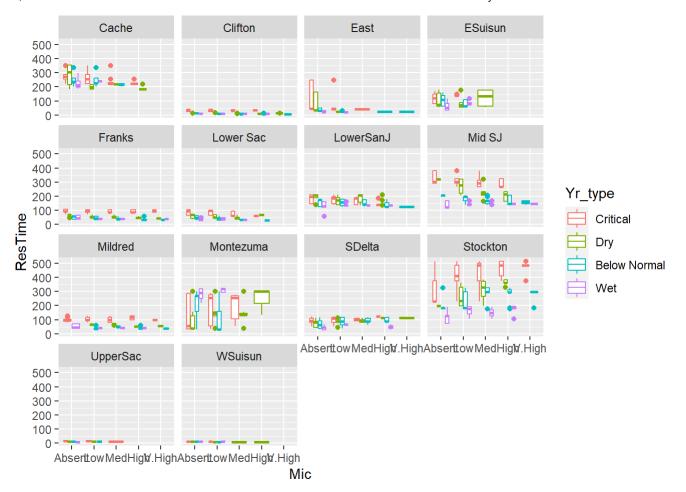
More DSM residence time analyss

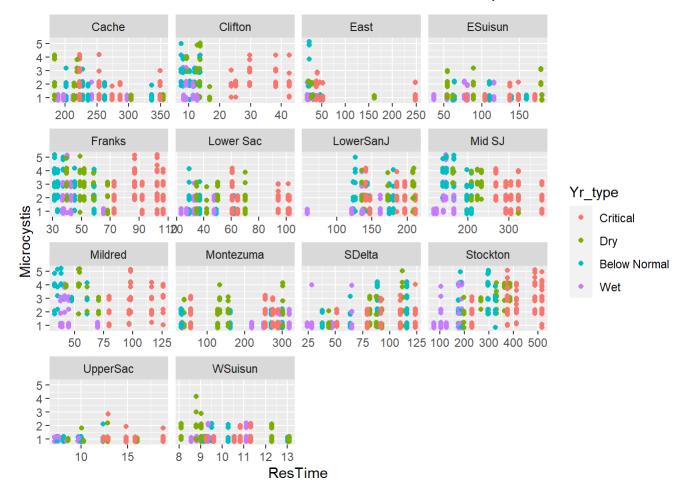
Rosie

2023-08-08

My models were all a bit of a mess. It's not really working very well with the residence time values. The salinity relationship makes no sense. I'm going to limit the datast to just the past five years. I may want to limit it to just the delta and kick out SUisun



```
#scatterplot of microcystis scores by residence time
ggplot(HABres2, aes(x =ResTime, y = Microcystis, color = Yr_type)) +
geom_point(position = "jitter")+
facet_wrap(~Region, scales = "free_x")
```



Now an ordered regression

```
m1 = clmm(MicF ~ ResTime+ Temperature+ Month + (1|Station)+ (1|Year), data = HABres2)
summary(m1)
```

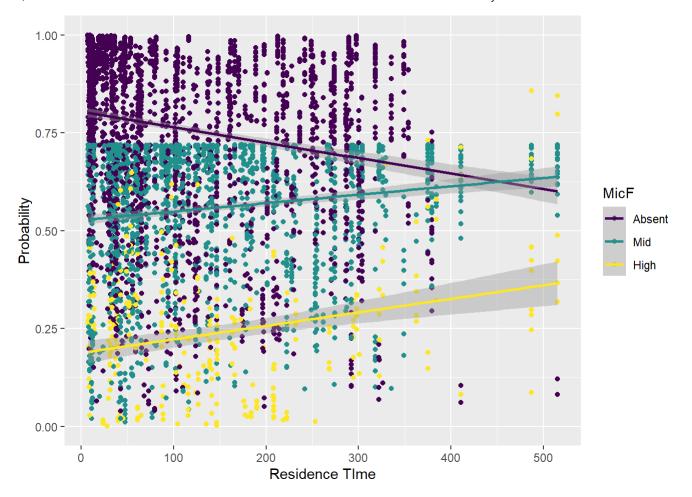
```
## Cumulative Link Mixed Model fitted with the Laplace approximation
##
## formula: MicF ~ ResTime + Temperature + Month + (1 | Station) + (1 | Year)
           HABres2
## data:
##
   link threshold nobs logLik
                                 AIC
                                          niter
                                                     max.grad cond.H
##
##
   logit flexible 4750 -2902.43 5818.87 1399(6983) 1.12e+02 4.8e+06
##
## Random effects:
##
   Groups Name
                        Variance Std.Dev.
   Station (Intercept) 1.946
                                1.395
##
   Year
            (Intercept) 1.051
                                 1.025
##
## Number of groups: Station 169, Year 8
##
## Coefficients:
##
                Estimate Std. Error z value Pr(>|z|)
## ResTime
               0.0002216 0.0005381
                                     0.412
## Temperature 0.4599304 0.0232555 19.777
                                              <2e-16 ***
## Month
               0.3148033 0.0325501 9.671 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Threshold coefficients:
##
              Estimate Std. Error z value
## Absent | Mid 12.8466
                           0.7887
                                    16.29
## Mid|High
               16.4707
                           0.8111
                                    20.31
```

```
#not as exciting as I'd think
#Also, i need to work on how to plot this.

HABres2$fits = fitted(m1)

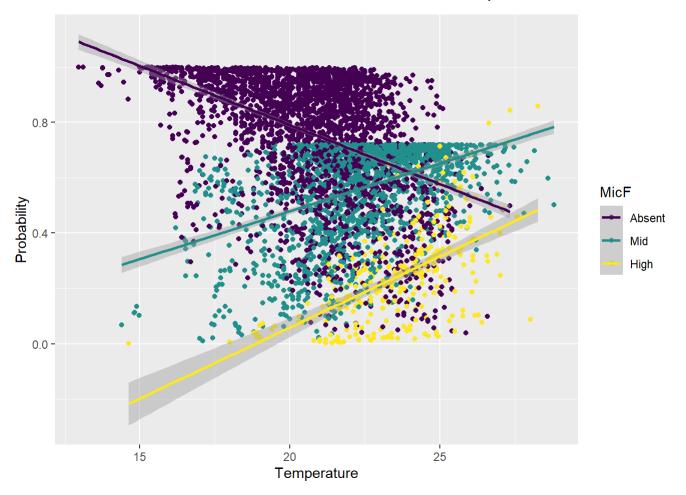
ggplot(HABres2, aes( ResTime, fits, color = MicF))+ geom_point()+ geom_smooth(method = "lm")+
   ylab("Probability")+ xlab("Residence TIme")
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

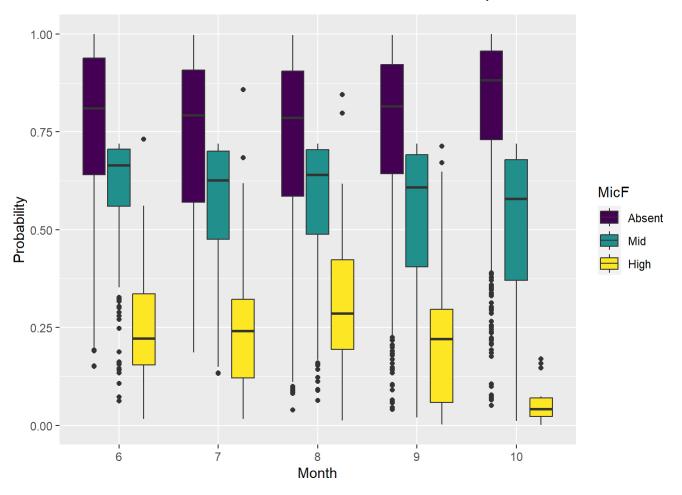


ggplot(HABres2, aes(Temperature, fits, color = MicF))+ geom_point()+ geom_smooth(method = "lm")+
ylab("Probability")+ xlab("Temperature")

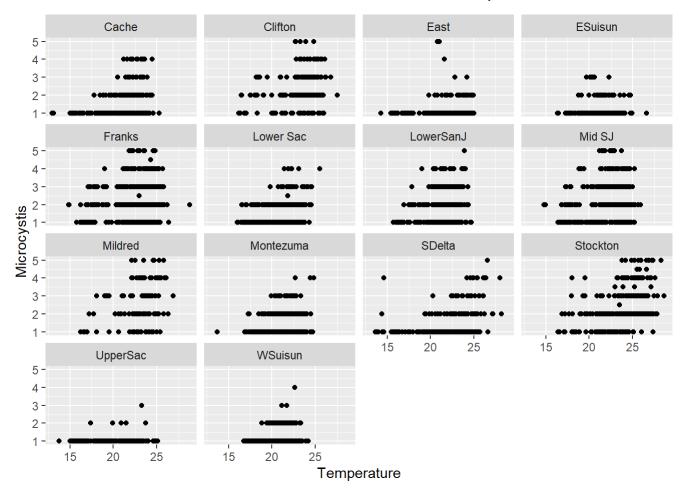
`geom_smooth()` using formula = 'y ~ x'



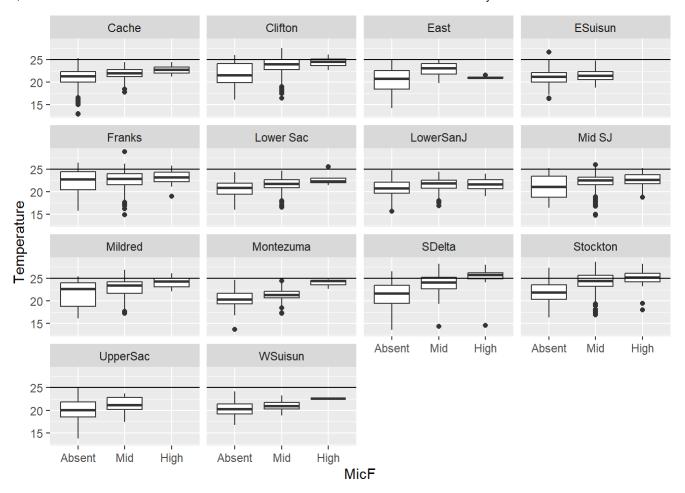
ggplot(HABres2, aes(as.factor(Month), fits, fill = MicF))+ geom_boxplot()+
 ylab("Probability")+ xlab("Month")



ggplot(HABres2, aes(x = Temperature, y = Microcystis))+ geom_point()+
 facet_wrap(~Region)



ggplot(HABres2, aes(x = MicF, y = Temperature))+ geom_boxplot()+
facet_wrap(~Region)+ geom_hline(yintercept = 25)



UGH. Still not great

Binomial presence absence by region

`summarise()` has grouped output by 'Region', 'Year'. You can override using
the `.groups` argument.

```
library(MuMIn)
```

```
micbin2017 = filter(micbin, !is.na(Sal), Year > 2016) %>%
   mutate(logres = log(Res+1))
global17 = glmer(cbind(Present, Absent)~ Temp + logres + Month+ Sal+(1|Year2)+ (1|Region), data
= micbin2017, family = "binomial", na.action = "na.fail")
mods2017 = dredge(global17)
```

```
## Fixed term is "(Intercept)"
```

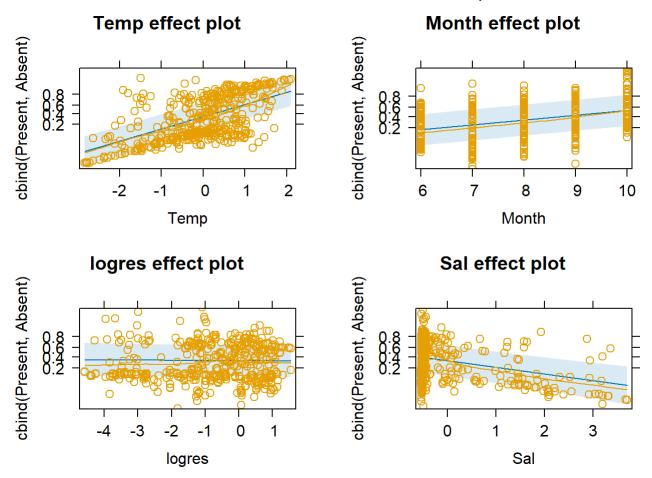
```
write.csv(as.data.frame(mods2017), "MicrocystisModels.csv")

#Month, salinity, and temperature are the best predictors, residence time close behind

best17.2 = glmer(cbind(Present, Absent)~ Temp + Month+ logres+Sal+ (1|Year2)+ (1|Region), data = micbin2017, family = "binomial", na.action = "na.fail")
summary(best17.2)
```

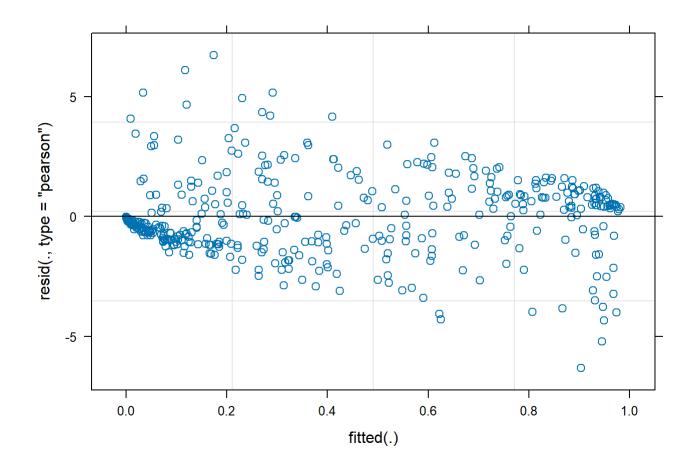
```
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
##
   Family: binomial (logit)
## Formula: cbind(Present, Absent) ~ Temp + Month + logres + Sal + (1 | Year2) +
##
       (1 | Region)
     Data: micbin2017
##
##
       AIC
                      logLik deviance df.resid
##
                BIC
    1747.4
                     -866.7
                               1733.4
                                           410
##
            1775.7
##
## Scaled residuals:
##
      Min
               10 Median
                               30
                                      Max
  -6.3253 -0.9828 -0.1827 0.8813 6.7363
##
##
## Random effects:
##
   Groups Name
                      Variance Std.Dev.
   Region (Intercept) 2.402
                               1.55
##
##
   Year2 (Intercept) 1.769
                               1.33
## Number of obs: 417, groups: Region, 14; Year2, 6
##
## Fixed effects:
##
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -4.08761
                          0.76999 -5.309 1.10e-07 ***
## Temp
               1.12780
                          0.06899 16.347 < 2e-16 ***
## Month
               0.41879
                          0.04094 10.230 < 2e-16 ***
## logres
              -0.01381
                          0.08620 -0.160
                                             0.873
                          0.13113 -4.412 1.03e-05 ***
## Sal
              -0.57849
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
          (Intr) Temp
                      Month logres
##
## Temp
         -0.262
## Month -0.447 0.595
## logres 0.141 0.007 -0.125
## Sal
          0.157 -0.039 -0.270 0.360
```

```
plot(allEffects(best17.2, residuals = TRUE))
```



Let's be sure our diagnostic plots look good before anything else

plot(best17.2)

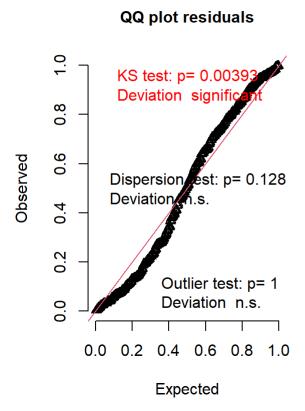


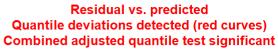
library(DHARMa)

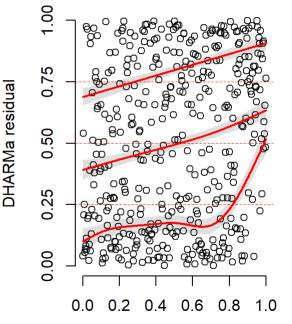
This is DHARMa 0.4.6. For overview type '?DHARMa'. For recent changes, type news(package = 'D
HARMa')

simres = simulateResiduals(best17.2)
plot(simres)

DHARMa residual



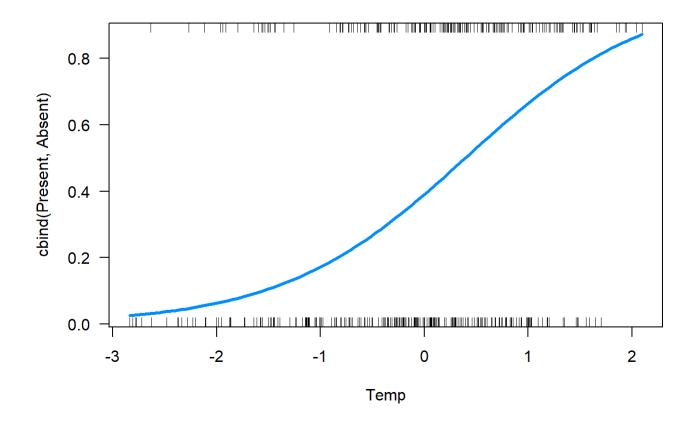


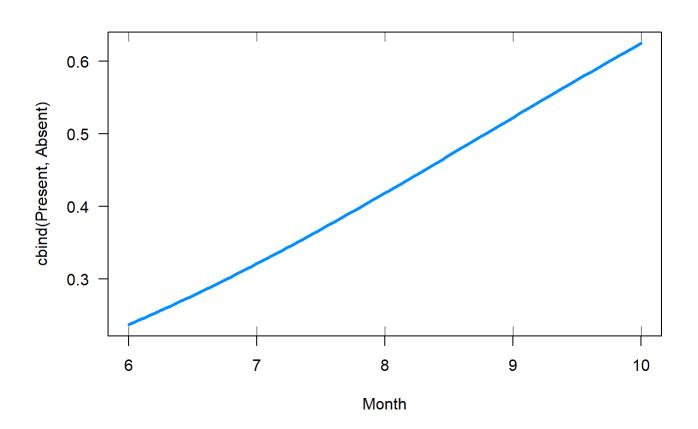


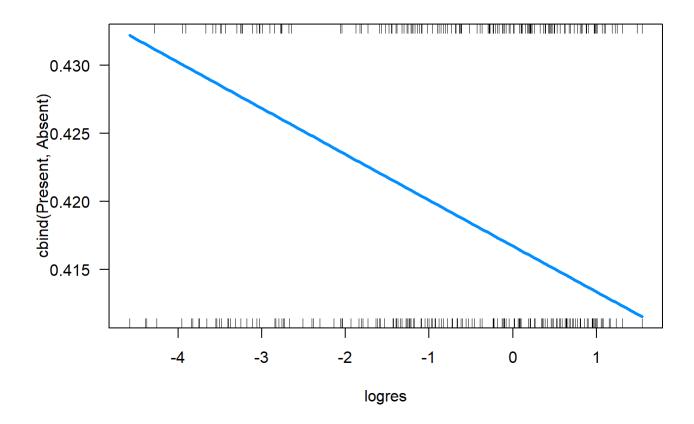
Model predictions (rank transformed)

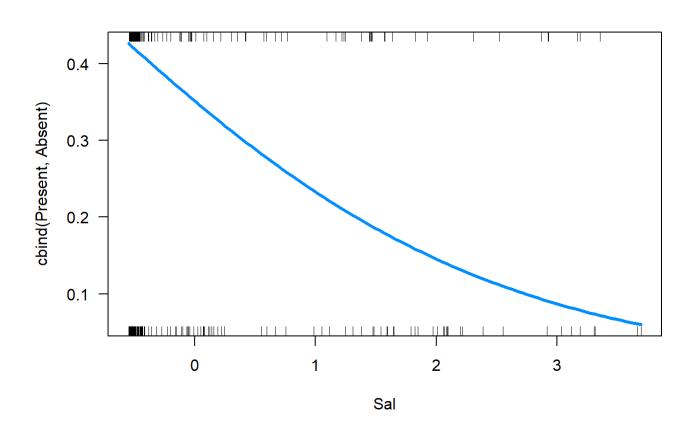
```
#not too bad, though some deviation
```

```
effs = allEffects(best17.2, partial.residuals = TRUE)
test = ggeffect(best17.2, terms = "Temp")
test2 = ggpredict(best17.2, terms = "Temp [all]")
#OK, need to find a pretty way to plot this
library(visreg)
visreg(best17.2, scale = "response")
```

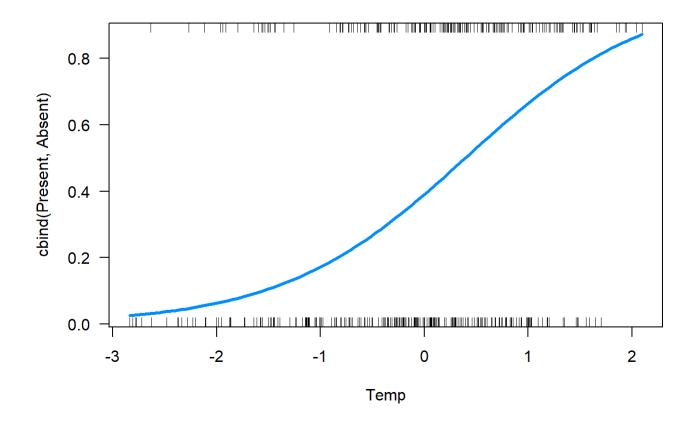


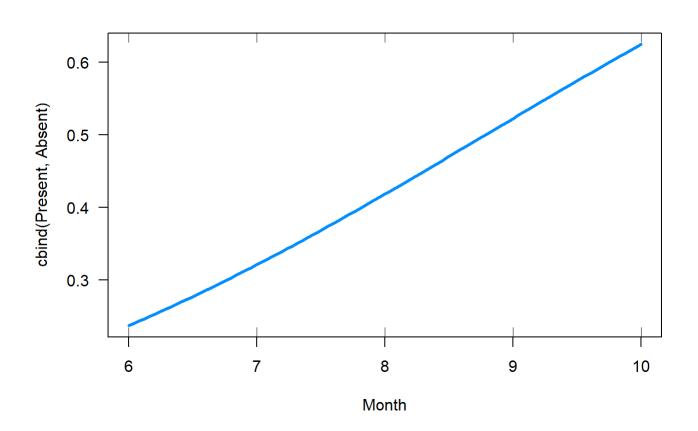


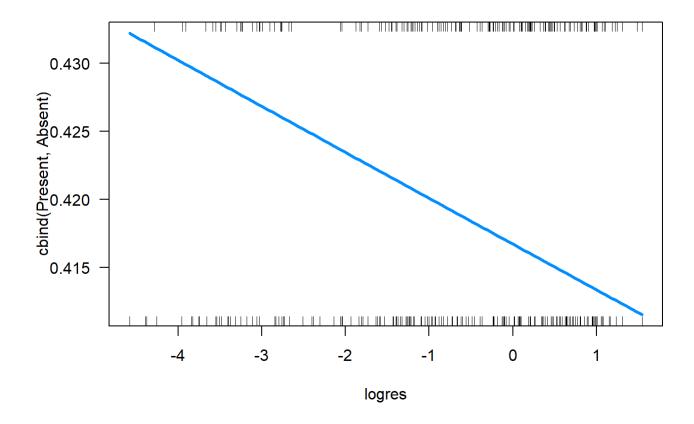


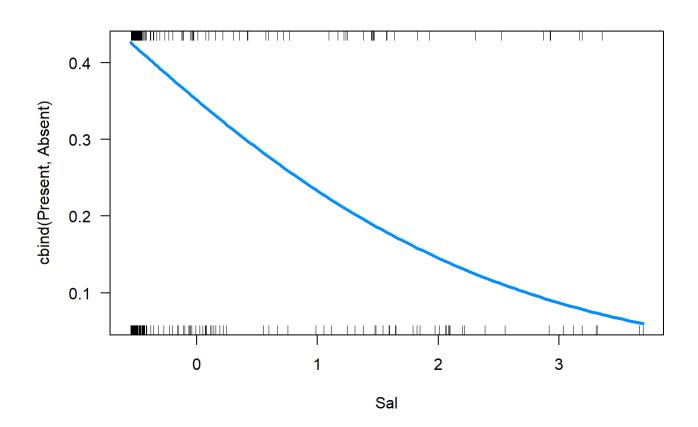


foo = visreg(best17.2, scale = "response")



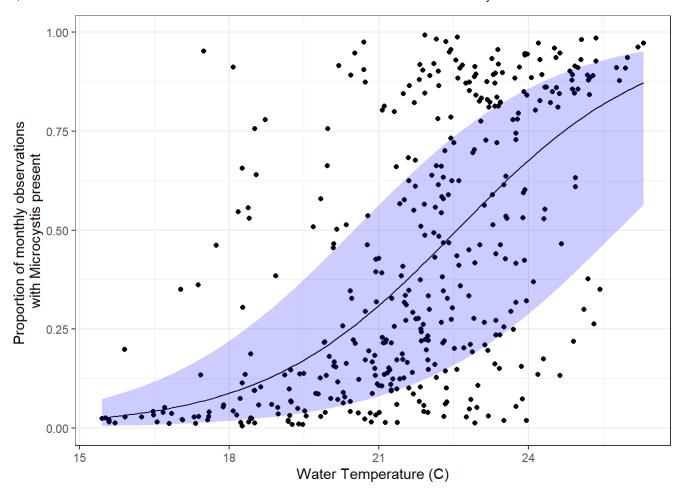


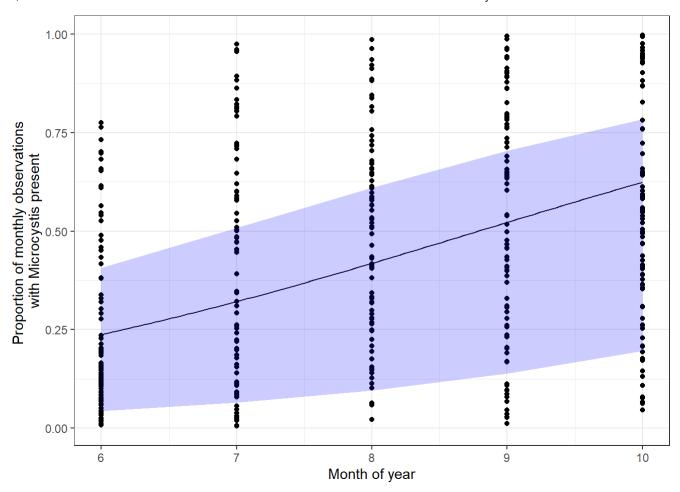


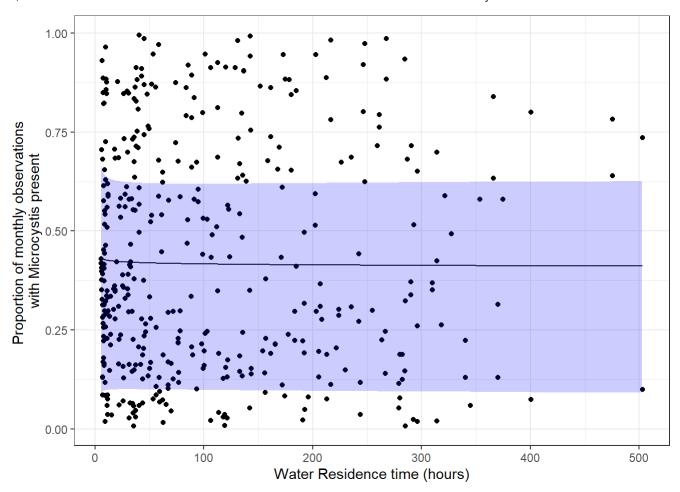


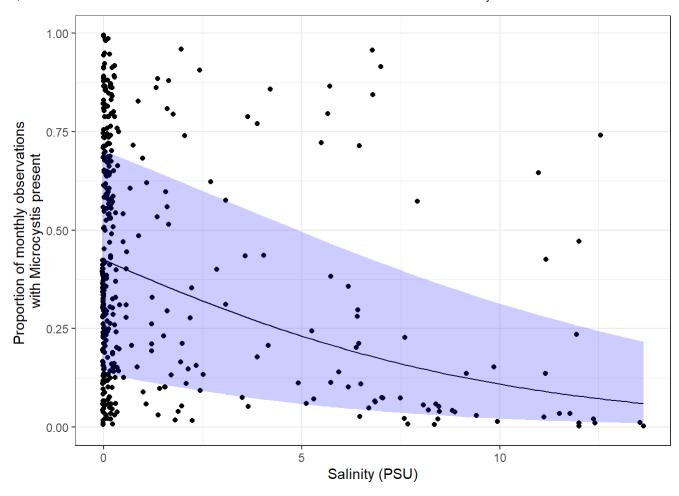
```
Backscale = function(X, dfvar){
    y = X* sd((dfvar-mean(dfvar, narm =t)), na.rm =T) + mean(dfvar, na.rm =T)
    return(y)
}
saltest = lm(Sal ~ Salinity, data = micbin2017)
summary(saltest)
```

```
##
## Call:
## lm(formula = Sal ~ Salinity, data = micbin2017)
##
## Residuals:
##
         Min
                     1Q
                            Median
                                           3Q
                                                     Max
## -4.037e-15 -3.610e-17 7.100e-18 2.280e-17 2.204e-15
##
## Coefficients:
##
                Estimate Std. Error
                                       t value Pr(>|t|)
## (Intercept) -5.592e-01 1.328e-17 -4.209e+16
                                                 <2e-16 ***
                                                 <2e-16 ***
## Salinity
               2.922e-01 3.638e-18 8.030e+16
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.391e-16 on 415 degrees of freedom
## Multiple R-squared:
                           1, Adjusted R-squared:
## F-statistic: 6.447e+33 on 1 and 415 DF, p-value: < 2.2e-16
```









```
ggsave("plots/Salplot.tiff", device = "tiff", width = 5, height =5)
```

```
#Pretty version of model results
library(broom.mixed)
summod = tidy(best17.2)
write.csv(summod, "Micmod.csv")
```

Mine wants a more balanced design

```
## `summarise()` has grouped output by 'Region', 'Year'. You can override using
## the `.groups` argument.
```

```
global2 = glmer(cbind(Present, Absent)~ Temp + logres + Month+ Sal+(1|Year2)+ (1|Region), data =
micbin2, family = "binomial", na.action = "na.fail")
mods2 = dredge(global2)
```

```
## Fixed term is "(Intercept)"

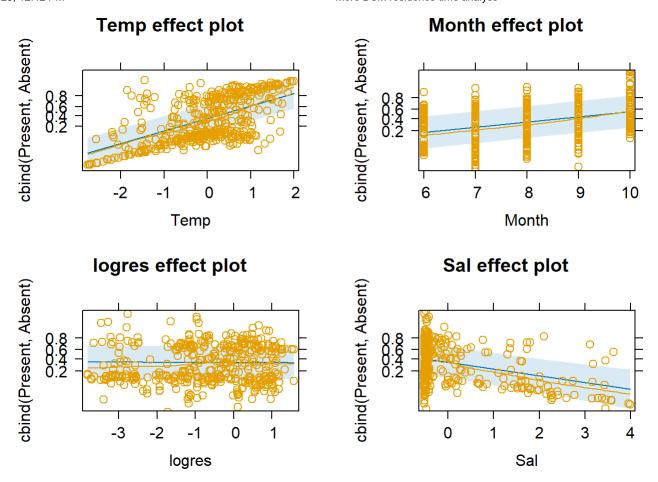
write.csv(as.data.frame(mods2), "MicrocystisModels2.csv")

#Month, salinity, and temperature are the best predictors, residence time close behind

best2 = glmer(cbind(Present, Absent)~ Temp + Month+ logres+Sal+ (1|Year2)+ (1|Region), data = m icbin2, family = "binomial", na.action = "na.fail")
summary(best2)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
     Approximation) [glmerMod]
##
##
   Family: binomial (logit)
## Formula: cbind(Present, Absent) ~ Temp + Month + logres + Sal + (1 | Year2) +
##
       (1 | Region)
      Data: micbin2
##
##
                      logLik deviance df.resid
##
        AIC
                 BIC
                      -837.6
##
     1689.1
              1717.4
                                1675.1
                                            410
##
## Scaled residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
  -6.3287 -0.9300 -0.1419 0.8788 6.5175
##
##
## Random effects:
##
   Groups Name
                       Variance Std.Dev.
   Region (Intercept) 2.264
                                1.505
   Year2 (Intercept) 1.818
                                1.348
##
## Number of obs: 417, groups: Region, 14; Year2, 6
##
## Fixed effects:
##
               Estimate Std. Error z value Pr(>|z|)
                           0.77332 -5.418 6.04e-08 ***
## (Intercept) -4.18967
                           0.07257 15.627 < 2e-16 ***
## Temp
                1.13407
## Month
                0.43760
                           0.04258 10.276 < 2e-16 ***
## logres
               -0.01120
                           0.09014 -0.124
                                              0.901
                           0.12480 -4.561 5.10e-06 ***
## Sal
               -0.56920
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
          (Intr) Temp
##
                       Month logres
## Temp
          -0.272
## Month -0.459 0.598
## logres 0.128 0.014 -0.114
## Sal
           0.151 -0.040 -0.275 0.362
```

```
plot(allEffects(best2, residuals = TRUE))
```

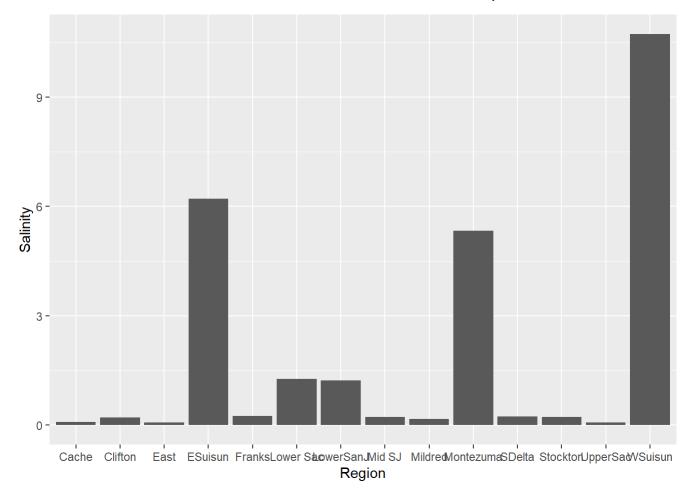


#pretty much the same result

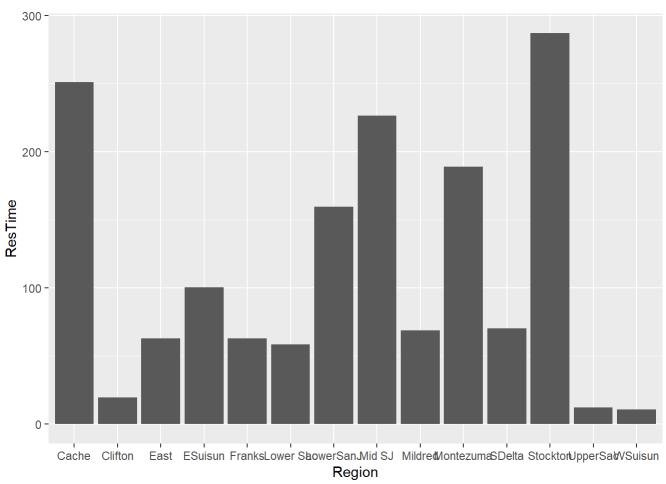
Look at data by region, compare physical variables.

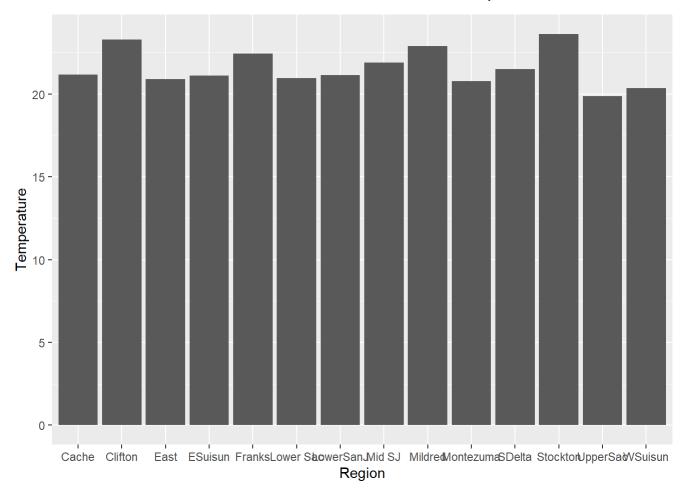
```
## `summarise()` has grouped output by 'Region'. You can override using the
## `.groups` argument.
```

```
ggplot(physreg, aes(x = Region, y = Salinity))+ geom_col()
```

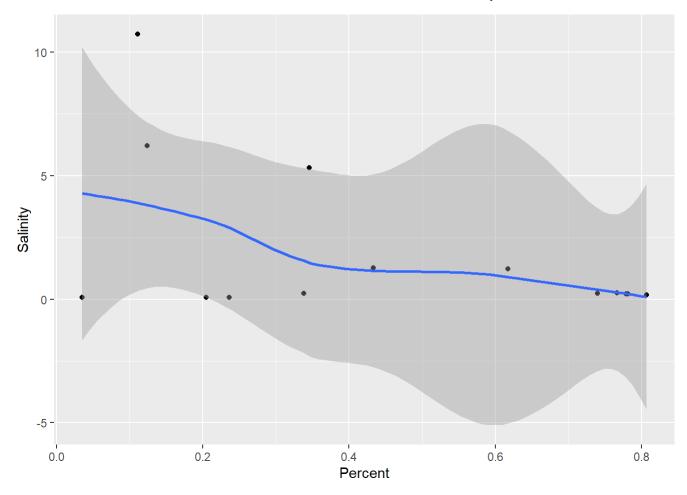


ggplot(physreg, aes(x = Region, y = ResTime))+ geom_col()

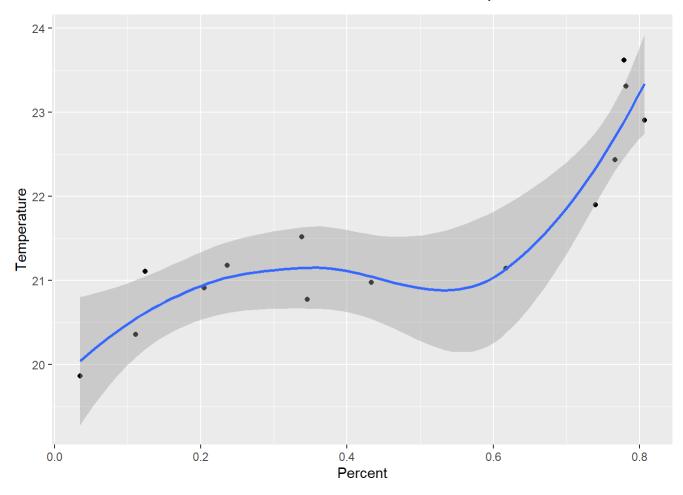




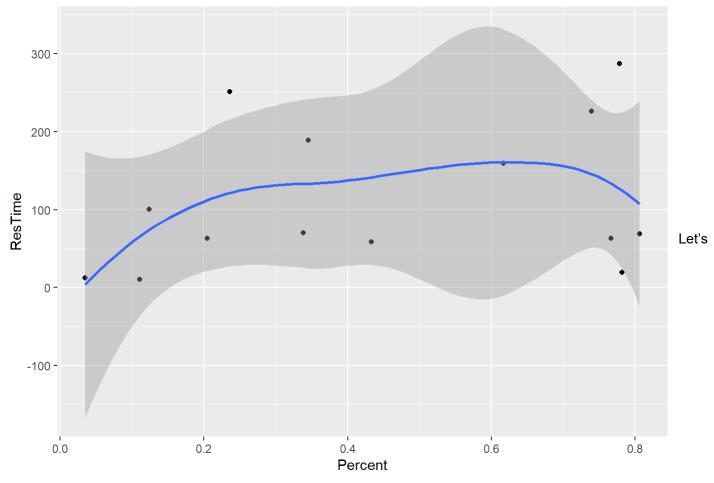
$geom_smooth()$ using method = 'loess' and formula = 'y ~ x'



$geom_smooth()$ using method = 'loess' and formula = 'y ~ x'



$geom_smooth()$ using method = 'loess' and formula = 'y ~ x'

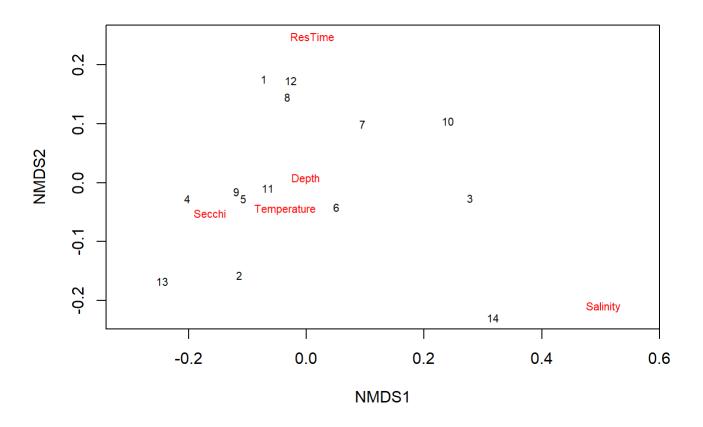


try an NMDS plot

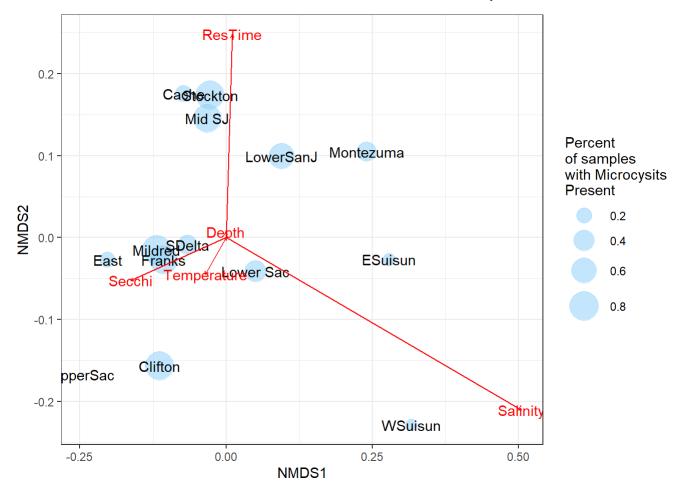
```
## Square root transformation
## Wisconsin double standardization
## Run 0 stress 0.03715665
## Run 1 stress 0.03715666
## ... Procrustes: rmse 2.646165e-05 max resid 6.39725e-05
## ... Similar to previous best
## Run 2 stress 0.0598041
## Run 3 stress 0.2804498
## Run 4 stress 0.0598041
## Run 5 stress 0.0598041
## Run 6 stress 0.03715665
## ... New best solution
## ... Procrustes: rmse 4.017872e-06 max resid 8.567225e-06
## ... Similar to previous best
## Run 7 stress 0.03715665
## ... New best solution
## ... Procrustes: rmse 9.1917e-06 max resid 2.234403e-05
## ... Similar to previous best
## Run 8 stress 0.03715666
## ... Procrustes: rmse 1.62986e-05 max resid 3.918239e-05
## ... Similar to previous best
## Run 9 stress 0.03715665
## ... New best solution
## ... Procrustes: rmse 4.761761e-06 max resid 1.133015e-05
## ... Similar to previous best
## Run 10 stress 0.03715665
## ... New best solution
## ... Procrustes: rmse 9.299208e-07 max resid 2.06624e-06
## ... Similar to previous best
## Run 11 stress 0.0598041
## Run 12 stress 0.03715665
## ... Procrustes: rmse 6.208699e-06 max resid 1.455219e-05
## ... Similar to previous best
## Run 13 stress 0.03715665
## ... Procrustes: rmse 4.643656e-06 max resid 1.048287e-05
## ... Similar to previous best
## Run 14 stress 0.0598041
## Run 15 stress 0.03715665
## ... Procrustes: rmse 1.630244e-05 max resid 3.881217e-05
## ... Similar to previous best
## Run 16 stress 0.03715665
## ... Procrustes: rmse 6.442109e-06 max resid 1.586081e-05
## ... Similar to previous best
## Run 17 stress 0.2607879
## Run 18 stress 0.0598041
## Run 19 stress 0.03715665
## ... Procrustes: rmse 3.806056e-06 max resid 8.80924e-06
## ... Similar to previous best
## Run 20 stress 0.03715665
## ... Procrustes: rmse 1.158419e-05 max resid 2.515813e-05
```

```
## ... Similar to previous best
## *** Best solution repeated 7 times
```

```
plot(mds1, type = "t")
```



```
data.scores <- as.data.frame(scores(mds1)$sites) #Using the scores function from vegan to extra
ct the site scores and convert to a
data.scores$Region = physreg$Region
data.scores$Percent = physreg$Percent
vars = as.data.frame(scores(mds1)$species)
vars$Label = row.names(vars)
ggplot(data.scores, aes(x = NMDS1, y = NMDS2))+ geom_point(aes(size = Percent), alpha = 0.5, col
or = "lightskyblue")+
  geom text(aes(label = Region))+
  geom_text(data = vars, aes(label = Label), color = "red")+
  geom\_segment(data = vars, aes(x = 0, y = 0, xend = NMDS1, yend = NMDS2),
               arrow = arrow(length = unit(0.1, "cm")), inherit.aes = FALSE,
               color = "red")+
  scale\_size\_continuous(range = c(1,10), name = "Percent\nof samples \nwith Microcysits\nPresen")
t")+
  theme_bw()
```



ggplot(physreg, aes(x = Region, y = Depth)) + geom_col()

