# Prelim Models

2025-02-26

## Load packages

```
library(tidyverse)
library(readr)
library(readxl)
library(splines)
library(mgcv)
```

## **Template**

```
knitr::opts_chunk$set(
   fig.width = 6,
   fig.asp = .6,
   out.width = "90%"
)

theme_set(theme_minimal() + theme(legend.position = "right"))

options(
   ggplot2.continuous.colour = "viridis",
   ggplot2.continuous.fill = "viridis"
)

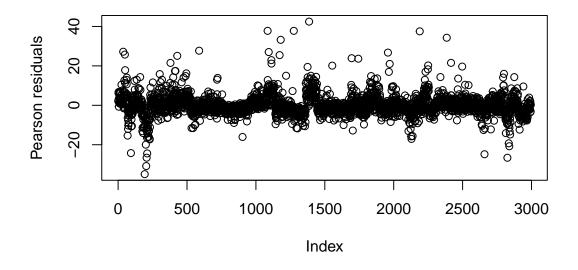
scale_colour_discrete = scale_colour_viridis_d
scale_fill_discrete = scale_fill_viridis_d
```

#Load dataset Clean dataset has unreliable low bw values and the NA low bw values removed
eqi\_lbw\_clean\_df <- read\_csv("data/eqi\_lbw\_clean\_df.csv")

#Prelim model just with EQI as exposure and low bw as outcome ##Poisson model

```
##
## Call:
## glm(formula = num_low_birthweight_births ~ eqi, family = poisson,
## data = eqi_lbw_clean_df, offset = log_live_births)
```

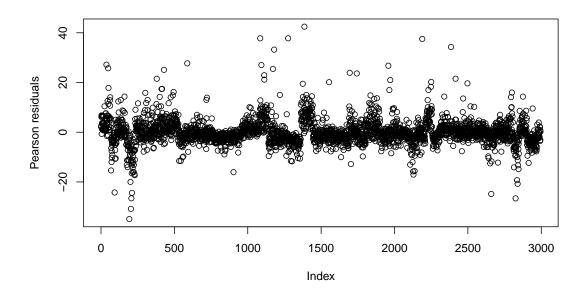
```
##
## Coefficients:
##
                 Estimate Std. Error z value Pr(>|z|)
  (Intercept) -2.4911216 0.0007558 -3295.84
                                                 <2e-16 ***
##
##
  eqi
               -0.0334156
                          0.0008549
                                       -39.09
                                                 <2e-16 ***
##
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
   (Dispersion parameter for poisson family taken to be 1)
##
##
##
       Null deviance: 77658
                             on 2994
                                      degrees of freedom
## Residual deviance: 76134
                             on 2993 degrees of freedom
  AIC: 97828
##
## Number of Fisher Scoring iterations: 4
sum(resid(mod1_p,type="pearson")^2)/mod1_p$df.residual
## [1] 26.26938
#yes dispersion is potential problem bc scale > 1
#qoodness of fit
pchisq(mod1_p$deviance, mod1_p$df.residual, lower.tail=F)
## [1] 0
#seeing a lack of fit for Poisson model
#Pearson residual plot
plot(resid(mod1_p,type="pearson"),ylab="Pearson residuals")
```



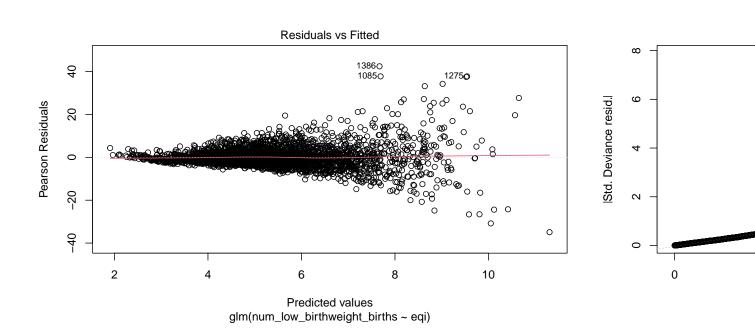
#### #also potential issue with outliers

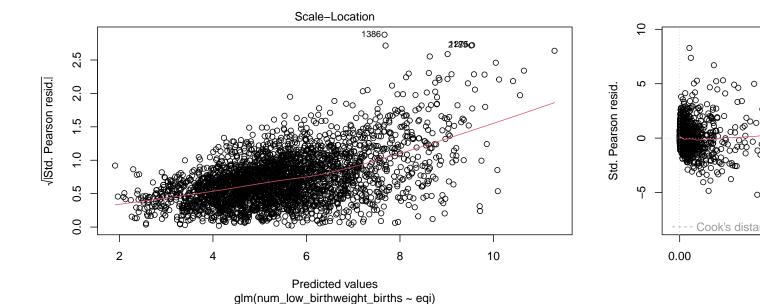
 $\# \mathrm{Prelim}$  model just with EQI as exposure and low bw as outcome  $\# \# \mathrm{Quasipoisson}$  model

```
#####Fit a Quasipoisson model (which acounts for overdispersion)
#fit the Poisson model (accounting for overdispersion)
#no offset bc exposure unit is already same here
mod1_qp = glm(num_low_birthweight_births ~ eqi,
         data=eqi_lbw_clean_df,
         family=quasipoisson,
         offset=log_live_births)
summary(mod1_qp)
##
## Call:
## glm(formula = num_low_birthweight_births ~ eqi, family = quasipoisson,
##
      data = eqi_lbw_clean_df, offset = log_live_births)
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.491122   0.003874 -643.046   < 2e-16 ***
              ## eqi
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for quasipoisson family taken to be 26.26938)
##
      Null deviance: 77658 on 2994 degrees of freedom
##
## Residual deviance: 76134 on 2993 degrees of freedom
## AIC: NA
##
## Number of Fisher Scoring iterations: 4
sum(resid(mod1_qp,type="pearson")^2)/mod1_qp$df.residual
## [1] 26.26938
#yes dispersion
#goodness of fit
pchisq(mod1_qp$deviance, mod1_qp$df.residual, lower.tail=F)
## [1] 0
#seeing a lack of fit with the quasipoisson
#Pearson residual plot
plot(resid(mod1_qp,type="pearson"),ylab="Pearson residuals")
```



#also potential issue with outliers
plot(mod1\_qp)





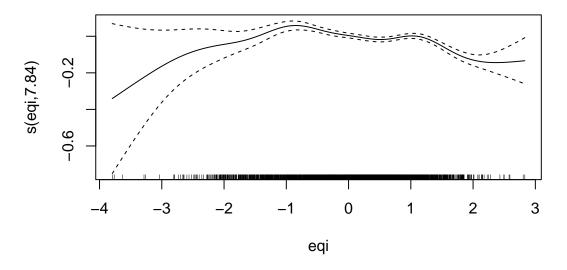
### #looks perhaps nonlinear?

#Prelim model just with EQI as exposure and low bw as outcome ##Penalized Spline

```
##
## Family: quasipoisson
## Link function: log
##
## Formula:
  num_low_birthweight_births ~ s(eqi)
##
##
## Parametric coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -2.499474
                          0.004047 -617.7
                                            <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
           edf Ref.df
                          F p-value
## s(eqi) 7.837 8.623 12.59 <2e-16 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## R-sq.(adj) = 0.98
                        Deviance explained = 3.84%
## GCV = 25.082 Scale est. = 25.783
```

```
#this is the penalty estimated by the model
mod1_qp_nl$sp

## s(eqi)
## 22.53531
plot(mod1_qp_nl)
```

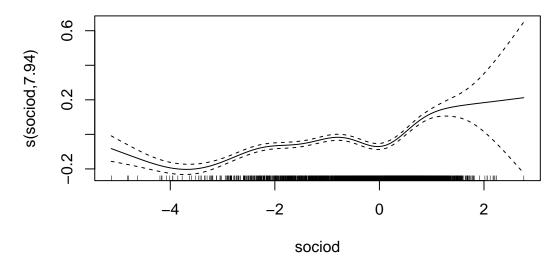


Seeing

nonlinear relationship between eqi and num\_low\_birthweight\_births

#Nonlinear checks of other indices and the low bw outcome; unadjusted

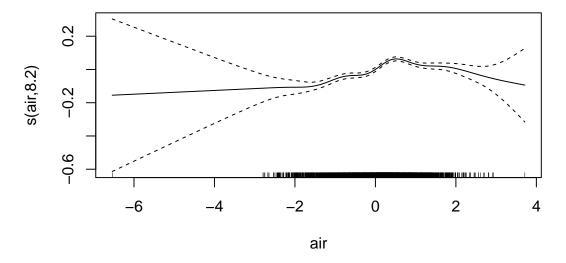
```
## Family: quasipoisson
## Link function: log
##
## Formula:
## num_low_birthweight_births ~ s(sociod)
##
## Parametric coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
  (Intercept) -2.446520
                          0.005652 - 432.8
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
             edf Ref.df
                            F p-value
##
```



```
#nonlinear
##AIR
mod3_qp_nl <- gam(num_low_birthweight_births ~ s(air),</pre>
                 family = "quasipoisson",
                  offset=log_live_births,
                  data = eqi_lbw_clean_df)
summary(mod3_qp_n1)
##
## Family: quasipoisson
## Link function: log
##
## Formula:
## num_low_birthweight_births ~ s(air)
##
## Parametric coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -2.509541 0.003389 -740.5 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

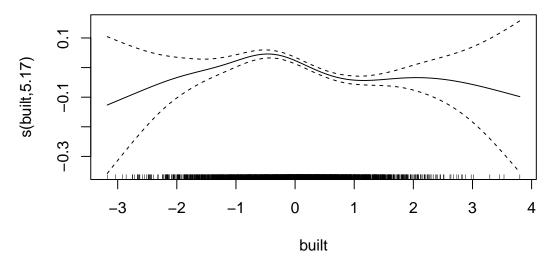
## Approximate significance of smooth terms:

```
## edf Ref.df F p-value
## s(air) 8.195 8.835 24.23 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.984 Deviance explained = 6.96%
## GCV = 24.274 Scale est. = 24.752 n = 2995
plot(mod3_qp_nl)</pre>
```



```
#nonlinear
##BUILT
mod4_qp_nl <- gam(num_low_birthweight_births ~ s(built),</pre>
                 family = "quasipoisson",
                  offset=log_live_births,
                  data = eqi_lbw_clean_df)
summary(mod4_qp_nl)
## Family: quasipoisson
## Link function: log
##
## Formula:
## num_low_birthweight_births ~ s(built)
## Parametric coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -2.504570 0.004227 -592.5 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
```

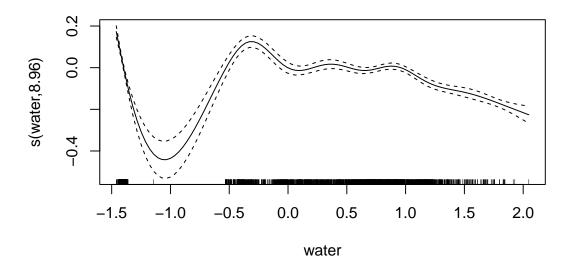
```
## Approximate significance of smooth terms:
## edf Ref.df F p-value
## s(built) 5.166 6.331 14.43 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.979 Deviance explained = 3.18%
## GCV = 25.208 Scale est. = 25.702 n = 2995
plot(mod4_qp_nl)</pre>
```



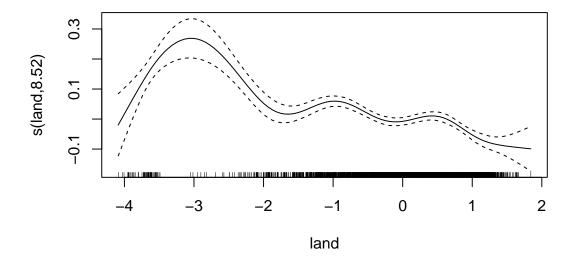
```
#nonlinear
##Water
mod5_qp_nl <- gam(num_low_birthweight_births ~ s(water),</pre>
                  family = "quasipoisson",
                  offset=log_live_births,
                  data = eqi_lbw_clean_df)
summary(mod5_qp_nl)
##
## Family: quasipoisson
## Link function: log
## Formula:
## num_low_birthweight_births ~ s(water)
##
## Parametric coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.484860 0.003393 -732.4
                                             <2e-16 ***
## ---
```

## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' ' 1

```
##
## Approximate significance of smooth terms:
## edf Ref.df F p-value
## s(water) 8.963 9 70.08 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.986 Deviance explained = 18.3%
## GCV = 21.332 Scale est. = 22.134 n = 2995
plot(mod5_qp_nl)</pre>
```



```
#nonlinear
#water index was weird distribution; establish a cutoff?
##Land
mod6_qp_nl <- gam(num_low_birthweight_births ~ s(land),</pre>
                  family = "quasipoisson",
                  offset=log_live_births,
                  data = eqi_lbw_clean_df)
summary(mod6_qp_n1)
## Family: quasipoisson
## Link function: log
##
## Formula:
## num_low_birthweight_births ~ s(land)
##
## Parametric coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -2.516455 0.003486 -721.9 <2e-16 ***
```



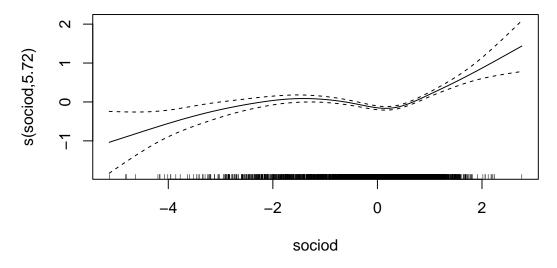
### #nonlinear

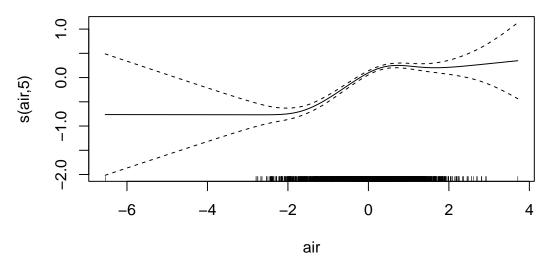
##

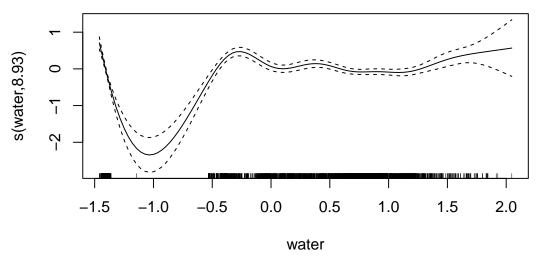
#Are there nonlinear relationships between the 5 subdomain indices? ##penalized spline #using built index as the outcome for now

Estimate Std. Error t value Pr(>|t|)

```
## (Intercept) 0.04333     0.01672     2.592     0.0096 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Approximate significance of smooth terms:
## edf Ref.df     F p-value
## s(sociod) 5.72     6.945 16.81     <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.0381     Deviance explained = 3.99%
## GCV = 0.83928     Scale est. = 0.83739     n = 2995
plot(mod7_qp_nl)</pre>
```



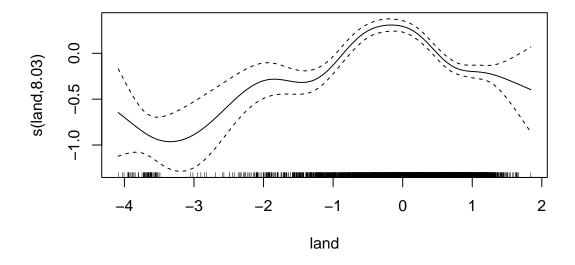




```
#nonlinear; again water has weird distribution; use cutoff?
##Land
mod10_qp_nl <- gam(built ~ s(land),</pre>
                  data = eqi_lbw_clean_df)
summary(mod10_qp_nl)
##
## Family: gaussian
## Link function: identity
## Formula:
## built ~ s(land)
##
## Parametric coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.04333
                          0.01642
                                    2.639 0.00835 **
## ---
```

## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' ' 1

```
##
## Approximate significance of smooth terms:
## edf Ref.df F p-value
## s(land) 8.028 8.756 27 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## R-sq.(adj) = 0.0725 Deviance explained = 7.5%
## GCV = 0.80985 Scale est. = 0.8074 n = 2995
plot(mod10_qp_nl)</pre>
```



## #very nonlinear

"