# Evaluation of Energy Intake from BXD Datasets

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### Contents

nalysis	<b>2</b>
Comparason of Datasets	2
Adjusting for Lean Mass	3
Adaptive Thermogenesis	8
ession Information	1 1
ESSION INIOCHIATION	14

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

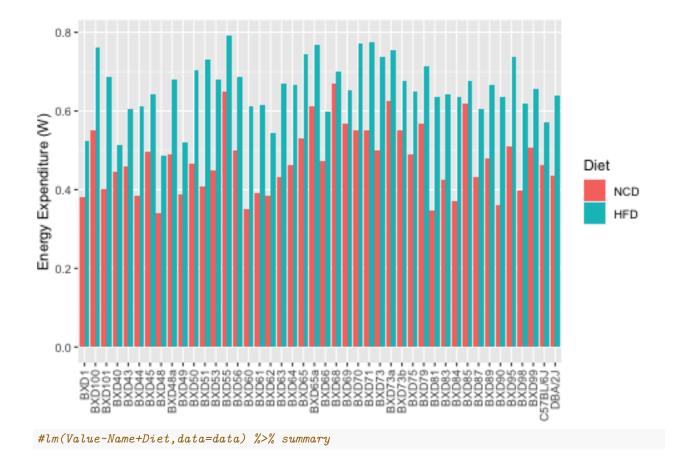
```
library(readr)
ncd.pirinen <- read_csv("BXD_17621.csv", skip=9) %>%
  mutate(Diet="NCD", Age=16, Dataset="Prinen")
williams.ncd.ee <- read_csv("BXD_17618.csv", skip=9)%>% #may be mislabelled on genenetwork, assigned ba
  mutate(Diet="NCD", Age=16, Dataset="Williams")
williams.ncd.bw <- read_csv("BXD_17559.csv", 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"))
```

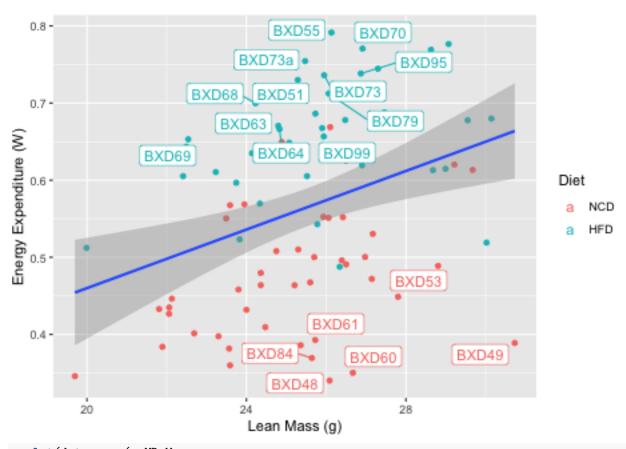
These data can be found in /Users/davebrid/Documents/GitHub/TissueSpecificTscKnockouts/Other Published Data/Systems Biology. This script was most recently updated on Thu Jan 20 15:21:02 2022.

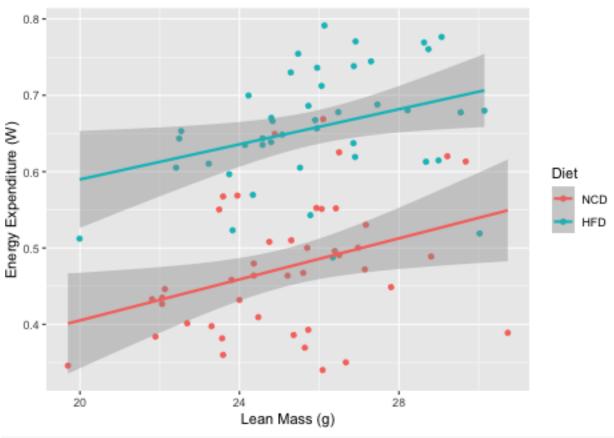
### **Analysis**

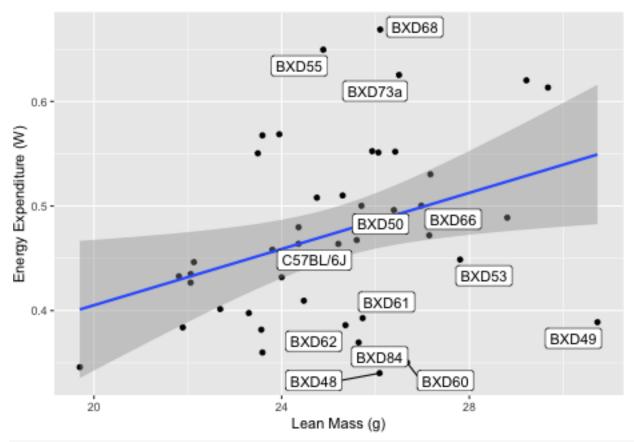
### Comparason of Datasets



### 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")

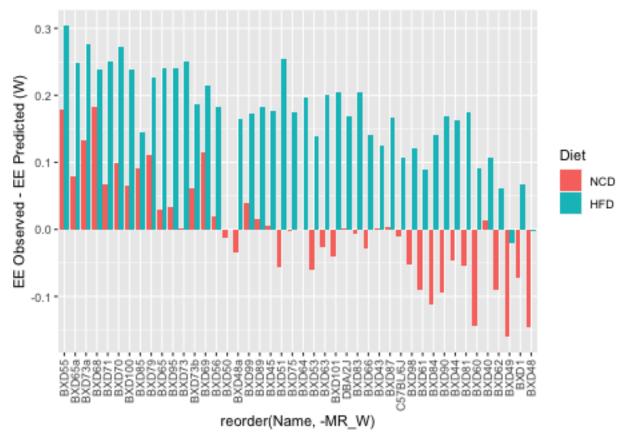
Table 3: Model 2 summary for adjusting for lean mass

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

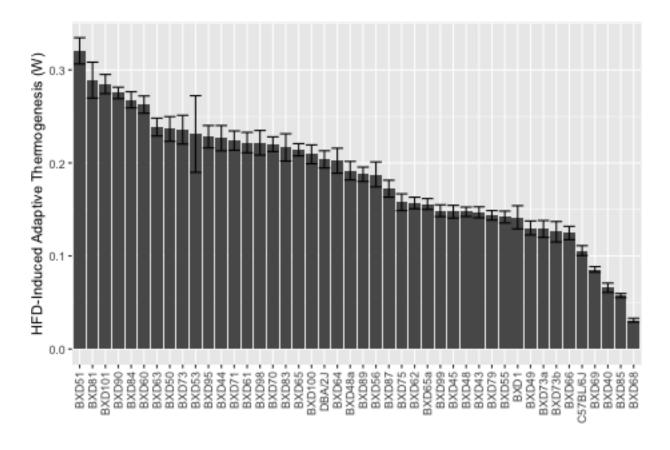


based on this modelling after adjusting for lean mass, HFD increases thermogenesis by (coef(lm.model.2)["(Intercept)"]-c

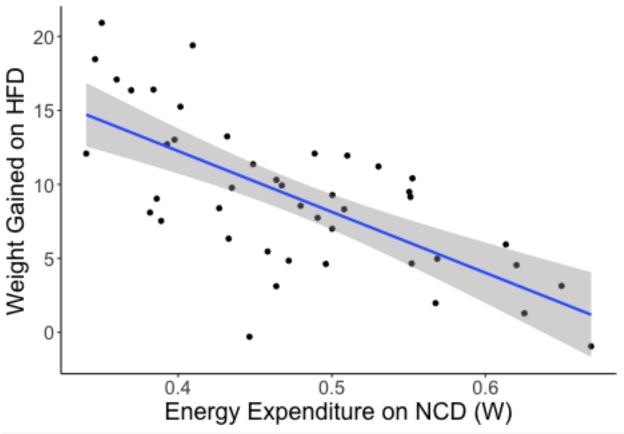
### Adaptive Thermogenesis

Defined as lean mass adjusted VO2 from HFD - NCD

```
data.wide <-
  data %>%
  select(Value_lm, Value_bw, MR_W, MR_W_SE, Name, Diet) %>%
  pivot_wider(names_from=Diet,id_cols=Name,values_from=c(Value_lm,Value_bw, MR_W,MR_W_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.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	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							

### 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

```
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 6: Overall heritability of energy expenditure on NCD mice

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Name	43	1.547	0.036	22	0	0.038	95.66
Residuals	148	0.242	0.002	NA	NA	0.038	4.34

Table 7: Overall heritability of energy expenditure on NCD including lean mass

term	df	$\operatorname{sumsq}$	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.207	0.207	127.8	0	0.24	86.303
Name	43	1.343	0.031	19.3	0	0.24	13.021
Residuals	147	0.238	0.002	NA	NA	0.24	0.676

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

term	df	sumsq	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.207	0.207	127.8	0	0.033	630.08
Name	43	1.343	0.031	19.3	0	0.033	95.07
Residuals	147	0.238	0.002	NA	NA	0.033	4.93

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

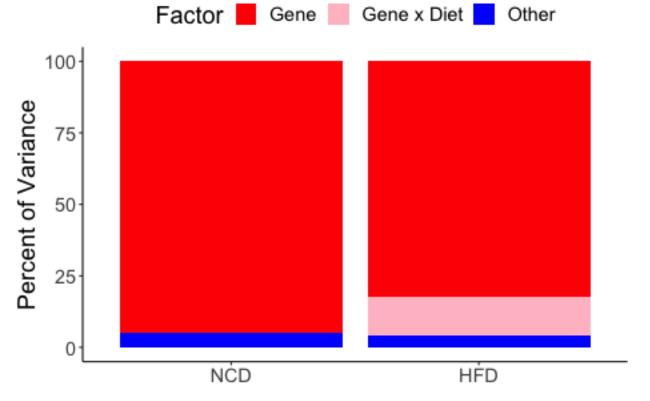
Table 9: Overall heritability of energy expenditure including diet and lean mass

term	df	$\operatorname{sumsq}$	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.697	0.697	260.67	0	3.39	20.577
Name	43	2.296	0.053	19.98	0	3.39	1.577
Diet	1	2.624	2.624	981.81	0	3.39	77.503
Name:Diet	42	0.375	0.009	3.34	0	3.39	0.264
Residuals	291	0.778	0.003	NA	NA	3.39	0.079

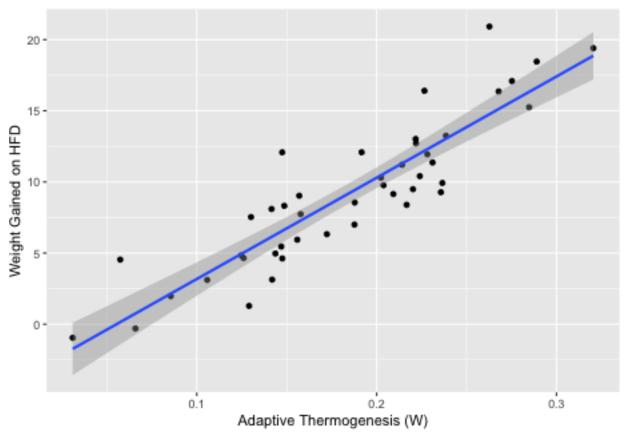
Table 10: Overall heritability of energy expenditure adjusted for diet and lean mass

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

term	df	$\operatorname{sumsq}$	meansq	statistic	p.value	Total.Var	Pct.Var
Lean	1	0.697	0.697	260.67	0	0.065	1071.68
Name	43	2.296	0.053	19.98	0	0.065	82.14
Diet	1	2.624	2.624	981.81	0	0.065	4036.53
Name:Diet	42	0.375	0.009	3.34	0	0.065	13.74
Residuals	291	0.778	0.003	NA	NA	0.065	4.11



### Adaptive Thermogenesis vs Weight Gain



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

Table 11: 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] broom_0.7.11 ggrepel_0.9.1 ggplot2_3.3.5 readr_2.1.1
                                                              dplyr_1.0.7
## [6] tidyr_1.1.4 knitr_1.37
##
## loaded via a namespace (and not attached):
## [1] tidyselect_1.1.1 xfun_0.29
                                                           splines_4.0.2
                                          purrr_0.3.4
   [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
                                                          stringr_1.4.0
                                          lifecycle_1.0.1
                                          evaluate_0.14
## [21] munsell_0.5.0
                         gtable_0.3.0
                                                           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
                        magick_2.7.3
                                          farver_2.1.0
                                                           bit_4.0.4
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                                          stringi_1.7.6
                                                           grid_4.0.2
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                         tools_4.0.2
                                          magrittr_2.0.1
                                                           tibble 3.1.6
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                        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
```