

data wrangling and plots

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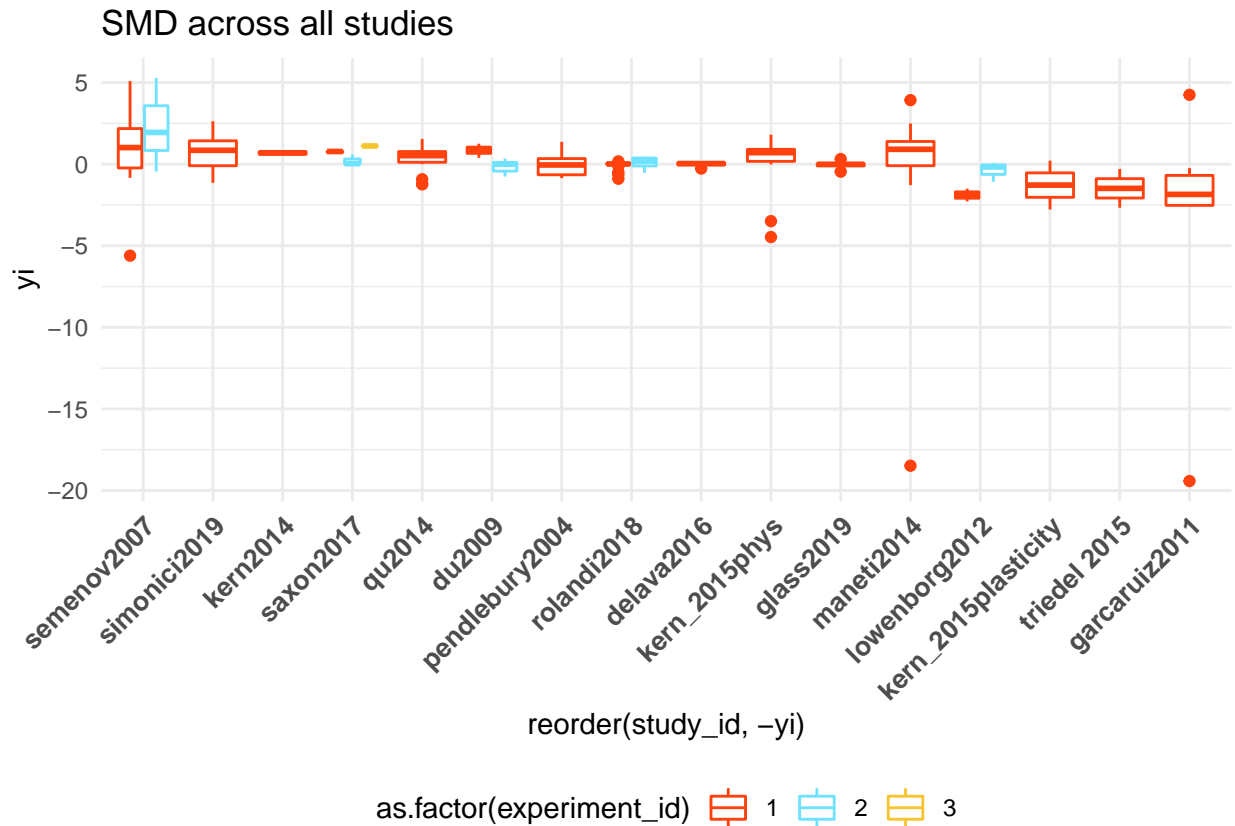
2/22/2021

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")
```



```
#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  0.0015     3    no      experiment_id
## sigma^2.2 11.1645  3.3413    23    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 202) = 6886.4086, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb  ci.ub
## -0.5649  0.6981 -0.8092  0.4184 -1.9331  0.8033
##
## ---
## 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): 96.33%
## H2 (unaccounted variability / sampling variability): 27.22
##
## Test for Residual Heterogeneity:
## QE(df = 186) = 5063.4195, p-val < .0001
##
## Test of Moderators (coefficients 2:17):
## QM(df = 16) = 1822.9891, 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  0.4404  0.0548   8.0347 <.0001   0.3330
## 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  0.5479 ***
## 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 ***
```

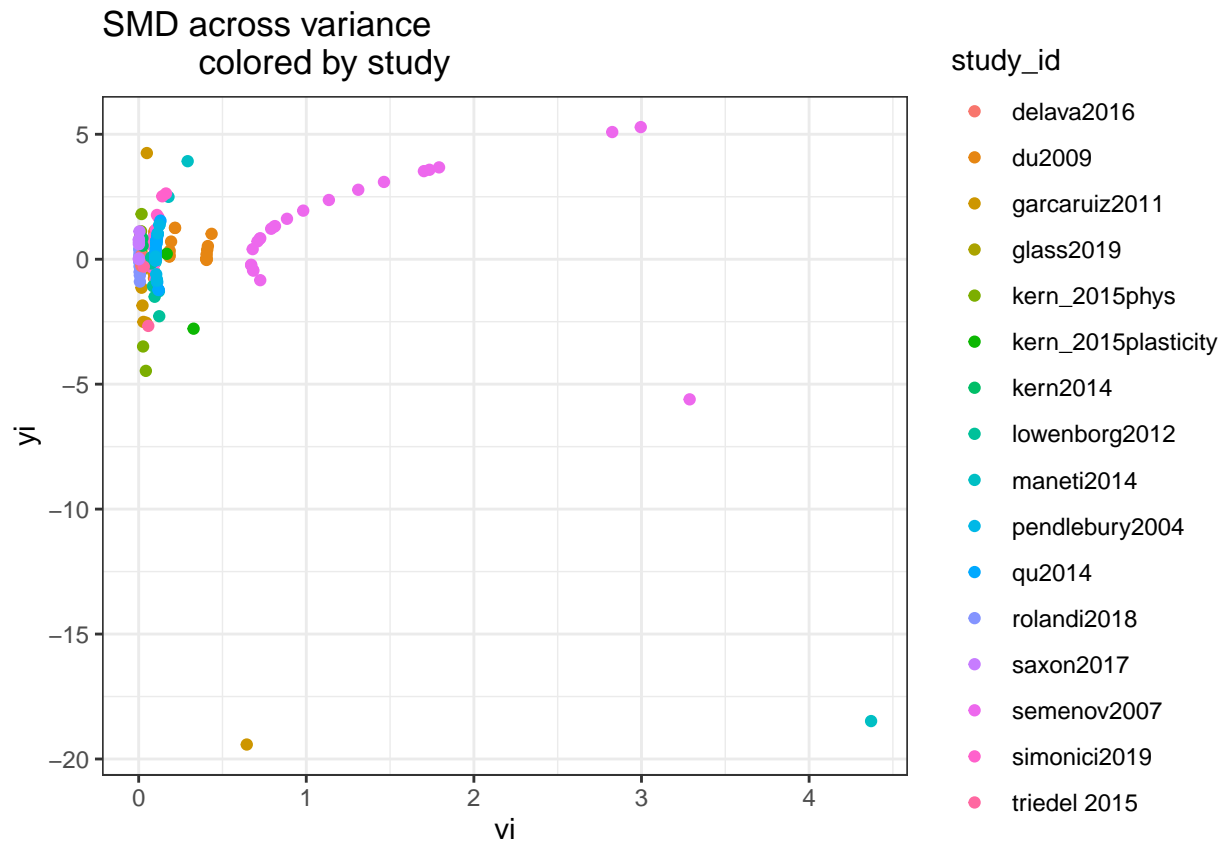
```
## study_idsimonici2019          0.7611 ***
## study_idtriedel 2015         -0.8339 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#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
## -2236.2768    4472.5536    4478.5536    4488.4784    4478.6748
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1  0.0000  0.0015     3    no      experiment_id
## sigma^2.2  11.1645  3.3413    23    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 202) = 6886.4086, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.5649  0.6981 -0.8092  0.4184 -1.9331  0.8033
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Figure 2.

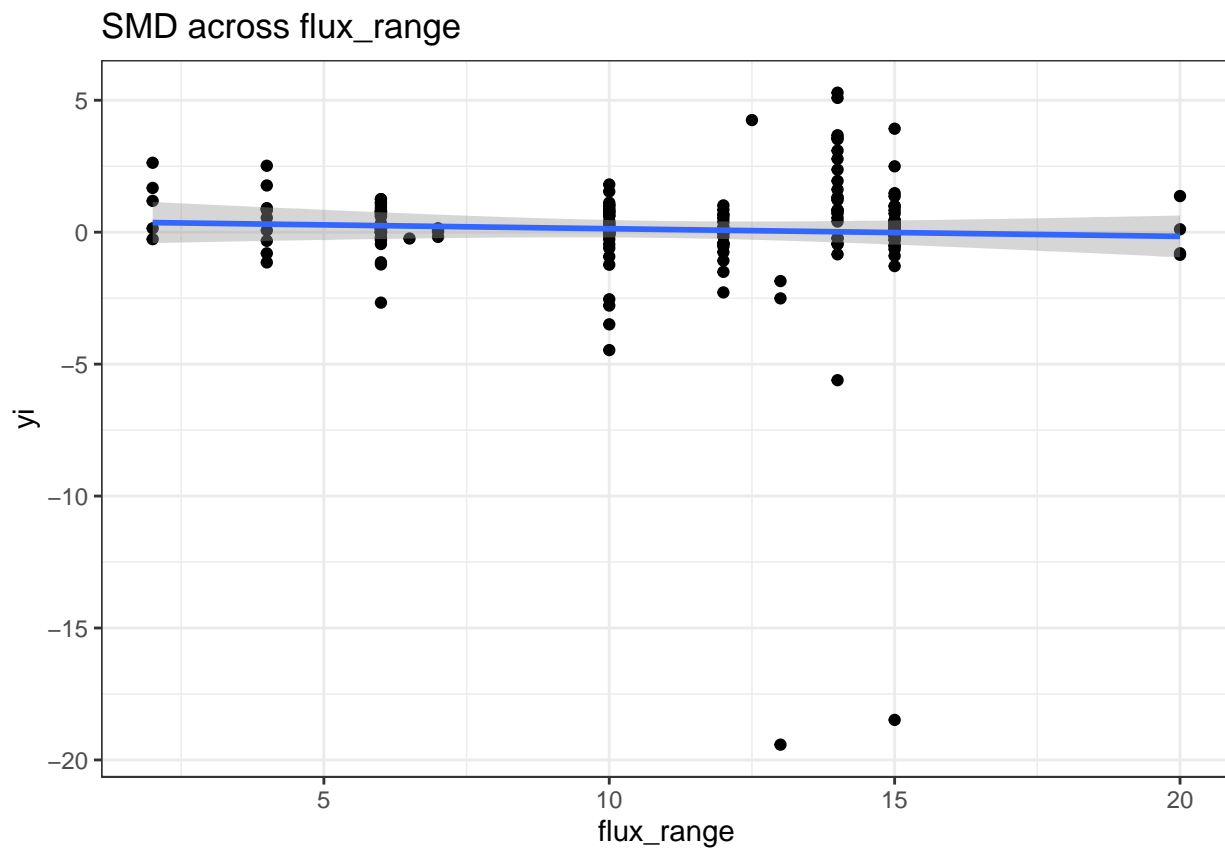


```
##
## Call:
## lm(formula = yi ~ vi, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.4410  -1.0808  -0.6359   0.4524  18.9554
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.1163     0.2675   4.173 4.48e-05 ***
## vi           -4.9353     0.2441 -20.219 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.662 on 201 degrees of freedom
## Multiple R-squared:  0.6704, Adjusted R-squared:  0.6688
## F-statistic: 408.8 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=flux_range, y=yi))+
  geom_point()+
  theme_bw()+
  geom_smooth(method="lm", formula = y~x)+
  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  0.0005     3    no  experiment_id
```

```
## sigma^2.2 10.8960 3.3009 23 no experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 6383.8992, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 18.2908, p-val < .0001
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      -0.0755  0.6991  -0.1080  0.9140  -1.4456   1.2947
## flux_range   -0.0466  0.0109  -4.2768  <.0001  -0.0679  -0.0252 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

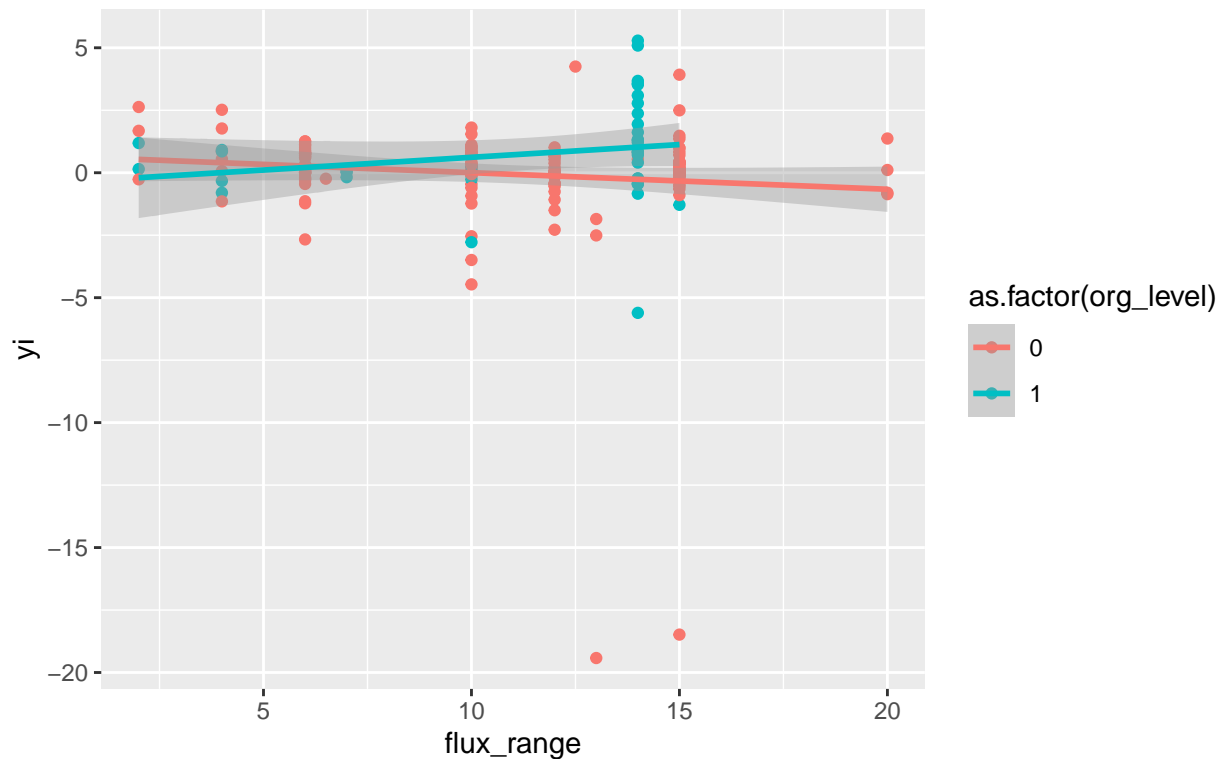
```
#simple linear model looking at how flux_range affects yi
simple7<-lm(yi~flux_range, data =dat_MA_ES)
summary(simple7)
```

```
##
## Call:
## lm(formula = yi ~ flux_range, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.316   0.044   0.656   1.147   5.976
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.7115     1.2885   0.552   0.581
## flux_range   -0.1001     0.1104  -0.906   0.366
##
## Residual standard error: 6.366 on 201 degrees of freedom
## Multiple R-squared:  0.004071, Adjusted R-squared:  -0.0008838
## F-statistic: 0.8216 on 1 and 201 DF, p-value: 0.3658
```

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  0.0012     3    no      experiment_id
## sigma^2.2 10.5105  3.2420    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 200) = 6318.2410, p-val < .0001
##
## Test of Moderators (coefficients 2:3):
## QM(df = 2) = 45.2966, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      0.0262  0.6873   0.0381  0.9696   -1.3208    1.3732
## flux_range  -0.0466  0.0109  -4.2821 <.0001   -0.0680   -0.0253 ***
```



```
## org_level    -0.4311  0.0830  -5.1950  <.0001  -0.5938  -0.2685  ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

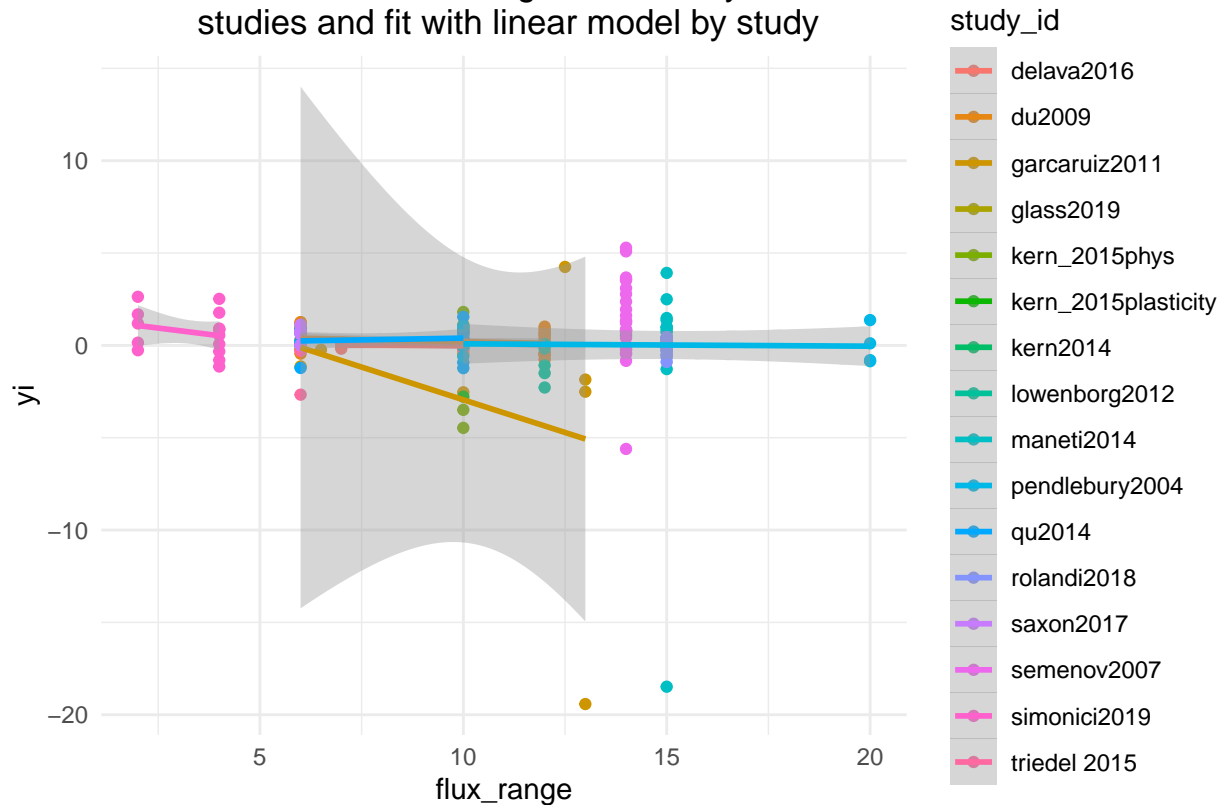
```
#simple linear model looking at how flux_range and org_level affect yi
simple8<-lm(yi~flux_range*org_level, data =dat_MA_ES)
summary(simple8)
```

```
##
## Call:
## lm(formula = yi ~ flux_range * org_level, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.053  -0.047   0.757   1.243   5.043
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.8933     1.4309   0.624   0.533
## flux_range      -0.1343     0.1239  -1.084   0.280
## org_level       -0.6589     3.3438  -0.197   0.844
## flux_range:org_level  0.1428     0.2779   0.514   0.608
##
## Residual standard error: 6.382 on 199 degrees of freedom
## Multiple R-squared:  0.008982,    Adjusted R-squared:  -0.005958
## F-statistic: 0.6012 on 3 and 199 DF,  p-value: 0.615
```

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) = 5046.4092, p-val < .0001
##
## Test of Moderators (coefficients 2:18):
## QM(df = 17) = 322.5804, 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  0.5094  0.8664   0.5880  0.5566  -1.1887
## 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  2.2075
## 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   0.539  22.869
##
## Coefficients: (11 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.091100  11.511151   0.008   0.9937
```

```

## flux_range -0.010684 1.333646 -0.008 0.9936
## study_iddu2009 0.575435 11.937810 0.048 0.9616
## study_idgarcaruiz2011 22.254823 13.975852 1.592 0.1131
## study_idglass2019 -0.020693 3.464914 -0.006 0.9952
## study_idkern_2015phys -0.010713 3.266753 -0.003 0.9974
## study_idkern_2015plasticity -1.266498 4.900129 -0.258 0.7963
## study_idkern2014 0.747017 8.591415 0.087 0.9308
## study_idlowenborg2012 -0.955077 5.673877 -0.168 0.8665
## study_idmaneti2014 -0.530046 9.045229 -0.059 0.9533
## study_idpendlebury2004 0.119910 13.134892 0.009 0.9927
## study_idpiccau2017 -16.120645 10.543399 -1.529 0.1280
## study_idqu2014 -0.077323 12.638013 -0.006 0.9951
## study_idrolandi2018 0.087188 8.943754 0.010 0.9922
## study_idsaxon2017 0.543332 4.437542 0.122 0.9027
## study_idsemenov2007 1.440235 7.693949 0.187 0.8517
## study_idsimonici2019 1.535025 12.701186 0.121 0.9039
## study_idtriedel 2015 -1.510015 5.579042 -0.271 0.7870
## flux_range:study_iddu2009 -0.035825 1.374691 -0.026 0.9792
## flux_range:study_idgarcaruiz2011 -3.266636 1.508620 -2.165 0.0317 *
## 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 1.392368 -0.002 0.9988
## flux_range:study_idpiccau2017 NA NA NA NA
## flux_range:study_idqu2014 0.048564 1.476076 0.033 0.9738
## 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 2.044441 -0.129 0.8972
## flux_range:study_idtriedel 2015 NA NA NA NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.658 on 180 degrees of freedom
## Multiple R-squared: 0.2953, Adjusted R-squared: 0.2092
## F-statistic: 3.429 on 22 and 180 DF, p-value: 2.439e-06

```

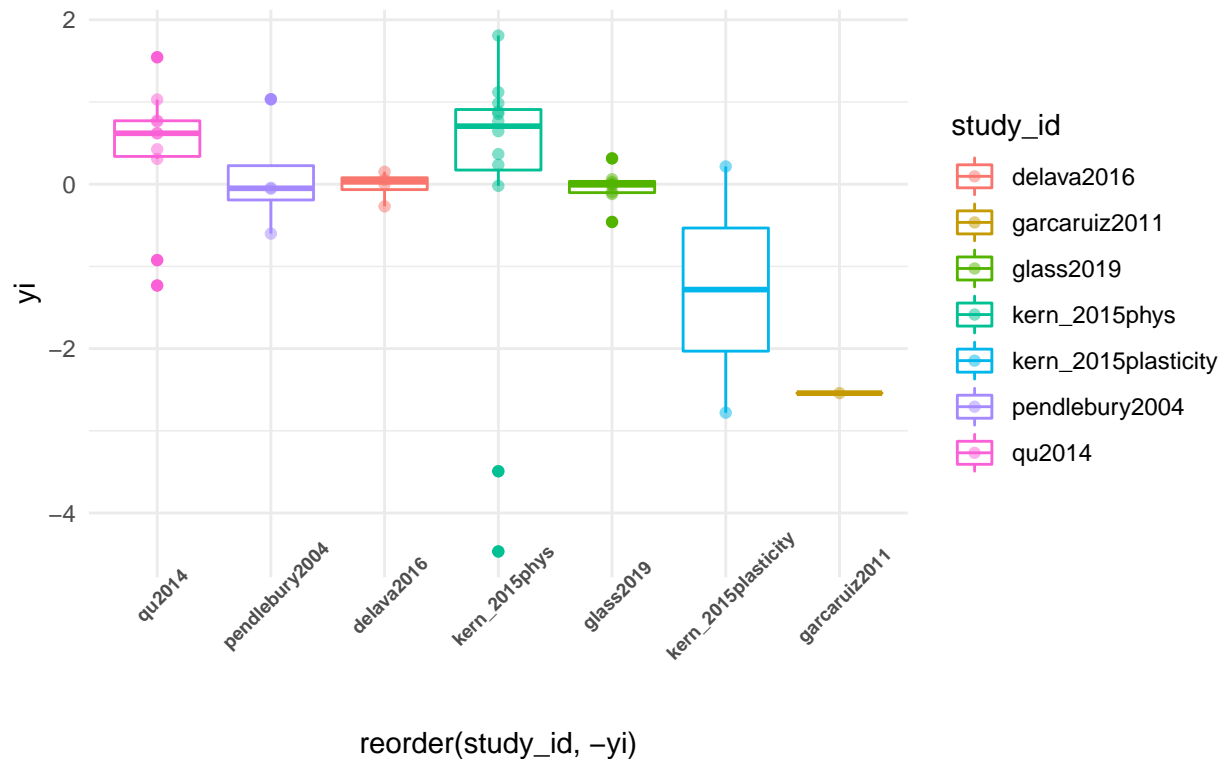
Figure 6.

```

#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()+
  geom_point(alpha = 0.5)+
  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 1.0453 1.0224    7   no      experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 40) = 1793.3191, p-val < .0001
##
## Model Results:
##
```

```
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.3405  0.3914 -0.8699  0.3843 -1.1078  0.4267
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

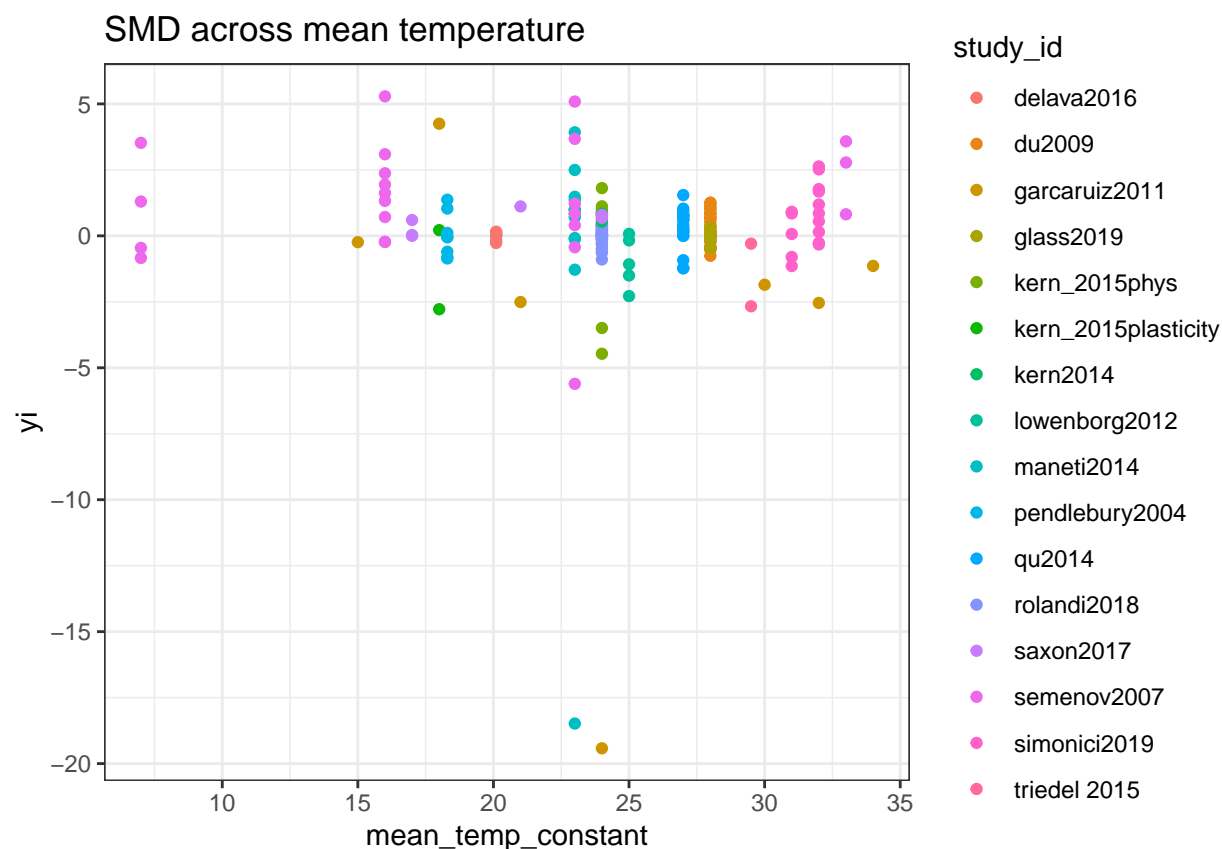
#simple linear model looking at how flux_range affects yi within subset data
simple12<-lm(yi~flux_range, data =common_range)
summary(simple12)

##
## Call:
## lm(formula = yi ~ flux_range, data = common_range)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4306 -0.0600  0.1875  0.7992  1.8452
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.03707    0.19822  -0.187   0.853
## flux_range      NA           NA      NA      NA
##
## Residual standard error: 1.269 on 40 degrees of freedom
```

Question how does mean temperature affect response variables?

Figure 7.

```
# scatterplot of how mean temperature influences SMD
ggplot(normalized, aes(x=mean_temp_constant, y=yi, color = study_id))+
  geom_point()+
  theme_bw()+
  ggtitle("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  0.0008     3    no      experiment_id
## sigma^2.2 12.8795  3.5888    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 6863.7808, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 112.3388, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt          1.4214  0.7728   1.8393  0.0659  -0.0932   2.9359
## mean_temp_constant -0.0863  0.0081 -10.5990 <.0001  -0.1022  -0.0703 ***
```

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

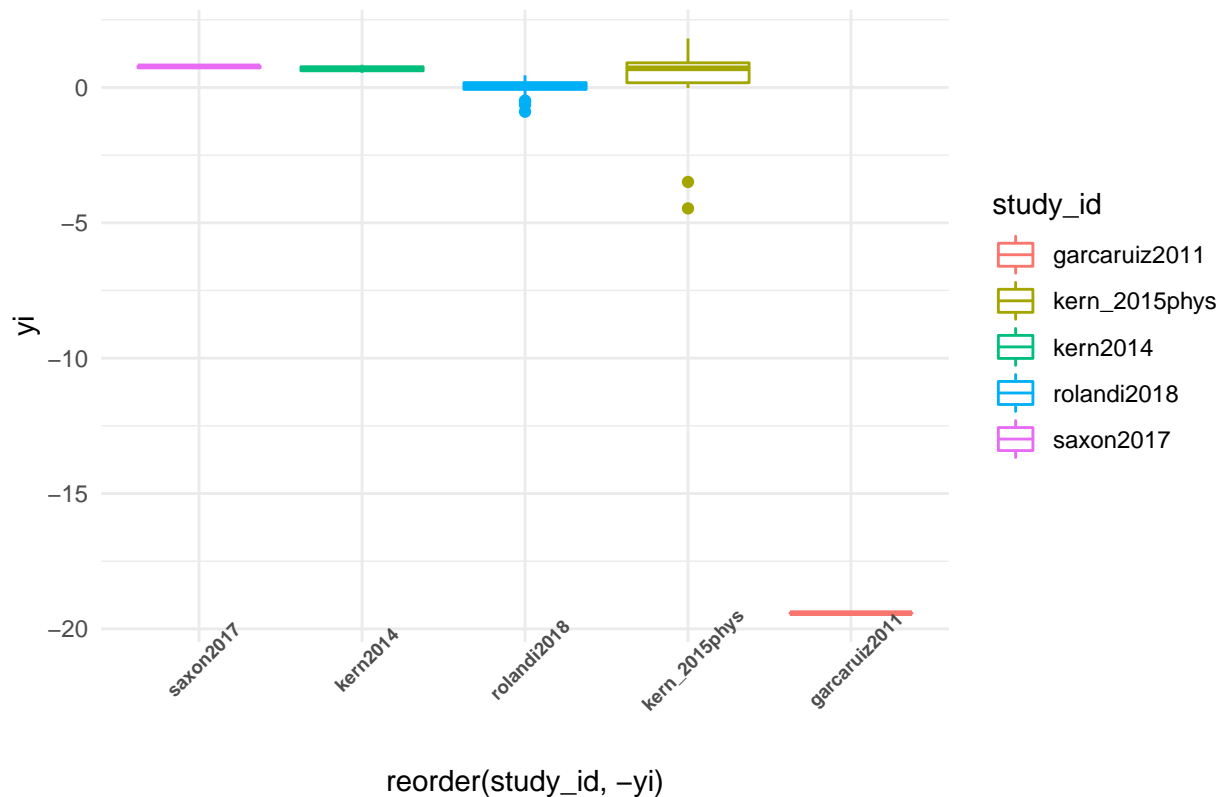
#simple linear model looking at how mean_temp_constant affects yi
simple6<-lm(yi~mean_temp_constant, data =dat_MA_ES)
summary(simple6)

##
## Call:
## lm(formula = yi ~ mean_temp_constant, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.499   0.216   0.511   1.180   5.597
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.170886    2.286937  -0.075    0.941
## mean_temp_constant -0.008699    0.091544  -0.095    0.924
##
## Residual standard error: 6.378 on 201 degrees of freedom
## Multiple R-squared:  4.492e-05, Adjusted R-squared:  -0.00493
## F-statistic: 0.00903 on 1 and 201 DF,  p-value: 0.9244
```

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  0.0052    2    no      experiment_id
## sigma^2.2 64.5106  8.0318    6    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 55) = 3403.3975, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -2.8890    3.2818   -0.8803  0.3787   -9.3211    3.5431
##
## ---
## 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)
```

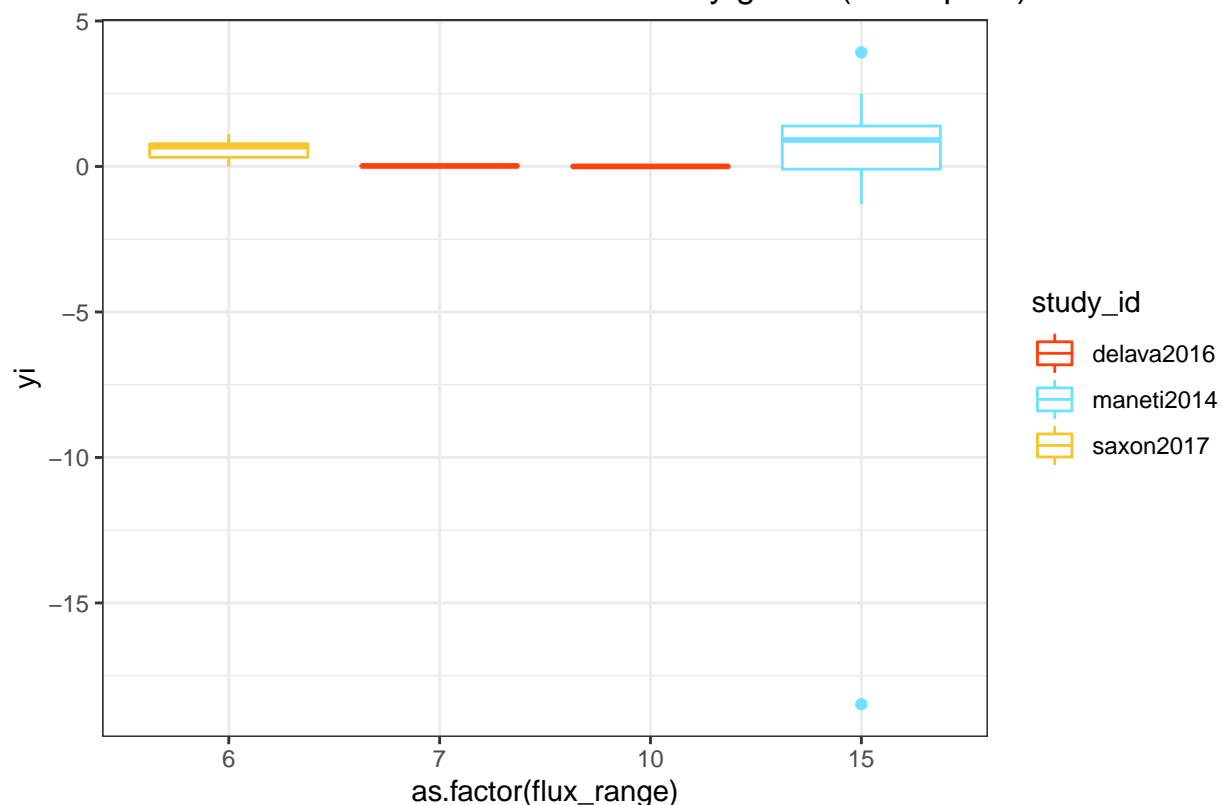
```
##
## Call:
## lm(formula = yi ~ mean_temp_constant, data = common_temp)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.1451   0.2517   0.3356   0.7096   2.0837
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.2755     0.3690  -0.747   0.458
## mean_temp_constant      NA         NA      NA     NA
##
## Residual standard error: 2.761 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
```

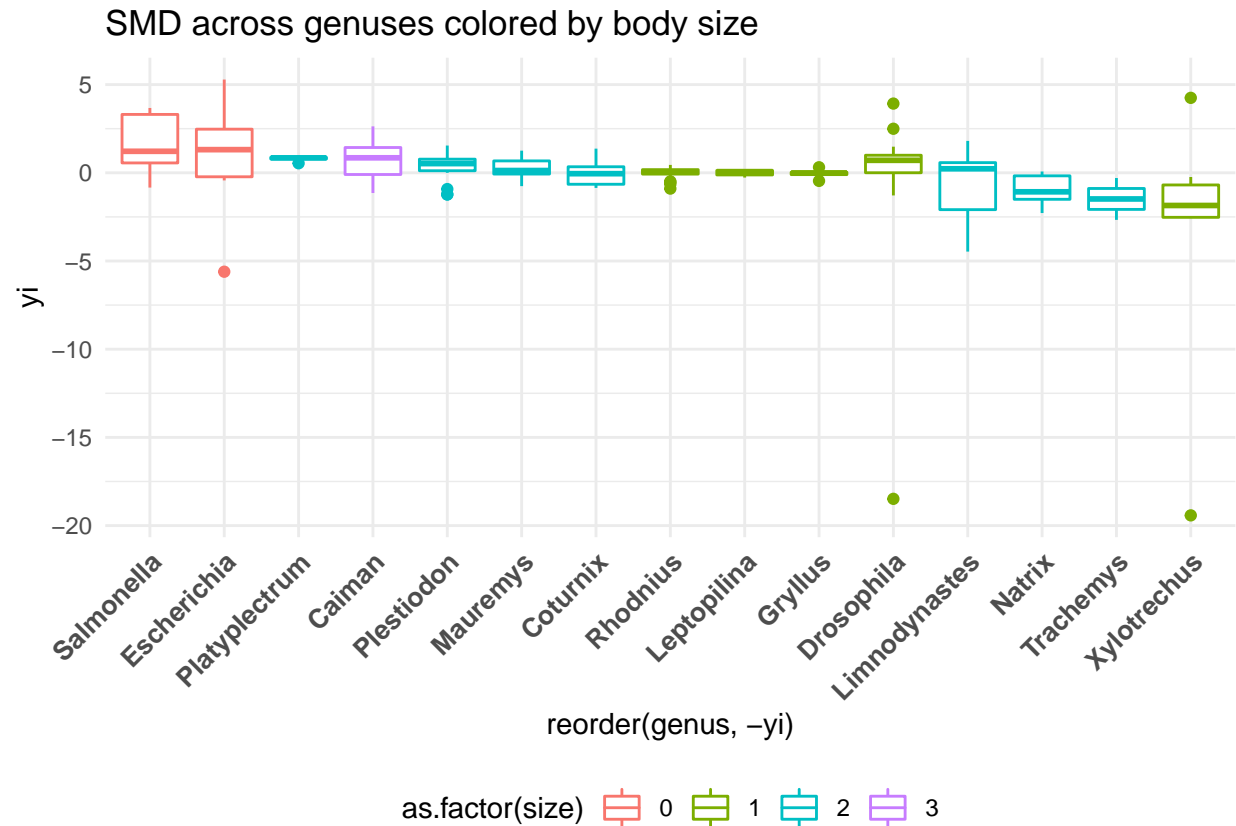
```
## study_idmaneti2014    0.7499  0.6134   1.2226  0.2215  -0.4523  1.9521
## study_idsaxon2017    0.6933  0.4505   1.5389  0.1238  -0.1897  1.5763
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#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")
```



```
#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  0.2888  0.5374    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 187) = 4957.0842, p-val < .0001
##
## Test of Moderators (coefficients 2:16):
## QM(df = 15) = 487.2880, p-val < .0001
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt          0.5791  0.5436   1.0653  0.2868   -0.4863   1.6445
## genusClematis    -16.7689  1.0710  -15.6574 <.0001  -18.8679  -14.6698 ***
```

```
## genusCoturnix      -0.5885  0.7731  -0.7612  0.4465  -2.1037  0.9267
## genusDrosophila    -0.0186  0.5952  -0.0312  0.9751  -1.1851  1.1480
## genusEscherichia    0.5307  0.7317   0.7252  0.4683  -0.9035  1.9649
## genusGryllus       -0.6144  0.7657  -0.8023  0.4224  -2.1151  0.8864
## genusLeptopilina   -0.0465  0.6013  -0.0773  0.9384  -1.2250  1.1320
## genusLimnodynastes -0.7919  0.6358  -1.2455  0.2130  -2.0382  0.4543
## genusMauremys      -0.2598  0.6666  -0.3898  0.6967  -1.5662  1.0466
## genusNatrix        -1.6507  0.6779  -2.4351  0.0149  -2.9793  -0.3221  *
## genusPlatyplectrum -0.0106  0.6357  -0.0167  0.9867  -1.2566  1.2353
## genusPlestiodon    -0.2612  0.7679  -0.3401  0.7337  -1.7662  1.2438
## genusRhodnius      -0.5584  0.6634  -0.8417  0.4000  -1.8585  0.7418
## genusSalmonella     0.4786  0.7238   0.6612  0.5085  -0.9401  1.8972
## genusTrachemys     -1.7031  0.7774  -2.1909  0.0285  -3.2267  -0.1795  *
## genusXylotrechus   -1.6640  0.7672  -2.1690  0.0301  -3.1676  -0.1604  *
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

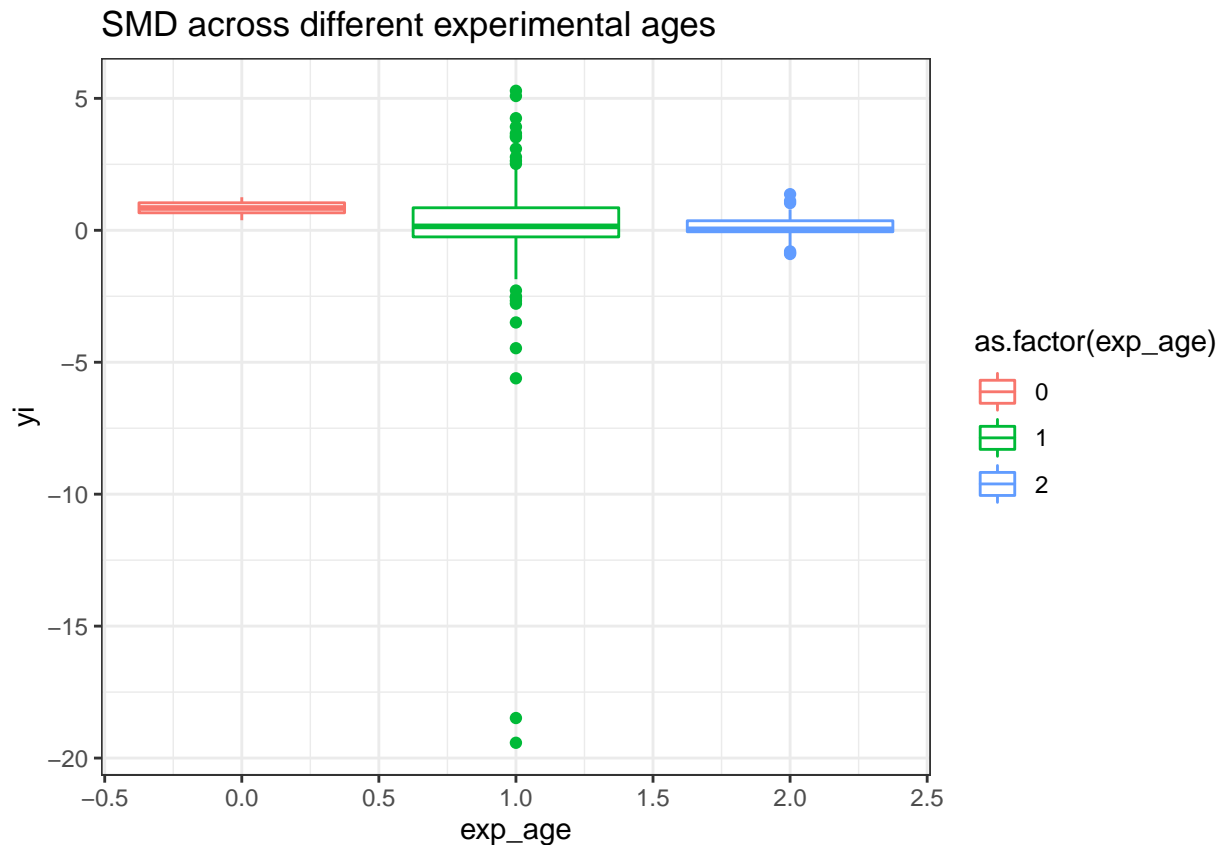
```
#simple linear model looking at how genus affects yi
simple11<-lm(yi~genus, data =dat_MA_ES)
summary(simple11)
```

```
##
## Call:
## lm(formula = yi ~ genus, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -69.610  -0.274   0.057   0.608  17.544
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.7084     1.5173   0.467  0.64111
## genusClematis  -16.8982     6.0691  -2.784  0.00592 **
## genusCoturnix   -0.6892     2.5727  -0.268  0.78907
## genusDrosophila -0.8597     1.9866  -0.433  0.66568
## genusEscherichia  0.4926     2.2759   0.216  0.82890
## genusGryllus    -0.7449     2.5727  -0.290  0.77250
## genusLeptopilina -0.7115     2.8386  -0.251  0.80235
## genusLimnodynastes -1.3455     2.3990  -0.561  0.57558
## genusMauremys   -0.4605     1.8388  -0.250  0.80254
## genusNatrix     -1.7006     3.0345  -0.560  0.57587
## genusPlatyplectrum  0.1025     2.8386   0.036  0.97123
## genusPlestiodon -0.3916     2.0072  -0.195  0.84552
## genusRhodnius   -0.6904     1.7919  -0.385  0.70046
## genusSalmonella  0.8706     2.3327   0.373  0.70942
## genusTrachemys  -2.1914     4.4236  -0.495  0.62090
## genusXylotrechus -14.0034     2.5727  -5.443 1.63e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.876 on 187 degrees of freedom
## Multiple R-squared:  0.2104, Adjusted R-squared:  0.147
## F-statistic: 3.321 on 15 and 187 DF, p-value: 6.022e-05
```

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  0.0004     3    no      experiment_id
## sigma^2.2  9.3934  3.0649    23    no  experiment_id/study_id
```

```
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 6860.5383, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 5.1960, p-val = 0.0226
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt    -3.6196  1.4875  -2.4333  0.0150  -6.5350  -0.7041  *
## exp_age     2.5997  1.1405   2.2795  0.0226   0.3644   4.8350  *
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

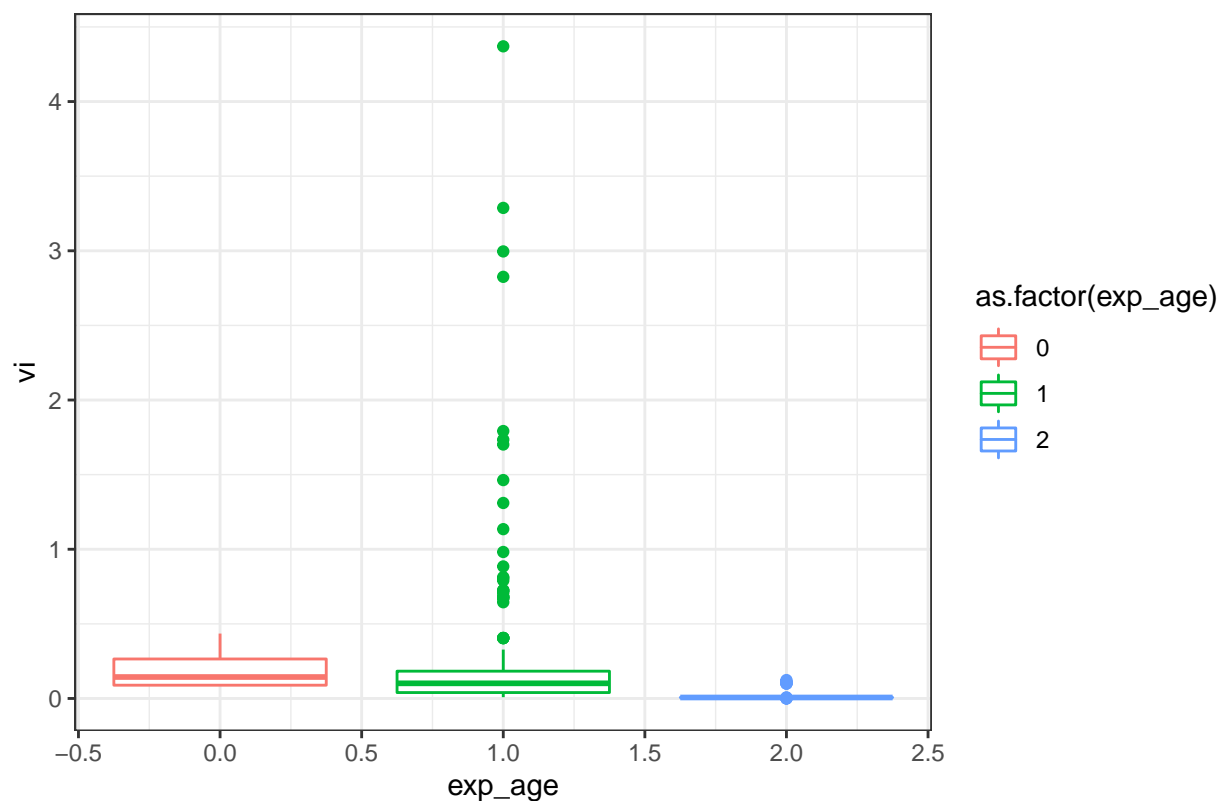
```
#simple linear model looking at how exp_age affects yi
simple3<-lm(yi~exp_age, data =dat_MA_ES)
summary(simple3)
```

```
##
## Call:
## lm(formula = yi ~ exp_age, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -82.433   0.032   0.472   1.306   5.759
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.9213     1.0963  -0.840   0.402
## exp_age       0.4489     0.8361   0.537   0.592
##
## Residual standard error: 6.374 on 201 degrees of freedom
## Multiple R-squared:  0.001432,    Adjusted R-squared:  -0.003536
## F-statistic: 0.2882 on 1 and 201 DF,  p-value: 0.5919
```

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 12
## 2      1 136
## 3      2  53
```

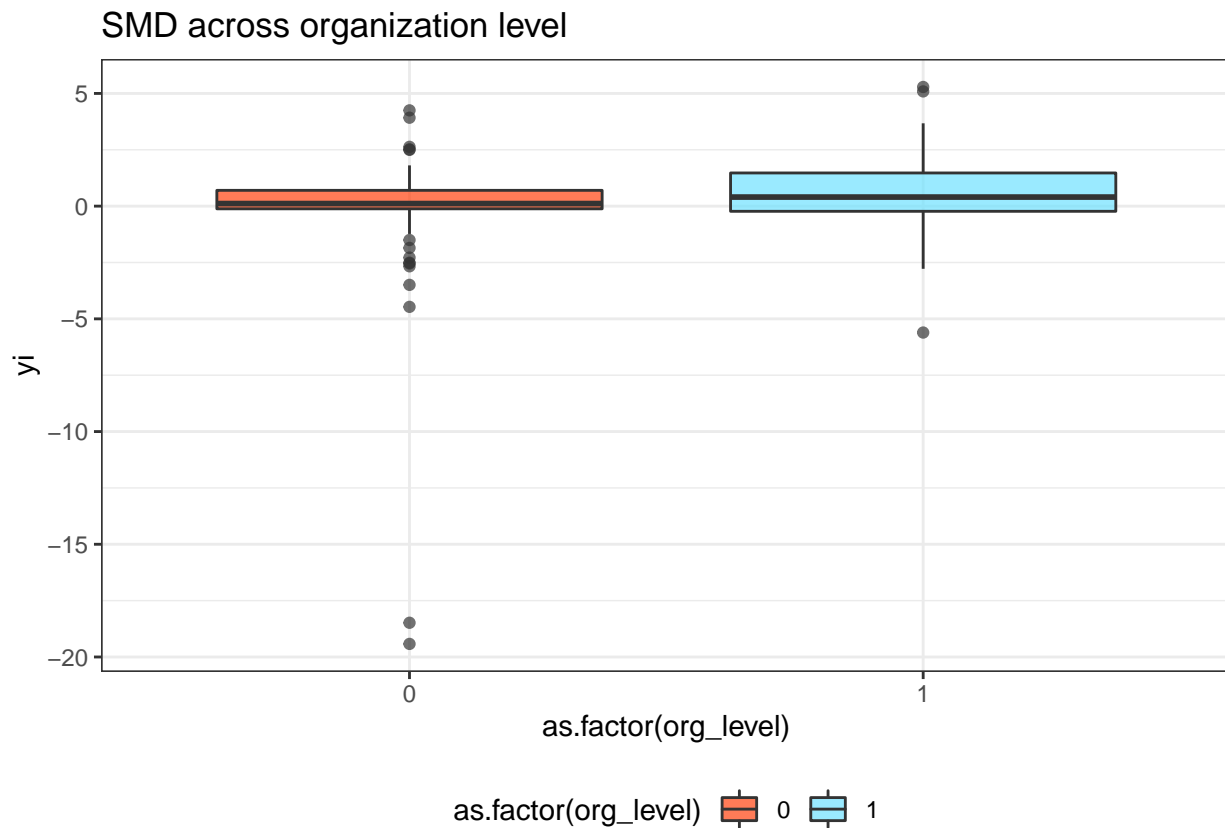
```
#simple linear model looking at how exp_age affects vi
simple4<-lm(vi~ exp_age, data =dat_MA_ES)
summary(simple4)
```

```
##
## Call:
## lm(formula = vi ~ exp_age, data = dat_MA_ES)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.5065 -0.2682 -0.2284 -0.1036 12.7339
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.5927     0.1806   3.281 0.00122 **
## exp_age      -0.2412     0.1378  -1.751 0.08155 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 1.05 on 201 degrees of freedom
## Multiple R-squared:  0.01502,    Adjusted R-squared:  0.01012
## F-statistic: 3.064 on 1 and 201 DF,  p-value: 0.08155
```

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  0.0006     3    no      experiment_id
## sigma^2.2 10.7655  3.2811    23    no  experiment_id/study_id
##
## Test for Residual Heterogeneity:
## QE(df = 201) = 6852.7599, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 26.9418, p-val < .0001
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      -0.4639  0.6858  -0.6764  0.4988  -1.8081   0.8803
## org_level     -0.4308  0.0830  -5.1906 <.0001  -0.5934  -0.2681 ***
##
## ---
## 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   0.131   0.620   1.254   4.954
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.5598     0.4988  -1.122   0.263
## org_level      0.8921     1.1237   0.794   0.428
##
## Residual standard error: 6.369 on 201 degrees of freedom
## Multiple R-squared:  0.003126, Adjusted R-squared:  -0.001834
## F-statistic: 0.6303 on 1 and 201 DF, p-value: 0.4282
```

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  0.0006     3    no          experiment_id
## sigma^2.2  0.7207  0.8489    23    no      experiment_id/study_id
## sigma^2.3  0.0366  0.1913    58    no  experiment_id/study_id/response_id
##
## Test for Residual Heterogeneity:
## QE(df = 156) = 4319.5235, p-val < .0001
##
## Test of Moderators (coefficients 2:47):
## QM(df = 46) = 421.7530, p-val < .0001
##
## Model Results:
##
##                                     estimate      se
## intrcpt                           0.3899  0.5166
## resp_defaquatic speed              -1.4687  1.0533
## resp_defaverage cumulative number of eggs laid per female -0.2757  1.0122
## resp_defaverage number of eggs laid per female          -0.4628  1.0122
## resp_defbody (centroid) size        -0.0128  0.8984
## resp_defbody length                 -0.1800  0.3808
## resp_defbody mass                   0.0850  0.3733
## resp_defcarapace height              0.1119  0.5218
## resp_defcarapace width               0.1797  0.5223
## resp_defdaily energy expenditure     -0.9579  1.0249
## resp_defdays to first slough        -2.6718  1.0709
## resp_defdessication tolerance       -0.7060  0.8906
## resp_defdevelopment time            -0.2640  0.7713
## resp_defdevelopment to stages 35-37 -1.4180  0.4312
## resp_defdevelopmental time           0.0554  0.8396
## resp_defdistance covered            -0.5621  1.0483
## resp_defegg mass                    0.1754  1.0253
## resp_defegg to adult viability       -2.4873  0.8883
## resp_deffore-limb length             0.8910  0.4270
## resp_defgermination                -16.5797  1.2597
## resp_defhatching success            0.0392  1.0276
## resp_defhead length                  0.2518  0.4203
## resp_defhead width                   0.5433  0.4210
## resp_defhind-limb length             0.3872  0.4194
## resp_defincubation period            -0.0271  1.0266
## resp_defincubation time              -1.8923  1.0581
## resp_definfestation rate             -0.3500  0.7712
## resp_defmass                        1.0584  1.0323
## resp_defmaximal length               -1.4584  0.4277
## resp_defoffspring per mating          0.0835  0.7296
## resp_defovary mass, dry              -0.4085  1.0140
## resp_defoxidative damage             -0.6869  1.0270
## resp_defpercent females              -0.1973  1.0265
## resp_defproductivity                 -0.9955  0.8882
## resp_defrate of change                1.1321  1.0681

```

## resp_defsnout-vent length	0.6062	0.4218
## resp_defsprint speed	-0.6694	0.4196
## resp_defstartvation tolerance	-1.4144	0.8873
## resp_defsuccess of parasitism	-0.6387	0.7713
## resp_defsurvival	-0.4057	0.7342
## resp_defTAC	-3.0590	1.0399
## resp_deftail length	0.1884	0.3749
## resp_defterrestrial speed	-0.3156	1.0482
## resp_deftestes mass, dry	-0.4427	1.0141
## resp_deftotal length	0.3417	1.0302
## resp_deftotal offspring	0.1470	0.7297
## resp_defwing centroid	0.4437	0.7212
##	zval	pval
## intrcpt	0.7548	0.4504
## resp_defaquatic speed	-1.3944	0.1632
## resp_defaverage cumulative number of eggs laid per female	-0.2723	0.7854
## resp_defaverage number of eggs laid per female	-0.4573	0.6475
## resp_defbody (centroid) size	-0.0142	0.9886
## resp_defbody length	-0.4727	0.6365
## resp_defbody mass	0.2278	0.8198
## resp_defcarapace height	0.2145	0.8302
## resp_defcarapace width	0.3440	0.7308
## resp_defdaily energy expenditure	-0.9346	0.3500
## resp_defdays to first slough	-2.4950	0.0126
## resp_defdessication tolerance	-0.7928	0.4279
## resp_defdevelopment time	-0.3423	0.7322
## resp_defdevelopment to stages 35-37	-3.2889	0.0010
## resp_defdevelopmental time	0.0660	0.9474
## resp_defdistance covered	-0.5362	0.5918
## resp_defegg mass	0.1711	0.8641
## resp_defegg to adult viability	-2.7999	0.0051
## resp_deffore-limb length	2.0868	0.0369
## resp_defgermination	-13.1616	<.0001
## resp_defhatching success	0.0382	0.9696
## resp_defhead length	0.5992	0.5490
## resp_defhead width	1.2905	0.1969
## resp_defhind-limb length	0.9232	0.3559
## resp_defincubation period	-0.0264	0.9790
## resp_defincubation time	-1.7885	0.0737
## resp_definfestation rate	-0.4538	0.6500
## resp_defmass	1.0253	0.3052
## resp_defmaximal length	-3.4097	0.0007
## resp_defoffspring per mating	0.1145	0.9088
## resp_defovary mass, dry	-0.4029	0.6870
## resp_defoxidative damage	-0.6688	0.5036
## resp_defpercent females	-0.1922	0.8476
## resp_defproductivity	-1.1208	0.2624
## resp_defrate of change	1.0599	0.2892
## resp_defsnout-vent length	1.4371	0.1507
## resp_defsprint speed	-1.5953	0.1106
## resp_defstartvation tolerance	-1.5940	0.1109
## resp_defsuccess of parasitism	-0.8281	0.4076
## resp_defsurvival	-0.5525	0.5806
## resp_defTAC	-2.9415	0.0033

## resp_deftail length	0.5026	0.6153
## resp_defterrestrial speed	-0.3011	0.7633
## resp_deftestes mass, dry	-0.4365	0.6625
## resp_deftotal length	0.3317	0.7401
## resp_deftotal offspring	0.2015	0.8403
## resp_defwing centroid	0.6152	0.5384
##	ci.lb	ci.ub
## intrcpt	-0.6226	1.4025
## resp_defaquatic speed	-3.5331	0.5957
## resp_defaverage cumulative number of eggs laid per female	-2.2595	1.7082
## resp_defaverage number of eggs laid per female	-2.4466	1.5210
## resp_defbody (centroid) size	-1.7737	1.7481
## resp_defbody length	-0.9263	0.5663
## resp_defbody mass	-0.6465	0.8166
## resp_defcarapace height	-0.9108	1.1346
## resp_defcarapace width	-0.8441	1.2035
## resp_defdaily energy expenditure	-2.9667	1.0509
## resp_defdays to first slough	-4.7708	-0.5729
## resp_defdessication tolerance	-2.4515	1.0394
## resp_defdevelopment time	-1.7756	1.2477
## resp_defdevelopment to stages 35-37	-2.2631	-0.5730
## resp_defdevelopmental time	-1.5901	1.7010
## resp_defdistance covered	-2.6167	1.4925
## resp_defegg mass	-1.8341	2.1850
## resp_defegg to adult viability	-4.2284	-0.7462
## resp_deffore-limb length	0.0541	1.7279
## resp_defgermination	-19.0487	-14.1108
## resp_defhatching success	-1.9748	2.0533
## resp_defhead length	-0.5719	1.0756
## resp_defhead width	-0.2819	1.3686
## resp_defhind-limb length	-0.4349	1.2092
## resp_defincubation period	-2.0392	1.9851
## resp_defincubation time	-3.9661	0.1815
## resp_definfestation rate	-1.8616	1.1616
## resp_defmass	-0.9649	3.0818
## resp_defmaximal length	-2.2967	-0.6201
## resp_defoffspring per mating	-1.3465	1.5136
## resp_defovary mass, dry	-2.3959	1.5788
## resp_defoxidative damage	-2.6999	1.3261
## resp_defpercent females	-2.2092	1.8147
## resp_defproductivity	-2.7364	0.7454
## resp_defrate of change	-0.9614	3.2255
## resp_defsnout-vent length	-0.2205	1.4330
## resp_defsprint speed	-1.4918	0.1530
## resp_defstartvation tolerance	-3.1535	0.3247
## resp_defsuccess of parasitism	-2.1504	0.8730
## resp_defsurvival	-1.8447	1.0333
## resp_defTAC	-5.0973	-1.0208
## resp_deftail length	-0.5464	0.9233
## resp_defterrestrial speed	-2.3700	1.7388
## resp_deftestes mass, dry	-2.4302	1.5448
## resp_deftotal length	-1.6774	2.3608
## resp_deftotal offspring	-1.2831	1.5771
## resp_defwing centroid	-0.9698	1.8572

```

##
## 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

```

```

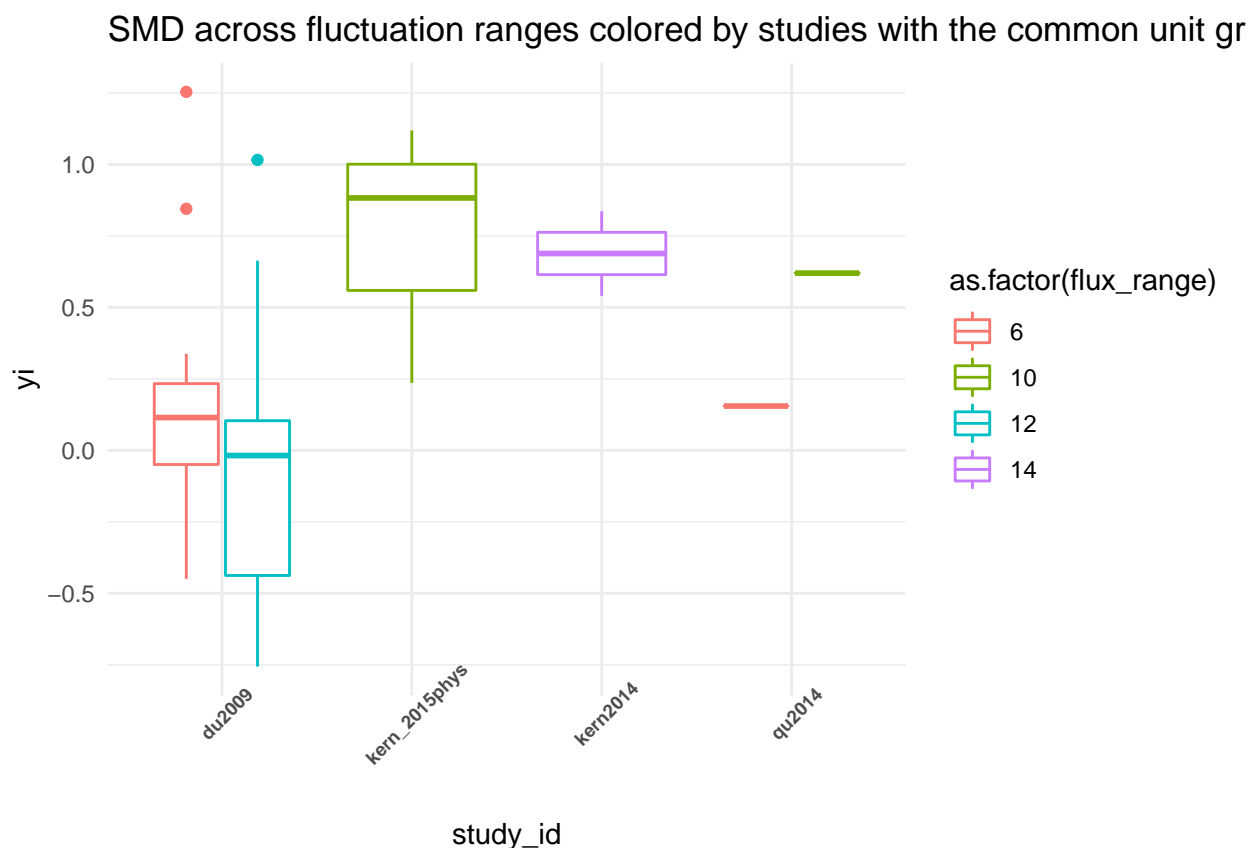
##
## Multivariate Meta-Analysis Model (k = 203; method: REML)
##
## Variance Components:
##
##          estim      sqrt  nlvls  fixed                                factor
## sigma^2.1  0.0000  0.0010     3    no                                experiment_id
## sigma^2.2 12.7637  3.5726    23    no                                experiment_id/study_id
## sigma^2.3  0.1127  0.3357    58    no  experiment_id/study_id/response_id
##
## Test for Residual Heterogeneity:
## QE(df = 183) = 5100.4215, p-val < .0001
##
## Test of Moderators (coefficients 2:20):
## QM(df = 19) = 77.1912, p-val < .0001
##
## Model Results:
##
##                                     estimate      se      zval      pval
## intrcpt                           1.5220  3.6046   0.4222  0.6729
## resp_unitsCFU * g dry weight manure^-1 -0.8243  5.0926  -0.1619  0.8714
## resp_unitscm                       -2.8677  3.7556  -0.7636  0.4451
## resp_unitdays                     -3.0170  3.7349  -0.8078  0.4192
## resp_unitseggs laid                 -1.5013  4.4081  -0.3406  0.7334
## resp_unitsg                         -1.8742  3.7345  -0.5019  0.6158
## resp_unitskJ *day^-1 *kg^-1         -2.9976  3.7687  -0.7954  0.4264
## resp_unitsm                        -3.2515  3.7523  -0.8665  0.3862
## resp_unitsm * s^-1                  -3.0594  3.7513  -0.8156  0.4148
## resp_unitsmg                       -1.5573  5.0864  -0.3062  0.7595
## resp_unitsmm                       -1.7278  3.7363  -0.4624  0.6438
## resp_unitsmol CHE / mgww            -1.8189  5.0892  -0.3574  0.7208
## resp_unitsoffspring per mating      -1.0423  4.1622  -0.2504  0.8023
## resp_unitsoffspring/female          -2.8275  3.7693  -0.7501  0.4532
## resp_unitspercent                   -3.5402  3.7363  -0.9475  0.3434
## resp_unitspixels                   -1.8448  3.7717  -0.4891  0.6248
## resp_unitstime to death (hour)      -2.8982  3.7583  -0.7711  0.4406
## resp_unitstotal offspring            -0.9788  4.1622  -0.2352  0.8141
## resp_unitsuM Trolox Equivalents/ mgww -4.1911  5.0918  -0.8231  0.4105
## resp_unitswing centroid              -0.6880  4.1577  -0.1655  0.8686
##                                     ci.lb   ci.ub
## intrcpt                           -5.5429  8.5869
## resp_unitsCFU * g dry weight manure^-1 -10.8056  9.1571
## resp_unitscm                      -10.2285  4.4931
## resp_unitdays                     -10.3373  4.3034
## resp_unitseggs laid                 -10.1410  7.1385
## resp_unitsg                        -9.1937  5.4453
## resp_unitskJ *day^-1 *kg^-1         -10.3842  4.3890
## resp_unitsm                        -10.6058  4.1028
## resp_unitsm * s^-1                  -10.4119  4.2931

```



```
## resp_unitmsg -11.5265 8.4119
## resp_unitsmm -9.0508 5.5953
## resp_unitsmol CHE / mgww -11.7936 8.1557
## resp_unitsoffspring per mating -9.2001 7.1155
## resp_unitsoffspring/female -10.2152 4.5603
## resp_unitspercent -10.8632 3.7829
## resp_unitspixels -9.2372 5.5477
## resp_unitstime to death (hour) -10.2643 4.4679
## resp_unitstotal offspring -9.1366 7.1789
## resp_unitsuM Trolox Equivalents/ mgww -14.1709 5.7888
## resp_unitswing centroid -8.8368 7.4609
##
## ---
## 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  0.4274  0.6537     2    no      experiment_id
## sigma^2.2  0.0000  0.0000     5    no  experiment_id/study_id
##
## Test for Heterogeneity:
## Q(df = 30) = 153.8790, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
##  0.2455  0.4647  0.5284  0.5972  -0.6652  1.1562
##
## ---
## 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
## -0.81541 -0.28023 -0.05915  0.14639  1.19456
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.05915    0.09796   0.604   0.5510
## study_idkern_2015phys  0.68669    0.29389   2.337   0.0271 *
## study_idkern2014      0.62940    0.35321   1.782   0.0860 .
## study_idqu2014        0.32825    0.35321   0.929   0.3610
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4799 on 27 degrees of freedom
## Multiple R-squared:  0.2352, Adjusted R-squared:  0.1502
## F-statistic: 2.768 on 3 and 27 DF,  p-value: 0.06101

```