

# OGC-biomass

## Primary Analyses

Kelly Murray Stoker, David Murray-Stoker

### Contents

<b>Raw Data Management</b>	<b>3</b>
<b>Load Processed Data</b>	<b>6</b>
<b>Biomass &amp; Density ANOVAs</b>	<b>7</b>
Biomass ANOVA . . . . .	7
Density ANOVA . . . . .	10
<b>Consumer &amp; Predators ANOVAs</b>	<b>13</b>
Consumer ANOVAs . . . . .	14
Biomass ANOVA . . . . .	14
Density ANOVA . . . . .	17
Predator ANOVAs . . . . .	20
Biomass ANOVA . . . . .	20
Density ANOVA . . . . .	23
<b>Size Groupings</b>	<b>26</b>
<b>Percent Biomass Change</b>	<b>28</b>
Biomass Change & Contribution of Dominant Caddisflies . . . . .	29
<b>Density Change</b>	<b>30</b>
Consumers . . . . .	30
Predators . . . . .	33
<b>Biomass Chi-Squared Tests</b>	<b>36</b>
Consumers . . . . .	36
Predators . . . . .	38
<b>Biomass Contributions</b>	<b>40</b>
Consumers . . . . .	41
Predators . . . . .	41
<b>Individual Body Size Change</b>	<b>42</b>
Body Size t-Tests . . . . .	42
<b>Community Composition</b>	<b>44</b>
<b>Indicator Taxa Analysis</b>	<b>45</b>
<b>Carbon &amp; Discharge ANCOVA</b>	<b>47</b>



## 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)
```

```

## 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)

## Set variables as factors
# Biomass data
biomass.data$UID <- as_factor(biomass.data$UID)
biomass.data$Year <- as_factor(biomass.data$Year)
biomass.data$Year_Recoded <- as_factor(biomass.data$Year_Recoded)
biomass.data$Season <- as_factor(biomass.data$Season)
biomass.data$Season_Recoded <- as_factor(biomass.data$Season_Recoded)
biomass.data$Period <- as_factor(biomass.data$Period)

# Density data
density.data$UID <- as_factor(density.data$UID)
density.data$Year <- as_factor(density.data$Year)
density.data$Year_Recoded <- as_factor(density.data$Year_Recoded)
density.data$Season <- as_factor(density.data$Season)
density.data$Season_Recoded <- as_factor(density.data$Season_Recoded)
density.data$Period <- as_factor(density.data$Period)

## Dataframe of sampling info
sampling.info <- biomass.data[, 1:6]

## Matrices of invertebrate biomass and density
invertebrate.biomass.matrix <- biomass.data[, 16:87]
invertebrate.density.matrix <- density.data[, 16:87]

## Filter biomass and density data by sampling period; no further subsetting
# Biomass data
biomass.1980s <- invertebrate.biomass.matrix[1:25, ]
biomass.2010s <- invertebrate.biomass.matrix[26:49, ]

# Density data
density.1980s <- invertebrate.density.matrix[1:25, ]
density.2010s <- invertebrate.density.matrix[26:49, ]
```

## 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

```
biomass.anova <- lm(
  log(Summed_Biomass) ~ Period * Season, data = biomass.data
)
```

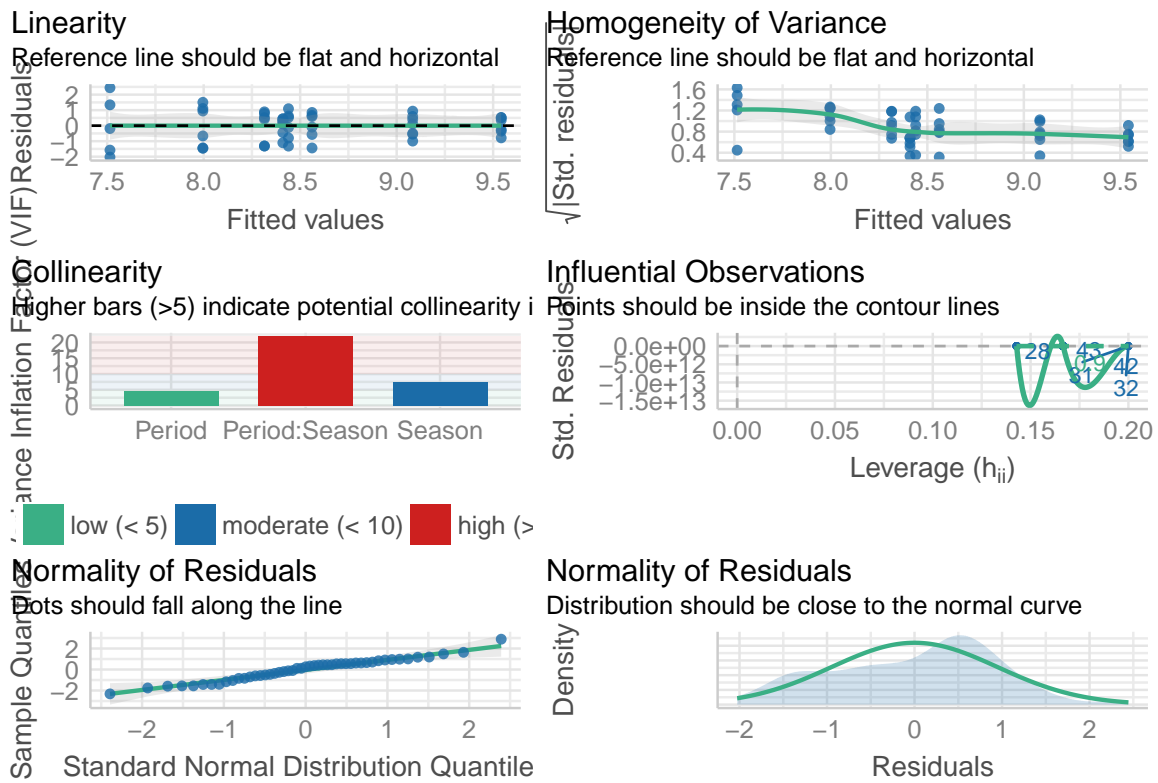


Figure 1: Diagnostic plots of the biomass ANOVA.

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



```
## Calculating partial eta-squared for each factor in the biomass ANOVA
biomass.anova.eta.squared <- eta_squared(
  Anova(biomass.anova, type = "II"),
  partial = TRUE
)
```

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
)
```

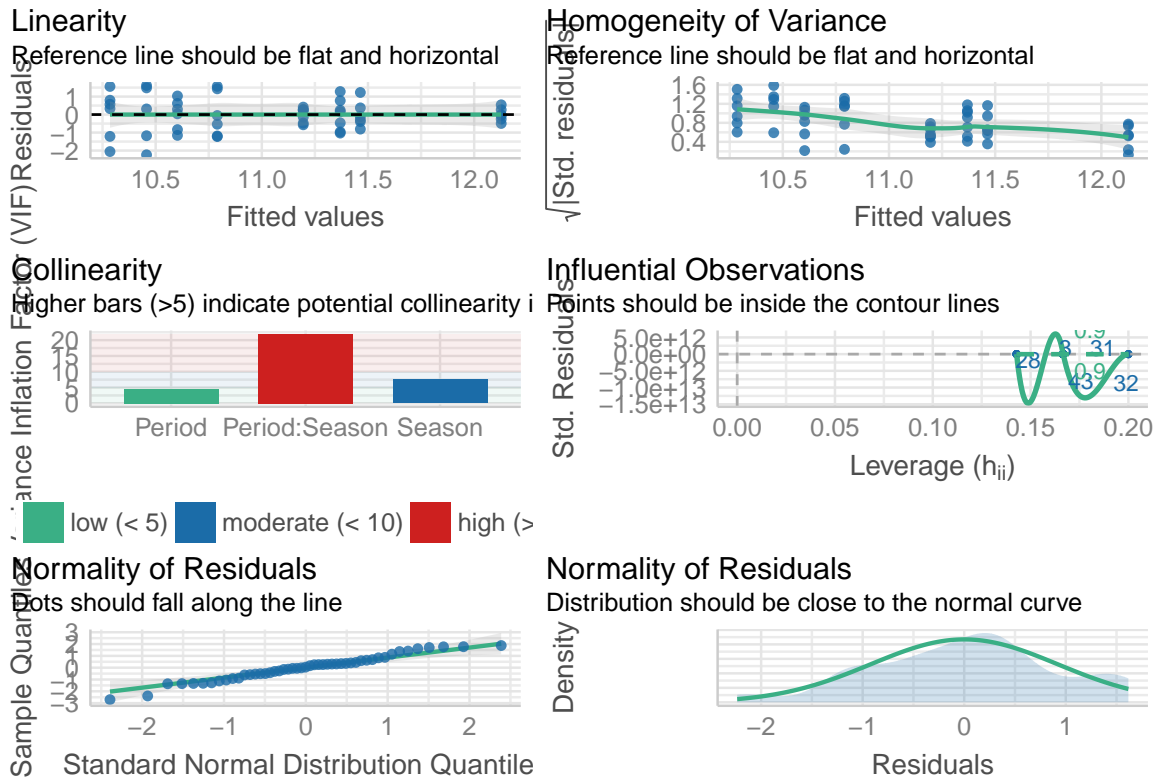


Figure 2: Diagnostic plots of the density ANOVA.

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.

	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

```
## Calculating partial eta-squared for each factor in the density ANOVA
density.anova.eta.squared <- eta_squared(
  Anova(density.anova, type = "II"),
  partial = TRUE
)
```

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)

# Density
consumer.density <- invertebrate.density.matrix[, 1:46] %>%
  rowSums()
consumer.density.data <- tibble(sampling.info, consumer.density)

## Subset data to only include predator taxa
# Biomass
predator.biomass <- invertebrate.biomass.matrix[, 47:72] %>%
  rowSums()
predator.biomass.data <- tibble(sampling.info, predator.biomass)

# Density
predator.density <- invertebrate.density.matrix[, 47:72] %>%
  rowSums()
predator.density.data <- tibble(sampling.info, predator.density)
```

## Consumer ANOVAs

### Biomass ANOVA

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

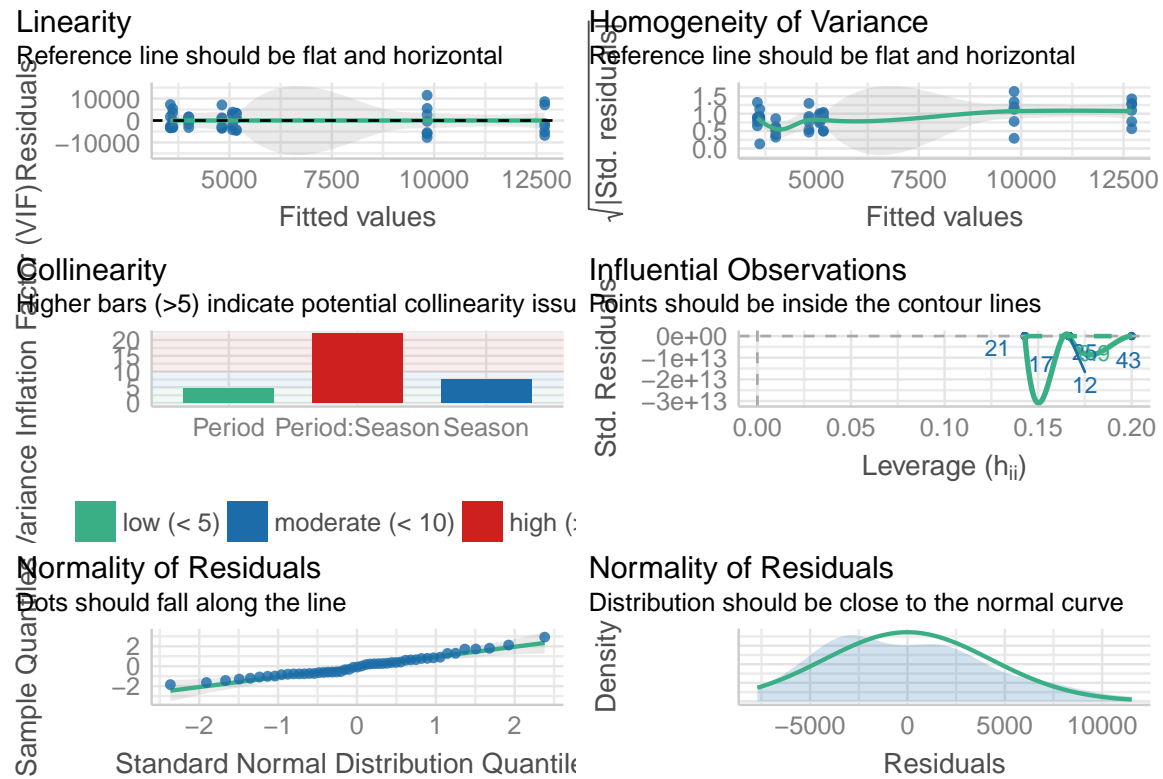


Figure 3: Diagnostic plots of the consumer biomass ANOVA.

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

```
## Calculating partial eta-squared for each factor in the biomass ANOVA
consumer.biomass.anova.eta.squared <- eta_squared(
  Anova(consumer.biomass.anova, type = "II"),
  partial = TRUE
)
```

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
)
```

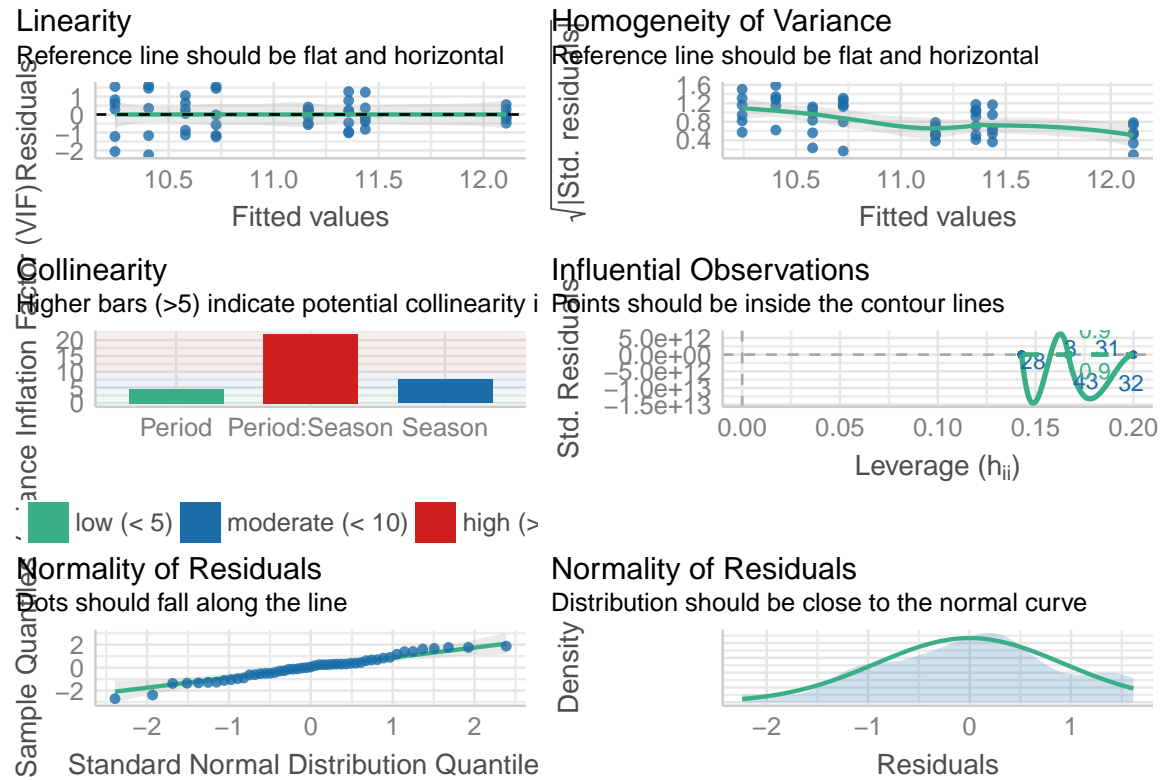


Figure 4: Diagnostic plots of the consumer density ANOVA.

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

```
## Calculating partial eta-squared for each factor in the density ANOVA
consumer.density.anova.eta.squared <- eta_squared(
  Anova(consumer.density.anova, type = "II"),
  partial = TRUE
)
```

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
)
```

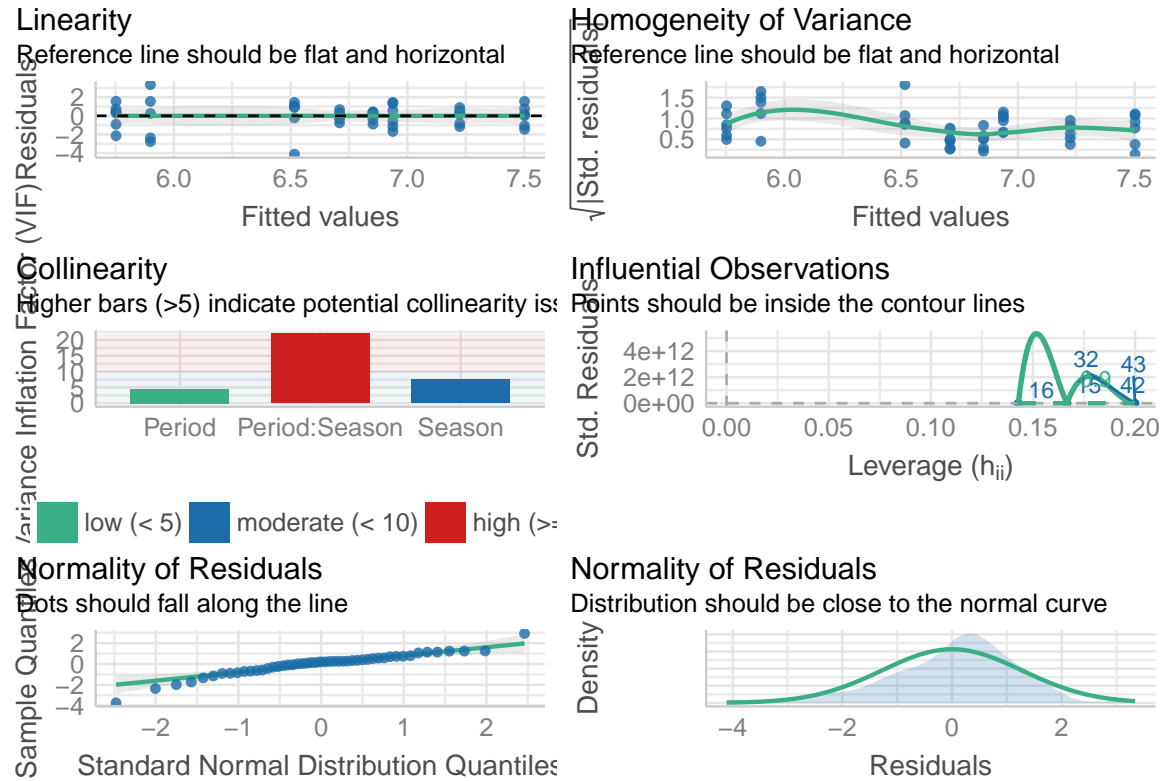


Figure 5: Diagnostic plots of the predator biomass ANOVA.

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

```
## Calculating partial eta-squared for each factor in the biomass ANOVA
predator.biomass.anova.eta.squared <- eta_squared(
  Anova(predator.biomass.anova, type = "II"),
  partial = TRUE
)
```

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
)
```

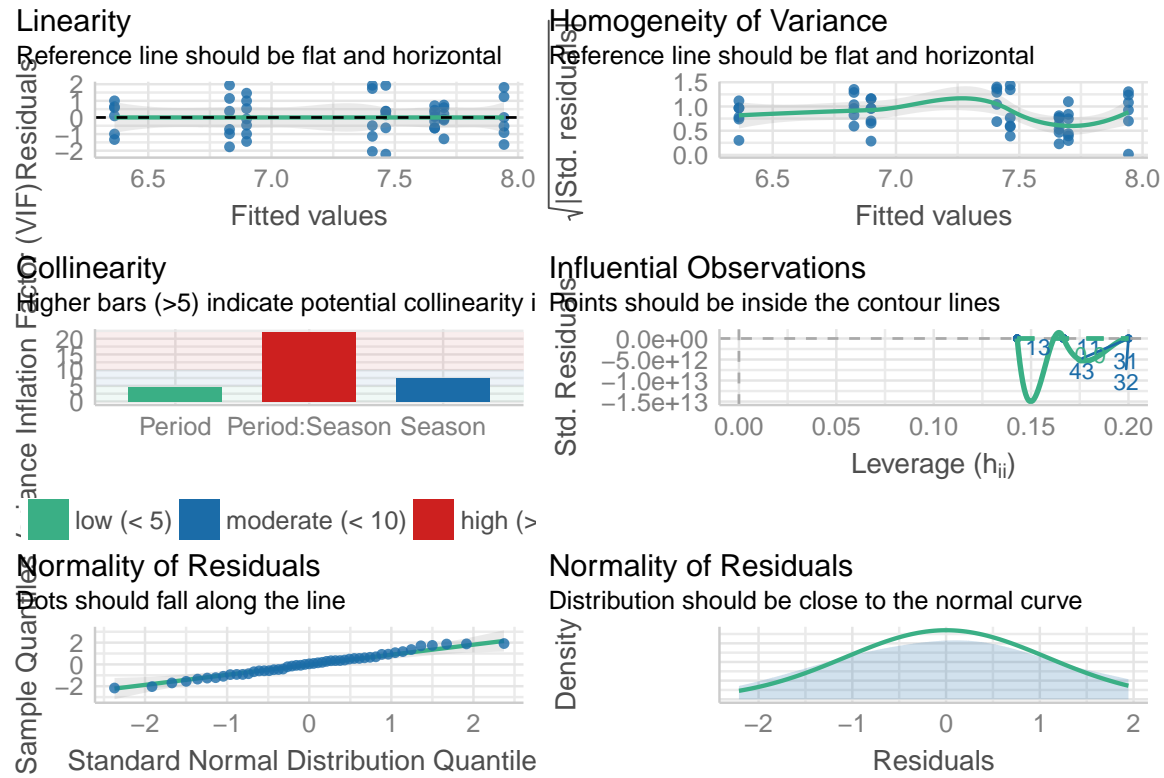


Figure 6: Diagnostic plots of the predator density ANOVA.

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



```
## Calculating partial eta-squared for each factor in the density ANOVA
predator.density.anova.eta.squared <- eta_squared(
  Anova(predator.density.anova, type = "II"),
  partial = TRUE
)
```

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

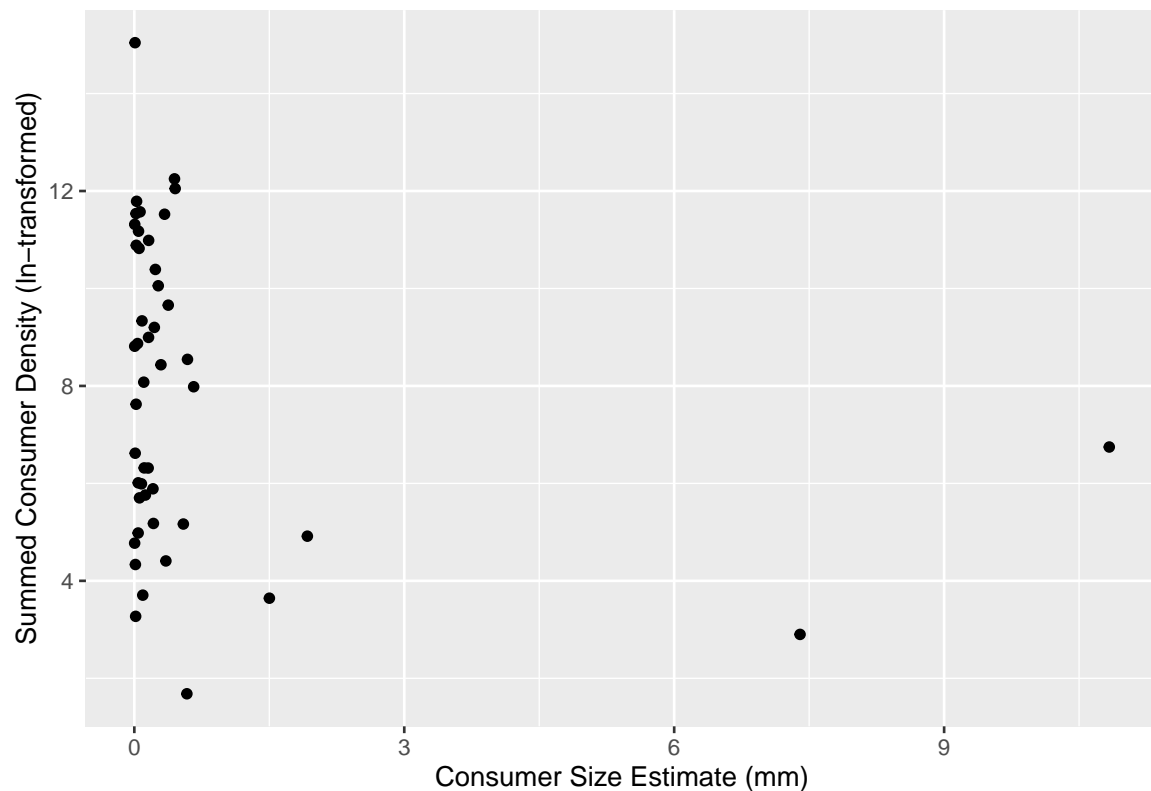
Term	Eta-squared	CI	CI_Low	CI_High
Period	0.121	0.95	0.01	1
Season	0.053	0.95	0.00	1
Period:Season	0.023	0.95	0.00	1

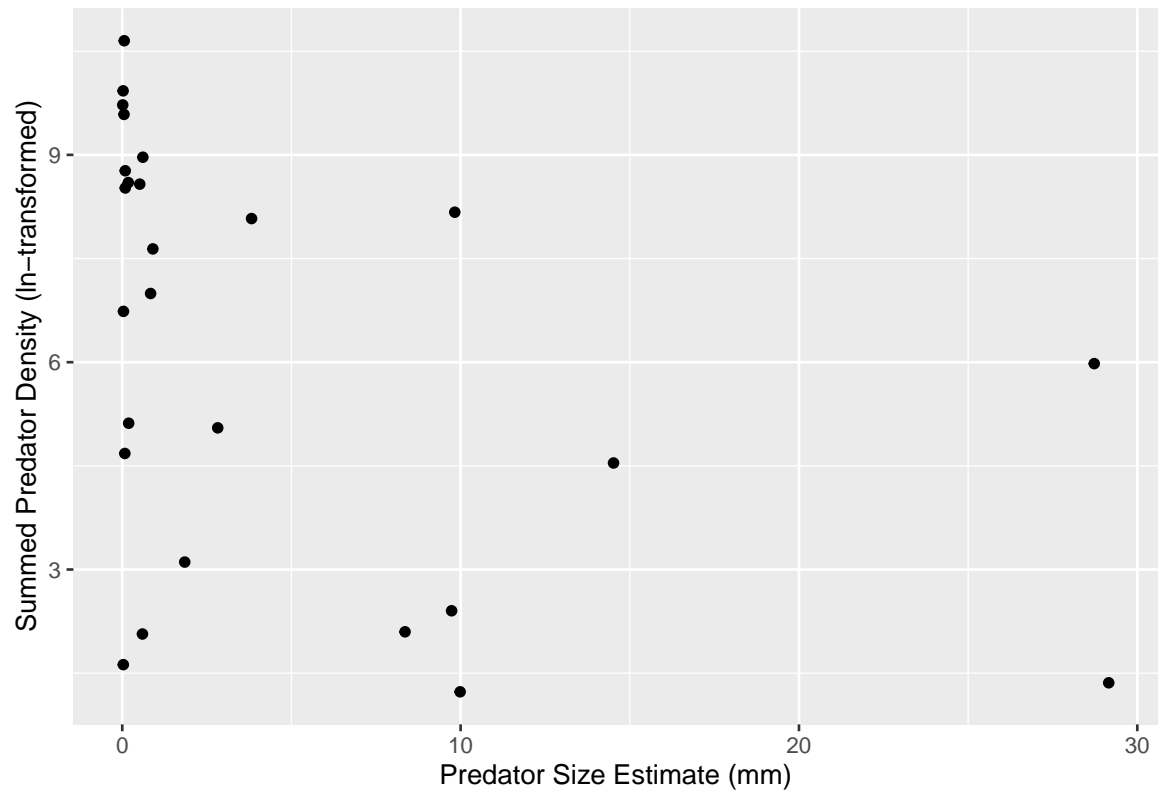
## Size Groupings

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.

```
## Define size classes
# Consumers
consumer.summed.biomass <- colSums(invertebrate.biomass.matrix[, 1:46])
consumer.summed.density <- colSums(invertebrate.density.matrix[, 1:46])
consumer.size.estimates <- consumer.summed.biomass/consumer.summed.density
consumer.size.data <- tibble(
  consumer.summed.biomass, consumer.summed.density, consumer.size.estimates
)

# Predators
predator.summed.biomass <- colSums(invertebrate.biomass.matrix[, 47:72])
predator.summed.density <- colSums(invertebrate.density.matrix[, 47:72])
predator.size.estimates <- predator.summed.biomass/predator.summed.density
predator.size.data <- tibble(
  predator.summed.biomass, predator.summed.density, predator.size.estimates
)
```





```
## Divide groups up by size as per plot groupings
# Consumers
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)
```

## Percent Biomass Change

We quantified the percent biomass change in size groupings between sampling periods by illustrating numerical differences. We also compared differences in biomass of three dominant caddisfly taxa. We did not conduct any null hypothesis significance tests for this component of the study.

```
## Calculate percent change in overall biomass
full.summed.biomass.vector <- biomass.data$Summed_Biomass
summed.biomass.1980s      <- full.summed.biomass.vector[1:25]
summed.biomass.2010s      <- full.summed.biomass.vector[26:49]

## Divide mean 2010s biomass by mean 1980s biomass
percent.biomass.change <- (mean(summed.biomass.2010s)/mean(summed.biomass.1980s))*100
# 2010s biomass = 60.28% 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
```

## 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, Cynellus, 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, Shpsa
- 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 biomass for the 1980s consumers
consumer.density.1980s.total.density <- sum(consumer.density.1980s$summed_density)

# 2010s
consumer.density.2010s <- density.data %>%
  filter(Period == "2010") %>%
  select(Amphinemura:Tricorythodes) %>%
  mutate(summed_density = rowSums(.))

# Sum all biomass for the 2010s consumers
consumer.density.2010s.total.density <- sum(consumer.density.2010s$summed_density)
```

```

## 1980s consumers group 1 taxa
group.1.consumer.density.1980s <- consumer.density.1980s %>%
  select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
         Cheumatopsyche, Chironomidae, Crambidae, Cynellus, 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)

## 1980s consumers group 2 taxa
group.2.consumer.density.1980s <- consumer.density.1980s %>%
  select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shippa) %>%
  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)

## 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)

```

```

## 2010s consumers group 1 taxa
group.1.consumer.density.2010s <- consumer.density.2010s %>%
  select(Amphinemura, Ancyronyx, Baetidae, Baetisca, Brachycentrus, Caenidae,
         Cheumatopsyche, Chironomidae, Crambidae, Cynellus, 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)

## 2010s consumers group 2 taxa
group.2.consumer.density.2010s <- consumer.density.2010s %>%
  select(Chimarra, Ectopria, Hydropsyche, Macrostemum, Paraponyx, Shippa) %>%
  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)

## 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)

```



## 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)

# 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)
```

```

## 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)

## 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)

## 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)

```

```

## 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)

## 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)

## 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)

```

## Biomass Chi-Squared Tests

We performed a chi-squared test to determine whether there were differences in density 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:

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

### Consumers

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

## Add biomass values by groups for the 1980s and 2010s
# 1980s
consumer.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.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
)

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

Table 22: Observed densities 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 23: Expected densities 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 24: Residual densities 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

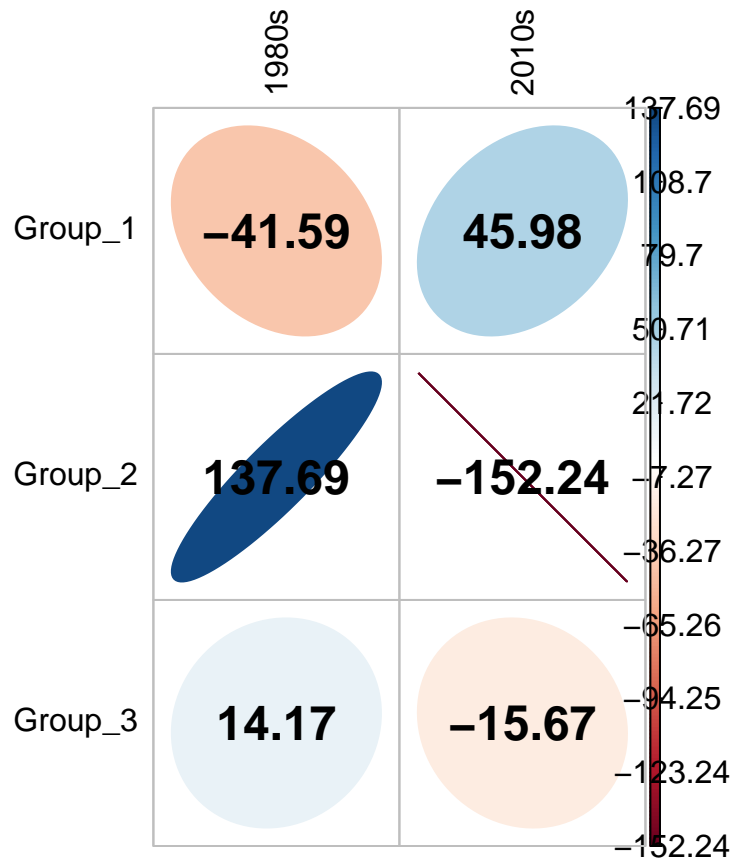


Figure 7: Plot of residuals from the consumer chi-squared test.

## Predators

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

## Add biomass values by groups for the 1980s and 2010s
# 1980s
predator.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.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
)

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

Table 25: Observed densities 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 26: Expected densities 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 27: Residual densities 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

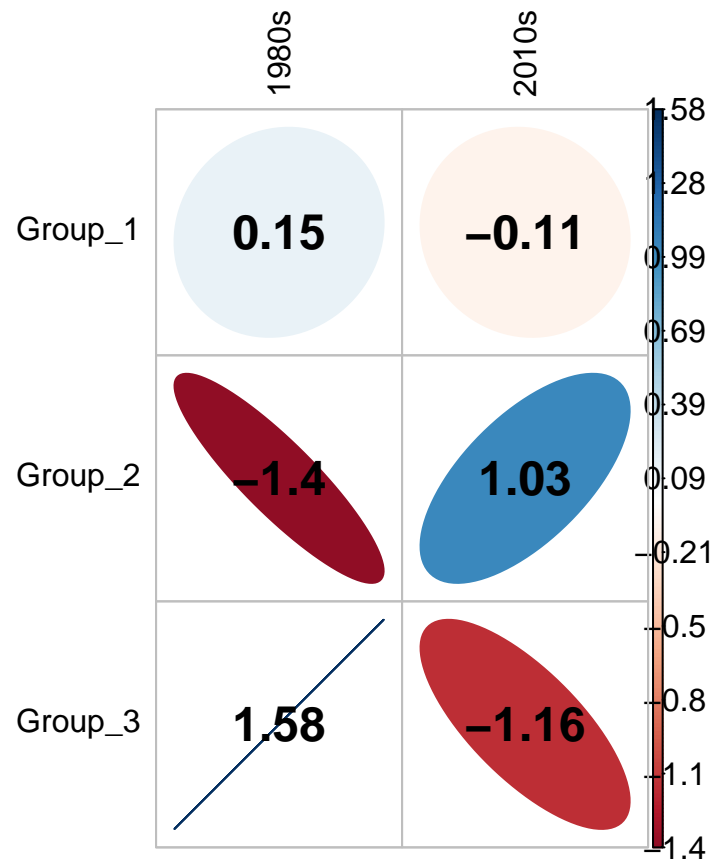


Figure 8: Plot of residuals from the predator chi-squared test.

## Biomass Contributions

Biomass contributions were calculated for consumer and predator groups. We again illustrated numerical differences without conducting any null hypothesis significance tests.

```
## Filter and format total biomass data by sampling period
# 1980s
total.1980s.biomass <- invertebrate.biomass.matrix[1:25, ]
total.1980s.mean.biomass <- t(colMeans(total.1980s.biomass))
total.1980s.percent.biomass.contribution <- (
  (total.1980s.mean.biomass/total.1980s.biomass)*100
)

# 2010s
total.2010s.biomass <- invertebrate.biomass.matrix[1:25, ]
total.2010s.mean.biomass <- t(colMeans(total.2010s.biomass))
total.2010s.percent.biomass.contribution <- (
  (total.2010s.mean.biomass/total.2010s.biomass)*100
)
```



## 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
)
```

## Predators

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

# 2010s
predator.2010s.biomass <- invertebrate.biomass.matrix %>%
  select(Acroneuria:Plecoptera) %>%
  slice(26:49)
predator.2010s.mean.biomass <- t(colMeans(predator.2010s.biomass))
predator.2010s.biomass.contribution <- (
  (predator.2010s.mean.biomass/predator.2010s.biomass)*100
)
```

## 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))
Heptageniidae.10.size <- ((biomass.2010s$Heptageniidae)/(density.2010s$Heptageniidae))

## 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))
Corydalus.10.size <- ((biomass.2010s$Corydalus)/(density.2010s$Corydalus))

## 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))
```

### ## Hydropsycha

```
t.test(log(Hydropsycha.80.size + 0.00001), log(Hydropsycha.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
```

### ## Corydalis

```
t.test(log(Corydalis.80.size + 0.00001), log(Corydalis.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)

## 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(
  BC.distance.biomass ~ Period * Season,
  data = distance.matrix.info,
  permutations = 10000
)
```

Table 28: 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	0.135
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)
)
```

Table 29: 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	0.224
Ancyronyx	1	1	3	0.857	NA
Baetidae	1	1	3	1.000	NA
Baetisca	0	1	2	0.552	0.030
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
Cynellus	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	0.004
Leptoceridae	0	1	2	0.886	0.000
Leptophlebiidae	0	1	2	0.599	0.004
Limnephilidae	0	1	2	0.284	0.352
Macronychus	1	1	3	0.958	NA
Macrostemum	0	1	2	0.839	0.000
Microcyloopus	0	1	2	0.704	0.000
Neargyractis	1	0	1	0.400	0.107
Nectopsyche	0	1	2	0.889	0.000

Table 29: 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	0.010
Triaenodes	0	1	2	0.733	0.000
Trichoptera	0	1	2	0.537	0.033
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	0.237
Amphipoda	0	1	2	0.875	0.000
Anisoptera	0	1	2	0.289	0.234
Argia	0	1	2	0.745	0.000
Boyeria	0	1	2	0.507	0.041
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	0.054
Corydalus	1	1	3	0.881	NA
Enallagma	0	1	2	0.537	0.006
Helopicus	1	0	1	0.200	1.000
Hemerodromia	1	1	3	0.881	NA
Hydroperla	0	1	2	0.289	0.238
Isoperla	0	1	2	0.426	0.107
Nasiaeschna	0	1	2	0.289	0.242
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	0.108
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:

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

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(
  log(Carbon) ~ log(Mean_Discharge) * Period,
  data = biomass.data
)
```

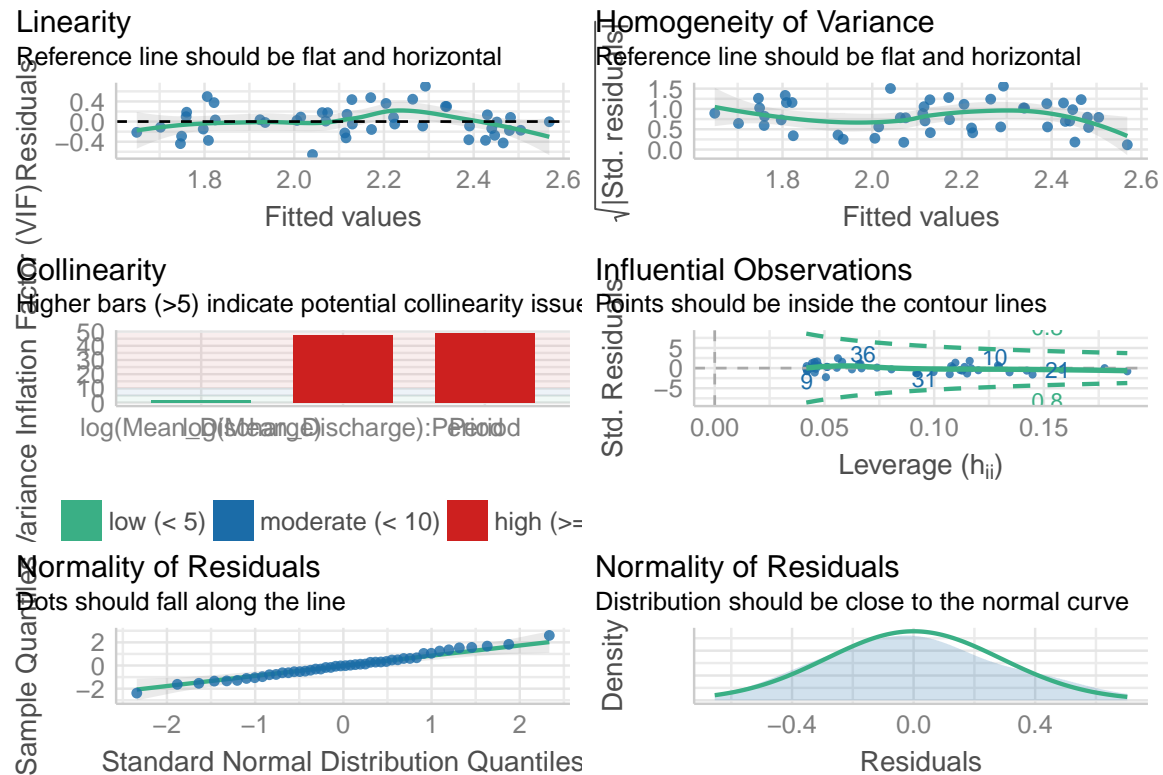


Figure 9: Diagnostic plots of the carbon by discharge ANCOVA.

Table 30: 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(Mean_Discharge):Period	0.007	1	0.076	0.784
Residuals	3.707	42	NA	NA



## R Session Information

Table 31: R session information for transparency and reproducing results.

Setting	Value
version	R version 4.1.2 (2021-11-01)
os	macOS Big Sur 10.16
system	x86_64, darwin17.0
ui	X11
language	(EN)
collate	en_CA.UTF-8
ctype	en_CA.UTF-8
tz	America/Toronto
date	2022-01-16
pandoc	2.14.0.3 @ /Applications/RStudio.app/Contents/MacOS/pandoc/ (via rmarkdown)

Table 32: Packages for data management and analysis.

Package	Loaded Version	Date
agricolae	1.3-5	2021-06-06
bayestestR	0.11.5	2021-10-30
broom	0.7.11	2022-01-03
car	3.0-12	2021-11-06
carData	3.0-5	2022-01-06
correlation	0.7.1	2021-10-06
datawizard	0.2.2	2022-01-04
dplyr	1.0.7	2021-06-18
easystats	0.4.3	2021-11-07
effectsize	0.5	2021-10-04
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	0.7.0.1	2021-11-17
parameters	0.15.0	2021-10-18
performance	0.8.0	2021-10-01
permute	0.9-5	2019-03-12
purrr	0.3.4	2020-04-17
readr	2.1.1	2021-11-30
report	0.4.0	2021-09-30
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