.3365  0.7365  -1.4104
## study_idrolandi2018   0.3188  0.7535   0.4231  0.6722  -1.1579
## study_idsaxon2017     0.5722  0.7082   0.8079  0.4192  -0.8160
## study_idsemenov2007   1.3393  0.7816   1.7134  0.0866  -0.1927
## study_idsimonici2019  0.3429  0.8713   0.3936  0.6939  -1.3648
## study_idtriedel 2015  -1.2394  0.8775  -1.4124  0.1578  -2.9594
##                      ci.ub
## intrcpt             1.6060
## flux_range          -0.0246 ***
## study_iddu2009      1.8068
## study_idgarcaruiz2011 0.6605
## study_idglass2019    1.7323
## study_idkern_2015phys 3.9029 *
## study_idkern_2015plasticity 1.0883
## study_idkern2014     2.6583
## study_idlowenborg2012 0.5796
## study_idmaneti2014    2.7375
## study_idpendlebury2004 2.0000
## study_idpiccau2017   -13.6423 ***
## study_idqu2014       1.9950
## study_idrolandi2018   1.7956
## study_idsaxon2017     1.9603
## study_idsemenov2007   2.8712 .
## study_idsimonici2019  2.0506
## study_idtriedel 2015  0.4805
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#simple linear model looking at how flux_range and study_id affect yi
simple9<-lm(yi~flux_range*study_id, data =dat_MA_ES)
summary(simple9)
```

```
##
## Call:
## lm(formula = yi ~ flux_range * study_id, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -62.646  -0.307   0.035   0.554  22.869
##
## Coefficients: (11 not defined because of singularities)
##                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)          0.091100  12.267945   0.007   0.9941
```

```

## flux_range -0.010684 1.421326 -0.008 0.9940
## study_iddu2009 0.575435 12.722653 0.045 0.9640
## study_idgarcaruiz2011 22.254823 14.894685 1.494 0.1369
## study_idglass2019 -0.020693 3.692713 -0.006 0.9955
## study_idkern_2015phys -0.016487 3.481523 -0.005 0.9962
## study_idkern_2015plasticity -1.266498 5.222285 -0.243 0.8087
## study_idkern2014 0.747017 9.156252 0.082 0.9351
## study_idlowenborg2012 -0.955077 6.046902 -0.158 0.8747
## study_idmaneti2014 -0.530046 9.639901 -0.055 0.9562
## study_idpendlebury2004 0.119910 13.998437 0.009 0.9932
## study_idpiccau2017 -16.120645 11.236568 -1.435 0.1531
## study_idqu2014 -0.077323 13.468891 -0.006 0.9954
## study_idrolandi2018 0.087188 9.531756 0.009 0.9927
## study_idsaxon2017 0.543332 4.729285 0.115 0.9087
## study_idsemenov2007 1.440235 8.199783 0.176 0.8608
## study_idsimonici2019 1.535025 13.536217 0.113 0.9098
## study_idtriedel 2015 -1.510015 5.945833 -0.254 0.7998
## flux_range:study_iddu2009 -0.035825 1.465069 -0.024 0.9805
## flux_range:study_idgarcaruiz2011 -3.266636 1.607803 -2.032 0.0437 *
## flux_range:study_idglass2019 NA NA NA NA
## flux_range:study_idkern_2015phys NA NA NA NA
## flux_range:study_idkern_2015plasticity NA NA NA NA
## flux_range:study_idkern2014 NA NA NA NA
## flux_range:study_idlowenborg2012 NA NA NA NA
## flux_range:study_idmaneti2014 NA NA NA NA
## flux_range:study_idpendlebury2004 -0.002104 1.483908 -0.001 0.9989
## flux_range:study_idpiccau2017 NA NA NA NA
## flux_range:study_idqu2014 0.048564 1.573119 0.031 0.9754
## flux_range:study_idrolandi2018 NA NA NA NA
## flux_range:study_idsaxon2017 NA NA NA NA
## flux_range:study_idsemenov2007 NA NA NA NA
## flux_range:study_idsimonici2019 -0.264625 2.178851 -0.121 0.9035
## flux_range:study_idtriedel 2015 NA NA NA NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.03 on 180 degrees of freedom
## Multiple R-squared: 0.2695, Adjusted R-squared: 0.1802
## F-statistic: 3.019 on 22 and 180 DF, p-value: 2.6e-05

```

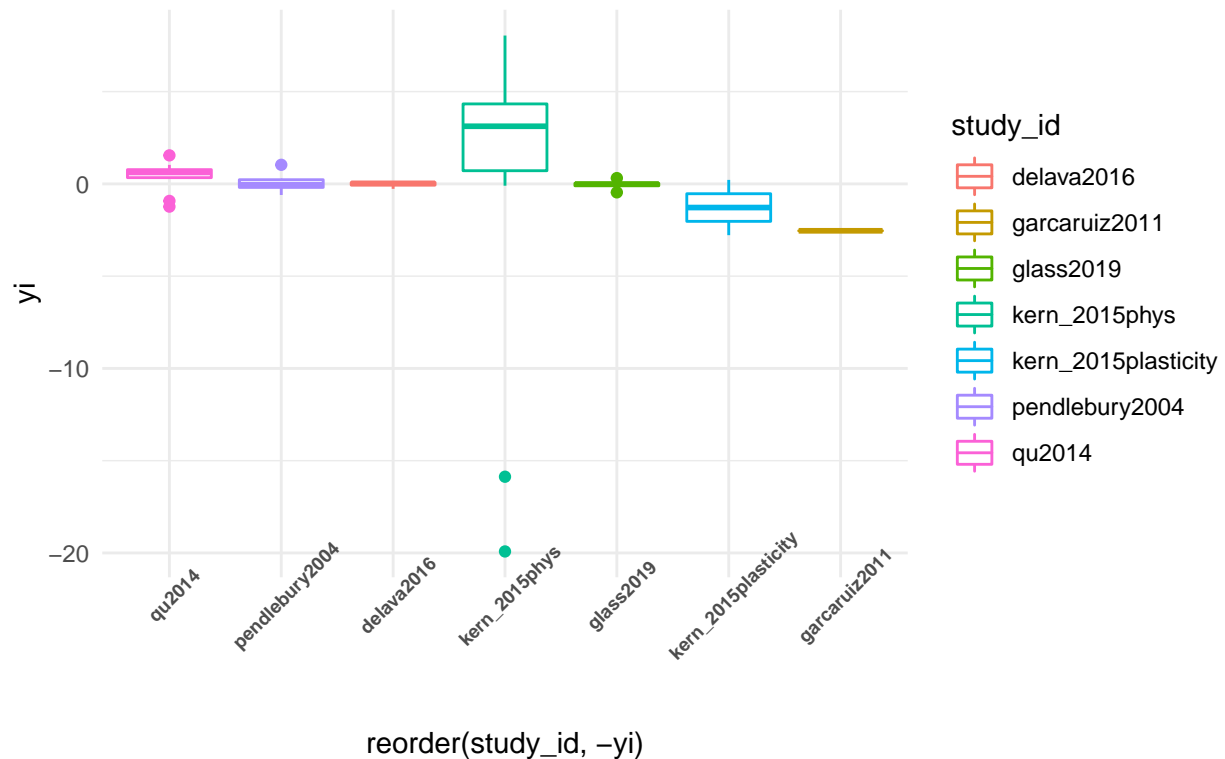
Figure 6.

```

#looking at yi across studies with the same temperature range (10 C)
ggplot(common_range, aes(y=yi, x=reorder(study_id, -yi), color = study_id))+
  geom_boxplot()+
  theme_minimal()+
  theme(axis.text.x = element_text(face = "bold",
                                    size = 7, angle = 45))+
  ggtitle("SMD across studies with the same
          temperature fluctuation range (10 C)")

```

SMD across studies with the same temperature fluctuation range (10 C)



```
#random effects model within subset data
fig12 <- rma.mv(yi, vi, data=common_range,
               random = ~1 | experiment_id/ study_id,
               method="REML")
```

```
## Warning in rma.mv(yi, vi, data = common_range, random = ~1 | experiment_id/
## study_id, : Single-level factor(s) found in 'random' argument. Corresponding
## 'sigma2' value(s) fixed to 0.
```

```
fig12
```

```
##
## Multivariate Meta-Analysis Model (k = 41; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0000     1   yes      experiment_id
## sigma^2.2  1.8372  1.3554     7    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 40) = 478.3083, p-val < .0001
##
## Model Results:
##
```

```
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.1210  0.5173  -0.2338  0.8151  -1.1349  0.8930
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#simple linear model looking at how flux_range affects yi within subset data
simple12<-lm(yi~flux_range, data =common_range)
summary(simple12)
```

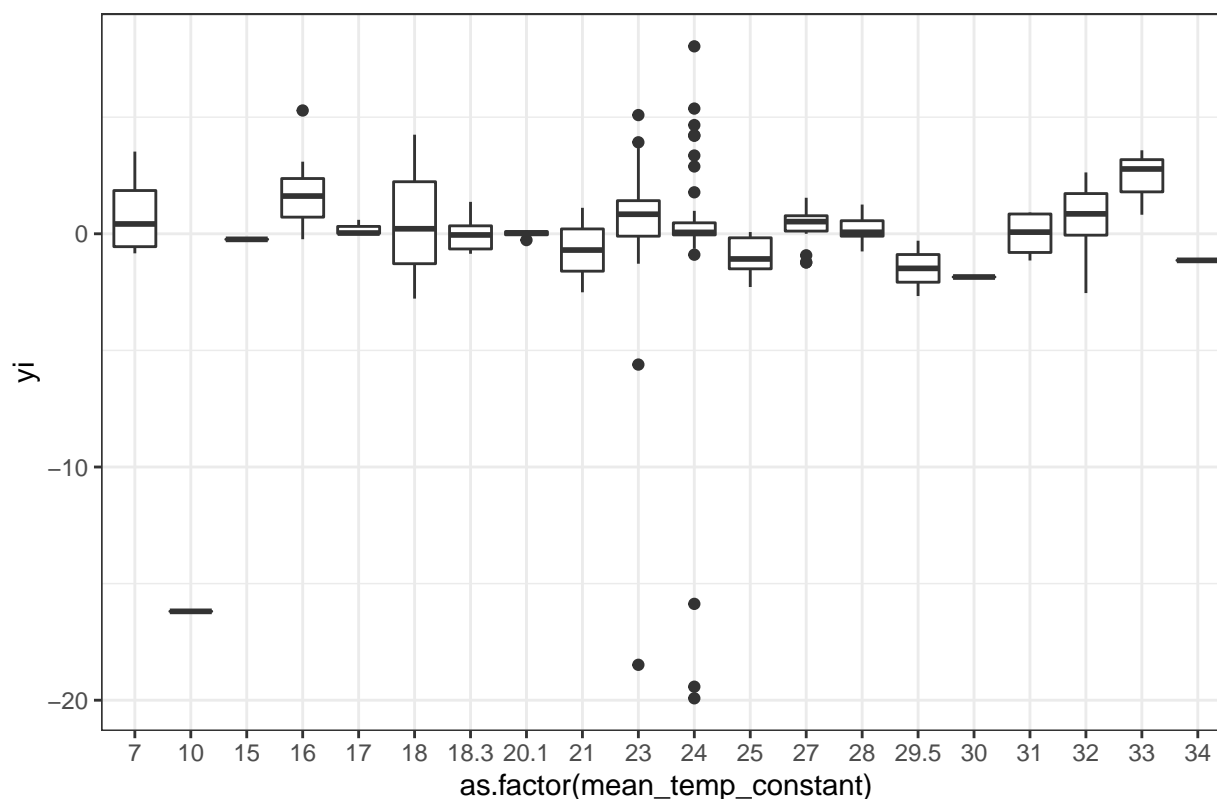
```
##
## Call:
## lm(formula = yi ~ flux_range, data = common_range)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.8796  -0.0606   0.1892   1.0683   8.0766
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.03876    0.71869  -0.054   0.957
## flux_range      NA           NA      NA      NA
##
## Residual standard error: 4.602 on 40 degrees of freedom
```

Question how does mean temperature affect response variables?

Figure 7.

```
# boxplots of how mean temperature influences SMD
ggplot(normalized, aes(x=as.factor(mean_temp_constant), y=yi))+
  geom_boxplot()+
  theme_bw()+
  ggtitle("SMD across mean temperature")
```

SMD across mean temperature



```
#random effects model including org_level as a random variable
fig6 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~mean_temp_constant,
              random = ~1 | experiment_id/ study_id,
              method="REML")
fig6
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##          estim  sqrt  nlvls  fixed          factor
## sigma^2.1  0.0000  0.0006    3    no          experiment_id
## sigma^2.2 13.1747  3.6297   23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 5585.5672, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 112.3797, p-val < .0001
##
## Model Results:
##
##          estimate      se      zval    pval    ci.lb    ci.ub
## intrcpt          1.4917  0.7811   1.9098  0.0562  -0.0392  3.0226
## mean_temp_constant -0.0863  0.0081 -10.6009 <.0001  -0.1022 -0.0703 ***
```

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

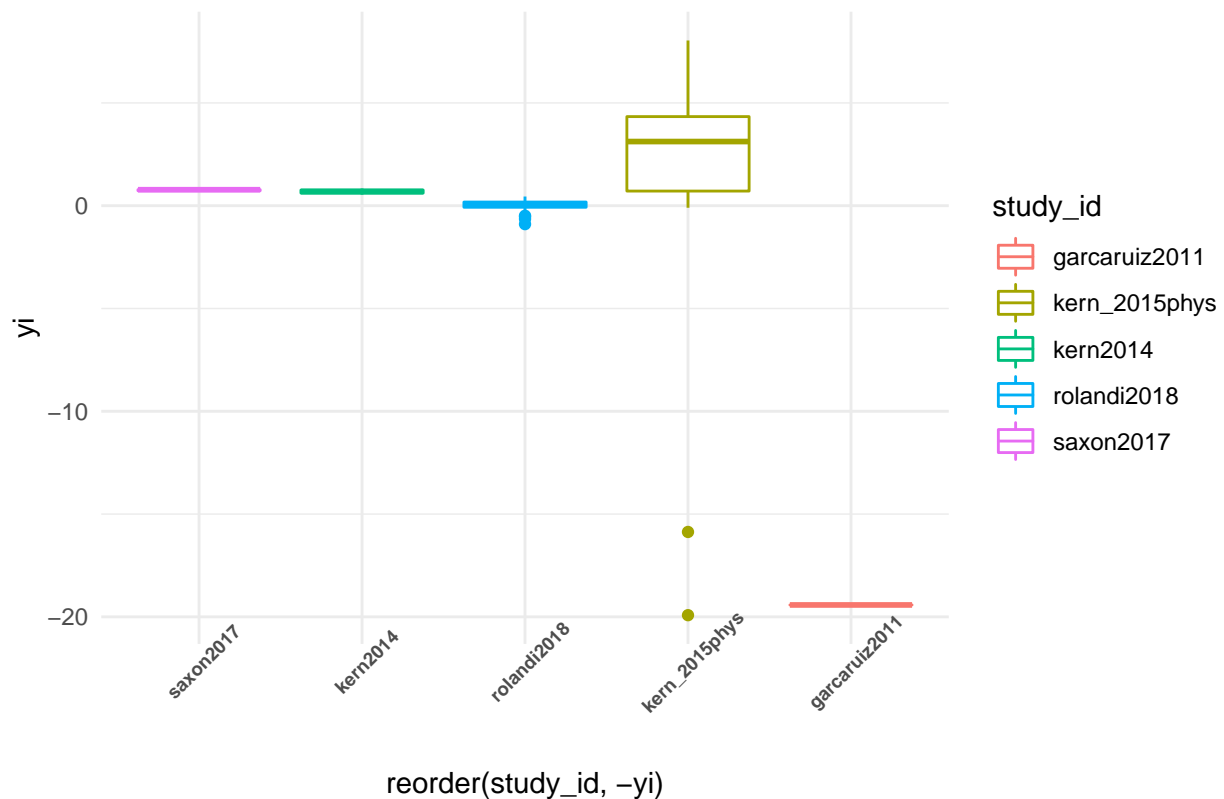
#simple linear model looking at how mean_temp_constant affects yi
simple6<-lm(yi~mean_temp_constant, data =dat_MA_ES)
summary(simple6)

##
## Call:
## lm(formula = yi ~ mean_temp_constant, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.499   0.217   0.512   1.238   8.418
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.171402    2.393869  -0.072    0.943
## mean_temp_constant -0.008692    0.095824  -0.091    0.928
##
## Residual standard error: 6.677 on 201 degrees of freedom
## Multiple R-squared:  4.093e-05, Adjusted R-squared:  -0.004934
## F-statistic: 0.008227 on 1 and 201 DF, p-value: 0.9278
```

Figure 8.

```
#looking at yi across studies with the same mean temperature (24 C)
ggplot(common_temp, aes(y=yi, x=reorder(study_id, -yi), color = study_id))+
  geom_boxplot()+
  theme_minimal()+
  theme(axis.text.x = element_text(face = "bold",
                                    size = 7, angle = 45))+
  ggtitle("SMD across studies with the same mean temperature (24 C)")
```


SMD across studies with the same mean temperature (24 C)



```
#random effects model within subset data
fig13 <- rma.mv(yi, vi, data=common_temp,
               random = ~1 | experiment_id/ study_id,
               method="REML")
fig13
```

```
##
## Multivariate Meta-Analysis Model (k = 56; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0024     2    no      experiment_id
## sigma^2.2  67.1521  8.1946     6    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 55) = 2146.3005, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -2.6186    3.3484   -0.7820  0.4342   -9.1814    3.9442
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#simple linear model looking at how flux_range affects yi within subset data
simple13<-lm(yi~mean_temp_constant, data =common_temp)
summary(simple13)
```

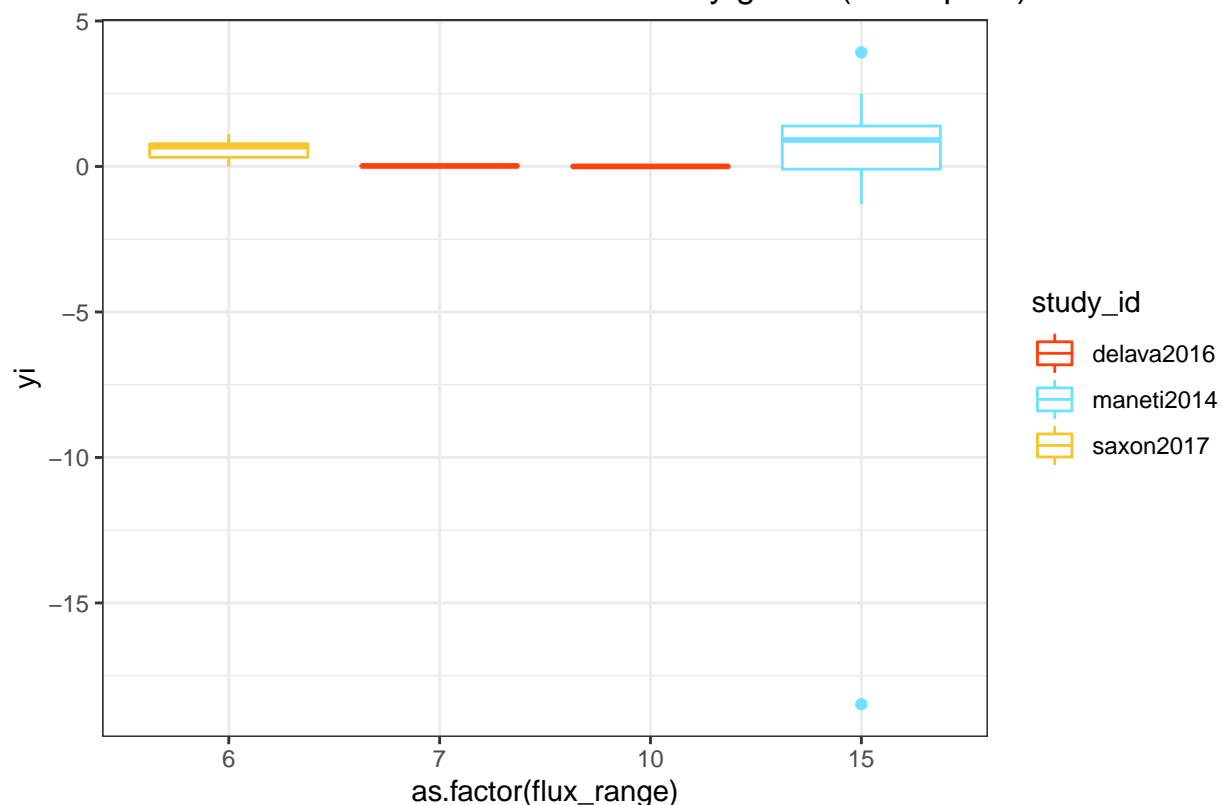
```
##
## Call:
## lm(formula = yi ~ mean_temp_constant, data = common_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.6416   0.2290   0.3368   0.7468   8.3146
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.2767     0.6247  -0.443   0.659
## mean_temp_constant      NA         NA      NA     NA
##
## Residual standard error: 4.675 on 55 degrees of freedom
```

Question: How does genus affect response?

Figure 9.

```
#trying to look at drosophila response across studies
ggplot(drosophila, aes(y=yi, x=as.factor(flux_range), color = study_id))+
  geom_boxplot()+
  scale_color_tron()+
  theme_bw()+
  ggtitle("SMD across studies with the same study genus (Drosophila)")
```

SMD across studies with the same study genus (Drosophila)



```
#random effects model including org_level as a random variable
fig10 <- rma.mv(yi, vi, data=drosophila, mods = ~flux_range + study_id,
               random = ~1 | experiment_id/ study_id,
               method="REML")
fig10
```

```
##
## Multivariate Meta-Analysis Model (k = 21; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.1034  0.3216     3    no      experiment_id
## sigma^2.2  0.1034  0.3216     5    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 17) = 798.4274, p-val < .0001
##
## Test of Moderators (coefficients 2:4):
## QM(df = 3) = 3.2667, p-val = 0.3523
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt          0.0259  0.6614   0.0392  0.9687  -1.2704  1.3223
## flux_range      -0.0055  0.0596  -0.0922  0.9265  -0.1224  0.1114
```

```
## study_idmaneti2014    0.7499  0.6134   1.2226  0.2215  -0.4523  1.9521
## study_idsaxon2017    0.6933  0.4505   1.5389  0.1238  -0.1897  1.5763
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

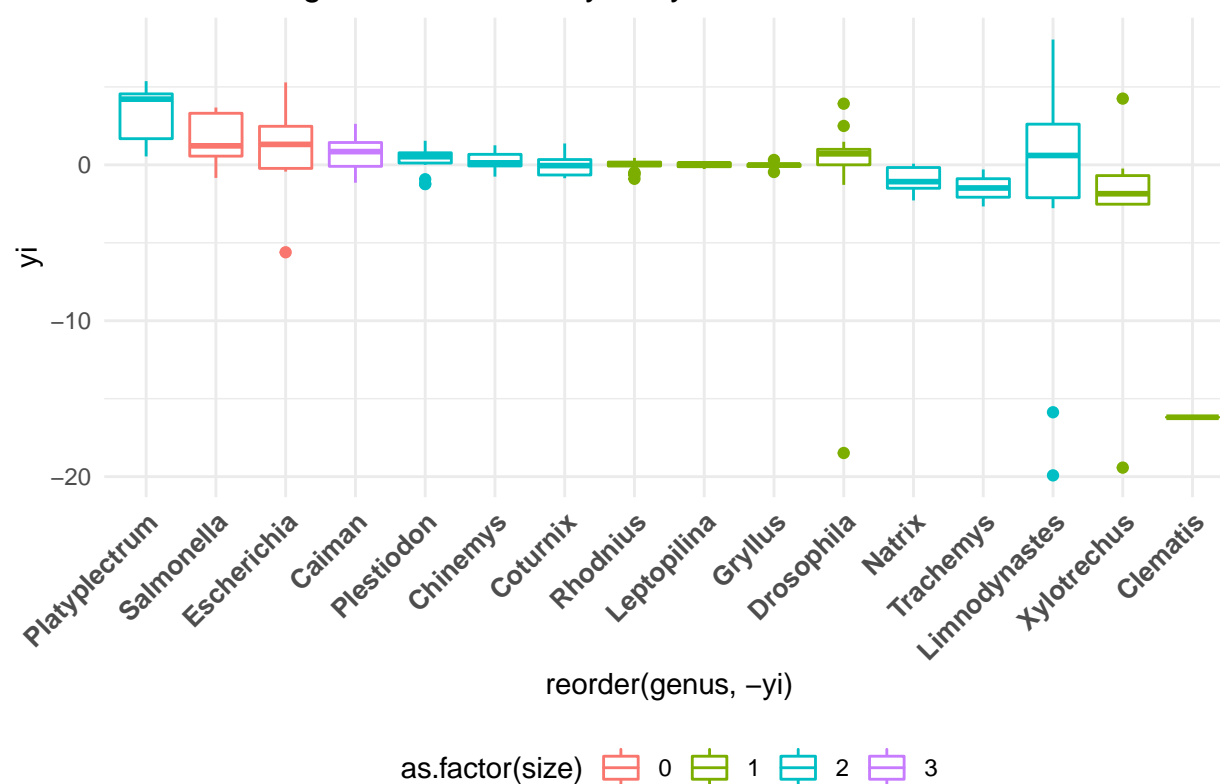
```
#simple linear model looking at how flux_range and study_id affect yi in drosophila specific studies
simple10<-lm(yi~flux_range + study_id, data =drosophila)
summary(simple10)
```

```
##
## Call:
## lm(formula = yi ~ flux_range + study_id, data = drosophila)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.8832   0.0000   0.4773   1.5850   4.5217
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.057126  18.951260   0.003   0.998
## flux_range     -0.005498   2.195634  -0.003   0.998
## study_idmaneti2014 -0.573860  14.708290  -0.039   0.969
## study_idsaxon2017  0.546191   6.638973   0.082   0.935
##
## Residual standard error: 4.658 on 17 degrees of freedom
## Multiple R-squared:  0.01628,    Adjusted R-squared:  -0.1573
## F-statistic: 0.09381 on 3 and 17 DF,  p-value: 0.9624
```

Figure 10.

```
#boxplots of SMD across different genres colored by body size
ggplot(normalized, aes(y=yi, x= reorder(genus, -yi), color = as.factor(size)))+
  geom_boxplot()+
  theme_minimal()+
  ggtitle("SMD across genres colored by body size")+
  theme(axis.text.x = element_text(face = "bold",
                                   size = 10, angle = 45,hjust = 1),
        legend.position = "bottom")
```

SMD across genres colored by body size



```
#random effects model including genus as a random variable
fig11 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~genus,
               random = ~1 | experiment_id/ study_id,
               method="REML")
fig11
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0002     3    no      experiment_id
## sigma^2.2  0.8920  0.9445    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 187) = 3871.5222, p-val < .0001
##
## Test of Moderators (coefficients 2:16):
## QM(df = 15) = 217.8816, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt          0.5791  0.9480   0.6108  0.5413  -1.2790   2.4371
## genusChinemys    -0.2569  1.1615  -0.2211  0.8250  -2.5334   2.0197
```

```
## genusClematis      -16.7689  1.5341 -10.9306 <.0001 -19.7757 -13.7620 ***
## genusCoturnix      -0.5885  1.3432 -0.4381  0.6613 -3.2211  2.0441
## genusDrosophila    -0.0185  1.0383 -0.0179  0.9858 -2.0536  2.0165
## genusEscherichia    0.5687  1.2012  0.4735  0.6359 -1.7855  2.9230
## genusGryllus       -0.6144  1.3390 -0.4588  0.6464 -3.2387  2.0100
## genusLeptopilina   -0.0365  1.0419 -0.0351  0.9720 -2.0786  2.0055
## genusLimnodynastes -1.0945  1.1257 -0.9723  0.3309 -3.3008  1.1118
## genusNatrix        -1.6709  1.1681 -1.4304  0.1526 -3.9604  0.6186
## genusPlatyplectrum  1.6995  1.1435  1.4863  0.1372 -0.5416  3.9407
## genusPlestiodon    -0.2612  1.3402 -0.1949  0.8455 -2.8879  2.3655
## genusRhodnius      -0.5584  1.1597 -0.4815  0.6302 -2.8314  1.7146
## genusSalmonella     0.4783  1.1953  0.4001  0.6890 -1.8645  2.8211
## genusTrachemys     -1.7031  1.3456 -1.2656  0.2057 -4.3405  0.9343
## genusXylotrechus   -1.6640  1.3398 -1.2420  0.2142 -4.2899  0.9619
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

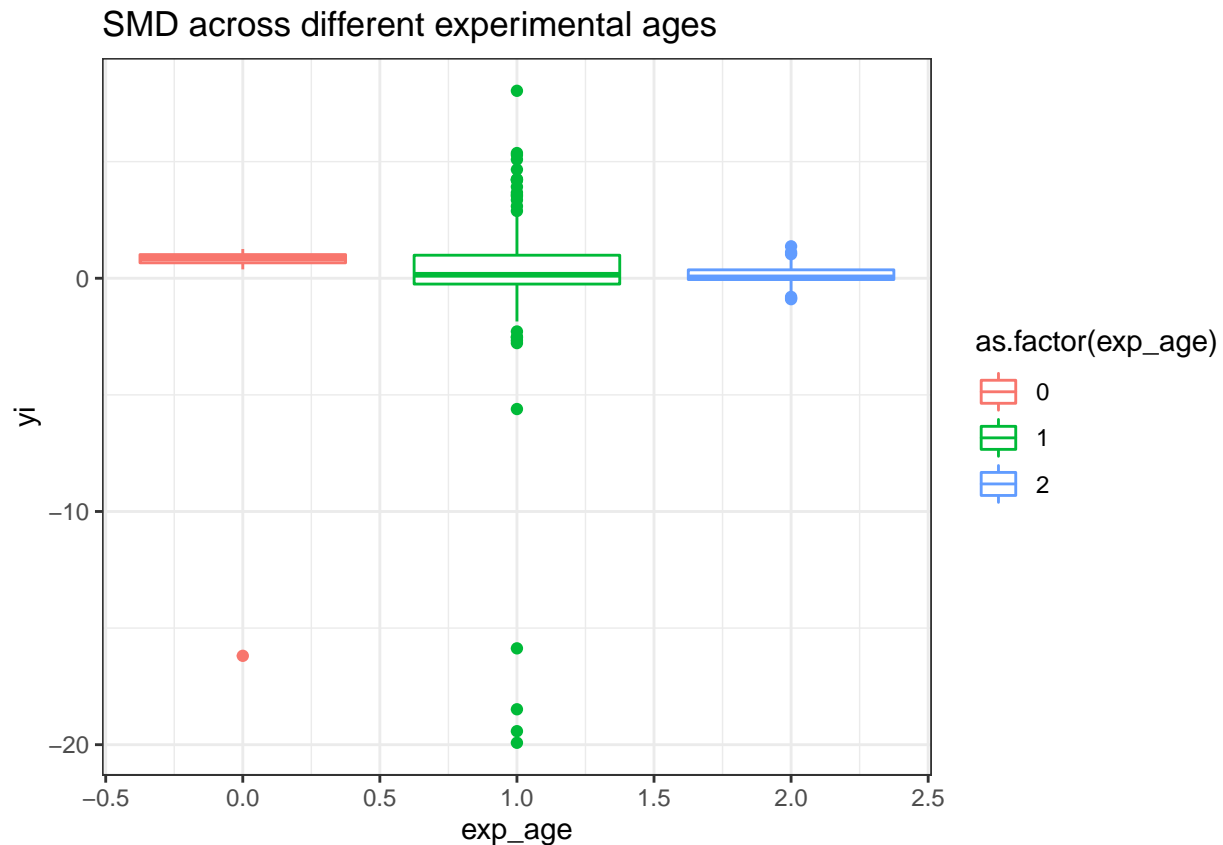
```
#simple linear model looking at how genus affects yi
simple11<-lm(yi~genus, data =dat_MA_ES)
summary(simple11)
```

```
##
## Call:
## lm(formula = yi ~ genus, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -69.610  -0.301   0.067   0.765  17.544
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.7084     1.5951   0.444  0.65746
## genusChinemys  -0.4605     1.9331  -0.238  0.81198
## genusClematis  -16.8982     6.3803  -2.648  0.00878 **
## genusCoturnix  -0.6892     2.7046  -0.255  0.79913
## genusDrosophila -0.8597     2.0884  -0.412  0.68106
## genusEscherichia  0.4926     2.3926   0.206  0.83712
## genusGryllus   -0.7449     2.7046  -0.275  0.78331
## genusLeptopilina -0.7115     2.9841  -0.238  0.81181
## genusLimnodynastes -2.8493     2.5220  -1.130  0.26002
## genusNatrix     -1.7006     3.1902  -0.533  0.59461
## genusPlatyplectrum  2.5973     2.9841   0.870  0.38520
## genusPlestiodon -0.3916     2.1101  -0.186  0.85297
## genusRhodnius   -0.6904     1.8838  -0.366  0.71441
## genusSalmonella  0.8706     2.4523   0.355  0.72299
## genusTrachemys  -2.1914     4.6504  -0.471  0.63802
## genusXylotrechus -14.0034     2.7046  -5.178 5.77e-07 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.178 on 187 degrees of freedom
## Multiple R-squared:  0.2035, Adjusted R-squared:  0.1396
## F-statistic: 3.186 on 15 and 187 DF,  p-value: 0.0001102
```

Question how does scale (life or org level) affect response?

Figure 11.

```
#boxplot
ggplot(normalized, aes(x=exp_age, y=yi, color = as.factor(exp_age)))+
  geom_boxplot()+
  theme_bw()+
  ggtitle("SMD across different experimental ages")
```



```
#random effects model including exp_age as a random variable
fig3 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~exp_age,
  random = ~1 | experiment_id/ study_id,
  method="REML")
fig3
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##      estim    sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0002    3    no      experiment_id
## sigma^2.2  9.7610  3.1243   23    no  experiment_id/study_id
```

```
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 5524.6959, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 4.8667, p-val = 0.0274
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt    -3.5084  1.5160  -2.3142  0.0207   -6.4798   -0.5371  *
## exp_age     2.5642  1.1624   2.2061  0.0274    0.2861    4.8424  *
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

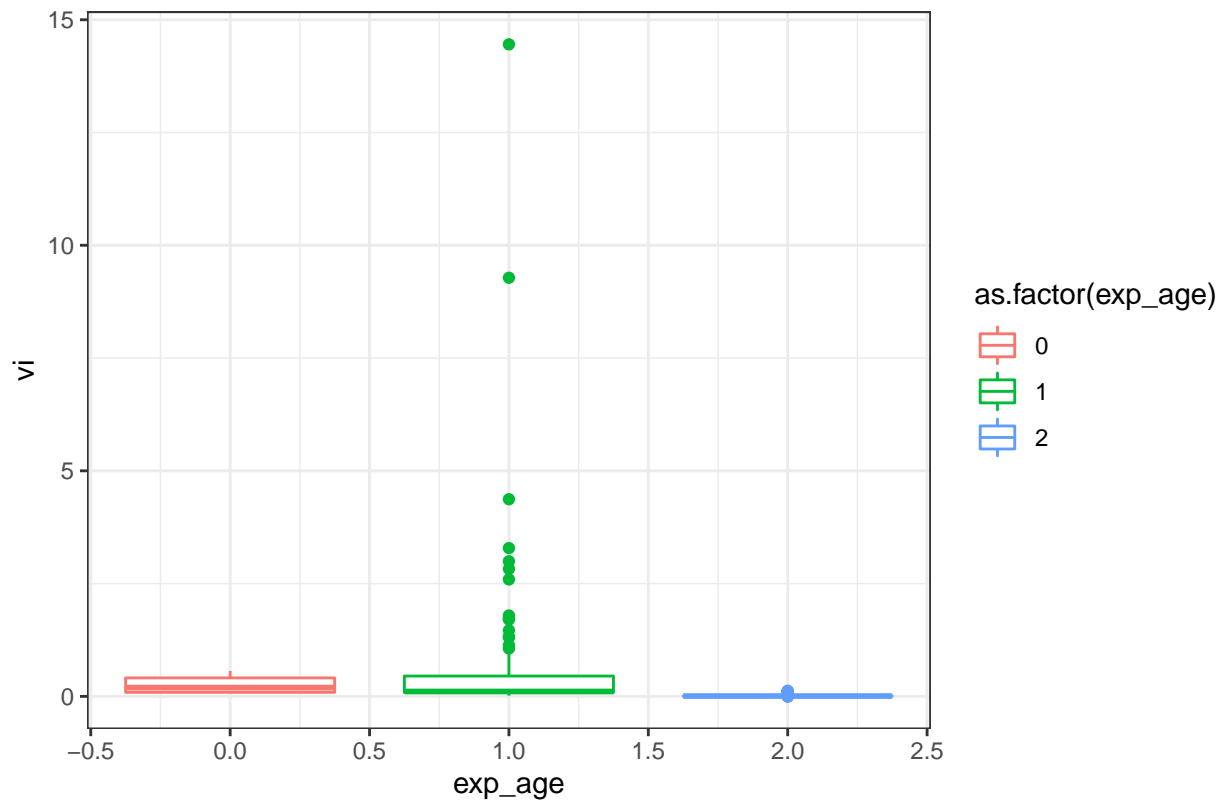
```
#simple linear model looking at how exp_age affects yi
simple3<-lm(yi~exp_age, data =dat_MA_ES)
summary(simple3)
```

```
##
## Call:
## lm(formula = yi ~ exp_age, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.432   0.032   0.473   1.324   8.511
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.9220     1.1476  -0.803   0.423
## exp_age       0.4491     0.8752   0.513   0.608
##
## Residual standard error: 6.672 on 201 degrees of freedom
## Multiple R-squared:  0.001308,    Adjusted R-squared:  -0.00366
## F-statistic: 0.2633 on 1 and 201 DF,  p-value: 0.6084
```

Figure 12.

```
ggplot(normalized, aes(x=exp_age, y=vi, color = as.factor(exp_age)))+
  geom_boxplot()+
  theme_bw()+
  ggtitle("Variance across different experimental ages")
```


Variance across different experimental ages



```
count(normalized, exp_age)
```

```
## exp_age n
## 1      0 13
## 2      1 136
## 3      2 53
```

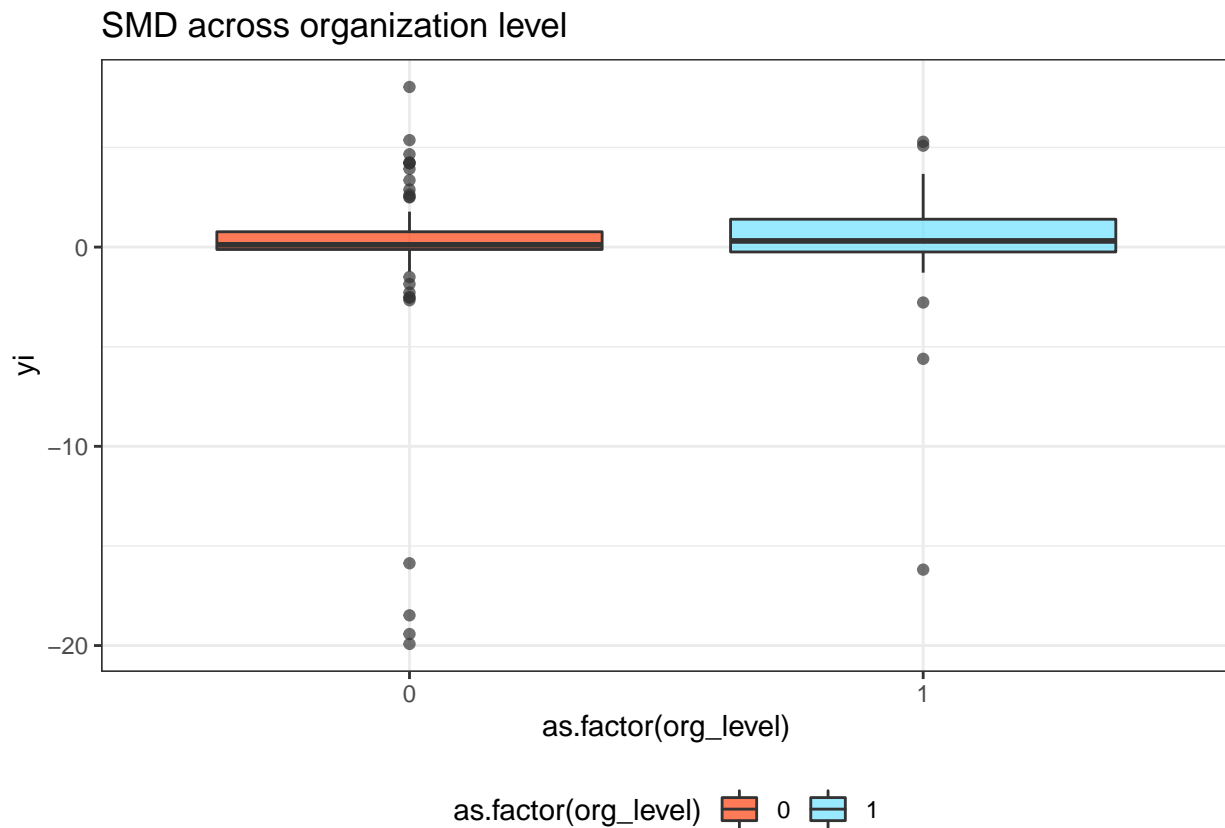
```
#simple linear model looking at how exp_age affects vi
simple4<-lm(vi~ exp_age, data =dat_MA_ES)
summary(simple4)
```

```
##
## Call:
## lm(formula = vi ~ exp_age, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.7998 -0.4468 -0.3512 -0.1753 13.9209
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.8859     0.2713   3.265 0.00129 **
## exp_age      -0.3518     0.2069  -1.700 0.09067 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 1.578 on 201 degrees of freedom
## Multiple R-squared:  0.01418,    Adjusted R-squared:  0.009271
## F-statistic:  2.89 on 1 and 201 DF,  p-value: 0.09067
```

Figure 13.

```
#boxplot of SMD across levels of organization
ggplot(normalized, aes(x=as.factor(org_level), y=yi, fill=as.factor(org_level)))+
  geom_boxplot(alpha =0.7)+
  scale_fill_tron()+
  theme_bw()+
  theme(legend.position = "bottom")+
  ggtitle("SMD across organization level")
```



```
#random effects model including org_level as a random variable
fig5 <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~org_level,
  random = ~1 | experiment_id/ study_id,
  method="REML")
fig5
```

```
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
```

```
##
## Variance Components:
##
##          estim      sqrt nlvls  fixed          factor
## sigma^2.1  0.0000  0.0005     3    no      experiment_id
## sigma^2.2 11.0340  3.3217    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 5584.5912, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 26.9530, p-val < .0001
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      -0.3940  0.6944  -0.5675  0.5704  -1.7549   0.9669
## org_level     -0.4309  0.0830  -5.1916  <.0001  -0.5935  -0.2682 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#simple linear model looking at how org_level affects yi
simple5<-lm(yi~org_level, data =dat_MA_ES)
summary(simple5)
```

```
##
## Call:
## lm(formula = yi ~ org_level, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.345   0.132   0.620   1.304   8.598
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.5602     0.5222  -1.073   0.285
## org_level      0.8926     1.1764   0.759   0.449
##
## Residual standard error: 6.667 on 201 degrees of freedom
## Multiple R-squared:  0.002856, Adjusted R-squared:  -0.002105
## F-statistic: 0.5756 on 1 and 201 DF, p-value: 0.4489
```

Question: How does response variable affect response?

Figure 14.

```
#full random effects model looking at all the responses in the entire dataset
response_variable_mod <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~resp_def,
                                random = ~1 | experiment_id/ study_id/ response_id,
                                method="REML")
response_variable_mod
```

```

##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##          estim      sqrt  nlvls  fixed          factor
## sigma^2.1  0.0000  0.0005     3    no          experiment_id
## sigma^2.2  1.1317  1.0638    23    no      experiment_id/study_id
## sigma^2.3  0.0494  0.2222    58    no  experiment_id/study_id/response_id
##
## Test for Residual Heterogeneity:
## QE(df = 156) = 3362.1739, p-val < .0001
##
## Test of Moderators (coefficients 2:47):
## QM(df = 46) = 301.2288, p-val < .0001
##
## Model Results:
##
##                                     estimate      se
## intrcpt                           0.7651  0.6179
## resp_defaquatic speed              -1.8439  1.2838
## resp_defaverage cumulative number of eggs laid per female -0.6509  1.2503
## resp_defaverage number of eggs laid per female          -0.8380  1.2503
## resp_defbody (centroid) size        -0.4839  1.0727
## resp_defbody length                 -0.8167  0.6643
## resp_defbody mass                   0.0207  0.4261
## resp_defcarapace height              0.0186  0.5831
## resp_defcarapace width               0.0864  0.5836
## resp_defdaily energy expenditure     -1.3331  1.2606
## resp_defdays to first slough        -3.0470  1.2983
## resp_defdessication tolerance        -1.1771  1.0661
## resp_defdevelopment time             -0.6439  0.9320
## resp_defdevelopment to stages 35-37 -1.2264  1.1748
## resp_defdevelopmental time          -0.2328  1.0174
## resp_defdistance covered            -0.9373  1.2797
## resp_defegg mass                    -0.1998  1.2609
## resp_defegg to adult viability       -2.9584  1.0642
## resp_deffore-limb length              0.8910  0.4560
## resp_defgermination                 -16.9549  1.4579
## resp_defhatching success           -0.3360  1.2628
## resp_defhead length                  0.2518  0.4497
## resp_defhead width                   0.5433  0.4504
## resp_defhind-limb length              0.3872  0.4489
## resp_defincubation period            -0.4023  1.2620
## resp_defincubation time              -2.2675  1.2877
## resp_definfestation rate             -0.7299  0.9319
## resp_defmass                         0.6832  1.2667
## resp_defmaximal length               -1.4584  0.4567
## resp_defoffspring per mating         -0.2905  0.8904
## resp_defovary mass, dry              -0.7837  1.2517
## resp_defoxidative damage             -1.0621  1.2624
## resp_defpercent females              -0.5725  1.2619
## resp_defproductivity                 -1.4666  1.0641
## resp_defrate of change                0.7569  1.2960

```

## resp_defsnout-vent length	0.6062	0.4512
## resp_defsprint speed	-0.6694	0.4491
## resp_defstartvation tolerance	-1.8855	1.0634
## resp_defsuccess of parasitism	-1.0187	0.9320
## resp_defsurvival	-0.7856	0.8919
## resp_defTAC	-3.4342	1.2729
## resp_deftail length	0.2805	0.4301
## resp_defterrestrial speed	-0.6908	1.2796
## resp_deftestes mass, dry	-0.8179	1.2518
## resp_deftotal length	-0.0335	1.2649
## resp_deftotal offspring	-0.2270	0.8904
## resp_defwing centroid	0.0686	0.8811
##	zval	pval
## intrcpt	1.2383	0.2156
## resp_defaquatic speed	-1.4363	0.1509
## resp_defaverage cumulative number of eggs laid per female	-0.5206	0.6027
## resp_defaverage number of eggs laid per female	-0.6703	0.5027
## resp_defbody (centroid) size	-0.4511	0.6519
## resp_defbody length	-1.2294	0.2189
## resp_defbody mass	0.0485	0.9613
## resp_defcarapace height	0.0319	0.9745
## resp_defcarapace width	0.1481	0.8823
## resp_defdaily energy expenditure	-1.0575	0.2903
## resp_defdays to first slough	-2.3470	0.0189
## resp_defdessication tolerance	-1.1042	0.2695
## resp_defdevelopment time	-0.6909	0.4896
## resp_defdevelopment to stages 35-37	-1.0439	0.2965
## resp_defdevelopmental time	-0.2288	0.8190
## resp_defdistance covered	-0.7325	0.4639
## resp_defegg mass	-0.1584	0.8741
## resp_defegg to adult viability	-2.7799	0.0054
## resp_deffore-limb length	1.9540	0.0507
## resp_defgermination	-11.6294	<.0001
## resp_defhatching success	-0.2661	0.7902
## resp_defhead length	0.5600	0.5755
## resp_defhead width	1.2063	0.2277
## resp_defhind-limb length	0.8625	0.3884
## resp_defincubation period	-0.3188	0.7499
## resp_defincubation time	-1.7609	0.0783
## resp_definfestation rate	-0.7833	0.4335
## resp_defmass	0.5394	0.5896
## resp_defmaximal length	-3.1935	0.0014
## resp_defoffspring per mating	-0.3263	0.7442
## resp_defovary mass, dry	-0.6261	0.5312
## resp_defoxidative damage	-0.8414	0.4002
## resp_defpercent females	-0.4536	0.6501
## resp_defproductivity	-1.3782	0.1681
## resp_defrate of change	0.5840	0.5592
## resp_defsnout-vent length	1.3437	0.1791
## resp_defsprint speed	-1.4906	0.1361
## resp_defstartvation tolerance	-1.7731	0.0762
## resp_defsuccess of parasitism	-1.0930	0.2744
## resp_defsurvival	-0.8808	0.3784
## resp_defTAC	-2.6980	0.0070

## resp_deftail length	0.6522	0.5142
## resp_defterrestrial speed	-0.5399	0.5893
## resp_deftestes mass, dry	-0.6533	0.5135
## resp_deftotal length	-0.0265	0.9789
## resp_deftotal offspring	-0.2550	0.7987
## resp_defwing centroid	0.0779	0.9379
##	ci.lb	ci.ub
## intrcpt	-0.4459	1.9762
## resp_defaquatic speed	-4.3601	0.6723
## resp_defaverage cumulative number of eggs laid per female	-3.1014	1.7997
## resp_defaverage number of eggs laid per female	-3.2885	1.6125
## resp_defbody (centroid) size	-2.5863	1.6185
## resp_defbody length	-2.1188	0.4853
## resp_defbody mass	-0.8146	0.8559
## resp_defcarapace height	-1.1242	1.1614
## resp_defcarapace width	-1.0573	1.2301
## resp_defdaily energy expenditure	-3.8039	1.1377
## resp_defdays to first slough	-5.5916	-0.5024
## resp_defdessication tolerance	-3.2666	0.9123
## resp_defdevelopment time	-2.4705	1.1827
## resp_defdevelopment to stages 35-37	-3.5288	1.0761
## resp_defdevelopmental time	-2.2267	1.7612
## resp_defdistance covered	-3.4455	1.5708
## resp_defegg mass	-2.6712	2.2716
## resp_defegg to adult viability	-5.0442	-0.8725
## resp_deffore-limb length	-0.0027	1.7847
## resp_defgermination	-19.8124	-14.0974
## resp_defhatching success	-2.8111	2.1391
## resp_defhead length	-0.6296	1.1333
## resp_defhead width	-0.3395	1.4262
## resp_defhind-limb length	-0.4927	1.2670
## resp_defincubation period	-2.8758	2.0712
## resp_defincubation time	-4.7914	0.2564
## resp_definfestation rate	-2.5565	1.0966
## resp_defmass	-1.7994	3.1659
## resp_defmaximal length	-2.3535	-0.5633
## resp_defoffspring per mating	-2.0356	1.4546
## resp_defovary mass, dry	-3.2371	1.6696
## resp_defoxidative damage	-3.5363	1.4121
## resp_defpercent females	-3.0458	1.9009
## resp_defproductivity	-3.5523	0.6191
## resp_defrate of change	-1.7833	3.2970
## resp_defsnout-vent length	-0.2781	1.4905
## resp_defsprint speed	-1.5496	0.2108
## resp_defstartvation tolerance	-3.9696	0.1987
## resp_defsuccess of parasitism	-2.8453	0.8080
## resp_defsurvival	-2.5336	0.9625
## resp_defTAC	-5.9290	-0.9394
## resp_deftail length	-0.5625	1.1236
## resp_defterrestrial speed	-3.1988	1.8172
## resp_deftestes mass, dry	-3.2714	1.6356
## resp_deftotal length	-2.5126	2.4457
## resp_deftotal offspring	-1.9722	1.5181
## resp_defwing centroid	-1.6583	1.7955

```

##
## intrcpt
## resp_defaquatic speed
## resp_defaverage cumulative number of eggs laid per female
## resp_defaverage number of eggs laid per female
## resp_defbody (centroid) size
## resp_defbody length
## resp_defbody mass
## resp_defcarapace height
## resp_defcarapace width
## resp_defdaily energy expenditure
## resp_defdays to first slough *
## resp_defdessication tolerance
## resp_defdevelopment time
## resp_defdevelopment to stages 35-37
## resp_defdevelopmental time
## resp_defdistance covered
## resp_defegg mass
## resp_defegg to adult viability **
## resp_deffore-limb length .
## resp_defgermination ***
## resp_defhatching success
## resp_defhead length
## resp_defhead width
## resp_defhind-limb length
## resp_defincubation period
## resp_defincubation time .
## resp_definfestation rate
## resp_defmass
## resp_defmaximal length **
## resp_defoffspring per mating
## resp_defovary mass, dry
## resp_defoxidative damage
## resp_defpercent females
## resp_defproductivity
## resp_defrate of change
## resp_defsnout-vent length
## resp_defsprint speed
## resp_defstartvation tolerance .
## resp_defsuccess of parasitism
## resp_defsurvival
## resp_defTAC **
## resp_deftail length
## resp_defterrestrial speed
## resp_deftestes mass, dry
## resp_deftotal length
## resp_deftotal offspring
## resp_defwing centroid
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

#full random effects model looking at all the response units in the entire dataset
response_units_mod <- rma.mv(yi, vi, data=dat_MA_ES, mods = ~resp_units,

```

```

        random = ~1 | experiment_id/ study_id /response_id,
        method="REML")
response_units_mod

```

```

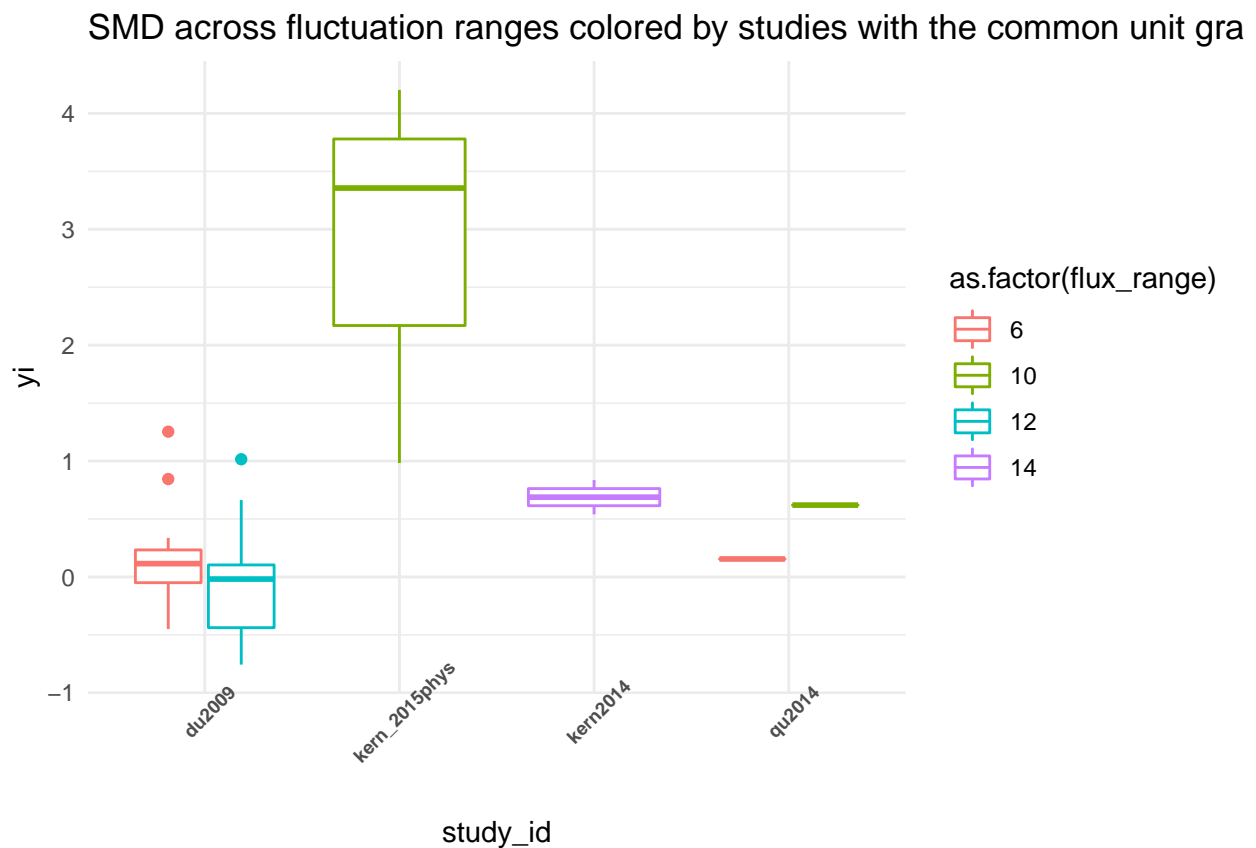
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##          estim      sqrt  nlvls  fixed                                factor
## sigma^2.1  0.0000  0.0004      3    no                                experiment_id
## sigma^2.2  13.2323  3.6376     23    no                                experiment_id/study_id
## sigma^2.3   0.1505  0.3880     58    no  experiment_id/study_id/response_id
##
## Test for Residual Heterogeneity:
## QE(df = 183) = 4135.6068, p-val < .0001
##
## Test of Moderators (coefficients 2:20):
## QM(df = 19) = 53.7881, p-val < .0001
##
## Model Results:
##
##                                     estimate      se      zval      pval
## intrcpt                           1.5220  3.6742   0.4142  0.6787
## resp_unitsCFU * g dry weight manure^-1 -0.8243  5.1911  -0.1588  0.8738
## resp_unitscm                       -2.7028  3.8344  -0.7049  0.4809
## resp_unitdays                     -2.7310  3.8135  -0.7161  0.4739
## resp_unitseggs laid                 -1.5013  4.4935  -0.3341  0.7383
## resp_unitsg                        -1.8582  3.8095  -0.4878  0.6257
## resp_unitskJ *day^-1 *kg^-1         -2.9796  3.8520  -0.7735  0.4392
## resp_unitsm                       -3.2358  3.8333  -0.8441  0.3986
## resp_unitsm * s^-1                  -3.0503  3.8323  -0.7959  0.4261
## resp_unitsmg                      -1.5573  5.1850  -0.3003  0.7639
## resp_unitsmm                      -1.7160  3.8152  -0.4498  0.6529
## resp_unitsmol CHE / mgww           -1.8189  5.1878  -0.3506  0.7259
## resp_unitsoffspring per mating      -1.0426  4.2440  -0.2457  0.8059
## resp_unitsoffspring/female          -2.6804  3.8516  -0.6959  0.4865
## resp_unitspercent                  -3.3689  3.8119  -0.8838  0.3768
## resp_unitspixels                   -1.6977  3.8539  -0.4405  0.6596
## resp_unitstime to death (hour)      -2.7500  3.8383  -0.7165  0.4737
## resp_unitstotal offspring           -0.9791  4.2440  -0.2307  0.8175
## resp_unitsuM Trolox Equivalents/ mgww -4.1911  5.1903  -0.8075  0.4194
## resp_unitswing centroid             -0.6880  4.2381  -0.1623  0.8710
##                                     ci.lb   ci.ub
## intrcpt                           -5.6793  8.7232
## resp_unitsCFU * g dry weight manure^-1 -10.9987  9.3502
## resp_unitscm                      -10.2181  4.8125
## resp_unitdays                     -10.2053  4.7433
## resp_unitseggs laid                 -10.3083  7.3057
## resp_unitsg                        -9.3247  5.6083
## resp_unitskJ *day^-1 *kg^-1         -10.5293  4.5701
## resp_unitsm                       -10.7490  4.2773
## resp_unitsm * s^-1                 -10.5613  4.4608

```



```
## resp_unitmsg -11.7197 8.6052
## resp_unitmm -9.1937 5.7618
## resp_unitnmol CHE / mgww -11.9868 8.3489
## resp_unitoffspring per mating -9.3607 7.2755
## resp_unitoffspring/female -10.2294 4.8685
## resp_unitspercent -10.8401 4.1024
## resp_unitpixels -9.2513 5.8559
## resp_unitstime to death (hour) -10.2730 4.7730
## resp_unitstotal offspring -9.2973 7.3390
## resp_unitsuM Trolox Equivalents/ mgww -14.3640 5.9818
## resp_unitswing centroid -8.9945 7.6185
##
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#plot across studies with common response variable body mass
ggplot(common_unit_g, aes(y=yi, color=as.factor(flux_range), x = study_id))+
  geom_boxplot()+
  theme_minimal()+
  theme(axis.text.x = element_text(face = "bold",
                                    size = 7, angle = 45))+
  ggtitle("SMD across fluctuation ranges colored by studies with the common unit grams")
```



```
#random effects model within subset data
fig14 <- rma.mv(yi, vi, data=common_unit_g,
```

```

random = ~1 | experiment_id/ study_id,
method="REML")
fig14

```

```

##
## Multivariate Meta-Analysis Model (k = 31; method: REML)
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.4301  0.6558     2     no      experiment_id
## sigma^2.2  0.4115  0.6415     5     no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 30) = 112.7702, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## 0.4964  0.5770  0.8603  0.3896  -0.6345  1.6273
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

#simple linear model looking at how flux_range affects yi within subset data
simple14<-lm(yi~study_id, data =common_unit_g)
summary(simple14)

```

```

##
## Call:
## lm(formula = yi ~ study_id, data = common_unit_g)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.86411 -0.28023 -0.05915  0.19022  1.35561
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.05915    0.13247   0.447   0.659
## study_idkern_2015phys 2.78814    0.39742  7.016 1.53e-07 ***
## study_idkern2014      0.62940    0.47764  1.318   0.199
## study_idqu2014       0.32825    0.47764  0.687   0.498
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.649 on 27 degrees of freedom
## Multiple R-squared:  0.6478, Adjusted R-squared:  0.6087
## F-statistic: 16.56 on 3 and 27 DF,  p-value: 2.655e-06

```