# SSHI sockeye exploratory analyses (stage 2)

2019-10-29

This document summarizes some quick exploratory analyses that relate Fraser sockeye survival indices to infectious agent data. All code and associated data can be found on Github here.

Infectious agent data provided by A. Teffer, Fraser sockeye data are described here. NA's in load (i.e., years with no infected individuals) replaced with 0.

For brevity I have turned off warnings. But they need to be carefully considered when digging into these models moving forward. See note below about great tool for understanding/visualizing MCMC samples from models fit using STAN.

## load required packages and functions

```
library(lme4)
library(rstanarm) # https://mc-stan.org/users/documentation/case-studies/tutorial_rstanarm.html
library(ggplot2)
library(plotrix)
```

## load data, standardize independent variables

```
##
     Year orderID Stock_Analysis brood_year
                                               metric resid_value N N. prev
## 1 2009
              546
                        L.Stuart
                                        2007 SR resid -0.9099142 24
## 2 2009
             1072
                                        2007 SR_resid
                          Weaver
                                                        0.3732061 24
                                                                            0
## 3 2009
              844
                            Raft
                                        2007 SR_resid
                                                       -1.6863897 24
                                                                            0
## 4 2009
             1023
                        Stellako
                                        2007 SR_resid
                                                       -0.2379979 24
                                                                            0
## 5 2009
              416
                                        2007 SR_resid
                                                        1.3025948 24
                                                                            0
                        Harrison
                                                                       0
## 6 2009
              481
                       L.Shuswap
                                        2007 SR_resid
                                                        0.3932027 24
                                                                            0
     mean_load prevload agent mean_load_all
##
                                                prev_std
                                                           load std
                                                                         Stock
## 1
            NA
                     NA arena2
                                            0 -0.6770613 -0.4500811 L.Stuart
## 2
            NA
                     NA arena2
                                            0 -0.6770613 -0.4500811
                                                                        Weaver
## 3
            NA
                     NA arena2
                                            0 -0.6770613 -0.4500811
                                                                          Raft
                     NA arena2
## 4
            NA
                                            0 -0.6770613 -0.4500811 Stellako
## 5
                     NA arena2
                                            0 -0.6770613 -0.4500811 Harrison
            NΑ
                     NA arena2
## 6
            NA
                                            0 -0.6770613 -0.4500811 L.Shuswap
```

## A couple of key decisons to ponder before proceeding:

1. There are two alternatives modelling approaches we can try at this point. Both are generalized linear mixed effects modelling approaches, but one is fit using Maximum Likelihood methods (via lme4 R package) and the other is fit via Bayesian methods (via rstanarm R package). The formula syntax

used in both is nearly identical which makes their use very compliemntary. I illustree both below but suggest going the rstanarm route because parameter estimates can be more intuitively interpreted probabilistically and because it allows for more complete estimation of random effects and uncertainty. Plus there are some great tools developed for model evaluation (e.g., see STAN shiny app here)

- 2. In addition to the decision about estimation framework, we also need to decide on the most sensible random effects structure. Initial thinking was that we want the effect of infectious agent to be able to vary by stock which is analogous to a random slopes and intercept structure like "+ (infectious agent |stock)". We also probably want to account for the non-independence of observations within years given that the current dataset applies a single common measure of infection to all stocks in a given year; this is analogous to a random intercept structure like "+ (1|year)".
- 3. Lastly we have to decide whether or not to fit a single model to all agents simultaneously, or do this iteratively by stock. Initial model runs suggested that the single global model had a hard time converging. So for now let's stick to independent models for each agent.

### Prevalence

w/lme4

#### plot raw data

```
ggplot(inf_agt_resid_data,aes(prev_std, resid_value, color=Stock))+
    geom point()+
    facet_wrap(~ agent,nrow=5)+
  xlab("prevalence")+
  ylab("residual")+
  theme bw()
                            c_b_cys
             arena2
                                                            de sal
                                            ce sha
                                                                           fa mar
                                                                                              Birkenhead
     2.5
     0.0
                                                                                              Bowron
    -2.5
                                                                                              Chilko
    -5.0
                                                                                              E.Stuart
              fl_psy
                             ic hof
                                            ic mul
                                                            ku_thy
                                                                            lo sal
     2.5
                                                                                              Fennell
     0.0
    -2.5
                                                                                              Gates
    -5.0
                                                                                              Harrison
                            pa kab
             my_arc
                                            pa_min
                                                           pa_ther
                                                                             prv
                                                                                              L.Shuswap
     2.5
 residual
     0.0
                                                                                              L.Stuart
    -2.5
    -5.0
                                                                                              Nadina
                               rlo
                                              sch
                                                           smallUK
                                                                           sp_des
              pspv
                                                                                              Pitt
     2.5
                                                                                              Portage
     0.0
    -2.5
                                                                                              Quesnel
    -5.0
                                                        -2-1 0
                                                                       -2-1 0
                                                                                              Raft
             te_bry
                                              ven
                             te mar
     2.5
                                                                                              Scotch
     0.0
                                                                                              Seymour
    -2.5
    -5.0
                                                                                              Stellako
                         -2-1 0 1
                                        -2-1 0 1
                                         prevalence
                                                                                              Weaver
```

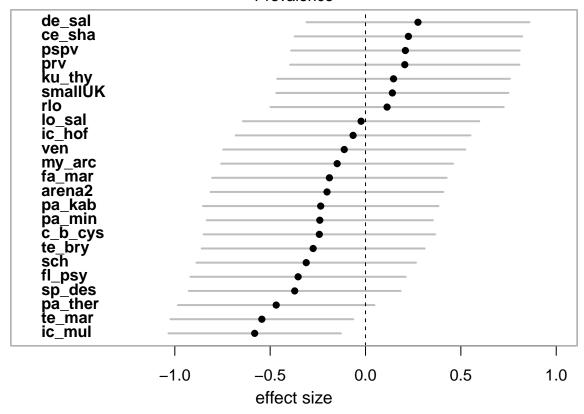
### loop through agents fitting individual models

## plot effect sizes

```
error bars are \pm 2 x SE
```

```
par(mfrow=c(1,1), mar=c(3,1,1,1), oma=c(0.5,0.5,0.5,0.5))
plotCI(x = coefs_order[,1],
   y = seq(1,length(agents)),
   ui = (coefs_order[,1]+(coefs_order[,2]*2)),
   li = (coefs_order[,1]-(coefs_order[,2]*2)),
   err = "x",
   sfrac = 0,
   gap = 0,
   yaxt = "n",
   xaxt = "n",
   ylab = "",
   xlab = "",
   xlim = c(-1.75,1),
   pch = 16,
   lwd = 2,
   scol = "grey")
text(rep(-1.75,length(agents)),
   seq(1,length(agents)),
   labels = rownames(coefs_order),
   pos =4,
   font = 2)
axis(1, at = c(-1, -0.5, 0, 0.5, 1))
abline(v = 0, lty = 2)
box(col="grey")
mtext("effect size",1,line=2.2, cex=1.1)
mtext("Prevalence",3,line=0.25)
```

## Prevalence



w/ STAN

loop through agents fitting individual models

change "eval = FALSE" in Rmd doc to actually run this code; takes 10-20 min.

```
coefs_stan <- matrix(NA,</pre>
                      nrow = length(agents),
                      ncol = 5,
                      dimnames = list(agents,c("lower","25","mid","75","upper")))
agents <- unique(inf_agt_resid_data$agent)</pre>
for(i in agents){
    data <- subset(inf_agt_resid_data, agent==i)</pre>
    model_ind_stan <- stan_lmer(resid_value ~ 0 + prev_std + (prev_std | Stock_Analysis) +(1 | Year),</pre>
                                   data = data,
                                   REML = F)
    ind_coef <- summary(model_ind_stan,</pre>
                          pars = c("prev_std"),
                          probs = c(0.025, 0.25, 0.5, 0.75, 0.975),
                          digits = 2)
    coefs_stan[i,] <- ind_coef[1,c(4:8)]</pre>
  }
write.csv(coefs_stan, "prev_coefs_stan.csv")
```

load parameter estimates if models above were not run.

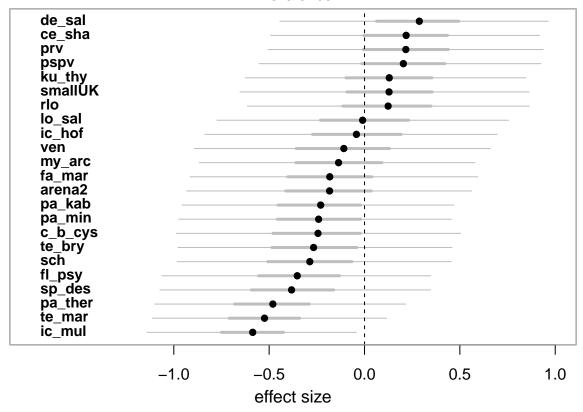
```
coefs_stan <- read.csv("prev_coefs_stan.csv")
rownames(coefs_stan) <- coefs_stan[,1]
coefs_stan <- coefs_stan[,-1] # drop first column with agent names</pre>
```

### plot effect sizes

error bars are 50% (thick grey) and 95% (thin grey) credibe interval

```
coefs_order <- coefs_stan[order(coefs_stan[,3]),]</pre>
par(mfrow=c(1,1), mar=c(3,1,1,1), oma=c(0.5,0.5,0.5,0.5))
plotCI(x = coefs_order[,3],
        y = seq(1,length(agents)),
        li = (coefs_order[,1]),
        ui = (coefs_order[,5]),
        err = "x",
        sfrac = 0,
        gap = 0,
        yaxt = "n",
        xaxt = "n",
        ylab = "",
        xlab = "",
        xlim = c(-1.75,1),
        pch = 16,
        scol = "grey")
plotCI(x = coefs_order[,3],
        y = seq(1,length(agents)),
        li = (coefs_order[,2]),
        ui = (coefs_order[,4]),
        err = "x",
        sfrac = 0,
        gap = 0,
        pch = 16,
        add = TRUE,
        lwd = 3,
        scol = "grey")
text(rep(-1.75,length(agents)),
        seq(1,length(agents)),
        labels = rownames(coefs_order),
        pos = 4,
        font = 2,
        cex=0.95)
axis(1, at = c(-1, -0.5, 0, 0.5, 1))
abline(v = 0, lty = 2)
box(col="grey")
mtext("effect size",1,line=2.2, cex=1.1)
mtext("Prevalence",3,line=0.25)
```

# Prevalence

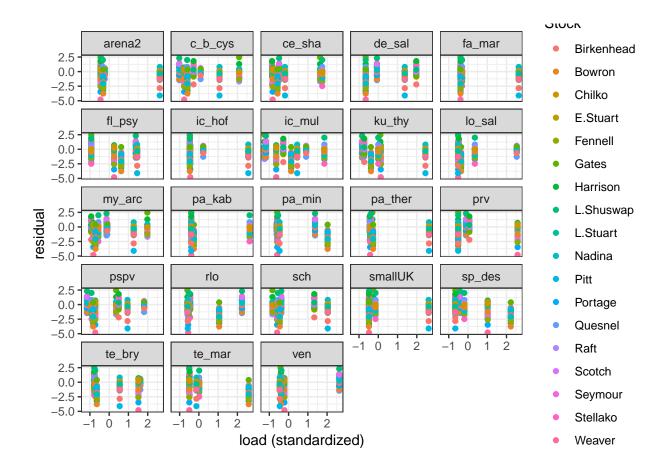


## Intensity

w/lme4

## plot raw data

```
ggplot(inf_agt_resid_data,aes(load_std, resid_value, color=Stock))+
   geom_point()+
   facet_wrap(~ agent,nrow=5)+
   xlab("load (standardized)")+
   ylab("residual")+
   theme_bw()
```

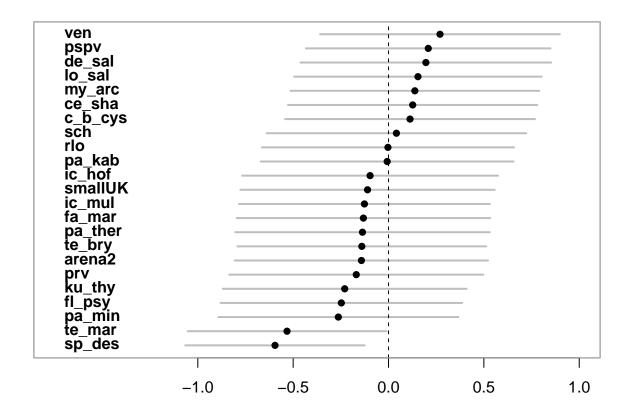


## loop through agents fitting individual models

```
coefs <-matrix(NA,</pre>
                 nrow = length(agents),
                 ncol = 2
                 dimnames = list(agents,c("est","se")))
agents <- unique(inf_agt_resid_data$agent)</pre>
for(i in agents){
    data <- subset(inf_agt_resid_data, agent==i)</pre>
    model_ind <- lmer(resid_value ~ 0 + load_std + (load_std | Stock_Analysis)+(1|Year),</pre>
                       data = data)
    coefs[i,] <- summary(model_ind)$coefficients[1, 1:2]</pre>
 }
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl =
## control$checkConv, : unable to evaluate scaled gradient
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl =
## control$checkConv, : Model failed to converge: degenerate Hessian with 1
## negative eigenvalues
coefs_order <- coefs[order(coefs[,1]),]</pre>
```

plot effect sizes

```
par(mfrow=c(1,1), mar=c(3,1,1,1),oma=c(0.5,0.5,0.5,0.5))
plotCI(x = coefs_order[,1],
        y = seq(1,length(agents)),
        ui = (coefs_order[,1]+(coefs_order[,2]*2)),
        li = (coefs_order[,1]-(coefs_order[,2]*2)),
        err = "x",
        sfrac = 0,
        gap = 0,
        yaxt = "n",
        xaxt = "n",
        ylab = "",
        xlab = "effect size",
        xlim = c(-1.75,1),
        pch = 16,
        lwd= 2,
        scol = "grey")
text(rep(-1.75,length(agents)),
    seq(1,length(agents)),
    labels = rownames(coefs_order),
    pos = 4, font = 2)
axis(1, at = c(-1, -0.5, 0, 0.5, 1))
abline(v = 0, lty = 2)
box(col="grey")
```



## w/STAN

loop through agents fitting individual models

change "eval = FALSE" in Rmd doc to actually run this code; takes 10-20 min.

```
write.csv(coefs_stan, "int_coefs_stan.csv")
```

load parameter estimates if models above were not run.

```
coefs_stan <- read.csv("int_coefs_stan.csv")
rownames(coefs_stan) <- coefs_stan[,1]
coefs_stan <- coefs_stan[,-1] # drop first column with agent names</pre>
```

### plot effect sizes

error bars are 50% (thick grey) and 95% (thin grey) credibe interval

```
coefs_order <- coefs_stan[order(coefs_stan[,3]),]</pre>
par(mfrow=c(1,1), mar=c(3,1,1,1),oma=c(0.5,0.5,0.5,0.5))
plotCI(x = coefs_order[,3],
        y = seq(1,length(agents)),
        li = (coefs_order[,1]),
        ui = (coefs_order[,5]),
        err = "x",
        sfrac = 0,
        gap = 0,
        yaxt = "n"
        xaxt = "n",
        ylab = "",
        xlab = "",
        xlim = c(-1.75,1),
        pch = 16,
        scol = "grey")
plotCI(x = coefs_order[,3],
        y = seq(1,length(agents)),
        li = (coefs_order[,2]),
        ui = (coefs_order[,4]),
        err = "x",
        sfrac = 0,
        gap = 0,
        pch = 16,
        add = TRUE,
        lwd = 3,
        scol = "grey")
text(rep(-1.75,length(agents)),
        seq(1,length(agents)),
        labels = rownames(coefs_order),
        pos = 4,
        font = 2,
        cex=0.95)
axis(1, at = c(-1, -0.5, 0, 0.5, 1))
abline(v = 0, ltv = 2)
box(col="grey")
mtext("effect size",1,line=2.2, cex=1.1)
```

