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# Competition and Selection

## Accounting for behavioral measurement uncertainty

```
library(MASS) # MASS clashes with dplyr... so always load first
library(pander) # pander clashes with dplyr... so always load first
library(ggplot2)
library(grid)
library(dplyr)
set.alignment('right', row.names = 'left')
```

```
load("data/analyses_data/sel_grads_mcmc.RData")
fitness <- read.table(file = "data/fitness+competition.csv", sep = ',',
   header = TRUE, stringsAsFactors = FALSE)
load("data/analyses_data/fit_raneff_data.RData")</pre>
```

We will examine the effect of competition on selection in two general steps.

- 1. Is there an interaction between competition and behavior on fitness?
- 2. Are there nonlinear effects of behavior on fitness?

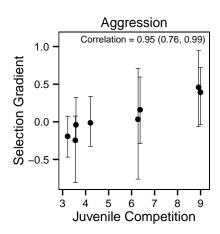
#### Correlations between selection gradients and competition

First a plot of the relationship between selection gradients and competition. The two study areas were pooled to calculate selection gradients for each year. Therefore we need to calculate competition for the combined study areas. Competition is the number of offspring produced during the year divided by the number of offspring that survived to spring (i.e. recruited into the population).

```
competition_year <- fitness %>%
  filter(grid_year != "SU2008") %>%
  select(Year, competition) %>%
  unique() %>%
  group by (Year, add = FALSE) %>%
  summarise(mean_competition = mean(competition))
n_year <- filter(fit_raneff_data, type == "blup") %>%
  group_by(Year, add = FALSE) %>% summarise(n = n())
competition_year <- left_join(competition_year, n_year, by = "Year")</pre>
competition_year$Year <- as.character(competition_year$Year)</pre>
load("data/analyses_data/sel_grads_mcmc.RData")
sel_grads_mcmc_comp <- left_join(</pre>
  filter(sel_grads_mcmc, standardization == "SD"), competition_year,
  by = "Year")
save(sel_grads_mcmc_comp, competition_year,
  file = "data/analyses_data/sel_grads_mcmc_comp.RData")
load("data/analyses_data/sel_grads_mcmc_comp.RData")
cor_sgrad_comp <- function(x){</pre>
  v <- x$variable[1]</pre>
  ct <- cor.test(x$post_mode, x$mean_competition)</pre>
  data.frame(variable = v, est = ct$estimate, lower = ct$conf.int[1],
    upper = ct$conf.int[2], stringsAsFactors = FALSE)
mcmc_cor <- sel_grads_mcmc_comp %>%
  group_by(variable, add = FALSE) %>%
  do(cor_sgrad_comp(.))
mcmc_cor[ ,2:4] <- round(mcmc_cor[ ,2:4], digits = 2)
mcmc_cor$print <- paste(mcmc_cor$est,</pre>
  " (", mcmc_cor$lower, ", ", mcmc_cor$upper, ")", sep = "")
```

## Aggression and Competition

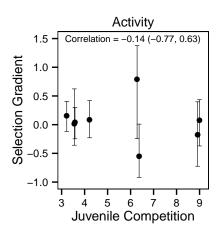
```
p <- ggplot(data = filter(sel_grads_mcmc_comp, variable == "Aggression"),</pre>
  aes(x = mean_competition, y = post_mode))
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()</pre>
p \leftarrow p + theme_bw(base_size = 10)
p \leftarrow p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")</pre>
p <- p + ggtitle("Aggression")</pre>
p <- p + xlab("Juvenile Competition")</pre>
p <- p + theme(panel.grid.major = element_blank(),</pre>
  panel.grid.minor = element_blank(), panel.background = element_blank(),
  strip.background = element_blank(), strip.text = element_text(size = 10),
  panel.border = element_rect(linetype = "solid", colour = "black"),
  plot.title = element_text(size = 10))
p <- p + geom_text(data = filter(mcmc_cor, variable == "Aggression"),</pre>
  aes(x = 6.5, y = 1.1, label = paste("Correlation = ", print, sep = '')),
  size = 2.5)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))
```



#### **Activity and Competition**

```
p <- ggplot(data = filter(sel_grads_mcmc_comp, variable == "Activity"),
    aes(x = mean_competition, y = post_mode))
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()
p <- p + theme_bw(base_size = 10)
p <- p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")
p <- p + ggtitle("Activity")
p <- p + xlab("Juvenile Competition")
p <- p + theme(panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(), panel.background = element_blank(),
    strip.background = element_blank(), strip.text = element_text(size = 10),</pre>
```

```
panel.border = element_rect(linetype = "solid", colour = "black"),
  plot.title = element_text(size = 10))
p <- p + ylim(c(-1, 1.5))
p <- p + geom_text(data = filter(mcmc_cor, variable == "Activity"),
  aes(x = 6.1, y = 1.5, label = paste("Correlation = ", print, sep = '')),
  size = 2.5)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))</pre>
```

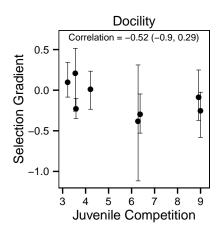


#### **Docility and Competition**

```
p <- ggplot(data = filter(sel_grads_mcmc_comp, variable == "Docility"),</pre>
  aes(x = mean\_competition, y = post\_mode))
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()</pre>
p <- p + theme_bw(base_size = 10)</pre>
p \leftarrow p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")</pre>
p <- p + ggtitle("Docility")</pre>
p <- p + xlab("Juvenile Competition")</pre>
p <- p + theme(panel.grid.major = element_blank(),</pre>
  panel.grid.minor = element_blank(), panel.background = element_blank(),
  strip.background = element_blank(), strip.text = element_text(size = 10),
  panel.border = element_rect(linetype = "solid", colour = "black"),
  plot.title = element_text(size = 10))
p <- p + geom_text(data = filter(mcmc_cor, variable == "Docility"),</pre>
  aes(x = 6.1, y = 0.65, label = paste("Correlation = ", print, sep = '')),
  size = 2.5)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))
```

#### Effect of competition on linear selection (glmms)

```
load("data/analyses_data/fit_raneff_data.RData")
library(dplyr)
fit_raneff_data <- tbl_df(fit_raneff_data)</pre>
```



#### library(lme4)

```
# Model with interactions between competition and the behavioral traits.
# grid_year and ID are random effects.
arsLinearCompetition <- function(dat){</pre>
  ars_linear_comp <- glmer(</pre>
    ars_all ~
    competition_s +
    aggression_s +
    competition_s:aggression_s +
    activity_s +
    competition_s:activity_s +
    docility_s +
    competition_s:docility_s +
    (1|Grid) + (1|ID),
    data = dat, family = poisson, control=glmerControl(optimizer="bobyqa")
  random_effect_variances <- VarCorr(ars_linear_comp)</pre>
  data.frame(t(summary(ars_linear_comp)$coefficients[ ,"Estimate"]),
  ID = random_effect_variances$ID[1], Grid = random_effect_variances$Grid[1])
library(foreach)
## foreach: simple, scalable parallel programming from Revolution Analytics
## Use Revolution R for scalability, fault tolerance and more.
## http://www.revolutionanalytics.com
library(doMC)
## Loading required package: iterators
## Loading required package: parallel
ncores = 12
registerDoMC(cores = ncores)
```

```
batches <- data.frame(start = seq(1, 1000, round(1000/ncores))[1:ncores])</pre>
batches$stop <- c(batches$start[2:length(batches$start)] - 1, 1000)</pre>
start_time <- Sys.time()</pre>
ars_linear_comp_posterior <- foreach(i = 1:ncores, .combine = rbind) %dopar% {</pre>
  results <- fit_raneff_data %>%
    filter(type == "raneff", itt %in% batches$start[i]:batches$stop[i]) %>%
    group_by(itt, add = FALSE) %>%
    do(arsLinearCompetition(.))
}
run_time <- Sys.time() - start_time</pre>
print(run_time)
## Time difference of 2.539 mins
save(ars_linear_comp_posterior,
  file = "data/analyses_data/ars_linear_comp_posterior.RData")
load("data/analyses_data/ars_linear_comp_posterior.RData")
library(MCMCglmm)
library(lme4)
library(data.table)
## data.table 1.9.2 For help type: help("data.table")
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##
       between, last
getPosteriorParams <- function(x){</pre>
  require(MCMCglmm)
  dat_mcmc <- mcmc(x)</pre>
  pm <- posterior.mode(dat_mcmc)</pre>
  hpd <- HPDinterval(dat_mcmc, prob = 0.9)</pre>
  pm_table <- format(round(pm , digits = 2), digits = 1, nsmall = 2,</pre>
    scientific = FALSE)
  hpd_table <- format(round(hpd, digits = 2), digits = 1, nsmall = 2,
    scientific = FALSE)
  pm_hpd_table <- data.frame(cbind(pm_table, hpd_table))</pre>
  pm_hpd_table$pm_hpd <- paste(</pre>
    pm_table, " (", hpd_table[ ,1], ", ", hpd_table[ ,2], ")", sep = ''
  pm_hpd_table$sig[sign(hpd[ ,1]) == sign(hpd[ ,2])] <- "*"</pre>
  pm_hpd_table$sig[sign(hpd[ ,1]) != sign(hpd[ ,2])] <- " "</pre>
  pm_hpd_table$space[sign(hpd[ ,1]) == sign(hpd[ ,2])] <- "*"</pre>
  pm_hpd_table$space[sign(hpd[ ,1]) != sign(hpd[ ,2])] <- "&nbsp;"</pre>
  pm_hpd_table$pm_hpd <- paste(pm_hpd_table$pm_hpd, pm_hpd_table$sig, sep = "")</pre>
```

```
return(pm_hpd_table)
}
linear_hpd_ars <- getPosteriorParams(ars_linear_comp_posterior %>%
  ungroup() %>%
select(Intercept = X.Intercept., Competition = competition_s,
  Aggression = aggression_s, Activity = activity_s, docility = docility_s,
  "Competition x Aggression" = competition_s.aggression_s,
  "Competition x Activity" = competition_s.activity_s,
  "Competition x Docility" = competition_s.docility_s)
)
```

**Interaction between competition and linear selection results** The effect of competition on linear selection on female behavioral traits for annual reproductive success. Posterior modes are given with highest posterior density intervals in parentheses.

pandoc.table(linear\_hpd\_ars %>% select(pm\_hpd), justify="right")

|                           | pm_hpd                    |
|---------------------------|---------------------------|
| Intercept                 | -1.01 (-1.11, -0.93)*     |
| Competition               | -3.27 (-3.65, -3.09)*     |
| Aggression                | 0.36 ( 0.13, 0.63)*       |
| Activity                  | -0.12 (-0.37, 0.15)       |
| $\operatorname{docility}$ | -0.18 (-0.38, -0.06)*     |
| Competition x Aggression  | 1.30 ( 0.50, 1.99)*       |
| Competition x Activity    | $-0.45 \ (-1.26, \ 0.32)$ |
| Competition x Docility    | -0.54 (-1.13, -0.09)*     |

#### Competition and nonlinear selection (glmms)

```
load("data/analyses_data/fit_raneff_data.RData")
library(lme4)

arsNonlinearResults <- function(dat){
    ars_model <- glmer(ars_all ~ aggression_s*competition_s +
        activity_s*competition_s + docility_s*competition_s +
        aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
        I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
        (1 | grid_year) + (1|ID), data = dat, family = poisson,
        control=glmerControl(optimizer="bobyqa"))
kpd_model <- glmer(kprod ~ aggression_s*competition_s +
        activity_s*competition_s + docility_s*competition_s +
        aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
        I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +</pre>
```

```
(1 | grid_year) + (1|ID), data = dat, family = poisson,
    control=glmerControl(optimizer="bobyqa"))
  ows_model <- glmer(prop ~ aggression_s*competition_s +</pre>
    activity_s*competition_s + docility_s*competition_s +
    aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
    I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
    (1 | grid_year) + (1|ID), data = dat, weights = kprod, family = binomial,
    control=glmerControl(optimizer="bobyqa"))
  ars_vc <- VarCorr(ars_model)</pre>
  ars_t <- data.table(fitness = "ars",</pre>
    t(summary(ars_model)$coefficients[,"Estimate"]), ID = ars_vc$ID[1],
    grid_year = ars_vc$grid_year[1])
  kpd_vc <- VarCorr(kpd_model)</pre>
  kpd_t <- data.table(fitness = "kpd",</pre>
    t(summary(kpd_model)$coefficients[,"Estimate"]), ID = kpd_vc$ID[1],
    grid_year = kpd_vc$grid_year[1])
  ows_vc <- VarCorr(ows_model)</pre>
  ows_t <- data.table(fitness = "ows",</pre>
    t(summary(ows_model)$coefficients[ ,"Estimate"]), ID = ows_vc$ID[1],
    grid_year = ows_vc$grid_year[1])
  rbind(rbind(ars_t, kpd_t), ows_t)
}
library(foreach)
library(doMC)
ncores = 12
registerDoMC(cores = ncores)
batches <- data.frame(start = seq(1, 1000, round(1000/ncores))[1:ncores])</pre>
batches$stop <- c(batches$start[2:length(batches$start)] - 1, 1000)</pre>
start_time <- Sys.time()</pre>
nonlinear_mcmc <- foreach(i = 1:ncores, .combine = rbind) %dopar% {</pre>
  results <- fit_raneff_data %>%
    filter(type == "raneff", itt %in% batches$start[i]:batches$stop[i]) %>%
    group_by(itt, add = FALSE) %>%
    do(arsNonlinearResults(.))
}
run_time <- Sys.time() - start_time</pre>
print(run_time)
## Time difference of 20.4 mins
save(nonlinear_mcmc, file = "data/analyses_data/nonlinear_mcmc_models.RData")
load("data/analyses_data/nonlinear_mcmc_models.RData")
pm_hpd_ars <- getPosteriorParams(</pre>
  nonlinear_mcmc[nonlinear_mcmc$fitness =="ars", 3:18])
```

```
pm_hpd_ows <- getPosteriorParams(
    nonlinear_mcmc[nonlinear_mcmc$fitness =="ows", 3:18])
pm_hpd_kpd <- getPosteriorParams(
    nonlinear_mcmc[nonlinear_mcmc$fitness =="kpd", 3:18])

nonlinear_results_mcmc <- data.frame(ARS = pm_hpd_ars$pm_hpd,
    OWS = pm_hpd_ows$pm_hpd, Fecundity = pm_hpd_kpd$pm_hpd)

row.names(nonlinear_results_mcmc) <- c("Intercept", "Aggression",
    "Competition", "Activity", "Docility", "Aggression^2", "Activity^2",
    "Docility^2", "Aggression x Competition", "Activity x Competition",
    "Docility x Competition", "Aggression x Activity",
    "Aggression^2 x Competition", "Activity^2 x Competition",
    "Docility^2 x Competition", "Agg. x Act. x Competition"
)</pre>
```

#### Nonlinear results

The effect of competition on linear and nonlinear selection on female behavioral traits for annual reproductive success. Posterior modes are given with highest posterior density intervals in parentheses.

```
pandoc.table(nonlinear_results_mcmc[c(1,3,2,4:16), ],
    split.tables = 160)
```

|  | ARS                      | OWS                      | Fecundity                |
|--|--------------------------|--------------------------|--------------------------|
| Intercept                                | -0.76 (-1.18, -0.42)*    | -1.84 (-2.35, -1.42)*    | 1.34 ( 1.28, 1.41)*      |
| Competition                              | -2.58 (-3.98, -1.72)*    | -2.41 (-4.21, -1.27)*    | -0.06 (-0.19, 0.09)      |
| Aggression                               | 0.45 ( 0.08, 0.82)*      | 0.59 ( 0.17, 1.13)*      | -0.04 (-0.09, 0.02)      |
| Activity                                 | -0.17 (-0.53, 0.17)      | -0.06 (-0.68, 0.24)      | $0.03 \ (-0.05, \ 0.06)$ |
| Docility                                 | -0.29 (-0.55, -0.13)*    | -0.29 (-0.54, -0.03)*    | 0.04 ( 0.00, 0.07)*      |
| Aggression <sup>2</sup>                  | -0.12 (-0.51, 0.15)      | -0.28 (-0.70, 0.16)      | $0.02 \ (-0.05, \ 0.05)$ |
| Activity <sup>2</sup>                    | 0.00 (-0.36, 0.23)       | -0.07 (-0.42, 0.31)      | -0.02 (-0.07, 0.02)      |
| Docility^2                               | -0.15 (-0.33, 0.04)      | -0.20 (-0.41, 0.05)      | $0.01 \ (-0.03, \ 0.04)$ |
| Aggression x Competition                 | 1.71 ( 0.48, 2.66)*      | 2.16 ( 0.55, 3.53)*      | -0.03 (-0.16, 0.09)      |
| Activity x Competition                   | -0.72 (-1.77, 0.35)      | -0.24 (-1.70, 1.10)      | -0.03 (-0.15, 0.08)      |
| Docility x Competition                   | -0.70 (-1.55, -0.26)*    | -0.45 (-1.13, 0.49)      | -0.04 (-0.14, 0.02)      |
| Aggression x Activity                    | $0.14 \ (-0.22, \ 0.67)$ | 0.21 (-0.29, 0.84)       | 0.02 (-0.06, 0.08)       |
| Aggression <sup>2</sup> x<br>Competition | -0.27 (-1.36, 0.56)      | -0.35 (-1.98, 0.70)      | 0.03 (-0.11, 0.14)       |
| Activity <sup>2</sup> x Competition      | -0.06 (-1.07, 0.68)      | -0.11 (-1.32, 1.00)      | 0.00 (-0.11, 0.11)       |
| Docility <sup>2</sup> x Competition      | -0.51 (-1.05, 0.15)      | -0.34 (-1.29, 0.19)      | -0.02 (-0.11, 0.05)      |
| Agg. x Act. x Competition                | $0.44 \ (-0.75, \ 1.92)$ | $0.52 \ (-1.08, \ 2.57)$ | 0.01 (-0.20, 0.16)       |

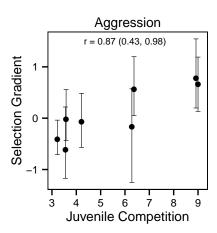
## Ignoring behavioural uncertainty

```
load("data/analyses_data/sel_grads_blup.RData")
load("data/analyses_data/fit_raneff_data.RData")
```

## Correlations between selection gradients and competition

```
competition_year <- fitness %>%
 filter(grid_year != "SU2008") %>%
 select(Year, competition) %>%
 unique() %>%
  group_by(Year, add = FALSE) %>%
  summarise(mean_competition = mean(competition))
n_year <- filter(fit_raneff_data, type == "blup") %>%
 group_by(Year, add = FALSE) %>%
 summarise(n = n())
competition_year <- left_join(competition_year, n_year, by = "Year")</pre>
competition_year$Year <- as.character(competition_year$Year)</pre>
load("data/analyses_data/sel_grads_blup.RData")
sel_grads_blup_competition <- left_join(</pre>
 filter(sel_grads_blup, standardization == "SD"), competition_year,
 by = "Year")
save(sel_grads_blup_competition, competition_year,
 file = "data/analyses_data/sel_grads_blup_competition.RData")
load("data/analyses_data/sel_grads_blup_competition.RData")
cor_sgrad_comp <- function(x){</pre>
 v <- x$variable[1]</pre>
 ct <- cor.test(x$coefficients, x$mean_competition)</pre>
 data.frame(variable = v, est = ct$estimate, lower = ct$conf.int[1],
    upper = ct$conf.int[2], stringsAsFactors = FALSE)
}
sg.comp <- sel_grads_blup_competition %>%
 group_by(variable, add = FALSE) %>%
 do(cor_sgrad_comp(x=.))
sg.comp[,2:4] \leftarrow round(sg.comp[,2:4], digits = 2)
sg.comp$print <- paste(sg.comp$est, " (", sg.comp$lower, ", ",</pre>
 sg.comp$upper, ")", sep = "")
Aggression and Competition
p <- ggplot(data = filter(sel_grads_blup_competition,</pre>
 variable == "Aggression"), aes(x = mean\_competition, y = coefficients))
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()</pre>
p <- p + theme_bw(base_size = 10)</pre>
```

```
p <- p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")
p <- p + ggtitle("Aggression")
p <- p + xlab("Juvenile Competition")
p <- p + theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(), panel.backgro
p <- p + geom_text(data = filter(sg.comp, variable == "Aggression"),
    aes(x = 6, y = 1.5, label = paste("r = ", print, sep = '')), size = 2.5)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))</pre>
```



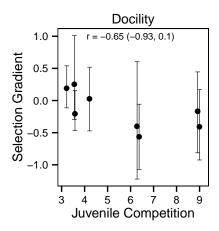
#### **Activity and Competition**

```
p <- ggplot(data = filter(sel_grads_blup_competition, variable == "Activity"),</pre>
  aes(x = mean\_competition, y = coefficients))
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()</pre>
p <- p + theme_bw(base_size = 10)</pre>
p \leftarrow p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")</pre>
p <- p + ggtitle("Activity")</pre>
p <- p + xlab("Juvenile Competition")</pre>
p <- p + theme(panel.grid.major = element_blank(),</pre>
  panel.grid.minor = element_blank(), panel.background = element_blank(),
  strip.background = element_blank(), strip.text = element_text(size = 10),
  panel.border = element_rect(linetype = "solid", colour = "black"),
  plot.title = element text(size = 10))
p <- p + geom_text(data = filter(sg.comp, variable == "Activity"),</pre>
  aes(x = 6, y = 2.2, label = paste("r = ", print, sep = '')), size = 2.5)
  pdf(file = "test.pdf", width = 2.17, height = 2.03)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))
```

## **Docility and Competition**

```
p <- ggplot(data = filter(sel_grads_blup_competition, variable == "Docility"),
   aes(x = mean_competition, y = coefficients))</pre>
```

```
p <- p + geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2, size = 0.2)
p <- p + geom_point()
p <- p + theme_bw(base_size = 10)
p <- p + scale_x_continuous(breaks = c(3,4,5,6,7,8,9))
p <- p + ylab("Selection Gradient")
p <- p + ggtitle("Docility")
p <- p + xlab("Juvenile Competition")
p <- p + theme(panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(), panel.background = element_blank(),
    strip.background = element_blank(), strip.text = element_text(size = 10),
    panel.border = element_rect(linetype = "solid", colour = "black"),
    plot.title = element_text(size = 10))
p <- p + geom_text(data = filter(sg.comp, variable == "Docility"),
    aes(x = 6, y = 1, label = paste("r = ", print, sep = '')), size = 2.5)
p + theme(plot.margin = unit(c(0.1, 0.1, 0.1, 0.1), "cm"))</pre>
```



## Effect of competition on linear selection

```
load("data/analyses_data/fit_raneff_data.RData")
library(dplyr)
fit_raneff_data <- tbl_df(fit_raneff_data)</pre>
library(lme4)
fit_raneff_data$oID <- 1:nrow(fit_raneff_data)</pre>
ars_linear_model <- glmer(ars_all ~ aggression_s*competition_s +
 activity_s*competition_s + docility_s*competition_s + (1 | grid_year) +
  (1|ID) + (1|oID), data = filter(fit_raneff_data, type == "blup"),
 family = poisson, control=glmerControl(optimizer="bobyqa"))
kpd_linear_model <- glmer(kprod ~ aggression_s*competition_s +</pre>
  activity_s*competition_s + docility_s*competition_s + (1 | grid_year) +
  (1|ID) + (1|oID), data = filter(fit_raneff_data, type == "blup"),
 family = poisson, control=glmerControl(optimizer="bobyqa"))
ows_linear_model <- glmer(prop ~ aggression_s*competition_s +</pre>
  activity_s*competition_s + docility_s*competition_s + (1 | grid_year) +
  (1|ID) + (1|oID), data = filter(fit_raneff_data, type == "blup"),
```

```
weights = kprod, family = binomial, control=glmerControl(optimizer="bobyqa"))
save(ars_linear_model, kpd_linear_model, ows_linear_model,
 file = "data/analyses_data/ars_linear_blup_models.RData")
Models with observation level random effect.
load("data/analyses_data/fit_raneff_data.RData")
library(dplyr)
fit_raneff_data <- tbl_df(fit_raneff_data)</pre>
library(lme4)
fit_raneff_data$oID <- 1:nrow(fit_raneff_data)</pre>
ars_linear_model_ <- glmer(ars_all ~ aggression_s*competition_s +
 activity_s*competition_s + docility_s*competition_s + (1 | oID) +
  (1 | grid_year) + (1|ID), data = filter(fit_raneff_data, type == "blup"),
 family = poisson, control=glmerControl(optimizer="bobyqa"))
kpd_linear_model_ <- glmer(kprod ~ aggression_s*competition_s +</pre>
 activity_s*competition_s + docility_s*competition_s + (1 | oID) +
  (1 | grid_year) + (1|ID), data = filter(fit_raneff_data, type == "blup"),
 family = poisson, control=glmerControl(optimizer="bobyqa"))
ows_linear_model_ <- glmer(prop</pre>
                                    ~ aggression_s*competition_s +
 activity_s*competition_s + docility_s*competition_s + (1 | oID) +
  (1 | grid_year) + (1|ID), data = filter(fit_raneff_data, type == "blup"),
 weights = kprod, family = binomial, control=glmerControl(optimizer="bobyqa"))
save(ars_linear_model, kpd_linear_model, ows_linear_model,
 file = "data/analyses_data/ars_linear_blup_models.RData")
load("data/analyses_data/ars_linear_blup_models.RData")
library(lme4)
getLmerParams <- function(x){</pre>
  coefs <- summary(x)$coefficients</pre>
  coef.table <- data.frame(format(coefs[ ,1:3], digits = 1, nsmall = 2,</pre>
    scientific = FALSE))
  coef.table$pval[coefs[,4] > 0.001] \leftarrow format(coefs[coefs[,4] > 0.001, 4],
    digits = 1, nsmall = 2)
  coef.table$pval[coefs[ ,4] < 0.001] <- "< 0.001"
  coef.table$coefs <- paste(coef.table$Estimate, " ±", coef.table$Std..Error,</pre>
    sep = '')
 return(coef.table)
}
ars_linear_blup_results <- getLmerParams(ars_linear_model)</pre>
row.names(ars_linear_blup_results) <- c("Intercept", "Aggression",</pre>
  "Competition", "Activity", "Docility", "Aggression x Competition",
  "Activity x Competition", "Docility x Competition")
names(ars_linear_blup_results) <- c("Estimate", "SE", "Z", "P", "Est ± se")</pre>
```

#### Interaction between competition and linear selection results

```
pandoc.table(ars_linear_blup_results[c(1,3,2,4:8), c(5,4,3)],
  justify = "right", split.tables = 160)
```

|                          | Est $\pm$ se     | P       | Z     |
|--------------------------|------------------|---------|-------|
| Intercept                | $-1.10 \pm 0.22$ | < 0.001 | -5.09 |
| Competition              | $-3.62 \pm 0.64$ | < 0.001 | -5.64 |
| Aggression               | $0.74 \pm 0.21$  | < 0.001 | 3.49  |
| Activity                 | $-0.32 \pm 0.20$ | 0.11    | -1.61 |
| Docility                 | $-0.27 \pm 0.16$ | 0.10    | -1.67 |
| Aggression x Competition | $2.62 \pm 0.64$  | < 0.001 | 4.07  |
| Activity x Competition   | $-1.36 \pm 0.61$ | 0.03    | -2.21 |
| Docility x Competition   | $-0.77 \pm 0.49$ | 0.11    | -1.58 |

## Competition and nonlinear selection

```
load("data/analyses_data/fit_raneff_data.RData")
library(lme4)
fit_blups_data <- filter(fit_raneff_data, type == "blup")</pre>
fit_blups_data$oID <- 1:nrow(fit_blups_data)</pre>
ars_nl_model <- glmer(ars_all ~ aggression_s*competition_s +</pre>
 activity_s*competition_s + docility_s*competition_s +
 aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
 I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
  (1 | grid_year) + (1|ID) + (1|oID),
 data = fit_blups_data, family = poisson,
 control=glmerControl(optimizer="bobyqa"))
ars_nl_model <- glmer(ars_all ~ aggression_s*competition_s +</pre>
 activity_s*competition_s + docility_s*competition_s +
 aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
 I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
  (1 | grid_year) + (1|ID),
 data = fit_blups_data, family = poisson,
 control=glmerControl(optimizer="bobyqa"))
kpd nl model <- glmer(kprod ~ aggression s*competition s +</pre>
 activity s*competition s + docility s*competition s +
 aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
  I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
  (1 | grid_year) + (1|ID) + (1|oID),
  data = fit_blups_data, family = poisson,
  control=glmerControl(optimizer="bobyqa"))
```

```
ows_nl_model <- glmer(prop ~ aggression_s*competition_s +</pre>
 activity_s*competition_s + docility_s*competition_s +
  aggression_s*activity_s*competition_s + I(aggression_s^2)*competition_s +
 I(activity_s^2)*competition_s + I(docility_s^2)*competition_s +
  (1 | grid_year) + (1|ID) + (1|oID),
 data = fit_blups_data, weights = kprod,
 family = binomial,
 control=glmerControl(optimizer="bobyqa"))
save(ars_nl_model,kpd_nl_model, ows_nl_model,
 file = "data/analyses_data/nl.blup_models.RData")
fit_raneff_data %>%
 ungroup() %>%
  summarise(
    mean_ars = mean(ars_all, na.rm = TRUE),
    var_ars = var(ars_all, na.rm = TRUE),
    mean_kpd = mean(kprod, na.rm = TRUE),
    var_kpd = var(kprod, na.rm = TRUE)
## Source: local data frame [1 x 4]
##
##
   mean_ars var_ars mean_kpd var_kpd
      0.8784 1.255 3.902 4.372
Format results of nonlinear selection for table
load("data/analyses_data/nl.blup_models.RData")
coef_p_ars <- getLmerParams(ars_nl_model)</pre>
coef_p_ows <- getLmerParams(ows_nl_model)</pre>
coef_p_kpd <- getLmerParams(kpd_nl_model)</pre>
term_names <- c("Intercept", "Aggression",</pre>
 "Competition", "Activity", "Docility", "Aggression^2", "Activity^2",
 "Docility^2", "Aggression x Competition", "Activity x Competition",
 "Docility x Competition", "Aggression x Activity",
 "Aggression^2 x Competition", "Activity^2 x Competition",
 "Docility^2 x Competition", "Agg. x Act. x Competition"
row.names(coef_p_ars) <- term_names</pre>
row.names(coef_p_ows) <- term_names</pre>
row.names(coef_p_kpd) <- term_names</pre>
names(coef_p_ars) <- c("Estimate", "SE", "Z", "P", "Est ± se")</pre>
names(coef_p_ows) <- c("Estimate", "SE", "Z", "P", "Est ± se")</pre>
names(coef_p_kpd) <- c("Estimate", "SE", "Z", "P", "Est ± se")</pre>
```

#### Nonlinear results

The effect of competition on linear and nonlinear selection on female behavioral traits for annual reproductive success. Posterior modes are given with highest posterior density intervals in parentheses.

 $\mathbf{ARS}$ 

```
pandoc.table(coef_p_ars[c(1,3,2,4:16), c(5,3,4)],
    split.tables = 160)
```

|                                       | Est $\pm$ se     | Z     | Р       |
|---------------------------------------|------------------|-------|---------|
| Intercept                             | $-0.68 \pm 0.31$ | -2.23 | 0.025   |
| Competition                           | $-2.24 \pm 0.91$ | -2.46 | 0.014   |
| Aggression                            | $0.96\pm0.31$    | 3.06  | 0.002   |
| Activity                              | $-0.66 \pm 0.28$ | -2.35 | 0.019   |
| Docility                              | $-0.51 \pm 0.23$ | -2.18 | 0.029   |
| ${f Aggression ^2}$                   | $-0.65 \pm 0.32$ | -2.07 | 0.039   |
| Activity <sup>2</sup>                 | $-0.30 \pm 0.26$ | -1.16 | 0.247   |
| Docility^2                            | $-0.18 \pm 0.12$ | -1.48 | 0.139   |
| Aggression x Competition              | $3.40\pm0.92$    | 3.68  | < 0.001 |
| Activity x Competition                | $-2.45 \pm 0.84$ | -2.90 | 0.004   |
| Docility x Competition                | $-1.41 \pm 0.69$ | -2.04 | 0.041   |
| Aggression x Activity                 | $1.01 \pm 0.43$  | 2.35  | 0.019   |
| Aggression <sup>2</sup> x Competition | $-1.86 \pm 0.96$ | -1.93 | 0.053   |
| Activity <sup>2</sup> x Competition   | $-0.81 \pm 0.77$ | -1.05 | 0.295   |
| Docility <sup>2</sup> x Competition   | $-0.63 \pm 0.39$ | -1.61 | 0.107   |
| Agg. x Act. x Competition             | $2.62 \pm 1.32$  | 1.99  | 0.046   |

#### ows

```
pandoc.table(coef_p_ows[c(1,3,2,4:16), c(5,3,4)],
    split.tables = 160)
```

|             | Est $\pm$ se     | $\mathbf{Z}$ | P       |
|-------------|------------------|--------------|---------|
| Intercept   | $-1.92 \pm 0.39$ | -4.87        | < 0.001 |
| Competition | $-2.48 \pm 1.18$ | -2.11        | 0.035   |
| Aggression  | $1.33\pm0.42$    | 3.17         | 0.001   |
| Activity    | $-0.82 \pm 0.38$ | -2.14        | 0.032   |

|                                       | Est $\pm$ se     | Z     | Р       |
|---------------------------------------|------------------|-------|---------|
| Docility                              | $-0.49 \pm 0.29$ | -1.70 | 0.088   |
| ${\bf Aggression \hat{2}}$            | $-1.02 \pm 0.43$ | -2.36 | 0.018   |
| Activity <sup>2</sup>                 | $-0.37 \pm 0.36$ | -1.03 | 0.304   |
| Docility^2                            | $-0.18 \pm 0.15$ | -1.17 | 0.242   |
| Aggression x Competition              | $4.75\pm1.28$    | 3.70  | < 0.001 |
| Activity x Competition                | $-2.89 \pm 1.20$ | -2.40 | 0.016   |
| Docility x Competition                | $-0.78 \pm 0.90$ | -0.86 | 0.388   |
| Aggression x Activity                 | $1.40 \pm 0.60$  | 2.35  | 0.019   |
| Aggression <sup>2</sup> x Competition | $-3.11 \pm 1.36$ | -2.29 | 0.022   |
| Activity <sup>2</sup> x Competition   | $-1.17 \pm 1.10$ | -1.06 | 0.291   |
| Docility <sup>2</sup> x Competition   | $-0.50 \pm 0.50$ | -0.98 | 0.325   |
| Agg. x Act. x Competition             | $3.95 \pm 1.86$  | 2.13  | 0.034   |

## Fecundity

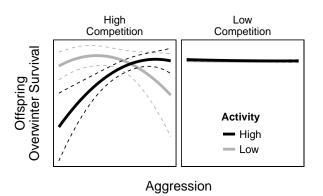
```
pandoc.table(coef_p_kpd[c(1,3,2,4:16), c(5,3,4)],
    split.tables = 160)
```

|                                       | Est $\pm$ se       | Z      | P       |
|---------------------------------------|--------------------|--------|---------|
| Intercept                             | $1.362 \pm 0.091$  | 14.937 | < 0.001 |
| Competition                           | $-0.017 \pm 0.161$ | -0.106 | 0.92    |
| Aggression                            | $-0.058 \pm 0.049$ | -1.176 | 0.24    |
| Activity                              | $0.030 \pm 0.050$  | 0.594  | 0.55    |
| Docility                              | $0.030 \pm 0.041$  | 0.724  | 0.47    |
| ${\bf Aggression \hat{2}}$            | $0.036 \pm 0.053$  | 0.672  | 0.50    |
| Activity <sup>2</sup>                 | $-0.064 \pm 0.051$ | -1.248 | 0.21    |
| Docility <sup>2</sup>                 | $0.005\pm0.026$    | 0.209  | 0.83    |
| Aggression x Competition              | $-0.067 \pm 0.117$ | -0.579 | 0.56    |
| Activity x Competition                | $0.008 \pm 0.088$  | 0.086  | 0.93    |
| Docility x Competition                | $-0.116 \pm 0.087$ | -1.325 | 0.19    |
| Aggression x Activity                 | $0.024\pm0.075$    | 0.319  | 0.75    |
| Aggression <sup>2</sup> x Competition | $0.153 \pm 0.149$  | 1.029  | 0.30    |
| Activity <sup>2</sup> x Competition   | $0.079\pm0.145$    | 0.545  | 0.59    |
| Docility <sup>2</sup> x Competition   | $-0.067 \pm 0.055$ | -1.218 | 0.22    |
| Agg. x Act. x Competition             | $-0.252 \pm 0.240$ | -1.051 | 0.29    |

#### Plot of quadratic interaction

```
library(effects)
## Loading required package: colorspace
## Attaching package: 'effects'
## The following object is masked from 'package:car':
##
       Prestige
library(ggplot2)
g.ows <- glm(prop ~ aggression_s * competition_s + activity_s *</pre>
    competition_s + docility_s * competition_s + aggression_s:activity_s *
    competition s + I(aggression s^2) * competition s, data = filter(fit raneff data,
    type == "blup"), weights = kprod, family = binomial)
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
e.ows <- effect(term = "aggression s:competition s:activity s",</pre>
    mod = g.ows)
e <- Effect(c("aggression_s", "activity_s", "competition_s"),</pre>
    g.ows, xlevels = list(aggression_s = 100, activity_s = 2,
        competition_s = 2))
s <- summary(e, type = "link")
se <- s$effect
su <- s$upper
sl <- s$lower
d <- as.data.frame(se)</pre>
du <- as.data.frame(su)
dl <- as.data.frame(sl)</pre>
names(d) <- c("la.lc", "ha.lc", "la.hc", "ha.hc")</pre>
names(du) <- c("la.lc", "ha.lc", "la.hc", "ha.hc")
names(dl) <- c("la.lc", "ha.lc", "la.hc", "ha.hc")</pre>
plot_d <- data.frame(Aggression = as.numeric(rep(row.names(s$effect),</pre>
    12)), OWS = c(d$1a.1c, d$ha.1c, d$1a.hc, d$ha.hc, du$1a.1c,
    du$ha.lc, du$la.hc, du$ha.hc, dl$la.lc, dl$ha.lc, dl$la.hc,
    dl$ha.hc), Competition = rep(rep(c("Low\nCompetition", "High\nCompetition"),
    each = 200), 3), Activity = rep(rep(c("Low", "High"), each = 100),
    6), type = rep(c("main", "upper", "lower"), each = 400))
plot_d$env <- paste(plot_d$Competition, plot_d$Activity, sep = ".")</pre>
quad_plot <- ggplot(plot_d, aes(x = Aggression, y = OWS)) + geom_line(aes(alpha = Activity,
    linetype = type, size = type)) + facet_wrap(~Competition) +
    scale_alpha_discrete(range = c(1, 0.3)) + scale_linetype_manual(values = c(2,
```

```
1, 2)) + scale_size_manual(values = c(0.3, 1, 0.3)) + ylab("Offspring\nOverwinter Survival") +
    xlab("Aggression") + theme_bw(base_size = 10) + theme(panel.grid.major = element_blank(),
   panel.grid.minor = element_blank(), panel.background = element_blank(),
   panel.border = element_rect(linetype = "solid", colour = "black"),
    axis.ticks = element_blank(), axis.text = element_text(size = 10),
    legend.key = element_blank(), strip.background = element_blank()) +
    guides(linetype = FALSE, size = FALSE, alpha = guide_legend(override.aes = list(size = 1))) +
    theme(legend.position = c(0.75, 0.25), legend.background = element_blank(),
        legend.key.size = unit(0.4, "cm")) + theme(plot.margin = unit(c(0.1,
    0.1, 0.1, 0.1), "cm")) + theme(axis.text.x = element_blank(),
    axis.text.y = element_blank())
pdf("figure/05_quad_print.pdf", width = 3.14, height = 2)
quad_plot
dev.off()
## pdf
##
quad_plot
```



## Tile plot of interaction

```
out <- paste(a, b, sep = ".")
    return(out)
}
d_hc <- data.frame(Aggression = rep(row.names(de_hc), n), Activity = rep(fix.names(de_hc),</pre>
    each = n), OWS = as.vector(as.matrix(de_hc)))
d_hc$Aggression <- as.numeric(as.character(d_hc$Aggression))</pre>
d_hc$Activity <- as.numeric(as.character(d_hc$Activity))</pre>
d_lc <- data.frame(Aggression = rep(row.names(de_lc), n), Activity = rep(fix.names(de_lc),</pre>
    each = n), OWS = as.vector(as.matrix(de_lc)))
d_lc$Aggression <- as.numeric(as.character(d_lc$Aggression))</pre>
d_lc$Activity <- as.numeric(as.character(d_lc$Activity))</pre>
tile_plot <- ggplot(d_lc, aes(x = Aggression, y = Activity, z = OWS)) +
    geom_tile(aes(alpha = OWS), fill = "black", size = 0) + scale_alpha_continuous(range = c(1,
    0)) + ylab("Activity") + xlab("Aggression") + theme_bw(base_size = 10) +
    theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
        panel.background = element_blank(), panel.border = element_rect(linetype = "solid",
            colour = "black"), axis.ticks = element_blank(),
        legend.key = element_blank(), strip.background = element_blank()) +
    guides(linetype = FALSE, size = FALSE, alpha = FALSE) + theme(plot.margin = unit(c(0.1,
    0.1, 0.1, 0.1), "cm"))
pdf("figure/05_tile_print.pdf", width = 3.14, height = 3.14)
tile_plot
dev.off()
## pdf
## 2
tile_plot
```

