

Lab Assignment 9

Summer Heschong

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(1) Logistic Regression

```
#load packages  
library(here)
```

```
## here() starts at /Users/summerheshong/stats_spring25
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr      1.1.4      v readr      2.1.5  
## v forcats    1.0.0      v stringr   1.5.1  
## v ggplot2    3.5.1      v tibble    3.2.1  
## v lubridate  1.9.4      v tidyr     1.3.1  
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(naniar)  
library(DHARMA)
```

```
## This is DHARMA 0.4.7. For overview type '?DHARMA'. For recent changes, type news(package = 'DHARMA')
```

```
library(gtsummary)  
library(broom)
```

```
#load data  
lizards <- read.csv(here('Data/Raw/jrn_lizard.csv'))  
  
#filter dataset for side-blotched lizards  
lizards <- lizards %>%  
  filter(spp == 'UTST') %>%  
  na.omit
```

Step 1 - Define research question

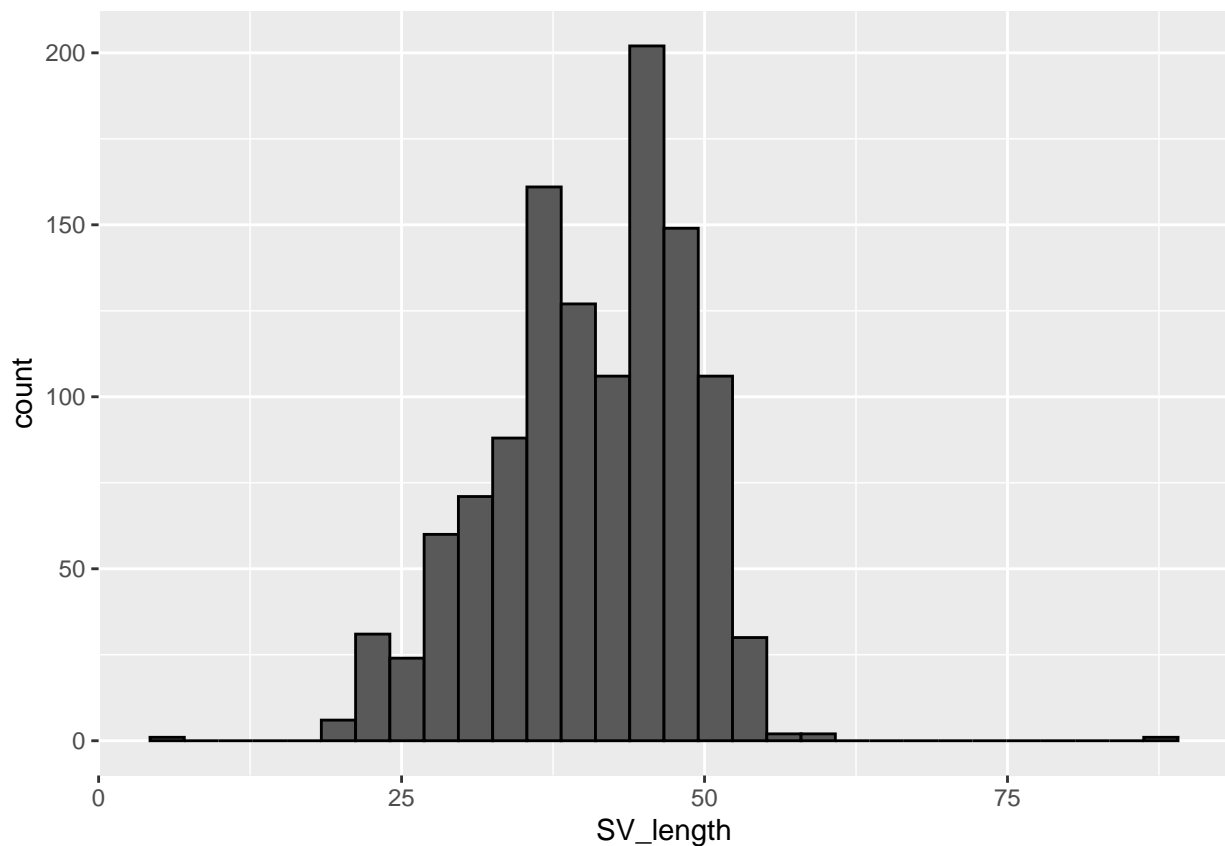
Do snout-to-vent length, sex, and vegetation zone at time of capture significantly predict if a lizard tail is recorded as whole?

Step 2 - Examine data

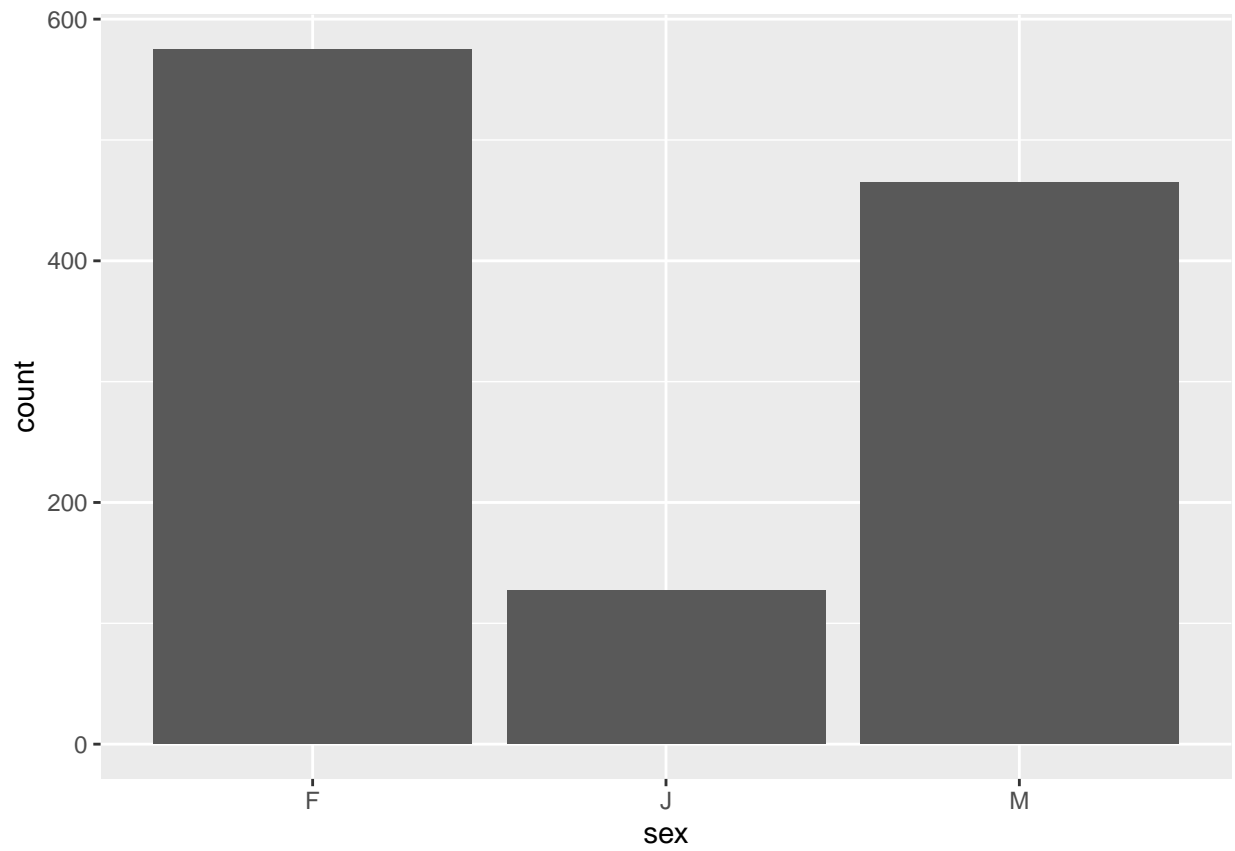
a. Display distributions or raw counts of data

```
#create histograms displaying data  
  
#snout-to-vent length  
ggplot(lizards, aes(x = SV_length)) +  
geom_histogram(color = 'black')
```

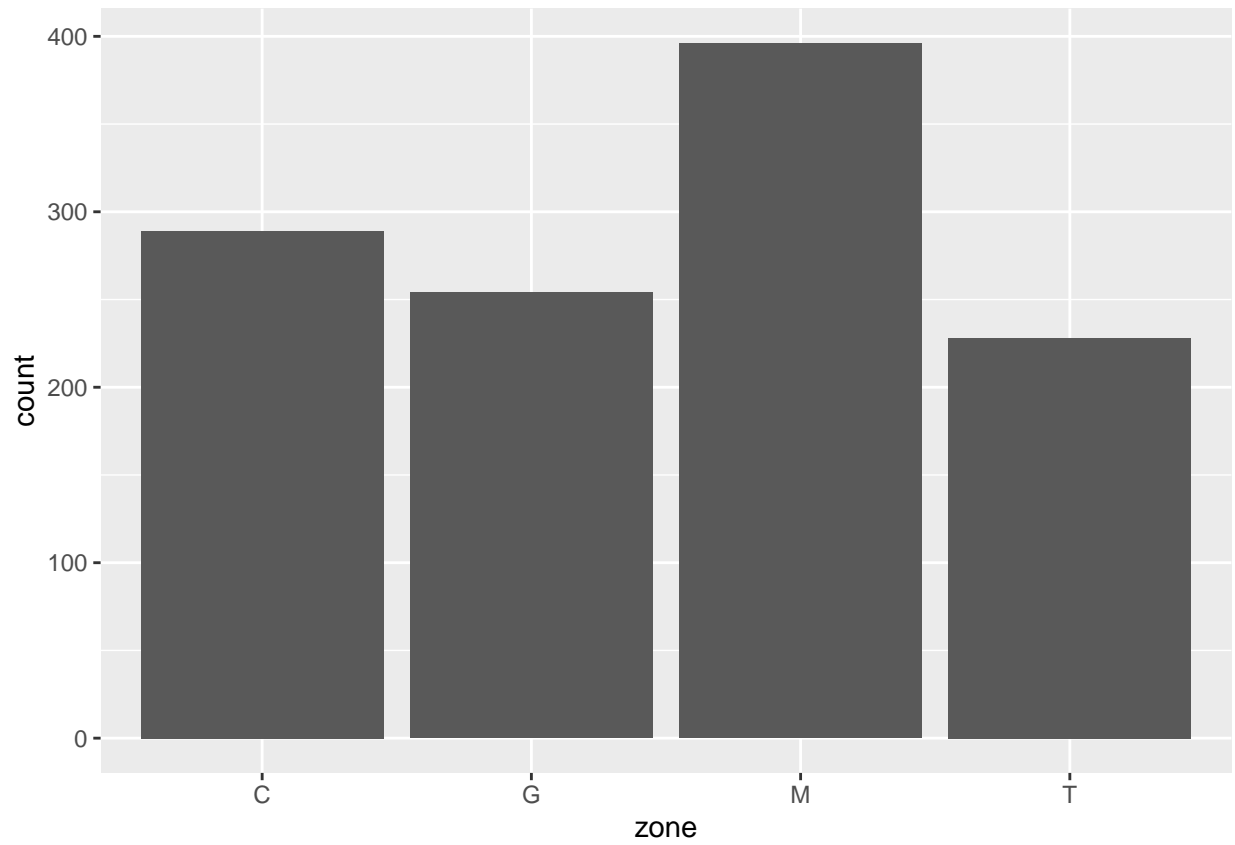
```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



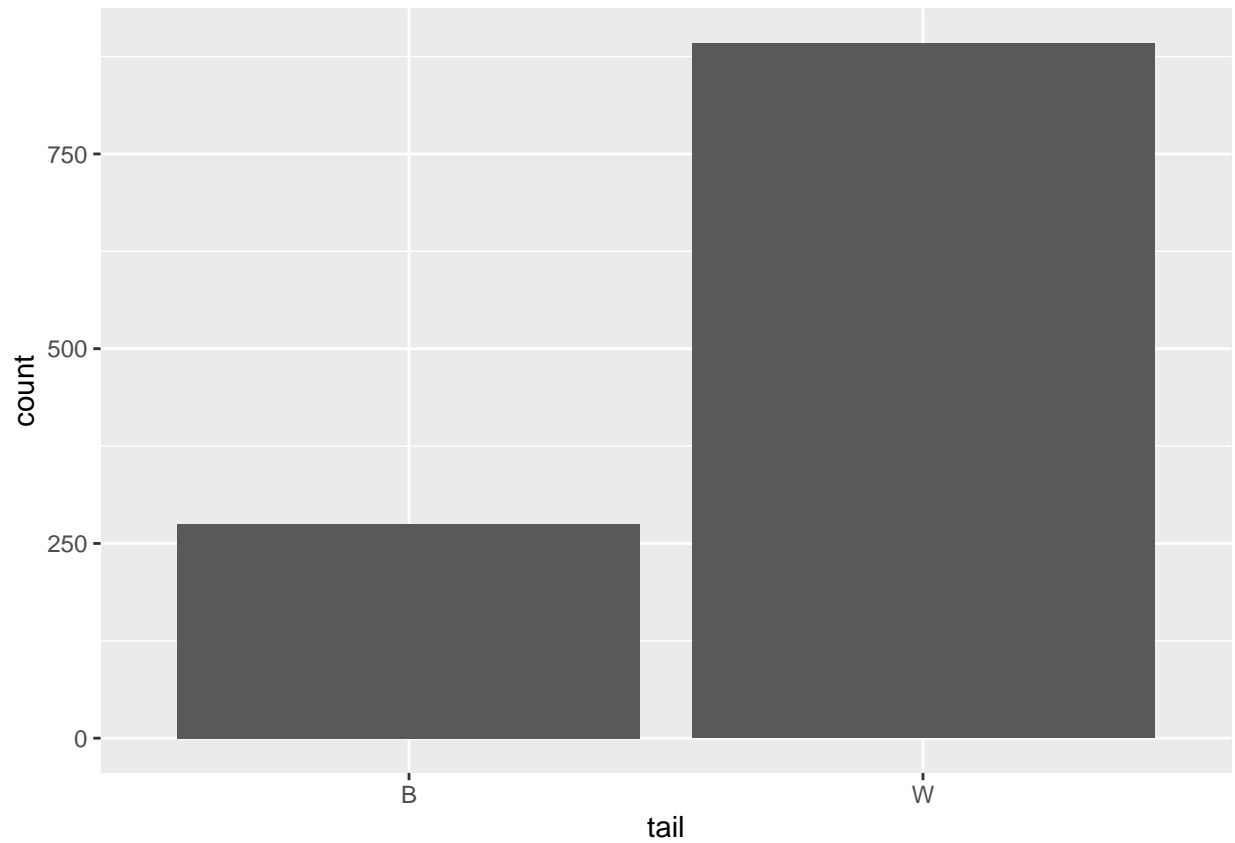
```
#sex  
ggplot(lizards, aes(x = sex)) +  
geom_bar ()
```



```
#vegetation zone  
ggplot(lizards, aes(x = zone)) +  
geom_bar ()
```

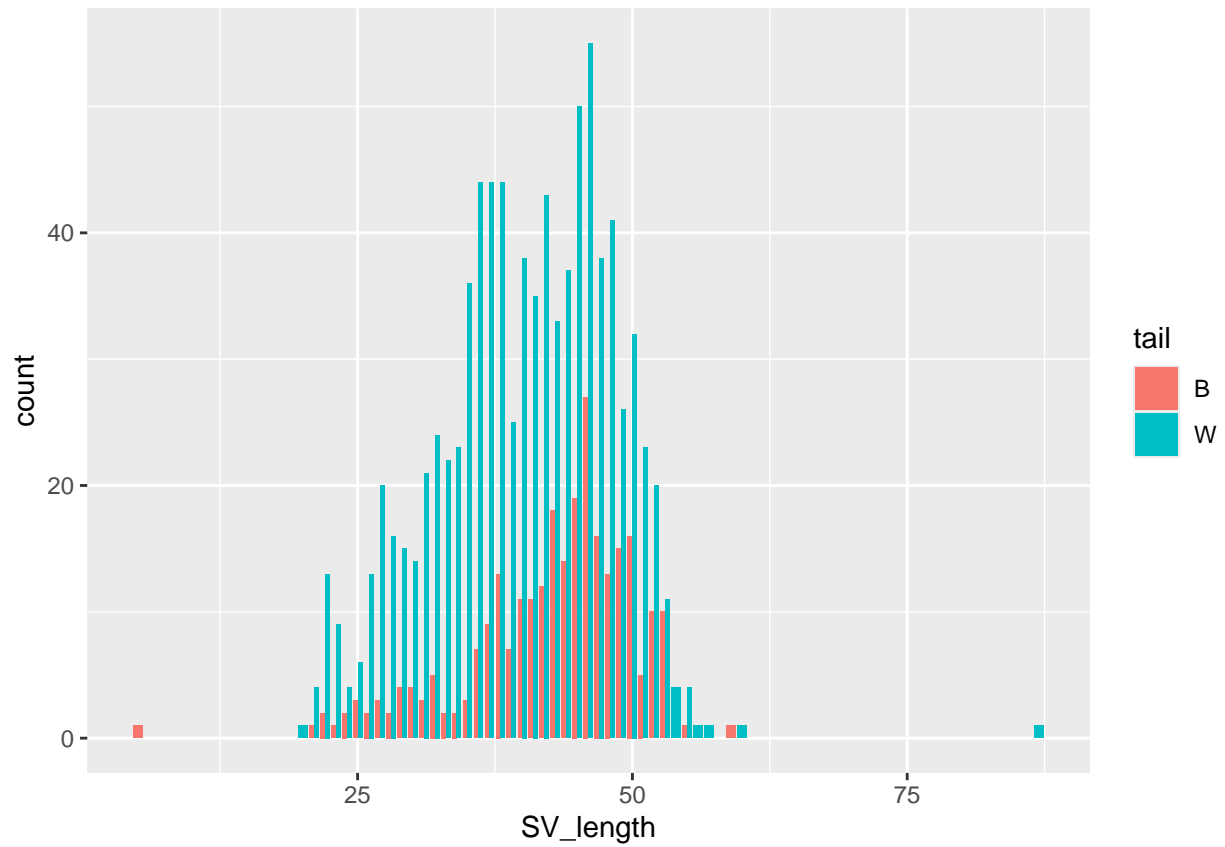


```
#lizard tail status  
ggplot(lizards, aes(x = tail)) +  
geom_bar ()
```

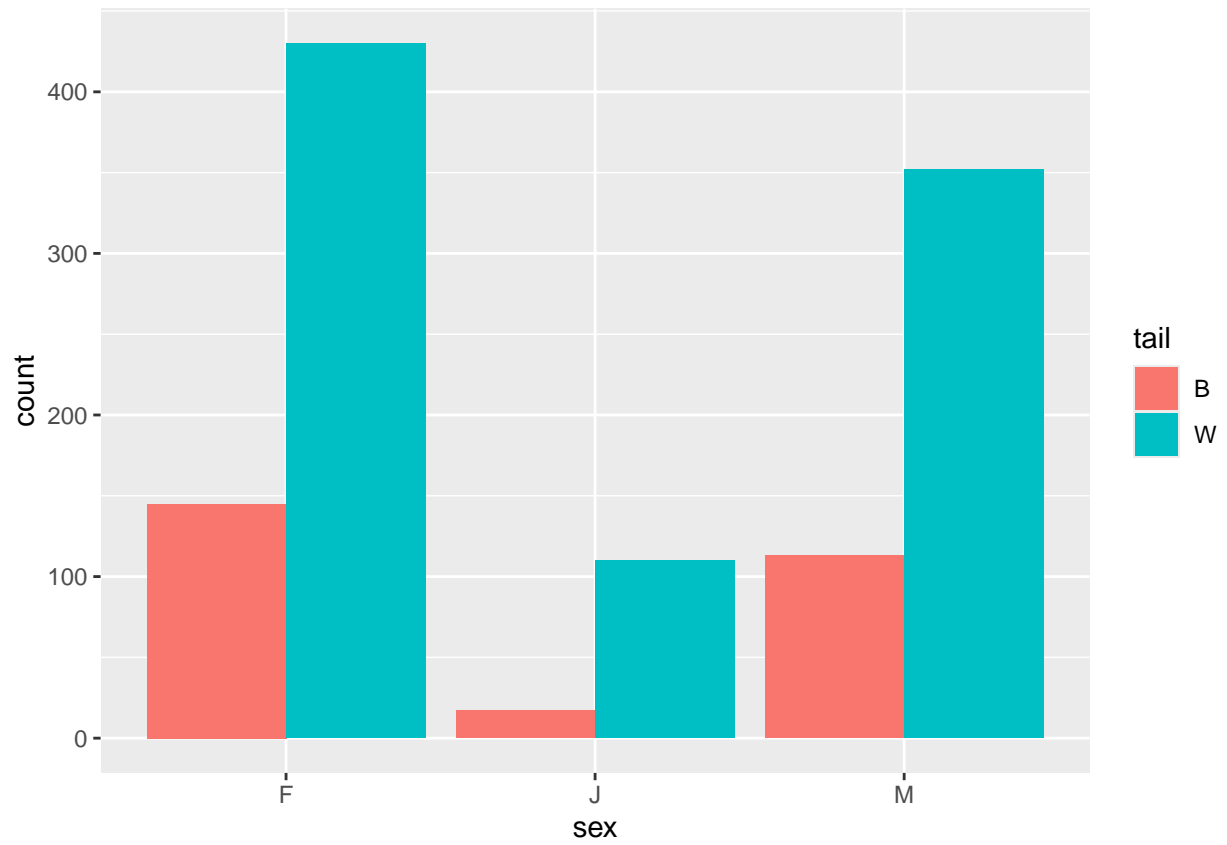


b. Display relationships between predictor and outcome variables

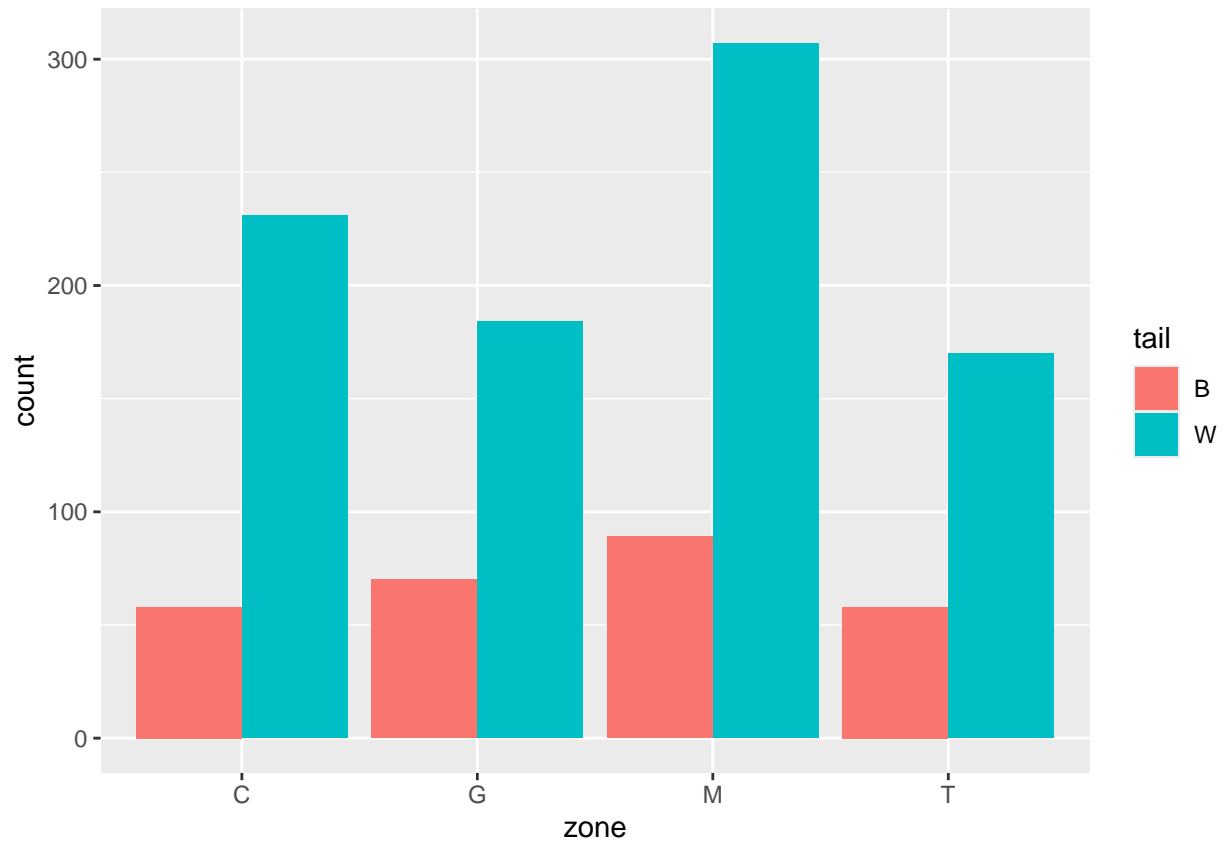
```
#create bar plots  
  
#snout-to-vent length  
ggplot(lizards, aes(x = SV_length, fill = tail)) +  
geom_bar (position = 'dodge')
```



```
#sex  
ggplot(lizards, aes(x = sex, fill = tail)) +  
geom_bar (position = 'dodge')
```



```
#zone  
ggplot(lizards, aes(x = zone, fill = tail)) +  
geom_bar (position = 'dodge')
```



Step 3 - Fit regression model

```
# change characters to factors
lizards$sex <- factor(lizards$sex, levels = c('J', 'F', 'M'))

lizards$zone <- factor(lizards$zone, levels = c('G', 'C', 'M', 'T'))

lizards$tail <- factor(lizards$tail, levels = c('B', 'W'))

#fit regression model
tail_mod <- glm(tail ~ SV_length + sex + zone,
  data = lizards,
  family = 'binomial')
```

Step 4 - Evaluate model diagnostics

```
#examine model output
summary(tail_mod)
```

```
##
## Call:
```

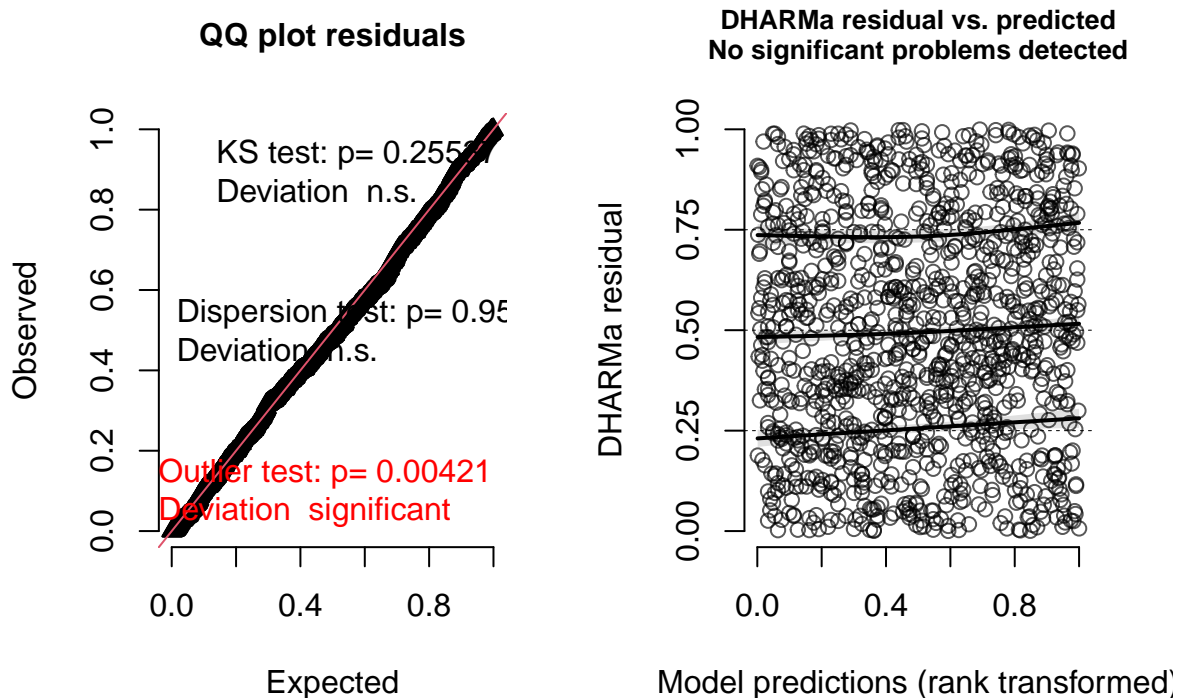


```
## glm(formula = tail ~ SV_length + sex + zone, family = "binomial",
##      data = lizards)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  2.85198    0.43630   6.537 6.29e-11 ***
## SV_length    -0.04075    0.01083  -3.762 0.000169 ***
## sexF         -0.33473    0.30241  -1.107 0.268356
## sexM         -0.20694    0.31869  -0.649 0.516114
## zoneC         0.50862    0.20800   2.445 0.014473 *
## zoneM         0.27847    0.18718   1.488 0.136834
## zoneT         0.25799    0.21165   1.219 0.222866
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1274.4  on 1166  degrees of freedom
## Residual deviance: 1245.7  on 1160  degrees of freedom
## AIC: 1259.7
##
## Number of Fisher Scoring iterations: 4
```

```
#examine modified residuals plot
simulateResiduals(tail_mod) %>% plot()
```

```
## DHARMA:testOutliers with type = binomial may have inflated Type I error rates for integer-valued dis
```

DHARMA residual



Step 5 - Interpret model and communicate results

a. Report results

The results of a logistic regression suggest that the log-odds of a lizard tail being whole are greater with shorter snout-to-vent length ($B = -0.04$, $p < 0.001$). There was no significant relationship between the log-odds of a lizard tail being whole and the sex of the lizard relative to juvenile lizards (male: $B = -0.21$, $p < 0.52$; female: $B = -0.33$, $p < 0.27$). There was a significant increase in the log-odds of a lizard tail being whole when the lizard is captured in creosotebush shrubland ($B = 0.51$, $p < 0.01$) relative to grama grassland. However, lizards caught in mesquite duneland and tarbush shrubland did not have significantly different log-odds of a lizard tail being whole relative to those caught in grama grassland (zoneM: $B = 0.28$, $p < 0.14$; zoneT: $B = 0.26$, $p < 0.22$).

b. Provide 2 predictive figures

```
# simulate snout-to-vent length data
SV_length_vector <- rep(seq(from = 0, to = 99), 12)

#simulate sex data
sex_vector <- c(rep('M', 400),
                rep('F', 400),
                rep('J', 400))
```

```

#simulate zone data
zone_vector <- c(rep('G', 300),
                 rep('C', 300),
                 rep('M', 300),
                 rep('T', 300))

#join data
pred_data <- data.frame(SV_length_vector, sex_vector, zone_vector)
colnames(pred_data) <- c('SV_length', 'sex', 'zone')

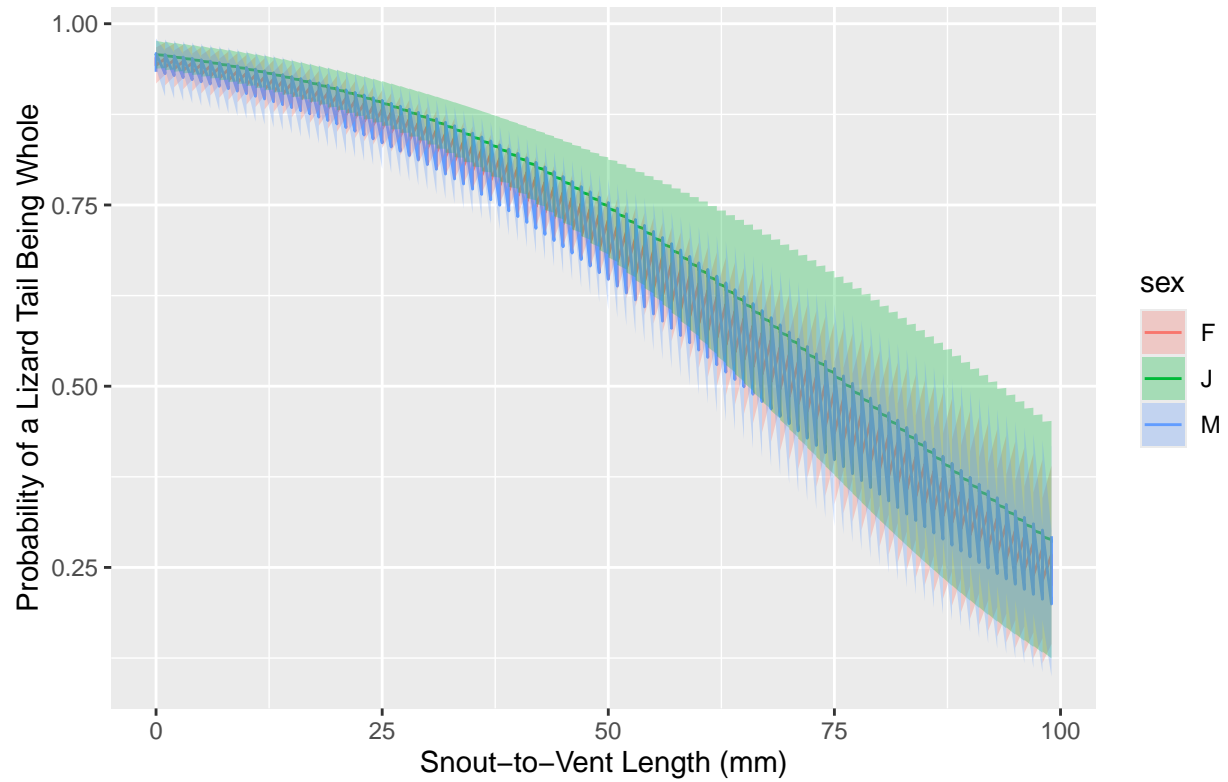
#Use original model to predict outcomes
prediction <- predict(tail_mod,
                     newdata = pred_data,
                     type = 'respons',
                     se.fit = TRUE)

#pull out predictions
prob_data <- data.frame(pred_data,
                       prediction$fit,
                       prediction$se.fit)

#rename columns
colnames(prob_data) <- c('SV_length', 'sex', 'zone', 'probability', 'se')

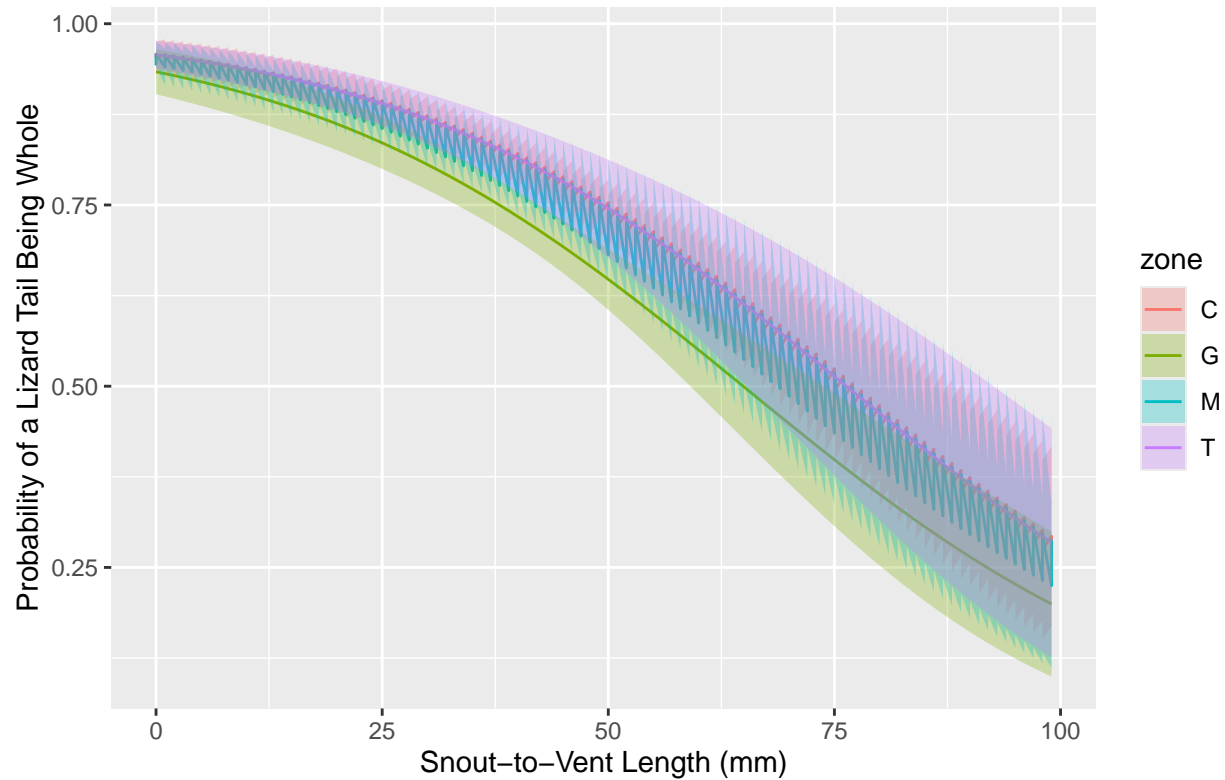
#Graph probabilities of lizard tail being whole
ggplot(prob_data, aes(x = SV_length, y = probability)) +
  geom_line(aes(color = sex)) +
  geom_ribbon(aes(ymin = probability - se,
                ymax = probability + se,
                fill = sex), alpha = 0.3) +
  labs(x = 'Snout-to-Vent Length (mm)',
       y = 'Probability of a Lizard Tail Being Whole',
       caption = 'Predictive figure showing change in probability of a lizard tail being whole \n as lizard ')

```



Predictive figure showing change in probability of a lizard tail being whole as lizard length increases, with lines colored by sex.

```
ggplot(prob_data, aes(x = SV_length, y = probability)) +
  geom_line(aes(color = zone)) +
  geom_ribbon(aes(ymin = probability - se,
                 ymax = probability + se,
                 fill = zone), alpha = 0.3) +
  labs(x = 'Snout-to-Vent Length (mm)',
       y = 'Probability of a Lizard Tail Being Whole',
       caption = 'Predictive figure showing change in probability of a lizard tail being whole \n as lizard
```



(2) Poisson Regression

Step 1 - Define research question

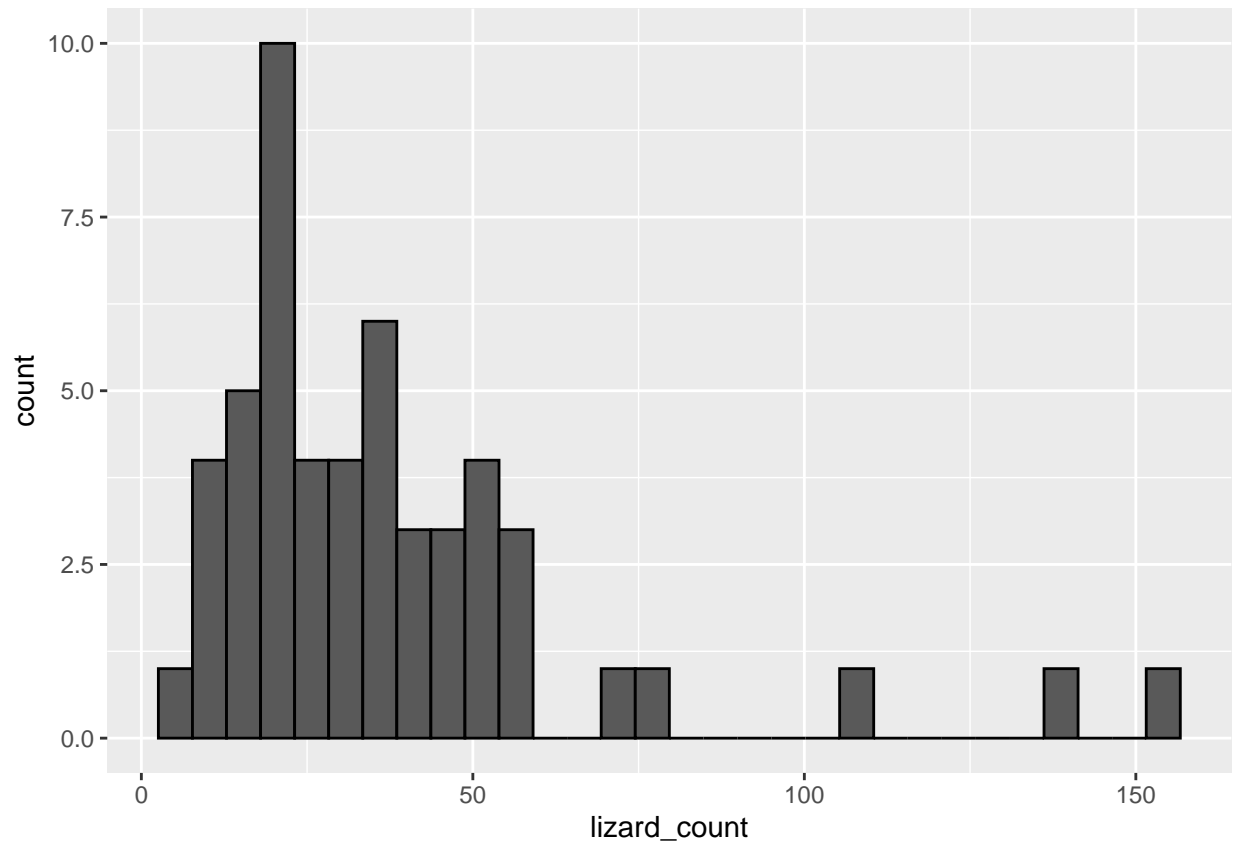
Do season and plant species percent cover significantly predict lizard counts?

```
lizard_counts <- read.csv(here('Data/Raw/jrn_lizard_npp.csv'))
```

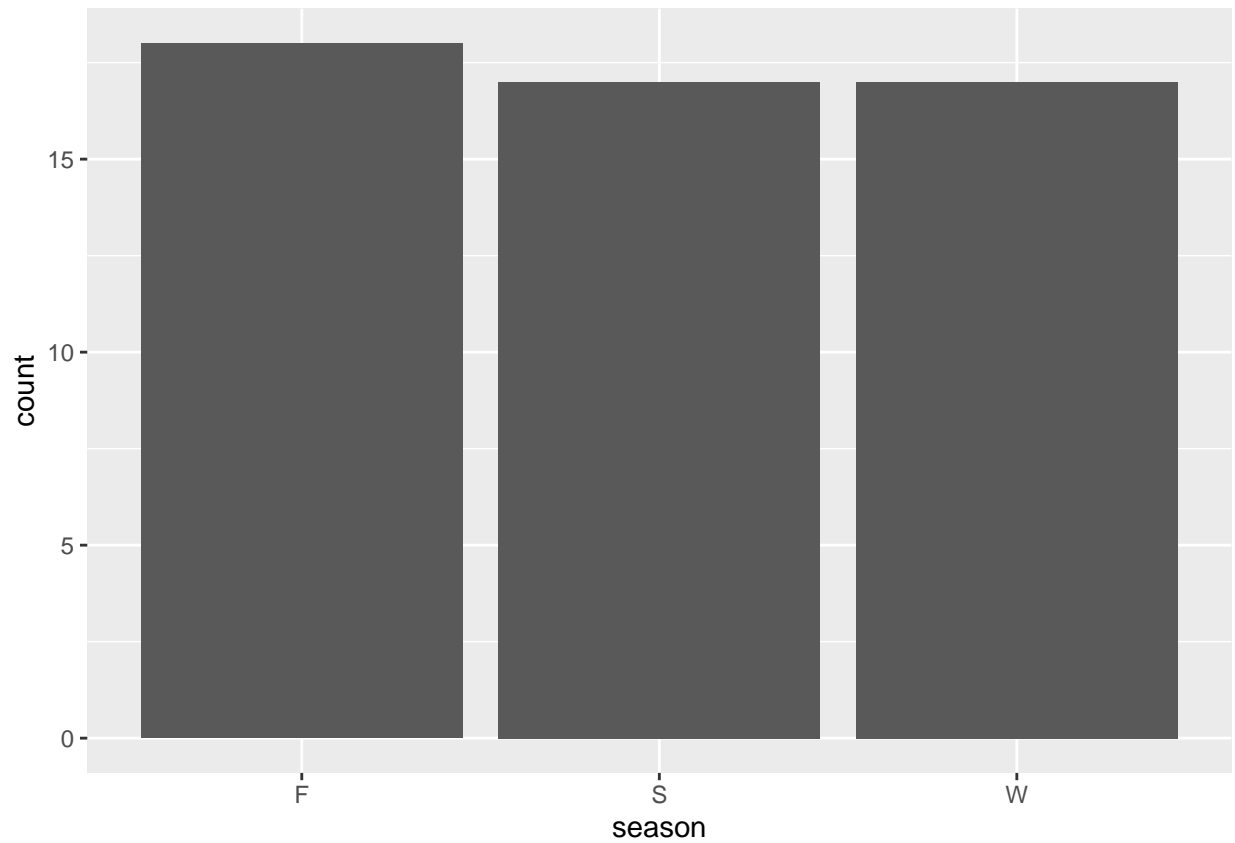
Step 2 - Examine data and possible correlations

```
#create histograms
ggplot(lizard_counts, aes(x = lizard_count)) +
  geom_histogram(color = 'black')
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

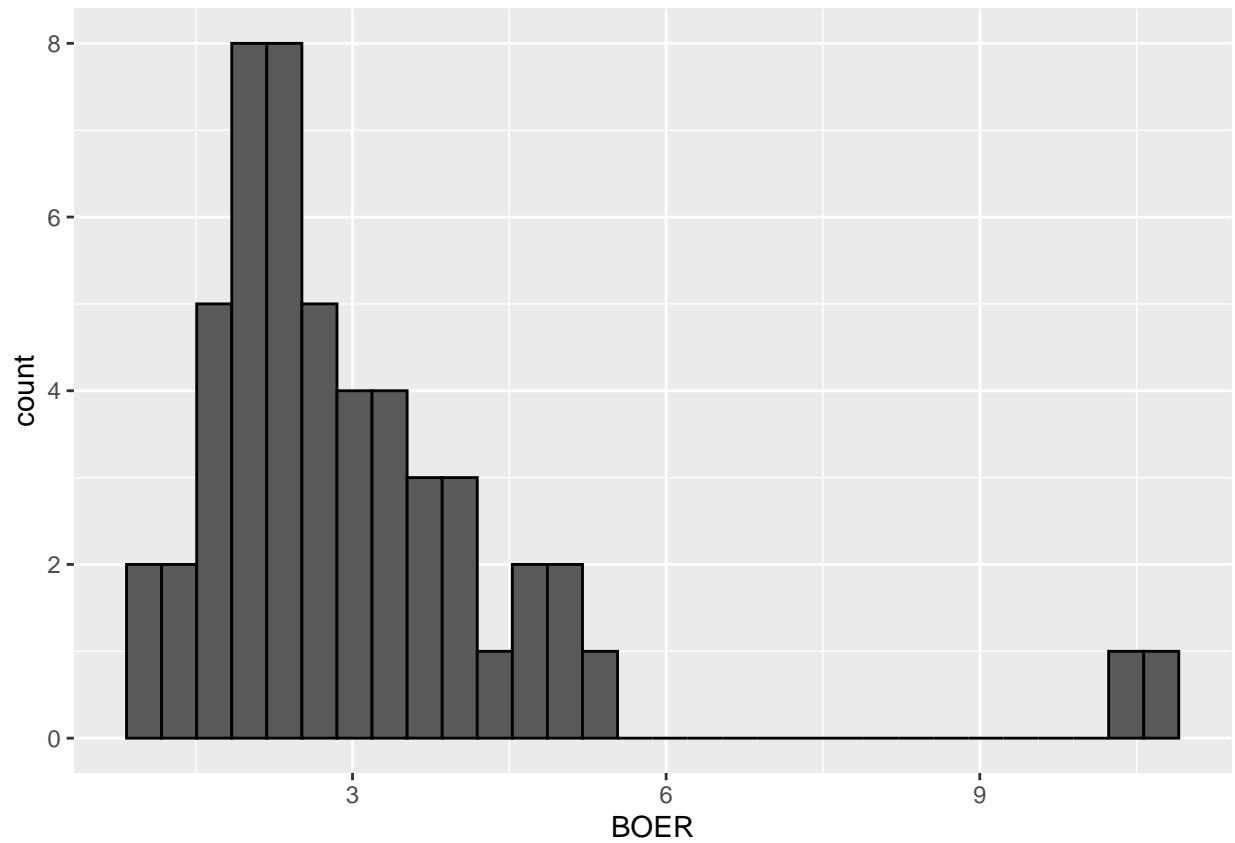


```
ggplot(lizard_counts, aes(x = season)) +  
geom_bar()
```



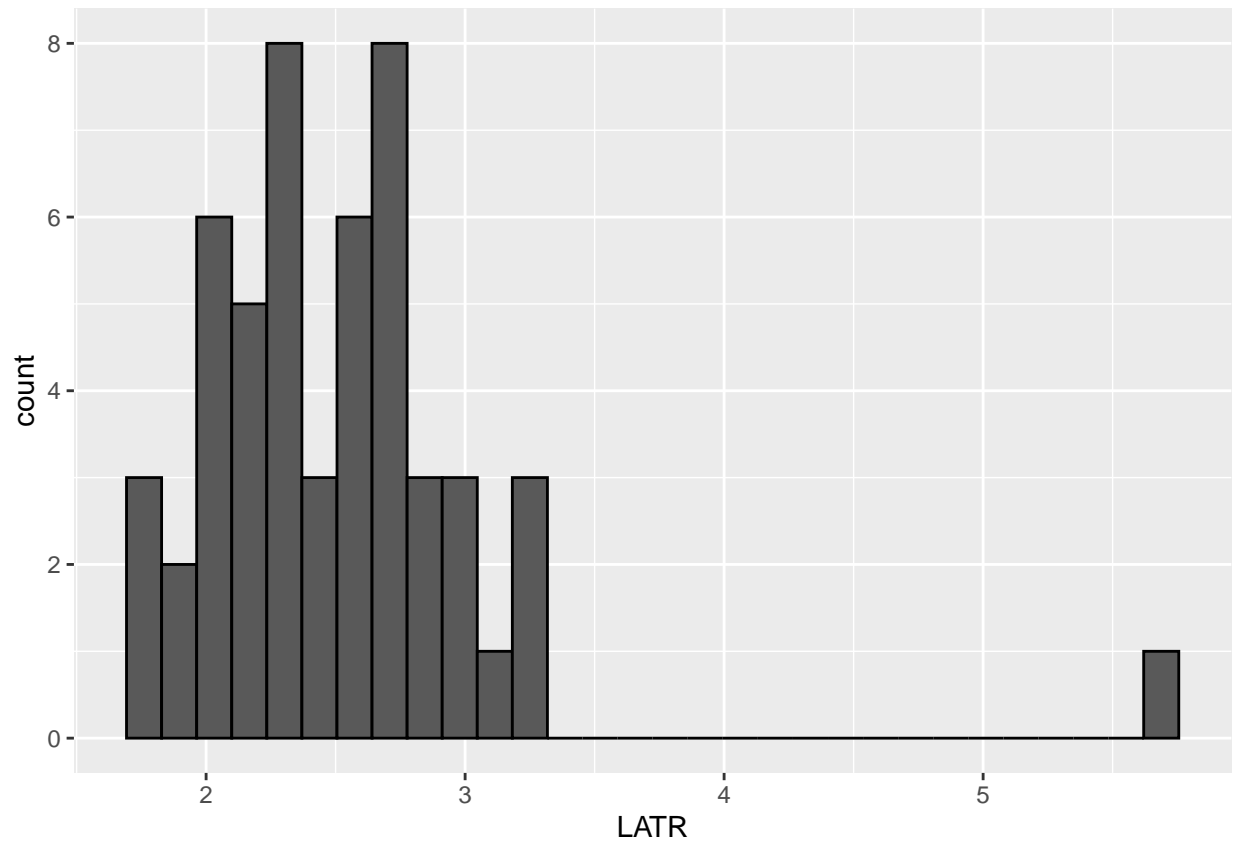
```
ggplot(lizard_counts, aes(x = BOER)) +  
geom_histogram(color = 'black')
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



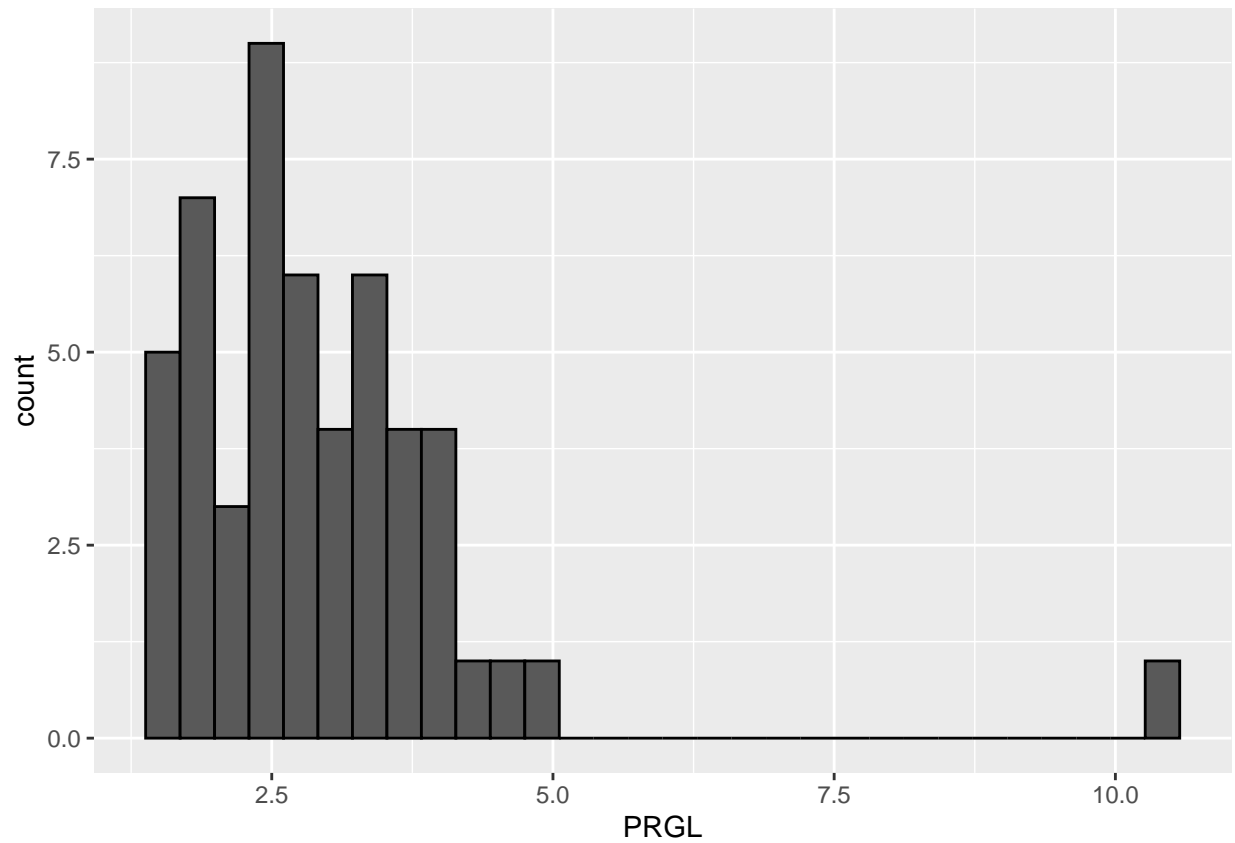
```
ggplot(lizard_counts, aes(x = LATR)) +  
geom_histogram(color = 'black')
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

```
ggplot(lizard_counts, aes(x = PRGL)) +  
  geom_histogram(color = 'black')
```

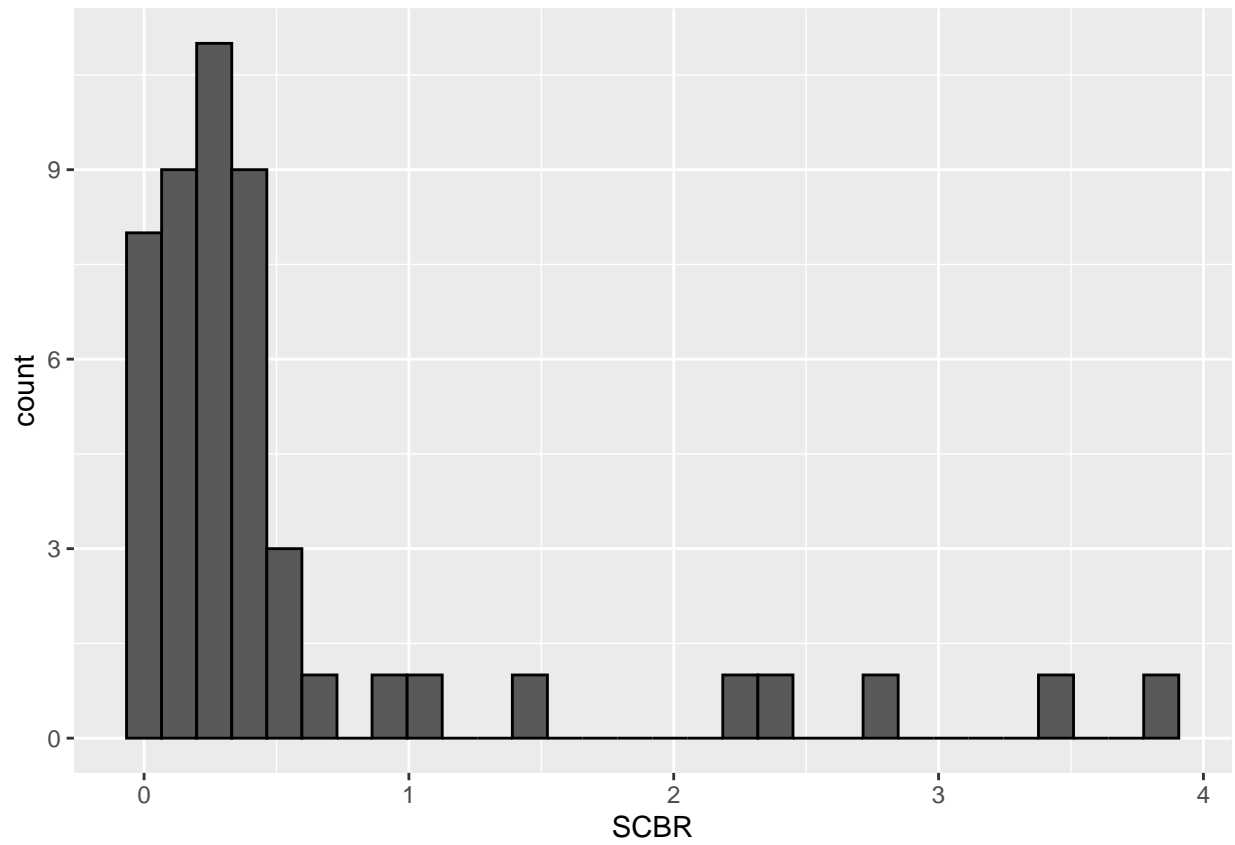
```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



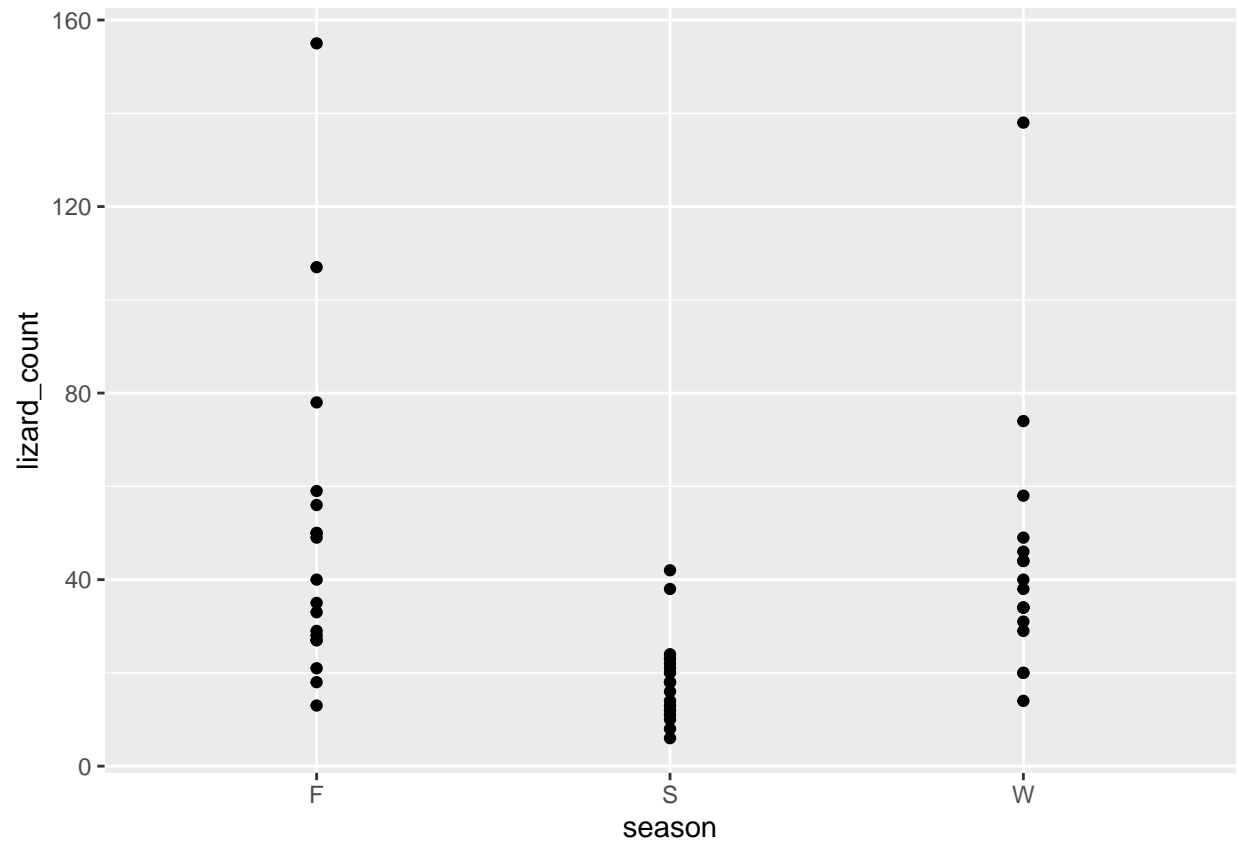
```
ggplot(lizard_counts, aes(x = SCBR)) +  
geom_histogram(color = 'black')
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

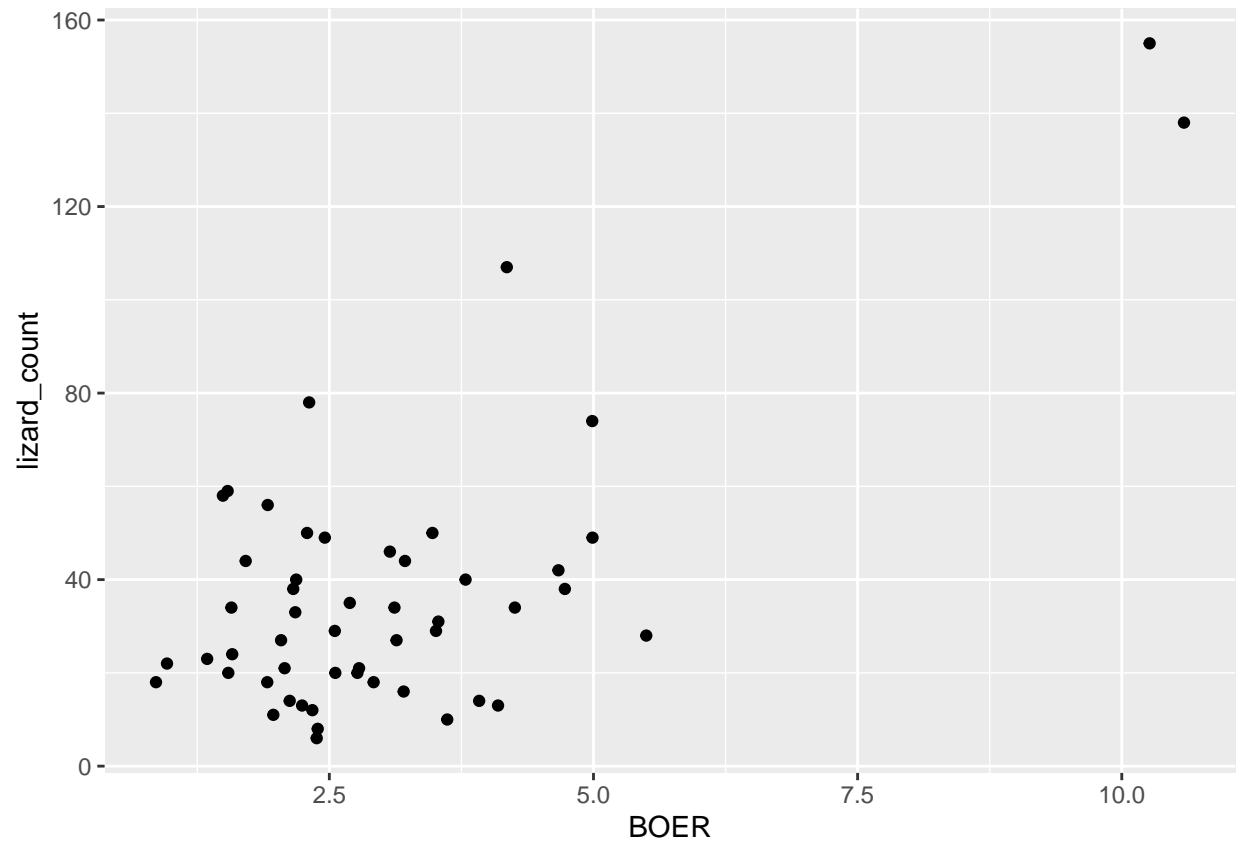
```
## Warning: Removed 3 rows containing non-finite outside the scale range  
## ('stat_bin()').
```



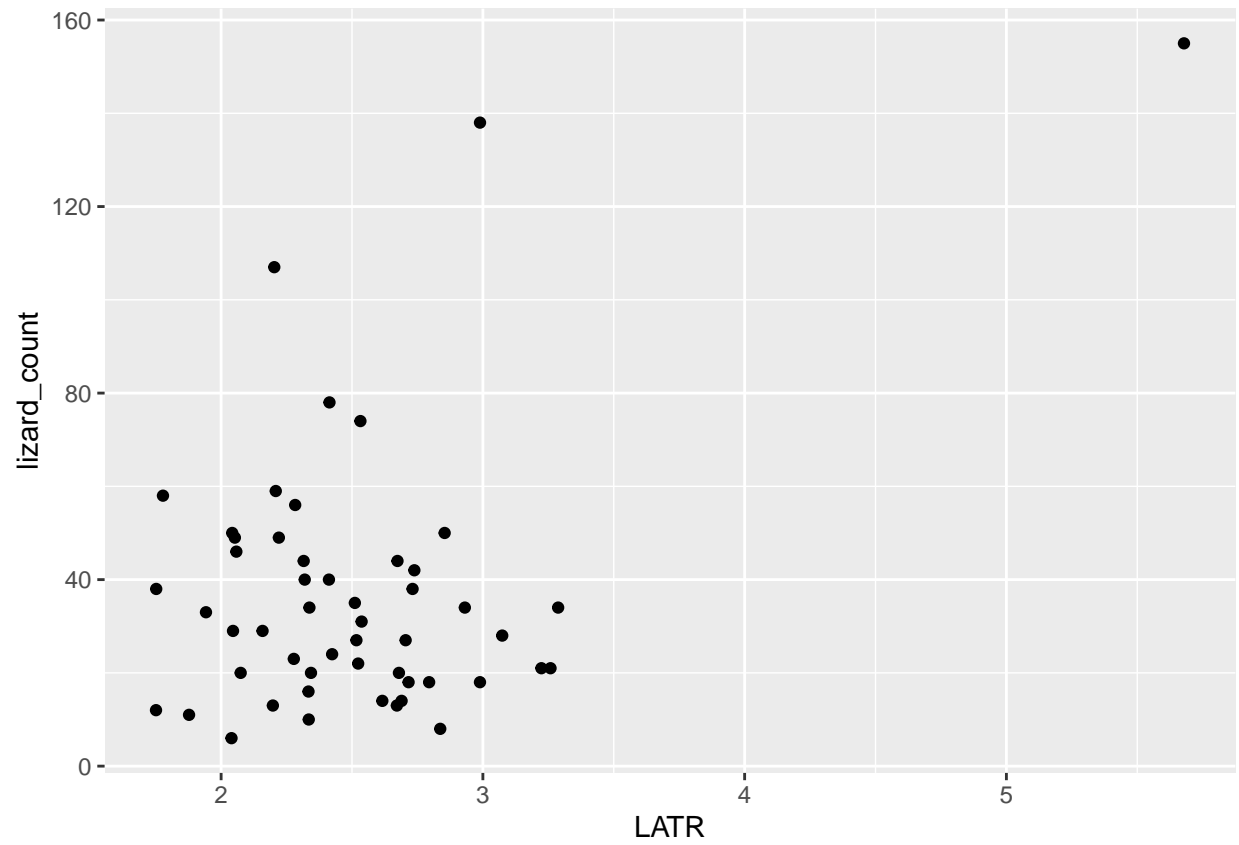
```
#Create scatter plots  
ggplot(lizard_counts, aes(x = season, y = lizard_count)) +  
geom_point(color = 'black')
```



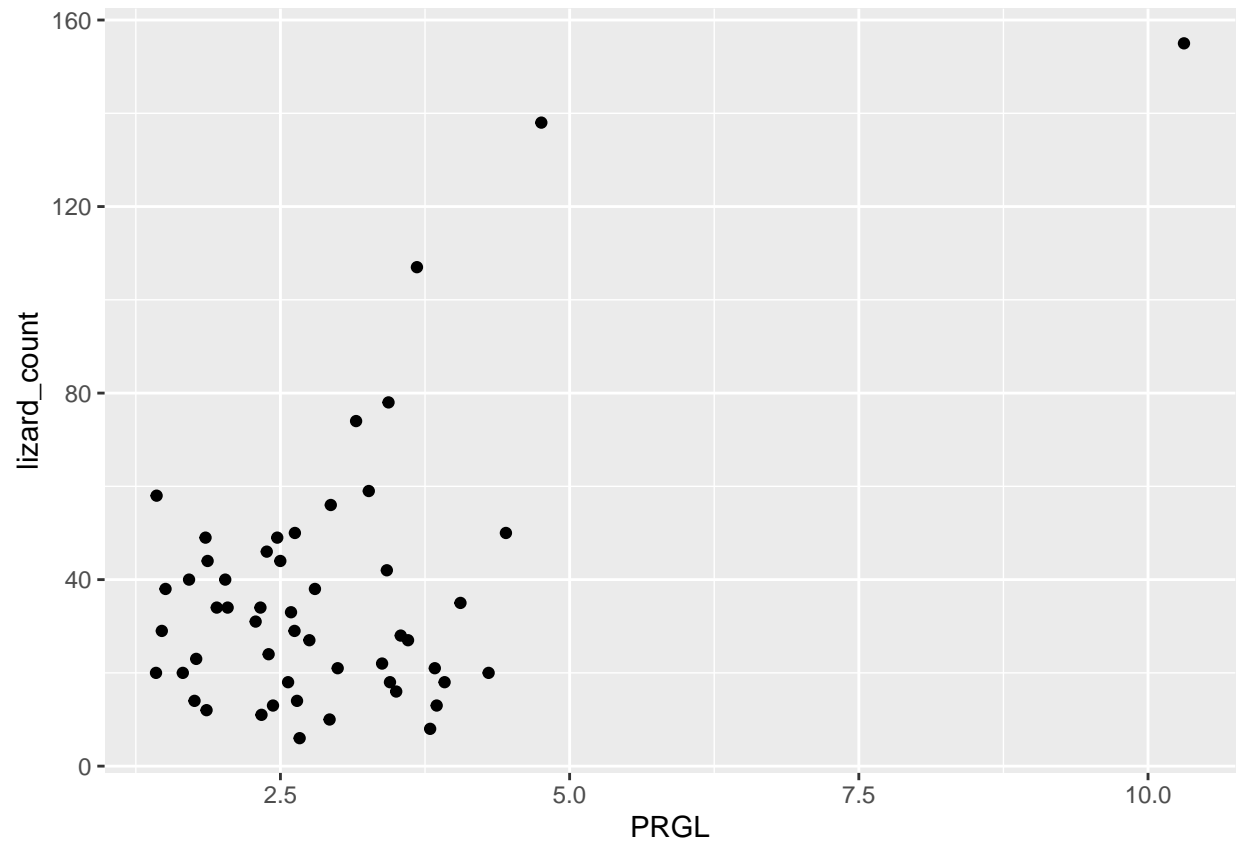
```
ggplot(lizard_counts, aes(x = BOER, y = lizard_count)) +  
geom_point(color = 'black')
```



```
ggplot(lizard_counts, aes(x = LATR, y = lizard_count)) +  
geom_point(color = 'black')
```



```
ggplot(lizard_counts, aes(x = PRGL, y = lizard_count)) +  
geom_point(color = 'black')
```



```
ggplot(lizard_counts, aes(x = SCBR, y = lizard_count)) +  
geom_point(color = 'black')
```

```
## Warning: Removed 3 rows containing missing values or values outside the scale range  
## ('geom_point()').
```

