# Evaluation of Energy Expenditure from BXD Datasets

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The goal is to identify genetic determinants of energy expenditure and of adaptive thermogenesis from BXD mice. To start we searched gene network for energy expenditure data, ignoring those involved in exercise physiology.

- BXD\_17621 Oxygen intake over 24h on NCD at 16 w age. Also included light/dark. Not adjusted for lean mass. Has mean +/- SE in mL/kg/h. From Prinen 2014 (https://doi.org/10.1016/j.cmet.2014.04.002)
- BXD\_17618 Oxygen intake over 24h at 16 w age on HFD?, males. Not adjusted for lean mass. Has mean +/- SE in mL/kg/h. From Williams (2016). Body weight in BXD\_17559, lean mass in BXD\_17573
- BXD\_17622 Oxygen intake over 24h at 16 w age on HFD?, males. Not adjusted for lean mass. Has mean +/- SE in mL/kg/h. From Williams (2016). Body weight in BXD\_17560, lean mass in BXD\_17574

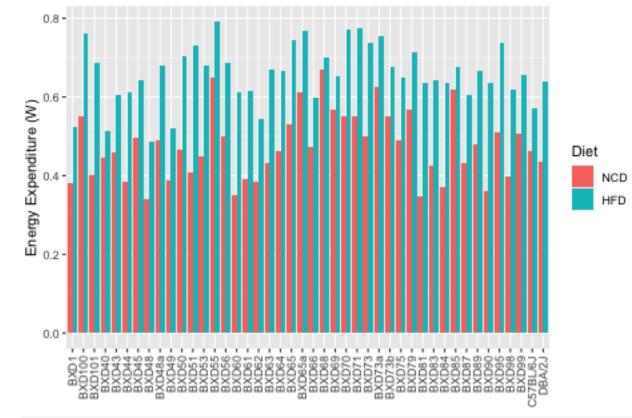
williams.hfd.bw <- read\_csv("BXD\_17560.csv", skip=9)%>%

```
mutate(Diet="HFD", Age=16, Dataset="Williams")
williams.hfd.lm <- read_csv("BXD_17574.csv", skip=9)%>%
  mutate(Diet="HFD", Age=16, Dataset="Williams")
williams.hfd <- full_join(williams.hfd.ee,williams.hfd.bw, suffix=c('_ee','_bw'), by=c("Name","Dataset"
  full_join(williams.hfd.lm) %>%
  mutate(Value_lm = Value,
         SE lm = SE)
data <- bind_rows(#ncd.pirinen,</pre>
                  williams.ncd,
                  williams.hfd) %>% # in mL/kg/h
  mutate(VO2 g min = Value ee/1000) %>% #in mL/q/h
  mutate(V02_min = V02_g_min*Value_bw/60) %>% # in mL/min #this seems reasonable
  mutate(MR_KJ_d = VO2_min * 60 * 24 / 1000 * 4.84 * 4.184,
         MR_KJ_d_SE = SE_ee/1000*Value_bw/60* 60 * 24 / 1000 * 4.84 * 4.184) %>% # 60min/h x 24h/day /
  mutate(MR_W = MR_KJ_d * 0.0115740741,
         MR_W_SE = MR_KJ_d_SE* 0.0115740741) %>% # in Watts
  mutate(Diet = relevel(factor(Diet), ref="NCD"))
```

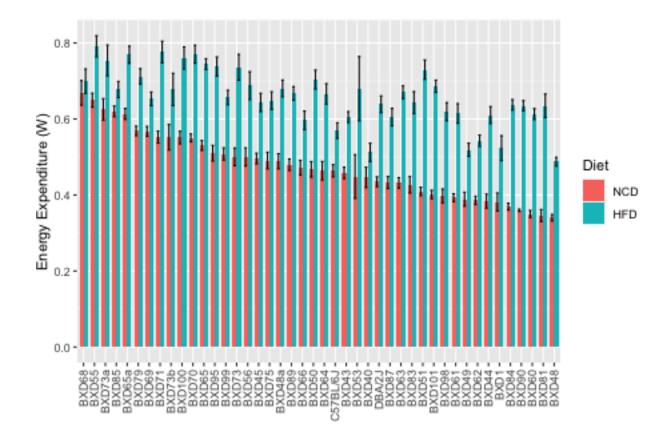
These data can be found in /Users/davebrid/Documents/GitHub/TissueSpecificTscKnockouts/Other Published Data/Systems Biology. This script was most recently updated on Wed Feb 16 11:11:15 2022.

## **Analysis**

## Comparason of Datasets



```
#lm(Value~Name+Diet,data=data) %>% summary
mr.order <-
  data %>%
  filter(Diet == "NCD") %>%
  arrange(desc(MR_W)) %>%
  mutate(Name = factor(Name))
data %>%
  filter(!is.na(MR_W)) %>% # complete cases only
  mutate(Name = factor(Name, levels = mr.order$Name, ordered = TRUE)) %>%
  ggplot(aes(y=MR_W,
         x=Name,
         ymin=MR_W-MR_W_SE,
         ymax=MR_W+MR_W_SE,
         fill=Diet)) +
  geom_bar(stat='identity',position='dodge', width=0.75) +
  geom_errorbar(position=position_dodge(width=0.75), width=0.5) +
  labs(y="Energy Expenditure (W)",x="") +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```



## **Estimating Effect Sizes**

Table 1: Summary statistics for thermogenesis from BXD mice

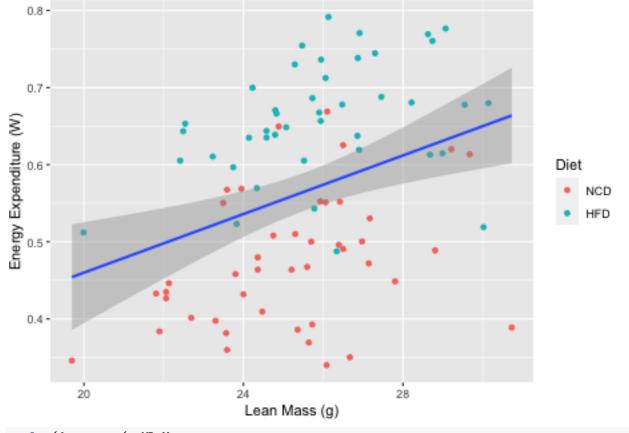
Diet	Max	Min	Mean	SE	N	SD	Rel.SD
			$0.475 \\ 0.659$				7.95 7.96

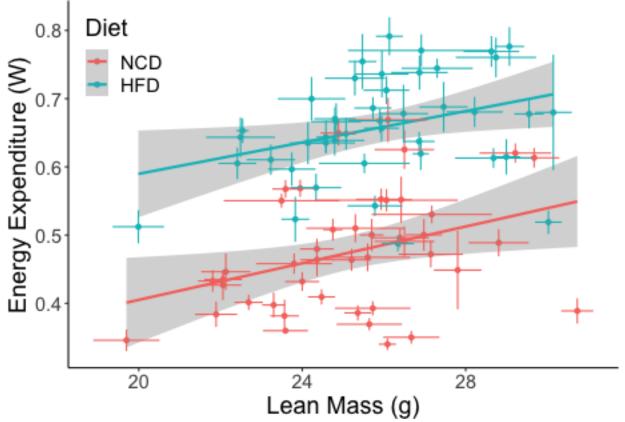
```
N = mean(N,na.rm=T)) %>%
mutate(SD = SE*sqrt(mean(N,na.rm=T))) %>%
mutate(Rel.SD = SD/Mean*100) %>%
kable(caption="Summary statistics for lean mass from BXD mice")
```

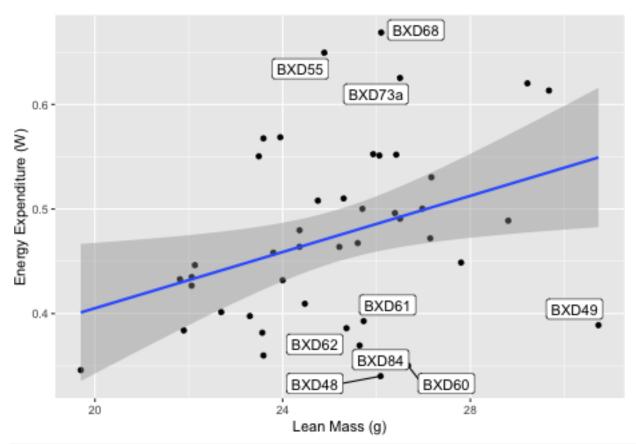
Table 2: Summary statistics for lean mass from BXD mice

Diet	Max	Min	Mean	SE	N	SD	Rel.SD
			25.4 26.0				5.19 4.97

## Adjusting for Lean Mass







lm.model.1 <- lm(MR\_W~Value\_lm,data=data %>% filter(Diet=="NCD")) #model built on only NCD
lm.model.2 <- lm(MR\_W~Value\_lm+Diet,data=data) #model built on NCD and AT
library(broom)
aov(lm.model.1) %>% tidy %>% kable(caption="Model 1 summary for adjusting for lean mass")

Table 3: Model 1 summary for adjusting for lean mass

term	df	sumsq	meansq	statistic	p.value
Value_lm	1	0.040	0.040	5.96	0.019
Residuals	42	0.279	0.007	NA	NA

summary(lm.model.1) %>% tidy %>% kable(caption="Model 1 coefficients for adjusting for lean mass")

Table 4: Model 1 coefficients for adjusting for lean mass

term	estimate	std.error	statistic	p.value
(Intercept)	0.136	0.139	0.981	0.332
$Value\_lm$	0.013	0.006	2.442	0.019

aov(lm.model.2) %>% tidy %>% kable(caption="Model 2 summary for adjusting for lean mass")

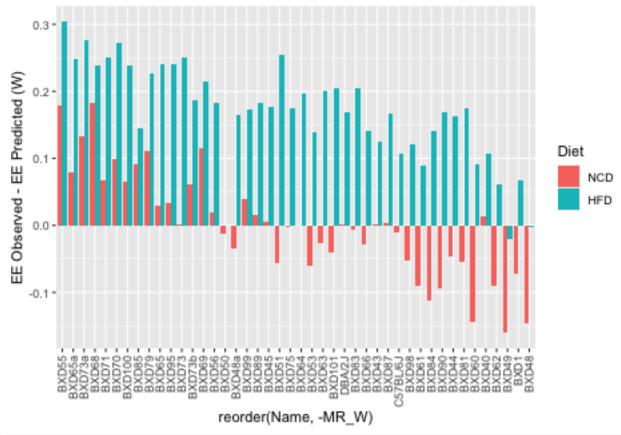
Table 5: Model 2 summary for adjusting for lean mass

term	df	sumsq	meansq	statistic	p.value
Value_lm	1	0.159	0.159	27.2	0
Diet	1	0.639	0.639	109.5	0
Residuals	84	0.490	0.006	NA	NA

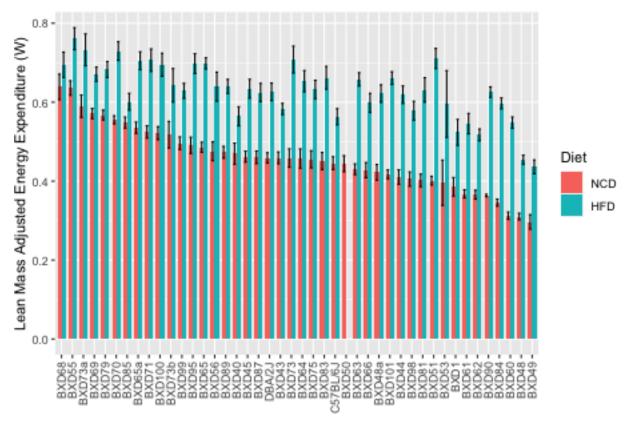
summary(lm.model.2) %>% tidy %>% kable(caption="Model 2 coefficients for adjusting for lean mass")

Table 6: Model 2 coefficients for adjusting for lean mass

term	estimate	std.error	statistic	p.value
(Intercept)	0.160	0.094	1.71	0.091
$Value\_lm$	0.012	0.004	3.38	0.001
DietHFD	0.174	0.017	10.47	0.000



```
mr.adj.order <-
  data %>%
  filter(Diet == "NCD") %>%
  arrange(desc(MR_adj)) %>%
  mutate(Name = factor(Name))
data %>%
  filter(!is.na(MR_W)) %>% # complete cases only
  mutate(Name = factor(Name, levels = mr.adj.order$Name, ordered = TRUE)) %>%
  ggplot(aes(y=MR_adj,
         x=Name,
         ymin=MR_adj-MR_W_SE,
         ymax=MR_adj+MR_W_SE,
         fill=Diet)) +
  geom_bar(stat='identity',position='dodge', width=0.75) +
  geom_errorbar(position=position_dodge(width=0.75), width=0.5) +
  labs(y="Lean Mass Adjusted Energy Expenditure (W)",x="") +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```

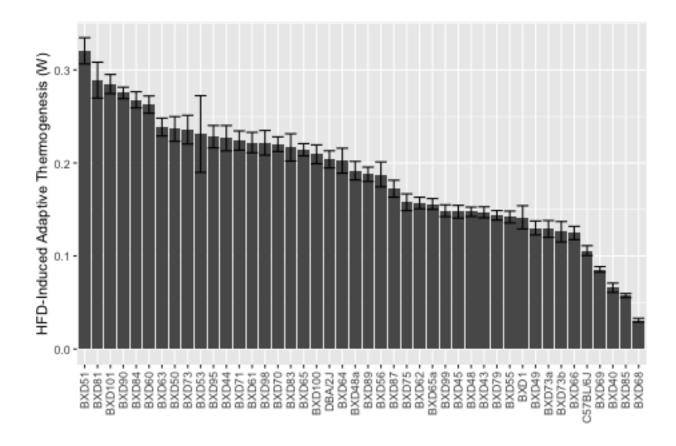


based on this modelling after adjusting for lean mass, HFD increases thermogenesis by (coef(lm.model.2)["(Intercept)"]-c

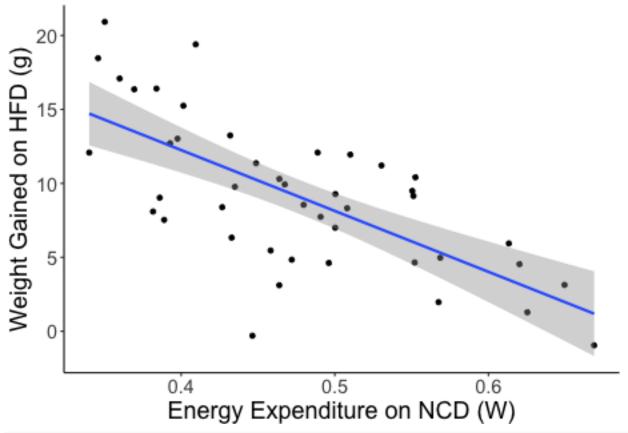
### Adaptive Thermogenesis

Defined as lean mass adjusted VO2 from HFD - NCD

```
data.wide <-
 data %>%
  select(Value_lm,SE,Value_bw, MR_W, MR_W_SE, MR_adj,Name,Diet) %>%
  pivot_wider(names_from=Diet,id_cols=Name,values_from=c(Value_lm,Value_bw, MR_W,MR_W_SE,MR_adj,SE)) %>
  mutate(AT = MR_W_HFD - MR_W_NCD,
         AT_SE = sqrt((MR_W_SE_NCD/MR_W_NCD)^2+(MR_W_SE_HFD/MR_W_HFD)^2)*AT,
         Wt_SE = sqrt((SE_NCD/Value_bw_NCD)^2+(SE_HFD/Value_bw_NCD)^2)*Value_bw_NCD,
         Wt.Gain = Value_bw_HFD-Value_bw_NCD)
data.wide %>%
  filter(!is.na(AT)) %>% # complete cases only
  ggplot(aes(y=AT,
         x=reorder(Name,-AT),
         ymin=AT-AT_SE,
         ymax=AT+AT_SE)) +
  geom_bar(stat='identity',position='dodge') +
    geom_errorbar() +
  labs(y="HFD-Induced Adaptive Thermogenesis (W)",
       x="") +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```



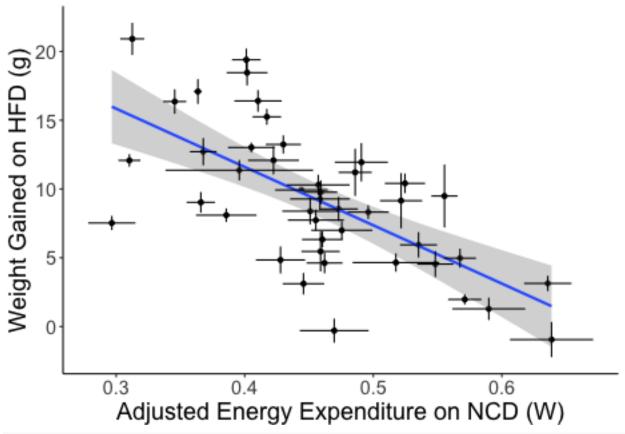
#### Thermogenesis on NCD as a Predictor of Weight Gain



lm(Wt.Gain~MR\_W\_NCD, data=data.wide) %>% glance %>% kable(caption="Summary of relationship between energy

Table 7: Summary of relationship between energy expenditure and diet-induced weight gain

r.squared	adj.r.squared	sigma	statistic	p.value	df	logLik	AIC	BIC	deviance	df.residual	nobs
0.474	0.462	3.77	37.9	2.4e-	1	-120	246	251	598	42	44
				07							



lm(Wt.Gain~MR\_adj\_NCD, data=data.wide) %>% glance %>% kable(caption="Summary of relationship between le

Table 8: Summary of relationship between lean mass adjusted energy expenditure and diet-induced weight gain

r.squared	adj.r.squared	$\operatorname{sigma}$	statistic	p.value	df	logLik	AIC	BIC	deviance	df.residual	nobs
0.441	0.427	3.89	33.1	9e-07	1	-121	248	254	636	42	44

```
gemma.phenotype.export <- 'Strain Level Energy Expenditure Data.csv'
data %>%
  filter(Diet=="NCD") %>%
  select(Name, MR_W, MR_adj) %>%
  write_csv(gemma.phenotype.export)
```

The data on lean mass adjusted thermogenesis was exported to Strain Level Energy Expenditure Data.csv

#### Heritability of NCD Thermogenesis

Since we don't have individual mouse data we will make fake data based on the mean and se of MR

```
new.sim.data <- data.frame(Name=NA, Diet=NA,EE=NA)

for (row in 1:dim(data)[1]) {
   strain.data <- data[row,]
   if(!(is.na(strain.data$MR_W))){
    sim.data <- with(strain.data,</pre>
```

```
rnorm(mean=MR_W,
                          sd=MR_W_SE * sqrt(N),
                          n=N_ee
                     ))
  sim.lean.data <- with(strain.data,</pre>
                    rnorm(mean=Value_lm,
                          sd=SE_lm * sqrt(N),
                          n=N ee
                     ))
  sim.dataset <- data.frame(Name=strain.data$Name,</pre>
                              Diet=strain.data$Diet,
                              EE=sim.data.
                              Lean=sim.lean.data)
  new.sim.data <- bind_rows(new.sim.data,sim.dataset)</pre>
  }
else{
    sim.dataset <- data.frame(Name=strain.data$Name,</pre>
                              Diet=strain.data$Diet,
                              EE=NA,
                              Lean=NA)
    new.sim.data <- bind_rows(new.sim.data,sim.dataset)</pre>
}
  }
aov(EE ~ Name, data=new.sim.data %>% filter(Diet=='NCD')) %>%
  tidy %>%
  mutate(Total.Var=sum(meansq),
         Pct.Var = meansq/Total.Var*100) %>%
  kable(caption="Overall heritability of energy expenditure on NCD mice")
```

Table 9: Overall heritability of energy expenditure on NCD mice

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Name	43	1.46	0.034	21.9	0	0.036	95.62
Residuals	148	0.23	0.002	NA	NA	0.036	4.38

Table 10: Overall heritability of energy expenditure on NCD including lean mass  $\,$ 

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.050	0.050	32.4	0	0.084	59.18
Name	43	1.415	0.033	21.4	0	0.084	39.00
Residuals	147	0.226	0.002	NA	NA	0.084	1.83

Table 11: Overall heritability of energy expenditure on NCD adjusting for lean mass

aov(EE ~ Lean + Name, data=new.sim.data %>% filter(Diet=='NCD')) %>%

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.050	0.050	32.4	0	0.034	144.96
Name	43	1.415	0.033	21.4	0	0.034	95.53
Residuals	147	0.226	0.002	NA	NA	0.034	4.47

Table 12: Overall heritability of energy expenditure including diet and lean mass

term	df	$\operatorname{sumsq}$	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.371	0.371	157.62	0	3.4	10.887
Name	43	2.307	0.054	22.82	0	3.4	1.576
Diet	1	2.969	2.969	1263.00	0	3.4	87.237
Name:Diet	42	0.330	0.008	3.34	0	3.4	0.231
Residuals	291	0.684	0.002	NA	NA	3.4	0.069

Table 13: Overall heritability of energy expenditure adjusted for diet and lean mass

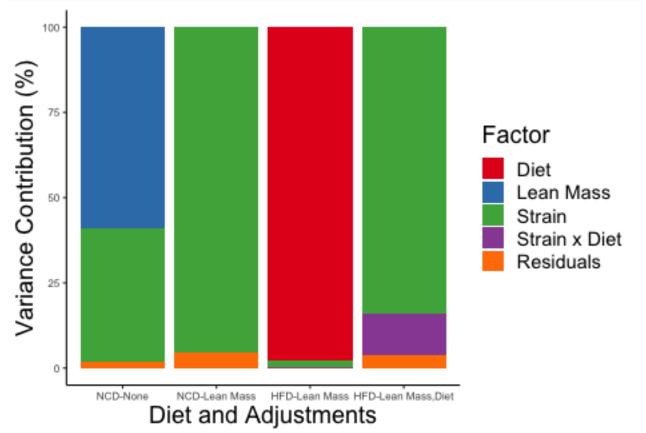
term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.371	0.371	157.62	0	0.064	580.33
Name	43	2.307	0.054	22.82	0	0.064	84.02
Diet	1	2.969	2.969	1263.00	0	0.064	4650.28
Name:Diet	42	0.330	0.008	3.34	0	0.064	12.30
Residuals	291	0.684	0.002	NA	NA	0.064	3.68

```
ee.var.data <- bind_rows(lean.adj.ee.lean %>% mutate(Diet="NCD"),hfd.adj.ee.adj %>% mutate(Diet="HFD"))
ggplot(ee.var.data %>% filter(term %in% c('Name','Name:Diet','Residuals')),
        aes(x=reorder(Diet,-Pct.Var),
            y=Pct.Var,
            fill=term)) +
  geom_bar(position="stack",stat='identity') +
  scale_fill_manual(labels = c("Strain", "Strain x Diet", "Other"), values = c("red", "pink", "blue"),
                    name="Factor") +
  labs(y="Percent of Variance",
      x="") +
  theme_classic() +
  theme(legend.position="top")+
  theme(text=element_text(size=18))
                   Factor Strain Strain x Diet
                                                                  Other
    100
Percent of Variance
     75
     50
```

0 NĊD HĖD ee.var.all.data <- bind\_rows(lean.adj.ee.all %>% mutate(Diet="NCD", Corr="None"), lean.adj.ee.lean %>% mutate(Diet="NCD", Corr="Lean Mass"), hfd.adj.ee.all %>% mutate(Diet="HFD", Corr="Lean Mass"), hfd.adj.ee.adj %>% mutate(Diet="HFD", Corr="Lean Mass,Diet")) %>%

25

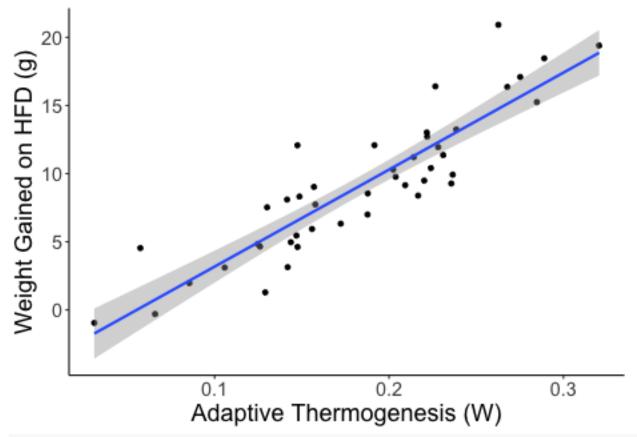
```
mutate(Group=paste(Diet,Corr,sep="-"))
library(RColorBrewer)
ee.var.all.data %>%
         filter(term %in% c('Lean','Name','Name:Diet','Diet','Residuals')) %>%
         filter(!(term=='Lean'&Corr=='Lean Mass')) %>%
         filter(!(term=='Lean'&Corr=="Lean Mass,Diet")) %>%
         filter(!(term=='Diet'&Corr=="Lean Mass,Diet")) %>%
ggplot(aes(x=ordered(Group, levels=c("NCD-None","NCD-Lean Mass", "HFD-Lean Mass", "HFD-Lean Mass, Diet")
            y=Pct.Var,
            fill=term)) +
  geom_bar(position="stack",stat='identity') +
  scale_fill_manual(name="Factor",
                    labels = c("Diet", "Lean Mass", "Strain", "Strain x Diet", "Residuals"),
                    values=brewer.pal(5, "Set1")) +
  labs(y="Variance Contribution (%)",
       x="Diet and Adjustments") +
  theme_classic() +
  theme(text=element_text(size=18),
        axis.text=element_text(size=8))
```



#### Adaptive Thermogenesis vs Weight Gain

```
data.wide %>%
  ggplot(aes(y=Wt.Gain,
```

```
x=AT)) +
labs(y="Weight Gained on HFD (g)",
    x="Adaptive Thermogenesis (W)") +
geom_point() +
geom_smooth(method="lm") + theme_classic() +
theme(text=element_text(size=18))
```



lm(Wt.Gain~AT, data=data.wide) %>% glance %>% kable(caption="Summary of relationship between energy exp

Table 14: Summary of relationship between energy expenditure and diet-induced weight gain

r.squared	${\it adj.r.} {\it squared}$	$\operatorname{sigma}$	statistic	p.value	df	logLik	AIC	BIC	deviance	${\rm df.residual}$	nobs
0.801	0.796	2.32	169	0	1	-98.5	203	208	226	42	44

## **Session Information**

```
## R version 4.0.2 (2020-06-22)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS 10.16
##
## Matrix products: default
```

```
/Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRblas.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
## attached base packages:
## [1] stats
                 graphics grDevices utils
                                               datasets methods
                                                                    base
##
## other attached packages:
## [1] RColorBrewer_1.1-2 broom_0.7.11
                                             ggrepel_0.9.1
                                                                ggplot2_3.3.5
## [5] readr_2.1.1
                          dplyr_1.0.7
                                             tidyr_1.1.4
                                                                knitr_1.37
## loaded via a namespace (and not attached):
## [1] tidyselect_1.1.1 xfun_0.29
                                          purrr_0.3.4
                                                           splines_4.0.2
   [5] lattice_0.20-45 colorspace_2.0-2 vctrs_0.3.8
                                                           generics_0.1.1
## [9] htmltools_0.5.2
                         yaml_2.2.1
                                          mgcv_1.8-38
                                                           utf8_1.2.2
## [13] rlang 0.4.12
                         pillar 1.6.4
                                          glue 1.6.0
                                                           withr 2.4.3
## [17] DBI_1.1.2
                         bit64_4.0.5
                                          lifecycle_1.0.1
                                                           stringr_1.4.0
## [21] munsell 0.5.0
                         gtable_0.3.0
                                          evaluate_0.14
                                                           labeling 0.4.2
## [25] tzdb_0.2.0
                         fastmap_1.1.0
                                          parallel_4.0.2
                                                           fansi_1.0.0
## [29] highr_0.9
                         Rcpp_1.0.7
                                          backports_1.4.1
                                                           scales 1.1.1
## [33] vroom_1.5.7
                                          farver_2.1.0
                                                           bit_4.0.4
                         magick_2.7.3
## [37] hms 1.1.1
                         digest_0.6.29
                                          stringi_1.7.6
                                                           grid 4.0.2
## [41] cli 3.1.0
                         tools_4.0.2
                                          magrittr_2.0.1
                                                           tibble_3.1.6
## [45] crayon_1.4.2
                         pkgconfig_2.0.3
                                          ellipsis_0.3.2
                                                           Matrix 1.4-0
## [49] assertthat_0.2.1 rmarkdown_2.11
                                          rstudioapi_0.13
                                                           R6_2.5.1
## [53] nlme_3.1-153
                         compiler_4.0.2
```