Evaluation of Energy Intake 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

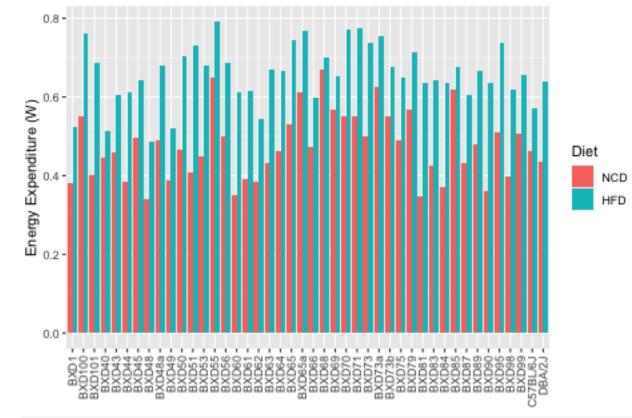
```
library(readr)
ncd.pirinen <- read_csv("BXD_17621.csv", skip=9) %>%
  mutate(Diet="NCD", Age=16, Dataset="Prinen")
williams.ncd.ee <- read_csv("BXD_17618.csv", skip=9)%>% #may be mislabelled on genenetwork, assigned ba
  mutate(Diet="NCD", Age=16, Dataset="Williams")
williams.ncd.bw <- read_csv("BXD_17559.csv", skip=9)%>%
  mutate(Diet="NCD", Age=16, Dataset="Williams")
williams.ncd.lm <- read_csv("BXD_17573.csv", skip=9)%>%
  mutate(Diet="NCD", Age=16, Dataset="Williams")
williams.ncd <- full_join(williams.ncd.ee,williams.ncd.bw, suffix=c('_ee','_bw'), by=c("Name","Dataset"
  full_join(williams.ncd.lm) %>%
  mutate(Value_lm = Value,
         SE_lm = SE)
williams.hfd.ee <- read_csv("BXD_17622.csv", skip=9)%% #may be mislabelled on genenetwork
  mutate(Diet="HFD", Age=16, Dataset="Williams")
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(V02_g_min = Value_ee/1000) %>% #in mL/g/h
  mutate(V02_min = V02_g_min*Value_bw/60) %>% # in mL/min #this seems reasonable
  mutate(MR_KJ_d = V02_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"))
```

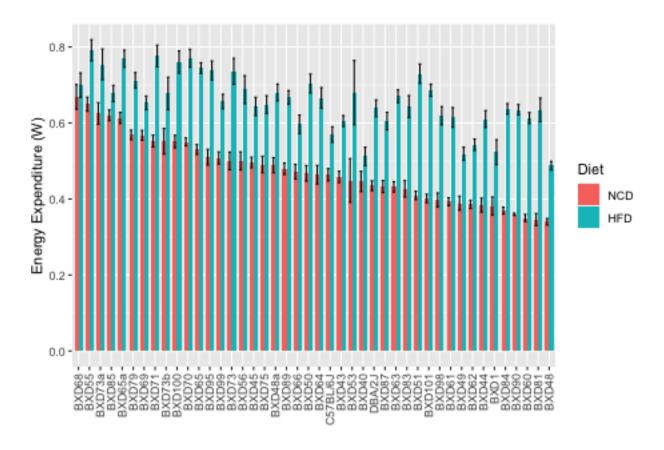
These data can be found in /Users/davebrid/Documents/GitHub/TissueSpecificTscKnockouts/Other Published Data/Systems Biology. This script was most recently updated on Sun Feb 13 10:20:35 2022.

Analysis

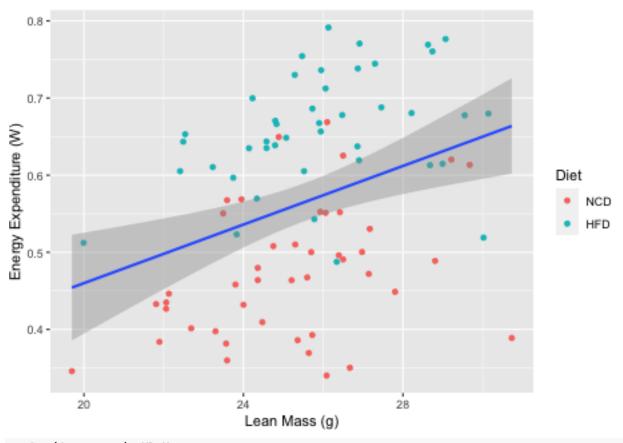
Comparason of Datasets

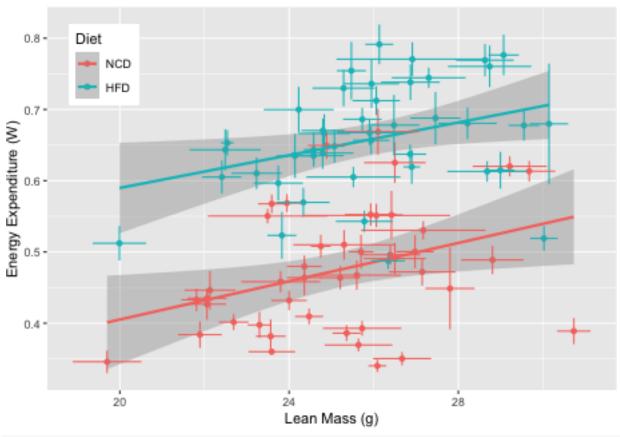


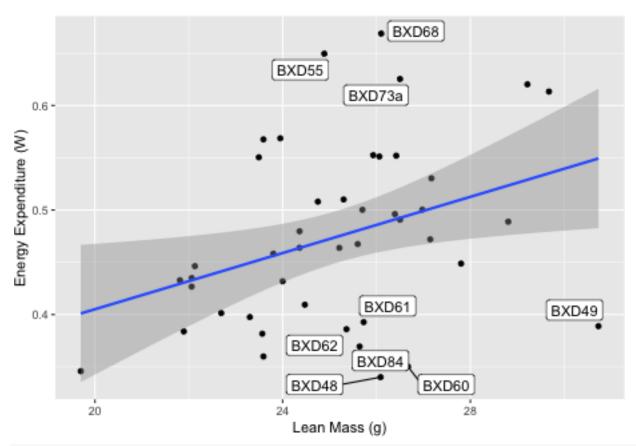
```
#lm(Value~Name+Diet,data=data) %>% summary
mr.order <-
  data %>%
  filter(Diet == "NCD") %>%
  arrange(desc(MR_W)) %>%
  mutate(Name = factor(Name))
data %>%
  filter(!is.na(MR_W)) %>% # complete cases only
  mutate(Name = factor(Name, levels = mr.order$Name, ordered = TRUE)) %>%
  ggplot(aes(y=MR_W,
         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))
```



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 1: 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 2: 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")

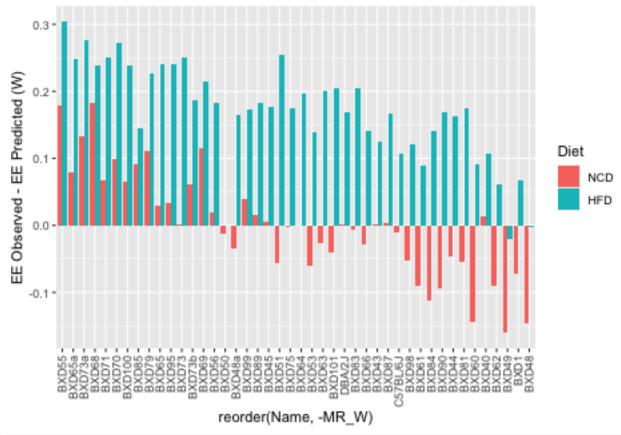
Table 3: Model 2 summary for adjusting for lean mass

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

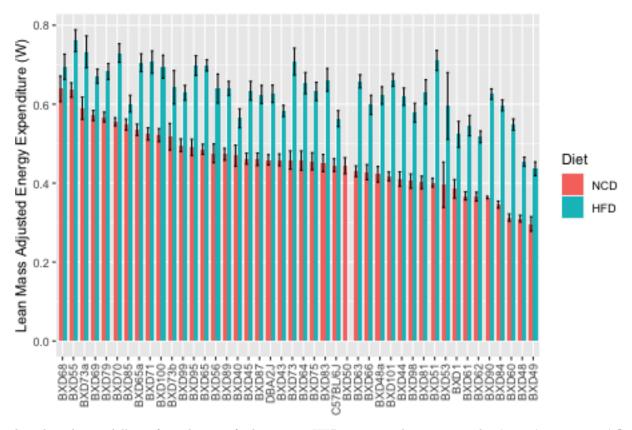
summary(lm.model.2) %>% tidy %>% kable(caption="Model 2 coefficients for adjusting for lean mass")

Table 4: 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))
```

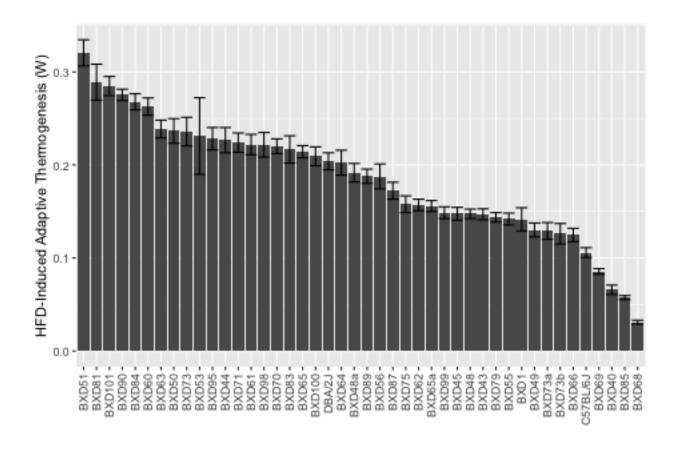


based on this modelling after adjusting for lean mass, HFD increases thermogenesis by (coef(lm.model.2)["(Intercept)"]-c

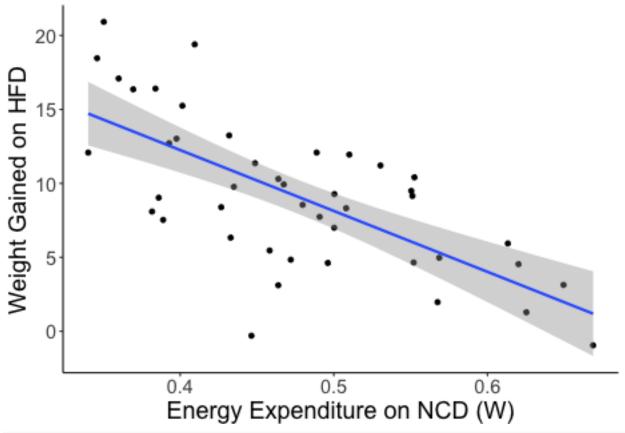
Adaptive Thermogenesis

Defined as lean mass adjusted VO2 from HFD - NCD

```
data.wide <-
 data %>%
  select(Value_lm, 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)) %>%
  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.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))
```



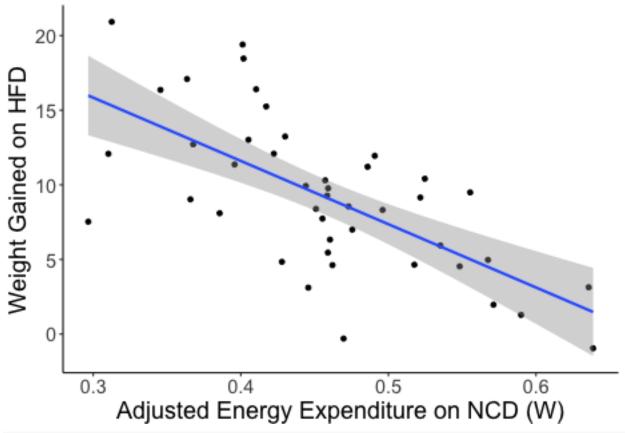
Thermogenesis on NCD as a Predictor of Weight Gain



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

Table 5: Summary of relationship between energy expenditure and diet-induced weight gain

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



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

Table 6: Summary of relationship between lean mass adjusted energy expenditure and diet-induced weight gain

r.squared	adj.r.squared	sigma	statistic	p.value	df	logLik	AIC	BIC	deviance	df.residual	nobs
0.441	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 7: Overall heritability of energy expenditure on NCD mice

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Name	43	1.431	0.033	16.9	0	0.035	94.43
Residuals	148	0.291	0.002	NA	NA	0.035	5.57

Table 8: Overall heritability of energy expenditure on NCD including lean mass

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.102	0.102	53.1	0	0.135	75.64
Name	43	1.335	0.031	16.1	0	0.135	22.93
Residuals	147	0.283	0.002	NA	NA	0.135	1.42

Table 9: 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	0.102	0.102	53.1	0	0.033	310.59
Name	43	1.335	0.031	16.1	0	0.033	94.16
Residuals	147	0.283	0.002	NA	NA	0.033	5.84

Table 10: Overall heritability of energy expenditure including diet and lean mass

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.718	0.718	236.47	0	3.48	20.620
Name	43	2.201	0.051	16.87	0	3.48	1.471
Diet	1	2.700	2.700	889.77	0	3.48	77.589
Name:Diet	42	0.341	0.008	2.68	0	3.48	0.234
Residuals	291	0.883	0.003	NA	NA	3.48	0.087

Table 11: Overall heritability of energy expenditure adjusted for diet and lean mass

term	df	sumsq	meansq	statistic	p.value	${\bf Total.Var}$	Pct.Var
Lean	1	0.718	0.718	236.47	0	0.062	1150.91
Name	43	2.201	0.051	16.87	0	0.062	82.09
Diet	1	2.700	2.700	889.77	0	0.062	4330.66
Name:Diet	42	0.341	0.008	2.68	0	0.062	13.04
Residuals	291	0.883	0.003	NA	NA	0.062	4.87

```
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
     25
       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")) %>%

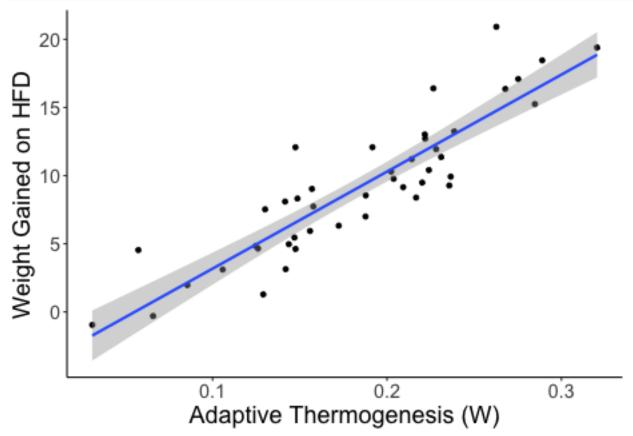
```
mutate(Group=paste(Diet,Corr,sep="-"))
library(RColorBrewer)
ee.var.all.data %>%
         filter(term %in% c('Lean','Name','Name:Diet','Diet','Residuals')) %>%
         filter(!(term=='Lean'&Corr=='Lean Mass')) %>%
         filter(!(term=='Lean'&Corr=="Lean Mass,Diet")) %>%
         filter(!(term=='Diet'&Corr=="Lean 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="Percent of Variance",
       x="Group - Adjustments") +
  theme_classic() +
  theme(text=element_text(size=12))
   100
    75
Percent of Variance
                                                                           Factor
                                                                                Diet
                                                                                Lean Mass
                                                                                Strain
                                                                                Strain x Diet
                                                                                Residuals
    25
     0
```

Adaptive Thermogenesis vs Weight Gain

NCD-None

NCD-Lean Mass HFD-Lean MassHFD-Lean Mass, Diet

Group - Adjustments



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

Table 12: 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
0.801	0.796	2.32	169	0	1	-98.5	203	208	226	42	44

Session Information

```
sessionInfo()
```

```
## R version 4.0.2 (2020-06-22)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS 10.16
##
## Matrix products: default
## BLAS: /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
                                                           fansi 1.0.0
## [25] tzdb 0.2.0
                         fastmap 1.1.0
                                          parallel 4.0.2
## [29] highr_0.9
                         Rcpp_1.0.7
                                          backports_1.4.1
                                                           scales_1.1.1
## [33] vroom 1.5.7
                         magick_2.7.3
                                          farver_2.1.0
                                                           bit_4.0.4
## [37] hms_1.1.1
                                                           grid_4.0.2
                         digest_0.6.29
                                          stringi_1.7.6
## [41] cli 3.1.0
                         tools_4.0.2
                                          magrittr_2.0.1
                                                           tibble 3.1.6
                                         ellipsis_0.3.2
## [45] crayon_1.4.2
                                                           Matrix_1.4-0
                         pkgconfig_2.0.3
## [49] assertthat 0.2.1 rmarkdown 2.11
                                          rstudioapi_0.13 R6_2.5.1
## [53] nlme_3.1-153
                         compiler_4.0.2
```