armor mr

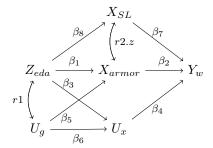
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History

```
# reboot Aug 9 2013
# reboot II April 5 2018
```

The original motivation was a Mendelian Randomization re-analysis of Marchinko, K.B., 2009. Predation's role in repeated phenotypic and genetic divergence of armor in threespine stickleback. Evolution, 63(1), pp.127-138. The new goal is to add the Schluter xxx data.

We need a DAG model of the potential paths



MR estimates β_2 . An unbiased Mendelian randomization (MR) estimate of β_2 assumes

- 1. $\beta_3 \beta_4 = 0$
- $2. r\beta_6\beta_4 = 0$

The estimate of the "direct selection on" X_{armor} (β_2), conditioning on Z_{eda} , is an unbiased estimate of β_2 if

1. $\beta_5 \beta_6 \beta_4 = 0$

Load libraries

```
library(readxl)
library(lme4)
library(ggplot2)
library(data.table)

source("../R/replicate.R") # for knit, drop the "../" for console
source("../R/mendelian_randomization.R") # for knit, drop the "../" for console
```

Import Rennison data

```
dir_path <- "data/" # for console
dir_path <- "../data/" # for knit</pre>
```

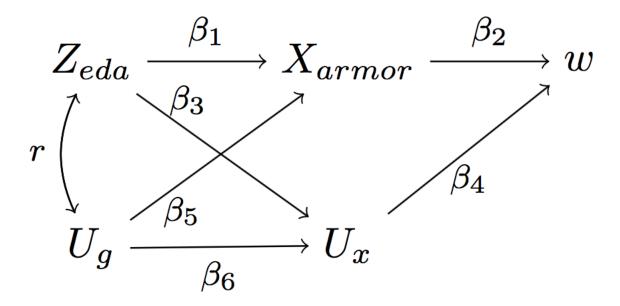


Figure 1: Full model.

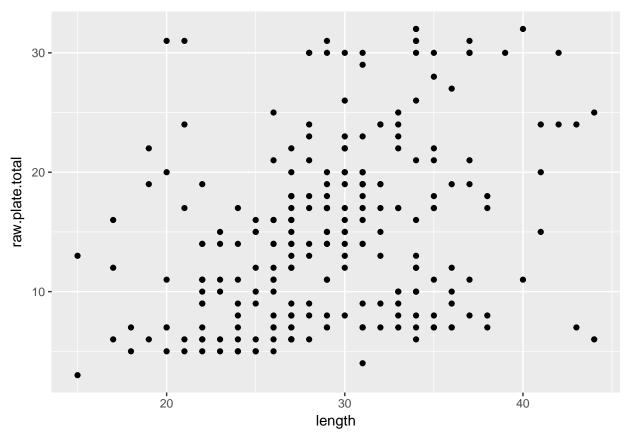
```
file_name <- "Rennison-am_nat-2016.txt"
file_path <- paste(dir_path, file_name, sep='')
rennison <- data.table(fread(file_path))
rennison[, month:=factor(month, c('06-Sep', '06-Oct', '06-Nov'))]</pre>
```

Replicate Rennison

Replicate size adjustment

It would be good to simulate process to know how sensitive result is to adjustment vs. simply using length as covariate in linear model.

```
# plot plates vs. size
qplot(x=length, y=raw.plate.total, data=rennison)
```



```
# check size adjustment - can't replicate
rennison[, logplates.s:=ifelse(length <=34.0, size_corrected(raw.logplates, length, origin=34.0), raw.l
# limit adjusted plates to max of log(32)
rennison[, logplates.s:=ifelse(logplates.s > log(32), log(32), logplates.s)]
x <- rennison[genotype=='CL', raw.logplates]
sl <- rennison[genotype=='CL', length]
fit <- lm(x~sl)
x_34 <- predict(fit, newdata=data.frame(sl=34.0))
logplates.s <- x_34 + residuals(fit)</pre>
```

Replicate multivariate selection differentials

The bootstrap implemented below does not bootstrap the whole process including the length standardization step

```
# this comes closer to replicating adj.plates than using eda_add column
Xa <- rennison[month==levels(month)[1] & eda_add!=-1, .SD, .SDcols=c('adj.plates', 'eda_dom')]
Xb <- rennison[month==levels(month)[2] & eda_add!=-1, .SD, .SDcols=c('adj.plates', 'eda_dom')]
res1 <- mv_selection_differential(Xa, Xb, std=TRUE, bootstrap=TRUE)
apply(res1,2, quantile, c(0.025, 0.5, 0.975))

## adj.plates eda_dom
## 2.5% -0.07807069 -0.6580128
## 50% 0.29447588 -0.2776027
## 97.5% 0.65601220 0.1005373</pre>
```

```
Xa <- rennison[month==levels(month)[2] & eda_add!=-1, .SD, .SDcols=c('adj.plates', 'eda_dom')]
Xb <- rennison[month==levels(month)[3] & eda_add!=-1, .SD, .SDcols=c('adj.plates', 'eda_dom')]
res2 <- mv_selection_differential(Xa, Xb, std=TRUE, bootstrap=TRUE)
apply(res2,2, quantile, c(0.025, 0.5, 0.975))

## adj.plates eda_dom
## 2.5% -0.7280472 -0.47655398
## 50% -0.2066366 0.03831118
## 97.5% 0.2220305 0.49700138
##</pre>
```

Rennison Mendelian Randomization

Import Marchinko data

```
dir_path <- "data/" # for console</pre>
dir_path <- "../data/" # for knit</pre>
file_name <- "Marchinko_PaxtonData.xls"</pre>
file_path <- paste(dir_path, file_name, sep='')
marchinko <- data.table(read_excel(file_path))</pre>
    # columns are
        # population
        # family
        # treatment {no = -insect, pred = +insect}
        # individual
        # standlength
        # ant.dorspine
        # sec.dorspine
        # pelspine.len
        # pelgirdle.len
        # plate.number
        # eda.genotype {AA, Aa, aa}
        # pelgirdle.presence {1=yes, 0 = no}
        # pelspine.presence {1 = yes, 0 = no}
```

```
marchinko[, family:=factor(family)]
marchinko[, treatment:=factor(treatment, c('no','pred'))]
```

Replicate Marchinko

Replication requires scaling armor variables within families and replacing values=0.0 with NA. Still, I cannot replicate ant.dorspine.s

```
table_1_replicate <- round(replicate_marchinko(marchinko),3)</pre>
```

Re-analysis using multi-level model

```
y_list <- c('ant.dorspine', 'sec.dorspine', 'pelspine.len', 'pelgirdle.len')
table_1_mlm <- numeric(length(y_list))
for(j in 1:length(y_list)){
    y <- y_list[j]
    # fit to get treatment effect
    model_formula <- formula(paste(y, " ~ standlength + treatment + (standlength|family)", sep=''))
    fit <- lmer(model_formula, data=marchinko)
    b_xy <- coefficients(summary(fit))['treatmentpred', 'Estimate']
    # get sd from control group only
    model_formula <- formula(paste(y, " ~ standlength + (standlength|family)", sep=''))
    fit <- lmer(model_formula, data=marchinko[treatment=='no',])
    sd_x <- unlist(as.data.table(VarCorr(fit))[grp=='Residual','sdcor'])
    sd_x <- 1
    table_1_mlm[j] <- b_xy/sd_x
}
table_1_mlm</pre>
```

[1] -0.04118469 -0.02889154 -0.02478541 -0.00759773

Mendelian randomization of Marchinko armor data

It doesn't make sense to compute this for each armor trait separately because then by definition there assumptions are violated (Z has a path to W independent of X). Possible solution here: "Multivariable Mendelian randomization: the use of pleiotropic genetic variants to estimate causal effects", doi: 10.1093/aje/kwu283.

 β_{zy} is the treatment effect on eda genotype.

```
# additive model of eda
marchinko[, eda_add:=ifelse(eda.genotype=="aa",-1,ifelse(eda.genotype=="Aa",0,1))]
# dominance model of eda
marchinko[, eda_dom:=ifelse(eda.genotype=="aa",0,1)]

fit <- lmer(eda_add ~ treatment + (1|family), data=marchinko)
b_zy <- coefficients(summary(fit))['treatmentpred', 'Estimate']

# get sd_z from control group only
fit <- lmer(eda_add ~ (1|family), data=marchinko[treatment=='no',])</pre>
```

```
sd_z<- unlist(as.data.table(VarCorr(fit))[grp=='Residual','sdcor'])</pre>
b_zy_prime <- b_zy/sd_z</pre>
\beta_{zx} is the effect of eda genotype on armor phenotype in the control group.
p <- length(y_list)</pre>
b_zx <- numeric(p)</pre>
sd_x <- numeric(p)</pre>
b_zx_prime <- numeric(p)</pre>
table_1_mr <- numeric(p)</pre>
table 1 mr prime <- numeric(p)
for(j in 1:length(y_list)){
 y <- y_list[j]</pre>
  # fit to get treatment effect
  model_formula <- formula(paste(y, " ~ standlength + eda_add + (standlength|family)", sep=''))</pre>
  fit <- lmer(model_formula, data=marchinko[treatment=='no'])</pre>
  b_zx[j] <- coefficients(summary(fit))['eda_add', 'Estimate']</pre>
  # get sd_x from control group only and from non-RM model
  model_formula <- formula(paste(y, " ~ standlength + (standlength|family)", sep=''))</pre>
  fit <- lmer(model_formula, data=marchinko[treatment=='no',])</pre>
  sd_x[j] <- unlist(as.data.table(VarCorr(fit))[grp=='Residual','sdcor'])</pre>
  table_1_mr[j] \leftarrow b_zy/b_zx[j]
  b_{zx_prime[j]} \leftarrow b_{zx[j]*(sd_z/sd_x[j])}
  table_1_mr_prime[j] <- b_zy_prime/b_zx_prime[j]</pre>
table 1 mr
## [1] -7.398726 -9.473280 -7.419083 -5.903709
table_1_mr_prime
## [1] -1.707147 -2.228926 -2.624435 -5.364176
table_1_mlm
## [1] -0.04118469 -0.02889154 -0.02478541 -0.00759773
table_1_replicate
##
       standlength ant.dorspine.s sec.dorspine.s pelspine.len.s
```

-0.065

##

0.403

0.011

pelgirdle.len.s

-0.225