# OGC-biomass

# Primary Analyses

# Kelly Murray Stoker, David Murray-Stoker

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## Raw Data Management

Raw data were imported and compiled for each of the two sampling periods, 1980s and 2010s. Subsets were merged or cast as needed for later analyses. Raw data were initially managed in the R environment, but final data management tasks were performed in Microsoft Excel after exporting the data; final data files are imported from csv files.

```
## Read in raw data files
raw.biomass.1982 <- read_csv("data/OGC_biomass_data-1982.csv", show_col_types = FALSE)
raw.biomass.1983 <- read_csv("data/OGC_biomass_data-1983.csv", show_col_types = FALSE)
raw.biomass.2010s <- read_csv("data/OGC_biomass_data-2010s.csv", show_col_types = FALSE)

## Set column structure for the raw data
raw.biomass.1982$Date <- as_factor(raw.biomass.1982$Date)
raw.biomass.1982$Snag <- as_factor(raw.biomass.1982$Snag)
raw.biomass.1983$Date <- as_factor(raw.biomass.1983$Date)
raw.biomass.1983$Snag <- as_factor(raw.biomass.1983$Snag)
raw.biomass.2010s$Date <- as_factor(raw.biomass.2010s$Date)
raw.biomass.2010s$Snag <- as_factor(raw.biomass.2010s$Snag)</pre>
```

```
## Biomass data management
## Create date-by-taxa matrices for biomass values
# 1982 raw biomass by taxa matrix
OGC.biomass.1982 <- dcast(
   raw.biomass.1982, Date ~ Genus, sum, value.var = "Biomass_Estimate"
   )
# 1983 raw biomass by taxa matrix
OGC.biomass.1983 <- dcast(
   raw.biomass.1983, Date ~ Genus, sum, value.var = "Biomass Estimate"
# 2010s raw biomass by taxa matrix
OGC.biomass.2010s <- dcast(
    raw.biomass.2010s, Date ~ Genus, sum, value.var = "Biomass_Estimate"
## Create dataframe of biomass values for each dataset
OGC.biomass.1982.values <- tibble(OGC.biomass.1982[, 2:59])
OGC.biomass.1983.values <- tibble(OGC.biomass.1983[, 2:48])
OGC.biomass.2010s.values <- tibble(OGC.biomass.2010s[, 2:64])
## Average by number of snags
OGC.biomass.1982.snag.correction
                                    <- OGC.biomass.1982.values/20
# Correct for 19 snags instead of 20 in first sample
OGC.biomass.1982.snag.correction[1, ] <- ((OGC.biomass.1982.snag.correction[1, ] * 20)/19)
OGC.biomass.1983.snag.correction <- OGC.biomass.1983.values/10
OGC.biomass.2010s.snag.correction <- OGC.biomass.2010s.values/10
## Add Date to the biomass data
OGC.biomass.1982.final <- OGC.biomass.1982.snag.correction %>%
    add_column(OGC.biomass.1982$Date) %>%
    rename(UID = "OGC.biomass.1982$Date")
OGC.biomass.1983.final <- OGC.biomass.1983.snag.correction %>%
    add_column(OGC.biomass.1983$Date) %>%
    rename(UID = "OGC.biomass.1983$Date")
OGC.biomass.2010s.final <- OGC.biomass.2010s.snag.correction %>%
    add_column(OGC.biomass.2010s$Date) %>%
   rename(UID = "OGC.biomass.2010s$Date")
## Export biomass data files
write_csv(OGC.biomass.1982.final, file = "data/OGC_1982_biomass_by_taxa.csv")
write_csv(OGC.biomass.1983.final, file = "data/OGC_1983_biomass_by_taxa.csv")
write csv(OGC.biomass.2010s.final, file = "data/OGC 2010s biomass by taxa.csv")
```

```
## Density data management
## Create date-by-taxa matrices for density values
# 1982 density by taxa matrix
OGC.density.1982 <- dcast(
   raw.biomass.1982, Date ~ Genus, sum, value.var = "Density"
   )
# 1983 density by taxa matrix
OGC.density.1983 <- dcast(
   raw.biomass.1983, Date ~ Genus, sum, value.var = "Density"
# 2010s density by taxa matrix
OGC.density.2010s <- dcast(
    raw.biomass.2010s, Date ~ Genus, sum, value.var = "Density"
## Create dataframe of density values for each dataset
OGC.density.1982.values <- OGC.density.1982[, 2:59]
OGC.density.1983.values <- OGC.density.1983[, 2:48]
OGC.density.2010s.values <- OGC.density.2010s[, 2:64]
## Average by number of snags
OGC.density.1982.snag.correction
                                     <- OGC.density.1982.values/20
# Correct for 19 snags instead of 20 in first sample
OGC.density.1982.snag.correction[1, ] <-((OGC.density.1982.snag.correction[1, ] * 20)/19)
OGC.density.1983.snag.correction <- OGC.density.1983.values/10
OGC.density.2010s.snag.correction <- OGC.density.2010s.values/10
## Add UID to density values
OGC.density.1982.final <- OGC.density.1982.snag.correction %>%
    add_column(OGC.density.1982$Date) %>%
   rename(UID = "OGC.density.1982$Date")
OGC.density.1983.final <- OGC.density.1983.snag.correction %>%
    add_column(OGC.density.1983$Date) %>%
    rename(UID = "OGC.density.1983$Date")
OGC.density.2010s.final <- OGC.density.2010s.snag.correction %>%
    add_column(OGC.density.2010s$Date) %>%
   rename(UID = "OGC.density.2010s$Date")
## Export density data files
write_csv(OGC.density.1982.final, file = "data/OGC_1982_density_by_taxa.csv")
write_csv(OGC.density.1983.final, file = "data/OGC_1983_density_by_taxa.csv")
write csv(OGC.density.2010s.final, file = "data/OGC 2010s density by taxa.csv")
```

### Load Processed Data

```
## Read in data
biomass.data <- read_csv("data/OGC_final_biomass_data.csv", show_col_types = FALSE)
density.data <- read_csv("data/OGC_final_density_data.csv", show_col_types = FALSE)</pre>
## Set variables as factors
# Biomass data
                      <- as_factor(biomass.data$UID)</pre>
biomass.data$UID
biomass.data$Year
                          <- as_factor(biomass.data$Year)</pre>
biomass.data$Year_Recoded <- as_factor(biomass.data$Year_Recoded)</pre>
biomass.data$Season_Recoded <- as_factor(biomass.data$Season_Recoded)
biomass.data$Period
                          <- as_factor(biomass.data$Period)</pre>
# Density data
density.data$UID
                            <- as_factor(density.data$UID)</pre>
                            <- as_factor(density.data$Year)</pre>
density.data$Year
density.data$Year_Recoded <- as_factor(density.data$Year_Recoded)</pre>
density.data$Season
                           <- as_factor(density.data$Season)</pre>
density.data$Season_Recoded <- as_factor(density.data$Season_Recoded)</pre>
density.data$Period
                            <- as_factor(density.data$Period)</pre>
## Dataframe of sampling info
sampling.info <- biomass.data[, 1:6]</pre>
## Matrices of invertebrate biomass and density
invertebrate.biomass.matrix <- biomass.data[, 16:87]</pre>
invertebrate.density.matrix <- density.data[, 16:87]</pre>
## Filter biomass and density data by sampling period; no further subsetting
# Biomass data
biomass.1980s <- invertebrate.biomass.matrix[1:25, ]</pre>
biomass.2010s <- invertebrate.biomass.matrix[26:49, ]
# Density data
density.1980s <- invertebrate.density.matrix[1:25, ]</pre>
density.2010s <- invertebrate.density.matrix[26:49, ]</pre>
```

# Biomass & Density ANOVAs

Biomass and density were compared by sampling period, season, and the interaction using an ANOVA with Type II sums-of-squares. ANOVA assumptions were inspected graphically using check\_model(), and effect sizes for the ANOVAs were calculated as  $\eta_P^2$  using eta\_squared(). Post-hoc Tukey's HSD tests were conducted using HSD.test() to examine for differences among groups for influential factors in the ANOVA.

#### Biomass ANOVA

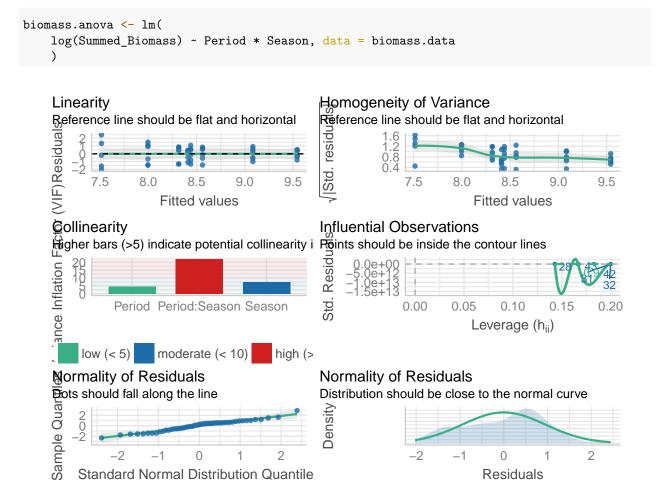


Figure 1: Diagnostic plots of the biomass ANOVA.

Residuals

Standard Normal Distribution Quantile

Table 1: ANOVA results for biomass by period, season, and the interaction.

|               | Sums-of-Squares | df | F     | P-value |
|---------------|-----------------|----|-------|---------|
| Period        | 6.348           | 1  | 6.014 | 0.019   |
| Season        | 5.432           | 3  | 1.716 | 0.179   |
| Period:Season | 4.199           | 3  | 1.326 | 0.279   |
| Residuals     | 43.275          | 41 | NA    | NA      |

Table 2: Tukey groups assigned to periods differing in biomass.

|      | Biomass  | Grouping |
|------|----------|----------|
| 1980 | 8.854356 | a        |
| 2010 | 8.157834 | b        |

Table 3: Table of the effect sizes in the biomass ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.128       | 0.95 | 0.013  | 1       |
| Season        | 0.112       | 0.95 | 0.000  | 1       |
| Period:Season | 0.088       | 0.95 | 0.000  | 1       |

### Density ANOVA

```
density.anova <- lm(
    log(Summed_Density) ~ Period * Season, data = density.data
)</pre>
```

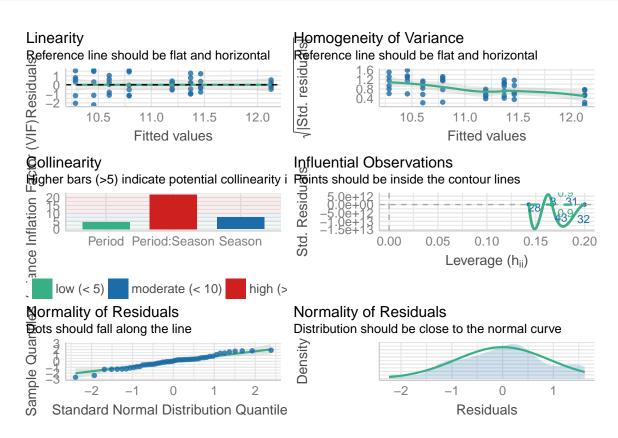


Figure 2: Diagnostic plots of the density ANOVA.

Table 4: ANOVA result for density by period, season, and the interaction.

|               | Sums-of-Squares | df | F     | P-value |
|---------------|-----------------|----|-------|---------|
| Period        | 0.201           | 1  | 0.204 | 0.654   |
| Season        | 8.029           | 3  | 2.712 | 0.057   |
| Period:Season | 7.575           | 3  | 2.558 | 0.068   |
| Residuals     | 40.465          | 41 | NA    | NA      |

Table 5: Tukey groups assigned to seasons differing in density.

|        | Density  | Grouping |
|--------|----------|----------|
| fall   | 11.45883 | a        |
| summer | 11.28221 | a        |
| spring | 11.03341 | a        |
| winter | 10.36318 | a        |

Table 6: Tukey groups assigned to seasons differing in density.

|              | D 11     | α .      |
|--------------|----------|----------|
|              | Density  | Grouping |
| fall:1980    | 12.12740 | a        |
| spring:2010  | 11.46404 | ab       |
| summer:1980  | 11.36928 | ab       |
| summer: 2010 | 11.19515 | ab       |
| fall:2010    | 10.79026 | ab       |
| spring:1980  | 10.60277 | ab       |
| winter:2010  | 10.45709 | ab       |
| winter:1980  | 10.28493 | b        |

Table 7: Table of the effect sizes in the density ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.005       | 0.95 | 0      | 1       |
| Season        | 0.166       | 0.95 | 0      | 1       |
| Period:Season | 0.158       | 0.95 | 0      | 1       |

### Consumer & Predators ANOVAs

Biomass and density values were subset and summed for consumer and predator taxa. Biomass and density were then compared by sampling period, season, and the interaction using an ANOVA with Type II sums-of-squares. ANOVA assumptions were inspected graphically using check\_model(), and effect sizes for the ANOVAs were calculated as  $\eta_P^2$  using eta\_squared(). Post-hoc Tukey's HSD tests were conducted using HSD.test() to examine for differences among groups for influential factors in the ANOVA.

```
## Subset data to only include consumer taxa
# Biomass
consumer.biomass <- invertebrate.biomass.matrix[, 1:46] %>%
    rowSums()
consumer.biomass.data <- tibble(sampling.info, consumer.biomass)</pre>
# Density
consumer.density <- invertebrate.density.matrix[, 1:46] %>%
consumer.density.data <- tibble(sampling.info, consumer.density)</pre>
## Subset data to only include predator taxa
# Biomass
predator.biomass <- invertebrate.biomass.matrix[, 47:72] %>%
    rowSums()
predator.biomass.data <- tibble(sampling.info, predator.biomass)</pre>
predator.density <- invertebrate.density.matrix[, 47:72] %>%
    rowSums()
predator.density.data <- tibble(sampling.info, predator.density)</pre>
```

### Consumer ANOVAs

#### Biomass ANOVA

```
consumer.biomass.anova <- lm(
  consumer.biomass ~ Period * Season, data = consumer.biomass.data
)</pre>
```

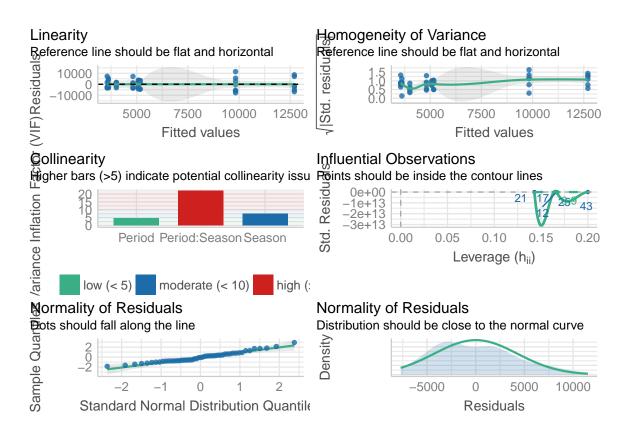


Figure 3: Diagnostic plots of the consumer biomass ANOVA.

Table 8: ANOVA results for consumer biomass by period, season, and the interaction.

|               | Sums-of-Squares | df | F      | P-value |
|---------------|-----------------|----|--------|---------|
| Period        | 214935664       | 1  | 10.005 | 0.003   |
| Season        | 113765041       | 3  | 1.765  | 0.169   |
| Period:Season | 158830701       | 3  | 2.465  | 0.076   |
| Residuals     | 880756410       | 41 | NA     | NA      |

Table 9: Tukey groups assigned to periods differing in consumer biomass.

|      | Consumer Biomass | Grouping |
|------|------------------|----------|
| 1980 | 8195.969         | a        |
| 2010 | 4080.589         | b        |

Table 10: Tukey groups assigned to season and period groups differing in consumer biomass.

|             | Consumer Biomass | Grouping |
|-------------|------------------|----------|
| fall:1980   | 12700.172        | a        |
| summer:1980 | 9827.673         | ab       |
| winter:1980 | 5168.723         | ab       |
| spring:2010 | 5075.846         | ab       |
| spring:1980 | 4815.356         | ab       |
| summer:2010 | 4003.295         | b        |
| fall:2010   | 3612.908         | b        |
| winter:2010 | 3555.710         | b        |

Table 11: Table of the effect sizes in the consumer biomass ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.196       | 0.95 | 0.046  | 1       |
| Season        | 0.114       | 0.95 | 0.000  | 1       |
| Period:Season | 0.153       | 0.95 | 0.000  | 1       |

### Density ANOVA

```
consumer.density.anova <- lm(
   log(consumer.density) ~ Period * Season, data = consumer.density.data
)</pre>
```

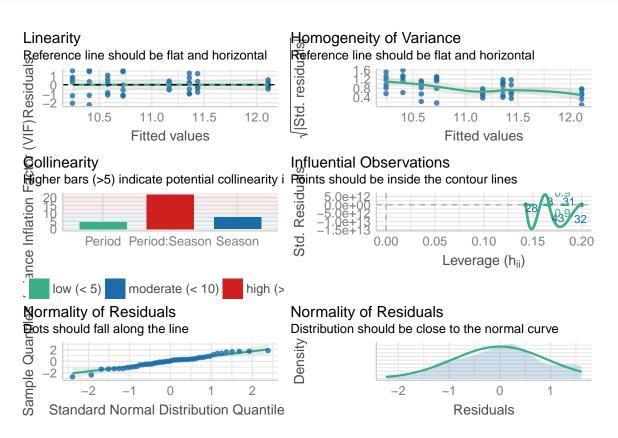


Figure 4: Diagnostic plots of the consumer density ANOVA.

Table 12: ANOVA results for consumer density by period, season, and the interaction.

|               | Sums-of-Squares | df | F     | P-value |
|---------------|-----------------|----|-------|---------|
| Period        | 0.269           | 1  | 0.271 | 0.605   |
| Season        | 8.236           | 3  | 2.768 | 0.054   |
| Period:Season | 7.867           | 3  | 2.644 | 0.062   |
| Residuals     | 40.666          | 41 | NA    | NA      |

Table 13: Tukey groups assigned to seasons differing in consumer density.

|        | Consumer Density | Grouping |
|--------|------------------|----------|
| fall   | 11.41674         | a        |
| summer | 11.26117         | a        |
| spring | 11.00815         | a        |
| winter | 10.31693         | a        |

Table 14: Tukey groups assigned to season and period groups differing in consumer density.

|              | Consumer Density | Grouping |
|--------------|------------------|----------|
| fall:1980    | 12.10793         | a        |
| spring:2010  | 11.43664         | ab       |
| summer:1980  | 11.35745         | ab       |
| summer: 2010 | 11.16490         | ab       |
| fall:2010    | 10.72555         | ab       |
| spring:1980  | 10.57966         | ab       |
| winter:2010  | 10.40433         | ab       |
| winter:1980  | 10.24409         | b        |

Table 15: Table of the effect sizes in the consumer density ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.007       | 0.95 | 0      | 1       |
| Season        | 0.168       | 0.95 | 0      | 1       |
| Period:Season | 0.162       | 0.95 | 0      | 1       |

### Predator ANOVAs

#### Biomass ANOVA

```
predator.biomass.anova <- lm(
    log(predator.biomass) ~ Period * Season, data = predator.biomass.data
)</pre>
```

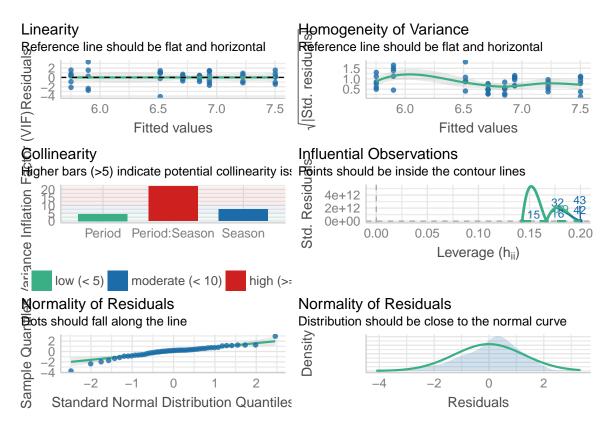


Figure 5: Diagnostic plots of the predator biomass ANOVA.

Table 16: ANOVA results for predator biomass by period, season, and the interaction.

|               | Sums-of-Squares | df | F     | P-value |
|---------------|-----------------|----|-------|---------|
| Period        | 0.145           | 1  | 0.076 | 0.784   |
| Season        | 12.100          | 3  | 2.121 | 0.112   |
| Period:Season | 2.442           | 3  | 0.428 | 0.734   |
| Residuals     | 77.975          | 41 | NA    | NA      |

Table 17: Tukey groups assigned to seasons differing in predator biomass.

|        | Predator Biomass | Grouping |
|--------|------------------|----------|
| fall   | 7.219654         | a        |
| spring | 6.870266         | a        |
| summer | 6.781070         | a        |
| winter | 5.819327         | a        |

Table 18: Table of the effect sizes in the predator biomass ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.002       | 0.95 | 0      | 1       |
| Season        | 0.134       | 0.95 | 0      | 1       |
| Period:Season | 0.030       | 0.95 | 0      | 1       |

### Density ANOVA

```
predator.density.anova <- lm(
    log(predator.density) ~ Period * Season, data = predator.density.data
)</pre>
```

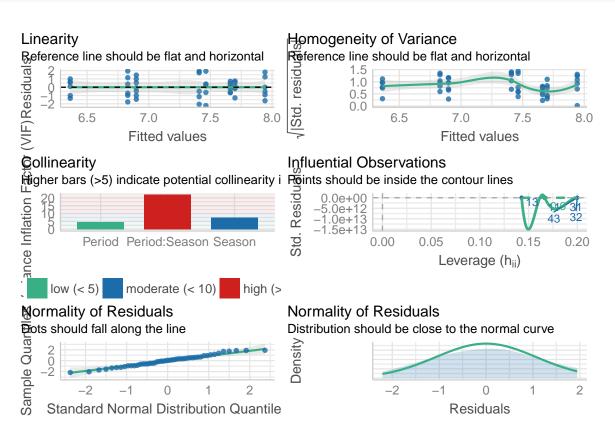


Figure 6: Diagnostic plots of the predator density ANOVA.

Table 19: ANOVA results for predator density by period, season, and the interaction.

|               | Sums-of-Squares | df | F     | P-value |
|---------------|-----------------|----|-------|---------|
| Period        | 7.688           | 1  | 5.651 | 0.022   |
| Season        | 3.138           | 3  | 0.769 | 0.518   |
| Period:Season | 1.308           | 3  | 0.320 | 0.811   |
| Residuals     | 55.780          | 41 | NA    | NA      |

Table 20: Tukey groups assigned to periods differing in predator density.

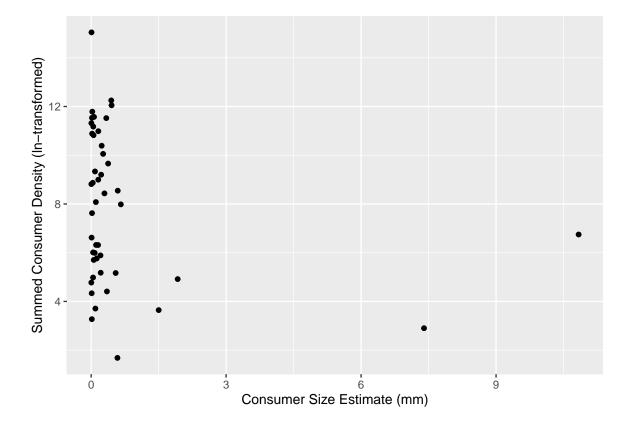
|      | Predator Density | Grouping |
|------|------------------|----------|
| 2010 | 7.688569         | a        |
| 1980 | 6.889090         | b        |

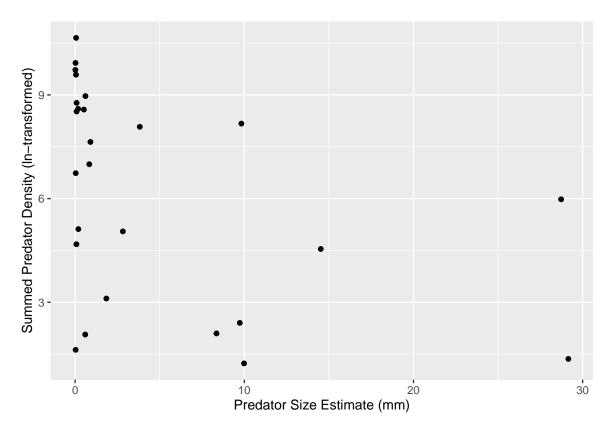
Table 21: Table of the effect sizes in the predator density ANOVA.

| Term          | Eta-squared | CI   | CI_Low | CI_High |
|---------------|-------------|------|--------|---------|
| Period        | 0.121       | 0.95 | 0.01   | 1       |
| Season        | 0.053       | 0.95 | 0.00   | 1       |
| Period:Season | 0.023       | 0.95 | 0.00   | 1       |

## Size Groupings

Body sizes were plotted to estimate size groupings. For each taxon, we calculated mean body mass across time periods by dividing total biomass by total abundance. Primary consumers were classified as Small: 0 - 0.4 mg, Medium: 0.41 - 1.0 mg, and Large: 1.1 - 12.0 mg. Predators were classified as Small: 0 - 5.0 mg, Medium: 5.1 - 15.0 mg, and Large: 15.1 - 30.0 mg.





```
## Divide groups up by size as per plot groupings
# Consumer.size.groups <- cut(
    consumer.size.data$consumer.size.estimates, c(0, 0.4, 1.0, 12.0, Inf),
    right = FALSE
    )
consumer.size.data$size.group <- as.numeric(consumer.size.groups)

# Predators
predator.size.groups <- cut(
    predator.size.data$predator.size.estimates, c(0, 5, 15, 30, Inf),
    right = FALSE
    )
predator.size.data$size.group <- as.numeric(predator.size.groups)</pre>
```

## **Biomass Change**

We numerically evaluated changes in biomass by size groupings for consumers and predators.

#### Consumers

Consumer taxa within each size grouping:

- Group 1: Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae, Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae, Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia, Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae, Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis, Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx, Tipulidae, Triaenodes, Trichoptera, Tricorythodes
- Group 2: Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa
- Group 3: Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche

```
## Consumer data management
# 1980s
consumer.biomass.1980s <- biomass.data %>%
   filter(Period == "1980") %>%
    select(Amphinemura:Tricorythodes) %>%
   mutate(summed biomass = rowSums(.))
# Sum all biomass for the 1980s consumers
consumer.biomass.1980s.total.biomass <- sum(consumer.biomass.1980s$summed_biomass)</pre>
# Summed biomass = 204899.2
# 2010s
consumer.biomass.2010s <- biomass.data %>%
   filter(Period == "2010") %>%
    select(Amphinemura:Tricorythodes) %>%
   mutate(summed_biomass = rowSums(.))
# Sum all biomass for the 2010s consumers
consumer.biomass.2010s.total.biomass <- sum(consumer.biomass.2010s$summed biomass)
# Summed biomass = 97934.14
```

```
## 1980s consumers group 1 taxa
group.1.consumer.biomass.1980s <- consumer.biomass.1980s %>%
    select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
                 Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae,
                 Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia,
                 Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae,
                 Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis,
                 Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx,
                 Tipulidae, Triaenodes, Trichoptera, Tricorythodes) %>%
   mutate(summed biomass = rowSums(.))
## 1980s consumers group 1 total biomass
group.1.consumer.biomass.1980s.total.biomass <- sum(group.1.consumer.biomass.1980s$summed_biomass)
# Summed biomass = 78968.09
## 1980s consumers group 2 taxa
group.2.consumer.biomass.1980s <- consumer.biomass.1980s %>%
    select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa) %>%
   mutate(summed_biomass = rowSums(.))
## 1980s consumers group 2 total biomass
group.2.consumer.biomass.1980s.total.biomass <- sum(group.2.consumer.biomass.1980s$summed biomass)
# Summed biomass = 116946.4
## 1980s consumers group 3 taxa
group.3.consumer.biomass.1980s <- consumer.biomass.1980s %>%
    select(Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche) %>%
   mutate(summed_biomass = rowSums(.))
## 1980s consumers group 3 total biomass
group.3.consumer.biomass.1980s.total.biomass <- sum(group.3.consumer.biomass.1980s$summed_biomass)
# Summed biomass = 8984.69
```

```
## 2010s consumers group 1 taxa
group.1.consumer.biomass.2010s <- consumer.biomass.2010s %>%
    select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
                 Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae,
                 Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia,
                 Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae,
                 Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis,
                 Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx,
                 Tipulidae, Triaenodes, Trichoptera, Tricorythodes) %>%
    mutate(summed biomass = rowSums(.))
## 2010s consumers group 1 total biomass
group.1.consumer.biomass.2010s.total.biomass <- sum(group.1.consumer.biomass.2010s$summed_biomass)
# Summed biomass = 38115.6
## 2010s consumers group 2 taxa
group.2.consumer.biomass.2010s <- consumer.biomass.2010s %>%
    select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa) %>%
   mutate(summed biomass = rowSums(.))
## 2010s consumers group 2 total biomass
group.2.consumer.biomass.2010s.total.biomass <- sum(group.2.consumer.biomass.2010s$summed biomass)
# Summed biomass = 59127.29
## 2010s consumers group 3 taxa
group.3.consumer.biomass.2010s <- consumer.biomass.2010s %>%
    select(Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche) %>%
    mutate(summed_biomass = rowSums(.))
## 2010s consumers group 3 total biomass
group.3.consumer.biomass.2010s.total.biomass <- sum(group.3.consumer.biomass.2010s$summed_biomass)
# Summed biomass = 691.2489
## 1980s Consumers (% of total biomass within sampling period)
(group.1.consumer.biomass.1980s.total.biomass/consumer.biomass.1980s.total.biomass)*100
# Group 1 = 38.54%
(group.2.consumer.biomass.1980s.total.biomass/consumer.biomass.1980s.total.biomass)*100
# Group 2 = 57.08%
(group.3.consumer.biomass.1980s.total.biomass/consumer.biomass.1980s.total.biomass)*100
# Group 3 = 4.38\%
## 2010s Consumers
(group.1.consumer.biomass.2010s.total.biomass/consumer.biomass.2010s.total.biomass)*100
# Group 1 = 38.92%
(group.2.consumer.biomass.2010s.total.biomass/consumer.biomass.2010s.total.biomass)*100
# Group 2 = 60.37\%
(group.3.consumer.biomass.2010s.total.biomass/consumer.biomass.2010s.total.biomass)*100
\# Group 3 = 0.71\%
```

### **Predators**

Predator taxa within each size grouping:

- Group 1: Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea, Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia, Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera
- Group 2: Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna
- Group 3: Helopicus, Neurocordulia

```
## Predator data management
# 1980s
predator.biomass.1980s <- biomass.data %>%
   filter(Period == "1980") %>%
    select(Acroneuria:Plecoptera) %>%
   mutate(summed_biomass = rowSums(.))
# Sum all biomass for the 1980s predators
predator.biomass.1980s.total.biomass <- sum(predator.biomass.1980s$summed_biomass)
# Summed biomass = 35683.48
# 2010s
predator.biomass.2010s <- biomass.data %>%
   filter(Period == "2010") %>%
   select(Acroneuria:Plecoptera) %>%
   mutate(summed_biomass = rowSums(.))
\# Sum all biomass for the 2010s predators
predator.biomass.2010s.total.biomass <- sum(predator.biomass.2010s$summed_biomass)</pre>
\# Summed biomass = 41286.46
```

```
## 1980s predators group 1 taxa
group.1.predator.biomass.1980s <- predator.biomass.1980s %>%
    select(Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea,
                 Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia,
                 Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera)
    mutate(summed_biomass = rowSums(.))
## 1980s predators group 1 total biomass
group.1.predator.biomass.1980s.total.biomass <- sum(group.1.predator.biomass.1980s$summed_biomass)
# Summed biomass = 13214.15
## 1980s predators group 2 taxa
group.2.predator.biomass.1980s <- predator.biomass.1980s %>%
    select(Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna) %>%
   mutate(summed_biomass = rowSums(.))
## 1980s predators group 2 total biomass
group.2.predator.biomass.1980s.total.biomass <- sum(group.2.predator.biomass.1980s$summed_biomass)</pre>
# Summed biomass = 17395.72
## 1980s predators group 3 taxa
group.3.predator.biomass.1980s <- predator.biomass.1980s %>%
    select(Helopicus, Neurocordulia) %>%
   mutate(summed_biomass = rowSums(.))
## 1980s predators group 3 total biomass
group.3.predator.biomass.1980s.total.biomass <- sum(group.3.predator.biomass.1980s$summed_biomass)
# Summed biomass = 5073.609
```

```
## 2010s predators group 1 taxa
group.1.predator.biomass.2010s <- predator.biomass.2010s %>%
    select(Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea,
                 Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia,
                 Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera)
    mutate(summed_biomass = rowSums(.))
## 2010s predators group 1 total biomass
group.1.predator.biomass.2010s.total.biomass <- sum(group.1.predator.biomass.2010s$summed_biomass)
# Summed biomass = 15968.22
## 2010s predators group 2 taxa
group.2.predator.biomass.2010s <- predator.biomass.2010s %>%
    select(Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna) %>%
   mutate(summed_biomass = rowSums(.))
## 2010s predators group 2 total biomass
group.2.predator.biomass.2010s.total.biomass <- sum(group.2.predator.biomass.2010s$summed_biomass)
# Summed biomass = 18922.44
## 2010s predators group 3 taxa
group.3.predator.biomass.2010s <- predator.biomass.2010s %>%
    select(Helopicus, Neurocordulia) %>%
    mutate(summed_biomass = rowSums(.))
## 2010s predators group 3 total biomass
group.3.predator.biomass.2010s.total.biomass <- sum(group.3.predator.biomass.2010s$summed_biomass)
# Summed biomass = 6395.791
## 1980s Predators (% of total biomass within sampling period)
(group.1.predator.biomass.1980s.total.biomass/predator.biomass.1980s.total.biomass)*100
# Group 1 = 37.03%
(group.2.predator.biomass.1980s.total.biomass/predator.biomass.1980s.total.biomass)*100
# Group 2 = 48.75\%
(group.3.predator.biomass.1980s.total.biomass/predator.biomass.1980s.total.biomass)*100
\# Group 3 = 14.22\%
## 2010s Predators
(group.1.predator.biomass.2010s.total.biomass/predator.biomass.2010s.total.biomass)*100
# Group 1 = 38.68\%
(group.2.predator.biomass.2010s.total.biomass/predator.biomass.2010s.total.biomass)*100
# Group 2 = 45.83\%
(\texttt{group.3.predator.biomass.2010s.total.biomass/predator.biomass.2010s.total.biomass)*100
# Group 3 = 15.49\%
## Divide mean 2010s biomass by mean 1980s biomass
percent.predator.biomass.change <- (</pre>
    mean(predator.biomass.2010s.total.biomass)/mean(predator.biomass.1980s.total.biomass)
    )*100
# 2010s biomass = 115.70% of 1980s biomass
```

## Biomass Change & Contribution of Dominant Caddisflies

```
## Hydropsyche
Hydropsyche.biomass.1980s <- biomass.1980s$Hydropsyche
Hydropsyche.biomass.2010s <- biomass.2010s$Hydropsyche

## Chimarra
Chimarra.biomass.1980s <- biomass.1980s$Chimarra
Chimarra.biomass.2010s <- biomass.2010s$Chimarra

## Cheumatopsyche
Cheumatopsyche.biomass.1980s <- biomass.1980s$Cheumatopsyche
Cheumatopsyche.biomass.2010s <- biomass.2010s$Cheumatopsyche</pre>
```

```
## Hydropsyche
(mean(Hydropsyche.biomass.2010s)/mean(Hydropsyche.biomass.1980s))*100
# 2010 biomass = 34.31% of 1980s
## Chimarra
(mean(Chimarra.biomass.2010s)/mean(Chimarra.biomass.1980s))*100
# 2010 biomass = 63.02% of 1980s
## Cheumatopsyche
(mean(Cheumatopsyche.biomass.2010s)/mean(Cheumatopsyche.biomass.1980s))*100
# 2010 biomass = 14.96% of 1980s
## Percent of total biomass made up by all 3 dominant caddisflies
# 1980s
((mean(Hydropsyche.biomass.1980s) + mean(Chimarra.biomass.1980s) +
         mean(Cheumatopsyche.biomass.1980s)) / mean(summed.biomass.1980s))*100
# 1980s = 60.80% of all biomass
# 2010s
((mean(Hydropsyche.biomass.2010s) + mean(Chimarra.biomass.2010s) +
          mean(Cheumatopsyche.biomass.2010s)) / mean(summed.biomass.2010s))*100
\# 2010s = 42.15\% of all biomass
```

## **Density Change**

We numerically evaluated changes in density by size groupings for consumers and predators.

#### Consumers

Consumer taxa within each size grouping:

- Group 1: Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae, Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae, Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia, Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae, Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis, Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx, Tipulidae, Triaenodes, Trichoptera, Tricorythodes
- Group 2: Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa
- Group 3: Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche

```
## Consumer data management
# 1980s
consumer.density.1980s <- density.data %>%
    filter(Period == "1980") %>%
    select(Amphinemura:Tricorythodes) %>%
    mutate(summed density = rowSums(.))
# Sum all density for the 1980s consumers
consumer.density.1980s.total.density <- sum(consumer.density.1980s$summed_density)</pre>
# Summed density = 2580454
# 2010s
consumer.density.2010s <- density.data %>%
    filter(Period == "2010") %>%
    select(Amphinemura:Tricorythodes) %>%
    mutate(summed_density = rowSums(.))
# Sum all density for the 2010s consumers
consumer.density.2010s.total.density <- sum(consumer.density.2010s$summed_density)</pre>
# Summed density = 2110836
```

```
## 1980s consumers group 1 taxa
group.1.consumer.density.1980s <- consumer.density.1980s %>%
    select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
                 Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae,
                 Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia,
                 Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae,
                 Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis,
                 Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx,
                 Tipulidae, Triaenodes, Trichoptera, Tricorythodes) %>%
   mutate(summed density = rowSums(.))
## 1980s consumers group 1 total density
group.1.consumer.density.1980s.total.density <- sum(group.1.consumer.density.1980s$summed_density)</pre>
# Summed density = 2302319
## 1980s consumers group 2 taxa
group.2.consumer.density.1980s <- consumer.density.1980s %>%
    select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa) %>%
   mutate(summed density = rowSums(.))
## 1980s consumers group 2 total density
group.2.consumer.density.1980s.total.density <- sum(group.2.consumer.density.1980s$summed_density)
# Summed density = 277220.8
## 1980s consumers group 3 taxa
group.3.consumer.density.1980s <- consumer.density.1980s %>%
    select(Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche) %>%
   mutate(summed_density = rowSums(.))
## 1980s consumers group 3 total density
group.3.consumer.density.1980s.total.density <- sum(group.3.consumer.density.1980s$summed_density)
\# Summed density = 913.8923
```

```
## 2010s consumers group 1 taxa
group.1.consumer.density.2010s <- consumer.density.2010s %>%
    select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
                 Cheumatopsyche, Chironomidae, Crambidae, Cyrnellus, Dubiraphia, Elmidae,
                 Ephemerellidae, Ephemeroptera, Heptageniidae, Hydropsychidae, Isonychia,
                 Isopoda, Lepidoptera, Leptoceridae, Leptophlebiidae, Limnephilidae,
                 Macronychus, Microcylloepus, Nectopsyche, Nemouridae, Neureclipsis,
                 Polycentropodidae, Psephenus, Simuliidae, Stenelmis, Taeniopteryx,
                 Tipulidae, Triaenodes, Trichoptera, Tricorythodes) %>%
   mutate(summed density = rowSums(.))
## 2010s consumers group 1 total density
group.1.consumer.density.2010s.total.density <- sum(group.1.consumer.density.2010s$summed_density)
# Summed density = 1999623
## 2010s consumers group 2 taxa
group.2.consumer.density.2010s <- consumer.density.2010s %>%
    select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shipsa) %>%
    mutate(summed density = rowSums(.))
## 2010s consumers group 2 total density
group.2.consumer.density.2010s.total.density <- sum(group.2.consumer.density.2010s$summed_density)
\# Summed density = 111083.3
## 2010s consumers group 3 taxa
group.3.consumer.density.2010s <- consumer.density.2010s %>%
    select(Ironoquia, Neargyractis, Pteronarcys, Pycnopsyche) %>%
   mutate(summed_density = rowSums(.))
## 2010s consumers group 3 total density
group.3.consumer.density.2010s.total.density <- sum(group.3.consumer.density.2010s$summed_density)
# Summed density = 130.0784
## 1980s Consumers (% of total density within sampling period)
(group.1.consumer.density.1980s.total.density/consumer.density.1980s.total.density)*100
# Group 1 = 89.22%
(group.2.consumer.density.1980s.total.density/consumer.density.1980s.total.density)*100
# Group 2 = 10.74%
(group.3.consumer.density.1980s.total.density/consumer.density.1980s.total.density)*100
# Group 3 = 0.04\%
## 2010s Consumers
(group.1.consumer.density.2010s.total.density/consumer.density.2010s.total.density)*100
# Group 1 = 94.73%
(group.2.consumer.density.2010s.total.density/consumer.density.2010s.total.density)*100
# Group 2 = 5.26\%
(group.3.consumer.density.2010s.total.density/consumer.density.2010s.total.density)*100
\# Group 3 = 0.01\%
## Divide mean 2010s density by mean 1980s density
percent.consumer.density.change <- (</pre>
   mean(consumer.density.2010s.total.density)/mean(consumer.density.1980s.total.density)
```

)\*100

# 2010s density = 81.80% of 1980s density

### **Predators**

Predator taxa within each size grouping:

- Group 1: Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea, Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia, Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera
- Group 2: Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna
- Group 3: Helopicus, Neurocordulia

```
## Predator data management
# 1980s
predator.density.1980s <- density.data %>%
   filter(Period == "1980") %>%
    select(Acroneuria:Plecoptera) %>%
   mutate(summed_density = rowSums(.))
# Sum all biomass for the 1980s predators
predator.density.1980s.total.density <- sum(predator.density.1980s$summed_density)
# Summed density = 47433.32
# 2010s
predator.density.2010s <- density.data %>%
   filter(Period == "2010") %>%
   select(Acroneuria:Plecoptera) %>%
   mutate(summed_density = rowSums(.))
# Sum all biomass for the 2010s predators
predator.density.2010s.total.density <- sum(predator.density.2010s$summed_density)</pre>
# Summed density = 88336.69
```

```
## 1980s predators group 1 taxa
group.1.predator.density.1980s <- predator.density.1980s %>%
    select(Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea,
                 Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia,
                 Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera)
    mutate(summed_density = rowSums(.))
## 1980s predators group 1 total density
group.1.predator.density.1980s.total.density <- sum(group.1.predator.density.1980s$summed_density)
# Summed density = 46049.37
## 1980s predators group 2 taxa
group.2.predator.density.1980s <- predator.density.1980s %>%
    select(Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna) %>%
   mutate(summed_density = rowSums(.))
## 1980s predators group 2 total density
group.2.predator.density.1980s.total.density <- sum(group.2.predator.density.1980s$summed_density)
# Summed density = 47433.32
## 1980s predators group 3 taxa
group.3.predator.density.1980s <- predator.density.1980s %>%
    select(Helopicus, Neurocordulia) %>%
   mutate(summed_density = rowSums(.))
## 1980s predators group 3 total density
group.3.predator.density.1980s.total.density <- sum(group.3.predator.density.1980s$summed_density)
# Summed density = 158.1566
```

```
## 2010s predators group 1 taxa
group.1.predator.density.2010s <- predator.density.2010s %>%
    select(Acroneuria, Aeshnidae, Amphipoda, Anisoptera, Argia, Ceraclea,
                 Ceratopogoninae, Cernotina, Coenagrionidae, Enallagma, Hemerodromia,
                 Isoperla, Neoperla, Oecetis, Paragnetina, Perlesta, Perlidae, Perlodidae, Plecoptera)
    mutate(summed_density = rowSums(.))
## 2010s predators group 1 total density
group.1.predator.density.2010s.total.density <- sum(group.1.predator.density.2010s$summed_density)</pre>
# Summed density = 85669.75
## 2010s predators group 2 taxa
group.2.predator.density.2010s <- predator.density.2010s %>%
    select(Aeshna, Boyeria, Corydalus, Hydroperla, Nasiaeschna) %>%
    mutate(summed_density = rowSums(.))
## 2010s predators group 2 total density
group.2.predator.density.2010s.total.density <- sum(group.2.predator.density.2010s$summed_density)
# Summed density = 2425.899
## 2010s predators group 3 taxa
group.3.predator.density.2010s <- predator.density.2010s %>%
    select(Helopicus, Neurocordulia) %>%
   mutate(summed_density = rowSums(.))
## 2010s predators group 3 total density
group.3.predator.density.2010s.total.density <- sum(group.3.predator.density.2010s$summed_density)
\# Summed density = 241.0455
## 1980s Predators (% of total density within sampling period)
(group.1.predator.density.1980s.total.density/predator.density.1980s.total.density)*100
# Group 1 = 97.08%
(group.2.predator.density.1980s.total.density/predator.density.1980s.total.density)*100
# Group 2 = 2.58%
(group.3.predator.density.1980s.total.density/predator.density.1980s.total.density)*100
\# Group 3 = 0.33\%
## 2010s Predators
(group.1.predator.density.2010s.total.density/predator.density.2010s.total.density)*100
# Group 1 = 96.98%
(group.2.predator.density.2010s.total.density/predator.density.2010s.total.density)*100
# Group 2 = 2.75%
(group.3.predator.density.2010s.total.density/predator.density.2010s.total.density)*100
# Group 3 = 0.27%
## Divide mean 2010s density by mean 1980s density
percent.predator.density.change <- (</pre>
   mean(predator.density.2010s.total.density)/mean(predator.density.1980s.total.density)
    )*100
# 2010s density = 186.23% of 1980s density
```

## Biomass & Density Chi-Squared Tests

We performed a chi-squared test to determine whether there were differences in biomass by size groupings between sampling period for consumers and predators. Chi-squared tests were performed using chisq.test(), with observed and expected frequencies examined to confirm test assumptions were met. Additionally, we assessed residuals to determine over- and under-represented size groupings by sampling period. Residuals were calculated as:

$$\frac{Observed-Expected}{\sqrt{Expected}}$$

#### **Consumer Biomass**

```
## Set consumer chi-square dataframe
consumer.biomass.chi.square.data <- as.data.frame(data.frame(matrix(0, nrow = 3, ncol = 2)))
colnames(consumer.biomass.chi.square.data) <- c("1980s", "2010s")
rownames(consumer.biomass.chi.square.data) <- c("Group_1", "Group_2", "Group_3")

## Add biomass values by groups for the 1980s and 2010s
# 1980s
consumer.biomass.chi.square.data$`1980s` <- c(
    group.1.consumer.biomass.1980s.total.biomass,
    group.2.consumer.biomass.1980s.total.biomass,
    group.3.consumer.biomass.2010s.total.biomass
)

# 2010s
consumer.biomass.chi.square.data$`2010s` <- c(
    group.1.consumer.biomass.2010s.total.biomass,
    group.3.consumer.biomass.2010s.total.biomass,
    group.3.consumer.biomass.2010s.total.biomass,
    group.3.consumer.biomass.2010s.total.biomass)
</pre>
```

```
consumer.biomass.chi.square.test <- chisq.test(consumer.biomass.chi.square.data) # chi-squared = 2933.7, df = 2, P < 0.00001
```

Table 22: Observed biomass by size groupings for the consumer chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 78968.09  | 38115.600 |
| $Group\_2$ | 116946.43 | 59127.287 |
| Group_3    | 8984.69   | 691.249   |

Table 23: Expected biomass by size groupings for the consumer chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 79219.67  | 37864.027 |
| $Group\_2$ | 119132.74 | 56940.979 |
| $Group\_3$ | 6546.81   | 3129.129  |

Table 24: Residual biomass by size groupings for the consumer chi-squared test.

|            | 1980s  | 2010s   |
|------------|--------|---------|
| Group_1    | -0.894 | 1.293   |
| $Group\_2$ | -6.334 | 9.162   |
| $Group\_3$ | 30.130 | -43.581 |

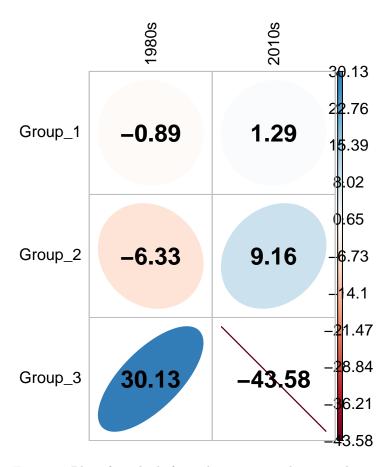


Figure 7: Plot of residuals from the consumer chi-squared test.

## Consumer Density

```
## Set consumer chi-square dataframe
consumer.density.chi.square.data <- as.data.frame(data.frame(matrix(0, nrow = 3, ncol = 2)))
colnames(consumer.density.chi.square.data) <- c("1980s", "2010s")
rownames(consumer.density.chi.square.data) <- c("Group_1", "Group_2", "Group_3")

## Add density values by groups for the 1980s and 2010s
# 1980s
consumer.density.chi.square.data$^1980s^ <- c(
    group.1.consumer.density.1980s.total.density,
    group.2.consumer.density.1980s.total.density,
    group.3.consumer.density.1980s.total.density
)

# 2010s
consumer.density.chi.square.data$^2010s^ <- c(
    group.1.consumer.density.2010s.total.density,
    group.2.consumer.density.2010s.total.density,
    group.3.consumer.density.2010s.total.density,
    group.3.consumer.density.2010s.total.density,
    group.3.consumer.density.2010s.total.density,
    group.3.consumer.density.2010s.total.density,
    group.3.consumer.density.2010s.total.density
)</pre>
```

```
consumer.density.chi.square.test <- chisq.test(consumer.density.chi.square.data) # chi-squared = 46424, df = 2, P < 0.00001
```

Table 25: Observed density by size groupings for the consumer chi-squared test.

|            | 1980s       | 2010s       |
|------------|-------------|-------------|
| Group_1    | 2302319.190 | 1999622.789 |
| $Group\_2$ | 277220.761  | 111083.255  |
| Group_3    | 913.892     | 130.078     |

Table 26: Expected density by size groupings for the consumer chi-squared test.

|            | 1980s       | 2010s       |
|------------|-------------|-------------|
| Group_1    | 2366292.170 | 1935649.809 |
| $Group\_2$ | 213587.435  | 174716.581  |
| Group_3    | 574.238     | 469.732     |

Table 27: Residual density by size groupings for the consumer chi-squared test.

|            | 1980s   | 2010s    |
|------------|---------|----------|
| Group_1    | -41.587 | 45.982   |
| $Group\_2$ | 137.688 | -152.236 |
| Group_3    | 14.174  | -15.672  |

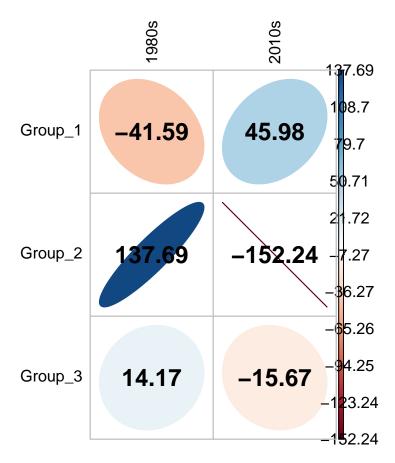


Figure 8: Plot of residuals from the consumer chi-squared test.

## **Predator Biomass**

```
## Set predator chi-square dataframe
predator.biomass.chi.square.data <- as.data.frame(data.frame(matrix(0, nrow = 3, ncol = 2)))
colnames(predator.biomass.chi.square.data) <- c("1980s", "2010s")
rownames(predator.biomass.chi.square.data) <- c("Group_1", "Group_2", "Group_3")

## Add biomass values by groups for the 1980s and 2010s
# 1980s
predator.biomass.chi.square.data$`1980s` <- c(
    group.1.predator.biomass.1980s.total.biomass,
    group.2.predator.biomass.1980s.total.biomass,
    group.3.predator.biomass.1980s.total.biomass
)

# 2010s
predator.biomass.chi.square.data$`2010s` <- c(
    group.1.predator.biomass.2010s.total.biomass,
    group.2.predator.biomass.2010s.total.biomass,
    group.3.predator.biomass.2010s.total.biomass,
    group.3.predator.biomass.2010s.total.biomass,
    group.3.predator.biomass.2010s.total.biomass
)</pre>
```

```
predator.biomass.chi.square.test <- chisq.test(predator.biomass.chi.square.data)
# chi-squared = 69.014, df = 2, P < 0.00001</pre>
```

Table 28: Observed biomass by size groupings for the predator chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 13214.148 | 15968.223 |
| $Group\_2$ | 17395.718 | 18922.441 |
| Group_3    | 5073.609  | 6395.791  |

Table 29: Expected biomass by size groupings for the predator chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 13529.029 | 15653.342 |
| $Group\_2$ | 16837.200 | 19480.959 |
| $Group\_3$ | 5317.246  | 6152.154  |

Table 30: Residual biomass by size groupings for the predator chi-squared test.

|            | 1980s  | 2010s  |
|------------|--------|--------|
| Group_1    | -2.707 | 2.517  |
| $Group\_2$ | 4.304  | -4.002 |
| Group_3    | -3.341 | 3.106  |

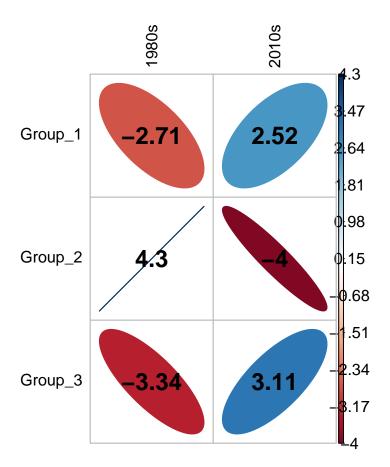


Figure 9: Plot of residuals from the predator chi-squared test.

## **Predator Density**

```
## Set predator chi-square dataframe
predator.density.chi.square.data <- as.data.frame(data.frame(matrix(0, nrow = 3, ncol = 2)))
colnames(predator.density.chi.square.data) <- c("1980s", "2010s")
rownames(predator.density.chi.square.data) <- c("Group_1", "Group_2", "Group_3")

## Add density values by groups for the 1980s and 2010s
# 1980s
predator.density.chi.square.data$`1980s` <- c(
    group.1.predator.density.1980s.total.density,
    group.2.predator.density.1980s.total.density,
    group.3.predator.density.1980s.total.density
)

# 2010s
predator.density.chi.square.data$`2010s` <- c(
    group.1.predator.density.2010s.total.density,
    group.2.predator.density.2010s.total.density,
    group.3.predator.density.2010s.total.density,
    group.3.predator.density.2010s.total.density,
    group.3.predator.density.2010s.total.density
)</pre>
```

```
predator.density.chi.square.test <- chisq.test(predator.density.chi.square.data)
# chi-squared = 6.8918, df = 2, P = 0.03188</pre>
```

Table 31: Observed density by size groupings for the predator chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 46049.372 | 85669.746 |
| $Group\_2$ | 1225.790  | 2425.899  |
| Group_3    | 158.157   | 241.046   |

Table 32: Expected density by size groupings for the predator chi-squared test.

|            | 1980s     | 2010s     |
|------------|-----------|-----------|
| Group_1    | 46018.078 | 85701.040 |
| $Group\_2$ | 1275.773  | 2375.916  |
| Group_3    | 139.467   | 259.735   |

Table 33: Residual density by size groupings for the predator chi-squared test.

|            | 1980s  | 2010s  |
|------------|--------|--------|
| Group_1    | 0.146  | -0.107 |
| $Group\_2$ | -1.399 | 1.025  |
| Group_3    | 1.583  | -1.160 |

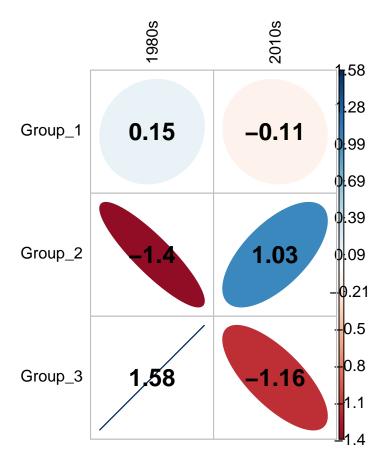


Figure 10: Plot of residuals from the predator chi-squared test.

## **Biomass Contributions**

Biomass contributions were calculated for consumer and predator groups.

#### Consumers

```
## Separate consumer biomass data by sampling period
# 1980s
consumer.1980s.biomass <- invertebrate.biomass.matrix %>%
        select(Amphinemura:Tricorythodes) %>%
        slice(1:25)
consumer.1980s.mean.biomass <- t(colMeans(consumer.1980s.biomass))
consumer.1980s.biomass.contribution <- (
        (consumer.1980s.mean.biomass/consumer.1980s.biomass)*100
)

# 2010s
consumer.2010s.biomass <- invertebrate.biomass.matrix %>%
        select(Amphinemura:Tricorythodes) %>%
        slice(26:49)
consumer.2010s.mean.biomass <- t(colMeans(consumer.2010s.biomass))
consumer.2010s.biomass.contribution <- (
        (consumer.2010s.mean.biomass/consumer.2010s.biomass)*100
)</pre>
```

#### **Predators**

## **Individual Body Size Change**

We tested whether mean body sizes of the most dominant primary consumers and predators were different between sampling periods. We calculated the average percent composition of biomass for each taxon within each sampling period and focused on the five most dominant consumers and the three most dominant predators. We calculated average individual mass for each sampling month by dividing biomass by abundance and performed two-sample Welch t-tests on average mass between the 1980s and the 2010s for each of the dominant taxa.

- Dominant consumers = Hydropsyche, Cheumatopsyche, Chimarra, Heptageniidae, Chironomidae
- Dominant predators = Corydalus, Paragnetina, Neurocordulia

#### **Body Size t-Tests**

```
## Data management for body size t-tests
## Hydropsyche
Hydropsyche.80.size <- ((biomass.1980s$Hydropsyche)/(density.1980s$Hydropsyche))
Hydropsyche.10.size <- ((biomass.2010s$Hydropsyche)/(density.2010s$Hydropsyche))
## Cheumatopsyche
Cheumatopsyche.80.size <- ((biomass.1980s$Cheumatopsyche)/(density.1980s$Cheumatopsyche))
Cheumatopsyche.10.size <- ((biomass.2010s$Cheumatopsyche)/(density.2010s$Cheumatopsyche))
## Chimarra
Chimarra.80.size <- ((biomass.1980s$Chimarra)/(density.1980s$Chimarra))
Chimarra.10.size <- ((biomass.2010s$Chimarra)/(density.2010s$Chimarra))
## Heptageniidae
Heptageniidae.80.size <- ((biomass.1980s$Heptageniidae)/(density.1980s$Heptageniidae))</pre>
Heptageniidae.10.size <- ((biomass.2010s$Heptageniidae)/(density.2010s$Heptageniidae))</pre>
## Chironomidae
Chironomidae.80.size <- ((biomass.1980s$Chironomidae)/(density.1980s$Chironomidae))
Chironomidae.10.size <- ((biomass.2010s$Chironomidae)/(density.2010s$Chironomidae))
## Corydalus
Corydalus.80.size <- ((biomass.1980s$Corydalus)/(density.1980s$Corydalus))</pre>
Corydalus.10.size <- ((biomass.2010s$Corydalus)/(density.2010s$Corydalus))</pre>
## Paragnetina
Paragnetina.80.size <- ((biomass.1980s$Paragnetina)/(density.1980s$Paragnetina))
Paragnetina.10.size <- ((biomass.2010s$Paragnetina)/(density.2010s$Paragnetina))
## Neurocordulia
Neurocordulia.80.size <- ((biomass.1980s$Neurocordulia)/(density.1980s$Neurocordulia))
Neurocordulia.10.size <- ((biomass.2010s$Neurocordulia)/(density.2010s$Neurocordulia))
```

```
## Hydrospyche
t.test(log(Hydropsyche.80.size + 0.00001), log(Hydropsyche.10.size + 0.00001),
       paired = FALSE)
# t = 1.8613, df = 46.739, P-value = 0.069
## Cheumatopsyche
t.test(log(Cheumatopsyche.80.size + 0.00001), log(Cheumatopsyche.10.size + 0.00001),
       paired = FALSE)
\# \ t = 5.1915, \ df = 46.86, \ P-value < 0.001
## Chimarra
t.test(log(Chimarra.80.size + 0.00001), log(Chimarra.10.size + 0.00001),
       paired = FALSE)
# t = 0.42278, df = 44.042, P-value = 0.675
## Heptageniidae
t.test(log(Heptageniidae.80.size + 0.00001), log(Heptageniidae.10.size + 0.00001),
       paired = FALSE)
# t = -0.40634, df = 46.633, P-value = 0.686
## Chironomidae
t.test(log(Chironomidae.80.size + 0.00001), log(Chironomidae.10.size + 0.00001),
       paired = FALSE)
# t = -0.17408, df = 46.334, P-value = 0.863
## Corydalus
t.test(log(Corydalus.80.size + 0.00001), log(Corydalus.10.size + 0.00001),
       paired = FALSE)
# t = 1.2464, df = 31.152, P-value = 0.222
## Paragnetina
t.test(Paragnetina.80.size + 0.00001, Paragnetina.10.size + 0.00001,
       paired = FALSE)
\# \ t = 0.2442, \ df = 26.9, \ P-value = 0.809
## Neurocordulia
t.test(Neurocordulia.80.size + 0.00001, Neurocordulia.10.size + 0.00001,
       paired = FALSE)
# t = 1.3378, df = 28.187, P-value = 0.192
```

## **Community Composition**

Community composition was evaluated by calculating a Bray-Curtis distance matrix on a taxon-by-sample biomass matrix. We then conducted a PERMANOVA on this distance matrix to test for effects of sampling period, season, and the interaction on community structure.

```
## Data management for the PERMANOVA
# Sample information for the distance matrix
distance.matrix.info <- biomass.data %>%
    select(UID, Year, Season, Period)
# Taxa-by-abundance community matrix
community.matrix <- log(invertebrate.biomass.matrix + 1)</pre>
## Calculate Bray-Curtis Distance
BC.distance.biomass <- vegdist(community.matrix, method = "bray")
## NMDS Ordination
BC.NMDS.biomass <- monoMDS(BC.distance.biomass, k = 2, model = "global")
# Stress = 0.1819
## PERMANOVA by period, season, and the interaction
community.composition.PERMANOVA <- adonis(</pre>
   BC.distance.biomass ~ Period * Season,
   data = distance.matrix.info,
   permutations = 10000
```

Table 34: Summary of the PERMANOVA comparing composition by period, season, and the interaction.

| Term          | df | Sums-of-Squares | Mean Square | F      | \$R^2\$ | P-value |
|---------------|----|-----------------|-------------|--------|---------|---------|
| Period        | 1  | 0.985           | 0.985       | 14.868 | 0.196   | 0.000   |
| Season        | 3  | 1.046           | 0.349       | 5.265  | 0.209   | 0.000   |
| Period:Season | 3  | 0.267           | 0.089       | 1.341  | 0.053   | 0.133   |
| Residuals     | 41 | 2.716           | 0.066       | NA     | 0.542   | NA      |
| Total         | 48 | 5.013           | NA          | NA     | 1.000   | NA      |

# Indicator Taxa Analysis

We conducted an indicator species analysis to determine which taxa characterized specific sampling period and season groups.

```
## Create groups based on season: Group 1 = 1980s, Group 2 = 2010s
period.groups <- c(rep(1, 25), rep(2, 24))

## Run indicator analysis comparing by period
indicator.analysis <- multipatt(
    community.matrix,
    period.groups,
    control = how(nperm = 10000)
    )</pre>
```

Table 35: Summary table of the sign of the relationship from the indicator taxa analysis.

|                 | 1980s | 2010s | Index | Test Statistic | P-value |
|-----------------|-------|-------|-------|----------------|---------|
| Amphinemura     | 1     | 0     | 1     | 0.346          | 0.235   |
| Ancyronyx       | 1     | 1     | 3     | 0.857          | NA      |
| Baetidae        | 1     | 1     | 3     | 1.000          | NA      |
| Baetisca        | 0     | 1     | 2     | 0.552          | 0.031   |
| Brachycentrus   | 0     | 1     | 2     | 0.764          | 0.000   |
| Caenidae        | 1     | 1     | 3     | 0.795          | NA      |
| Cheumatopsyche  | 1     | 1     | 3     | 1.000          | NA      |
| Chimarra        | 1     | 1     | 3     | 0.979          | NA      |
| Chironomidae    | 1     | 1     | 3     | 1.000          | NA      |
| Crambidae       | 0     | 1     | 2     | 0.281          | 0.359   |
| Cyrnellus       | 0     | 1     | 2     | 0.645          | 0.000   |
| Dubiraphia      | 1     | 1     | 3     | 0.319          | NA      |
| Ectopria        | 1     | 0     | 1     | 0.400          | 0.109   |
| Elmidae         | 0     | 1     | 2     | 0.945          | 0.000   |
| Ephemerellidae  | 1     | 1     | 3     | 0.795          | NA      |
| Ephemeroptera   | 0     | 1     | 2     | 0.996          | 0.000   |
| Heptageniidae   | 1     | 1     | 3     | 1.000          | NA      |
| Hydropsyche     | 1     | 1     | 3     | 1.000          | NA      |
| Hydropsychidae  | 1     | 1     | 3     | 0.881          | NA      |
| Ironoquia       | 1     | 1     | 3     | 0.319          | NA      |
| Isonychia       | 1     | 1     | 3     | 0.904          | NA      |
| Isopoda         | 1     | 1     | 3     | 0.728          | NA      |
| Lepidoptera     | 0     | 1     | 2     | 0.568          | 0.005   |
| Leptoceridae    | 0     | 1     | 2     | 0.886          | 0.000   |
| Leptophlebiidae | 0     | 1     | 2     | 0.599          | 0.003   |
| Limnephilidae   | 0     | 1     | 2     | 0.284          | 0.363   |
| Macronychus     | 1     | 1     | 3     | 0.958          | NA      |
| Macrostemum     | 0     | 1     | 2     | 0.839          | 0.000   |
| Microcylloepus  | 0     | 1     | 2     | 0.704          | 0.000   |
| Neargyractis    | 1     | 0     | 1     | 0.400          | 0.106   |
| Nectopsyche     | 0     | 1     | 2     | 0.889          | 0.000   |

Table 35: Summary table of the sign of the relationship from the indicator taxa analysis. (continued)

|                   | 1980s | 2010s | Index | Test Statistic | P-value |
|-------------------|-------|-------|-------|----------------|---------|
| Nemouridae        | 1     | 0     | 1     | 0.200          | 1.000   |
| Neureclipsis      | 1     | 1     | 3     | 0.589          | NA      |
| Paraponyx         | 1     | 1     | 3     | 0.429          | NA      |
| Polycentropodidae | 0     | 1     | 2     | 0.661          | 0.001   |
| Psephenus         | 1     | 0     | 1     | 0.200          | 1.000   |
| Pteronarcys       | 1     | 0     | 1     | 0.937          | 0.000   |
| Pycnopsyche       | 1     | 1     | 3     | 0.202          | NA      |
| Shipsa            | 1     | 0     | 1     | 0.200          | 1.000   |
| Simuliidae        | 1     | 1     | 3     | 0.926          | NA      |
| Stenelmis         | 1     | 1     | 3     | 0.990          | NA      |
| Taeniopteryx      | 1     | 1     | 3     | 0.589          | NA      |
| Tipulidae         | 1     | 0     | 1     | 0.529          | 0.009   |
| Triaenodes        | 0     | 1     | 2     | 0.733          | 0.000   |
| Trichoptera       | 0     | 1     | 2     | 0.537          | 0.039   |
| Tricorythodes     | 1     | 1     | 3     | 0.742          | NA      |
| Acroneuria        | 1     | 1     | 3     | 0.833          | NA      |
| Aeshna            | 1     | 0     | 1     | 0.200          | 1.000   |
| Aeshnidae         | 0     | 1     | 2     | 0.289          | 0.235   |
| Amphipoda         | 0     | 1     | 2     | 0.875          | 0.000   |
| Anisoptera        | 0     | 1     | 2     | 0.289          | 0.225   |
| Argia             | 0     | 1     | 2     | 0.745          | 0.000   |
| Boyeria           | 0     | 1     | 2     | 0.507          | 0.043   |
| Ceraclea          | 0     | 1     | 2     | 0.680          | 0.003   |
| Ceratopogoninae   | 1     | 1     | 3     | 0.958          | NA      |
| Cernotina         | 0     | 1     | 2     | 0.577          | 0.001   |
| Coenagrionidae    | 0     | 1     | 2     | 0.524          | 0.058   |
| Corydalus         | 1     | 1     | 3     | 0.881          | NA      |
| Enallagma         | 0     | 1     | 2     | 0.537          | 0.007   |
| Helopicus         | 1     | 0     | 1     | 0.200          | 1.000   |
| Hemerodromia      | 1     | 1     | 3     | 0.881          | NA      |
| Hydroperla        | 0     | 1     | 2     | 0.289          | 0.240   |
| Isoperla          | 0     | 1     | 2     | 0.426          | 0.109   |
| Nasiaeschna       | 0     | 1     | 2     | 0.289          | 0.233   |
| Neoperla          | 1     | 1     | 3     | 0.904          | NA      |
| Neurocordulia     | 1     | 1     | 3     | 0.821          | NA      |
| Oecetis           | 1     | 1     | 3     | 0.904          | NA      |
| Paragnetina       | 1     | 1     | 3     | 0.915          | NA      |
| Perlesta          | 1     | 1     | 3     | 0.795          | NA      |
| Perlidae          | 1     | 1     | 3     | 0.881          | NA      |
| Perlodidae        | 0     | 1     | 2     | 0.354          | 0.117   |
| Plecoptera        | 1     | 1     | 3     | 0.728          | NA      |

# Carbon & Discharge ANCOVA

We evaluated the relationship between dissolved organic carbon and discharge using an ANCOVA. Monthly averages of carbon and discharge during our sampling periods (i.e., 1980s = December 1981-November 1983, 2010s = July 2015-August 2017) were used for this analysis. We structured the ANCOVA as:

```
Carbon = \alpha + \beta_1(\ln(\text{Discharge})) + \beta_2(\text{Period}) + \beta_3(\ln(\text{Discharge}) \times \text{Period}) + \epsilon
```

with the ANCOVA fitted using lm() and model assumptions checked using check\_model(). Influence of terms within the model was estimated by Type II sums-of-squares using the Anova() function.

```
carbon.discharge.ancova <- lm(</pre>
                  log(Carbon) ~ log(Mean_Discharge) * Period,
                  data = biomass.data
                   Linearity
                                                                                                                                                                                                                 Homogeneity of Variance
                   Reference line should be flat and horizontal
                                                                                                                                                                                                                  Reference line should be flat and horizontal
                  (VIF)Residual
                                                                                                                                                                                                                   Std.
                                                                            1.8
                                                                                                         2.0
                                                                                                                                      2.2
                                                                                                                                                                                                                                                                          1.8
                                                                                                                                                                                                                                                                                                       2.0
                                                                                                                                                                                                                                                                                                                                     2.2
                                                                                                                                                                                                                                                                                                                                                                  2.4
                                                                                                                                                                                                                                                                                                                                                                                                2.6
                                                                                                    Fitted values
                                                                                                                                                                                                                                                                                                   Fitted values
                   ©ollinearity
                                                                                                                                                                                                                  Influential Observations
                                                                                                                                                                                                                Roints should be inside the contour lines
                   Higher bars (>5) indicate potential collinearity issue
                                                                                                                                                                                                                   Residu
                                       50
50
70
70
70
70
                   ariance Inflation
                                                                                                                                                                                                                   Std.
                                       log(Meanlog(Steamge)ischarge):Perioidd
                                                                                                                                                                                                                                                 0.00
                                                                                                                                                                                                                                                                                                                         0.10
                                                                                                                                                                                                                                                                                                                                                            0.15
                                                                                                                                                                                                                                                                                                  Leverage (hii)
                                                                                                   moderate (< 10)
                                                low (< 5)
                                                                                                                                                                               high (>=
                   Mormality of Residuals
                                                                                                                                                                                                                 Normality of Residuals
                 Export Should fall along the line

The standard Normal Discourage of the s
                                                                                                                                                                                                                 Distribution should be close to the normal curve
                                                                                                                                                                                                                   Density
                                                                                                                                                                                      2
                                                                                                                                                                                                                                                                          -0.4
                                                                                                                                                                                                                                                                                                                     0.0
                                                                                                                                                                                                                                                                                                                                                            0.4
                                               Standard Normal Distribution Quantiles
                                                                                                                                                                                                                                                                                                         Residuals
```

Figure 11: Diagnostic plots of the carbon by discharge ANCOVA.

Table 36: Summary of the carbon by discharge ANCOVA.

| Term                                   | Sums-of-Squares | df | F      | P-value |
|--|-----------------|----|--------|---------|
| log(Mean_Discharge)                    | 3.098           | 1  | 35.101 | 0.000   |
| Period                                 | 0.086           | 1  | 0.976  | 0.329   |
| $\log(\text{Mean\_Discharge})$ :Period | 0.007           | 1  | 0.076  | 0.784   |
| Residuals                              | 3.707           | 42 | NA     | NA      |

Table 37: Table of the effect sizes in the carbon by discharge ANCOVA.

| Term                                   | Eta-squared | CI   | CI_Low | CI_High |
|--|-------------|------|--------|---------|
| log(Mean_Discharge)                    | 0.455       | 0.95 | 0.27   | 1       |
| Period                                 | 0.023       | 0.95 | 0.00   | 1       |
| $\log(\text{Mean\_Discharge})$ :Period | 0.002       | 0.95 | 0.00   | 1       |

# R Session Information

Table 38: Packages for data management and analyses.

| Package      | Loaded Version | Date           |
|--------------|----------------|----------------|
| agricolae    | 1.3-5          | 2021-06-06     |
| bayestestR   | 0.11.5         | 2021-10-30     |
| broom        | 0.7.12         | 2022-01-28     |
| car          | 3.0-12         | 2021-11-06     |
| carData      | 3.0-5          | 2022-01-06     |
| correlation  | 0.7.1          | 2021-10-06     |
| datawizard   | 0.2.3          | 2022-01-26     |
| dplyr        | 1.0.7          | 2021-06-18     |
| easystats    | 0.4.3          | 2021-11-07     |
| effectsize   | 0.6.0.1        | 2022-01-26     |
| forcats      | 0.5.1          | 2021-01-27     |
| ggplot2      | 3.3.5          | 2021-06-25     |
| indicspecies | 1.7.9          | 2020-02-04     |
| insight      | 0.15.0         | 2022-01-07     |
| kableExtra   | 1.3.4          | 2021-02-20     |
| knitr        | 1.37           | 2021-12-16     |
| lattice      | 0.20 - 45      | 2021-09-22     |
| modelbased   | 0.7.1          | 2022-01-13     |
| parameters   | 0.16.0         | 2022-01-12     |
| performance  | 0.8.0          | 2021-10-01     |
| permute      | 0.9-7          | 2022-01-27     |
| purrr        | 0.3.4          | 2020 - 04 - 17 |
| readr        | 2.1.2          | 2022-01-30     |
| report       | 0.5.0          | 2022-01-30     |
| reshape2     | 1.4.4          | 2020-04-09     |
| see          | 0.6.8          | 2021-10-03     |
| stringr      | 1.4.0          | 2019-02-10     |
| tibble       | 3.1.6          | 2021 - 11 - 07 |
| tidyr        | 1.1.4          | 2021-09-27     |
| tidyverse    | 1.3.1          | 2021-04-15     |
| vegan        | 2.5-7          | 2020-11-28     |
| waterData    | 1.0.8          | 2017-04-28     |