

Univariate Modeling

```
TreeData<- read.csv("https://raw.githubusercontent.com/dmccglinn/quant_methods/gh-pages/data/tree  
data.csv", header = T)
```

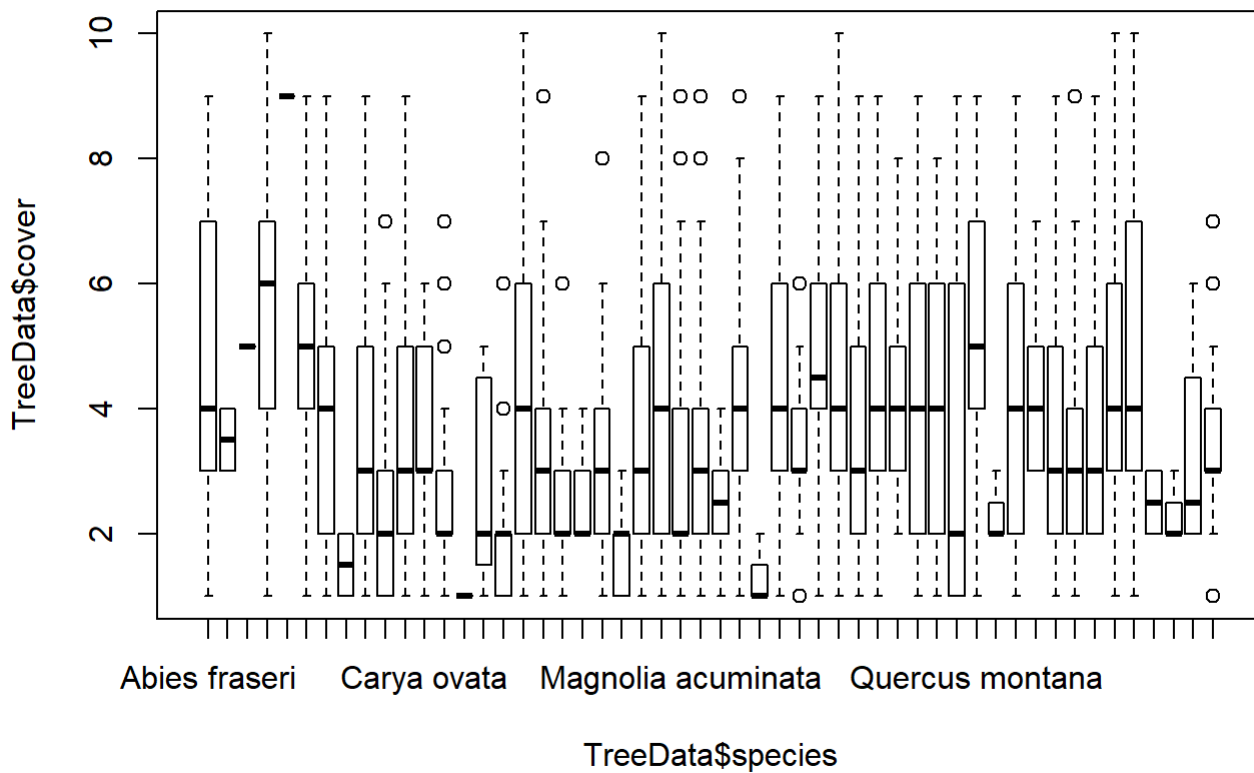
```
library("car")
```

```
## Warning: package 'car' was built under R version 3.5.2
```

```
## Loading required package: carData
```

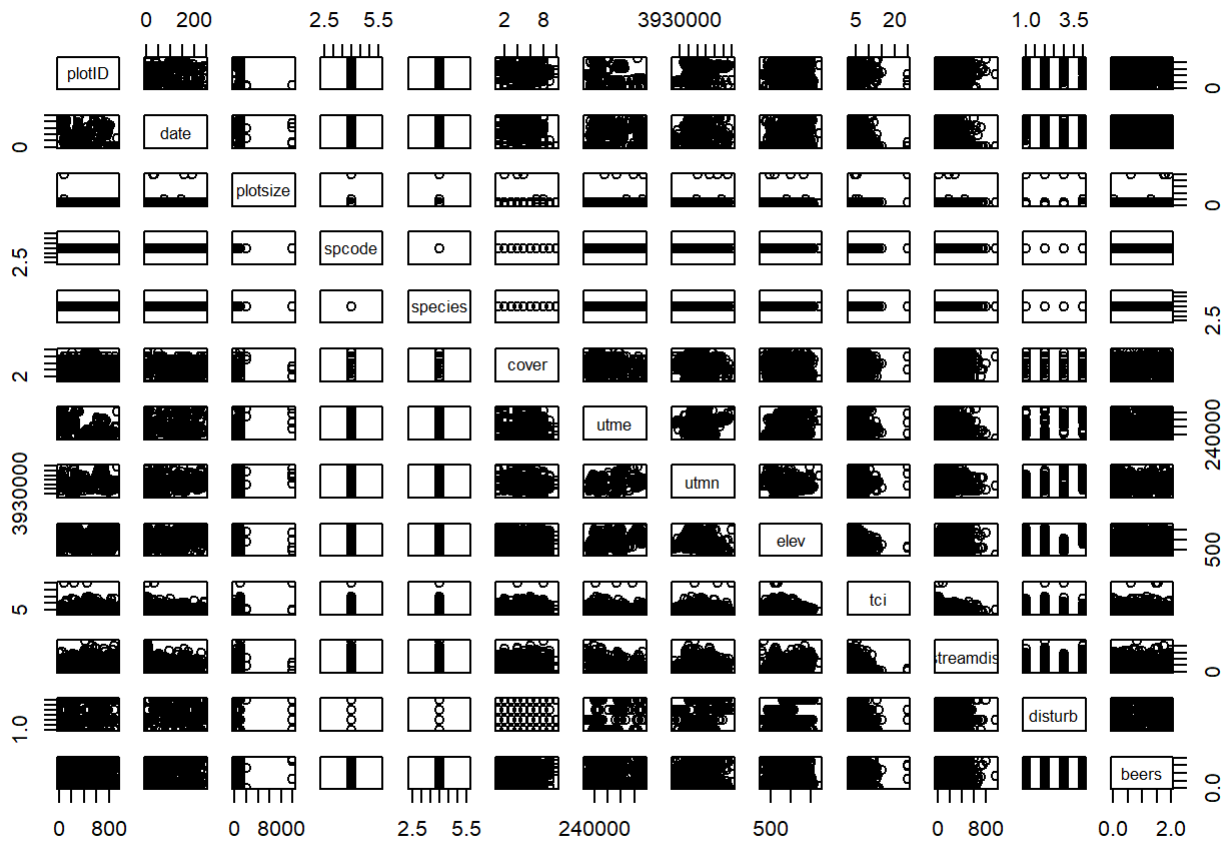
```
## Warning: package 'carData' was built under R version 3.5.2
```

```
plot(TreeData$cover~TreeData$species)
```

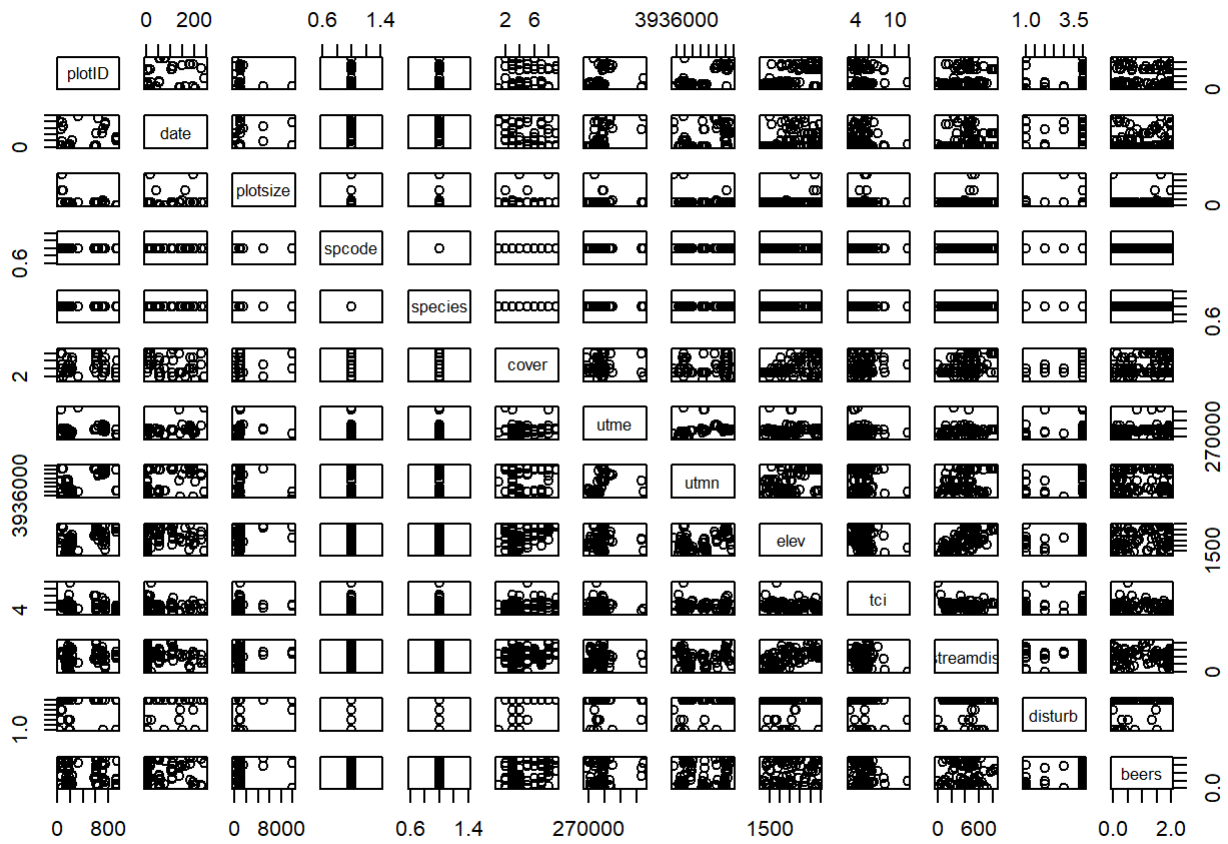


```
Acer<-TreeData[TreeData$spcode=="ACERRUB",]  
Abies<-TreeData[TreeData$spcode=="ABIEFRA",]
```

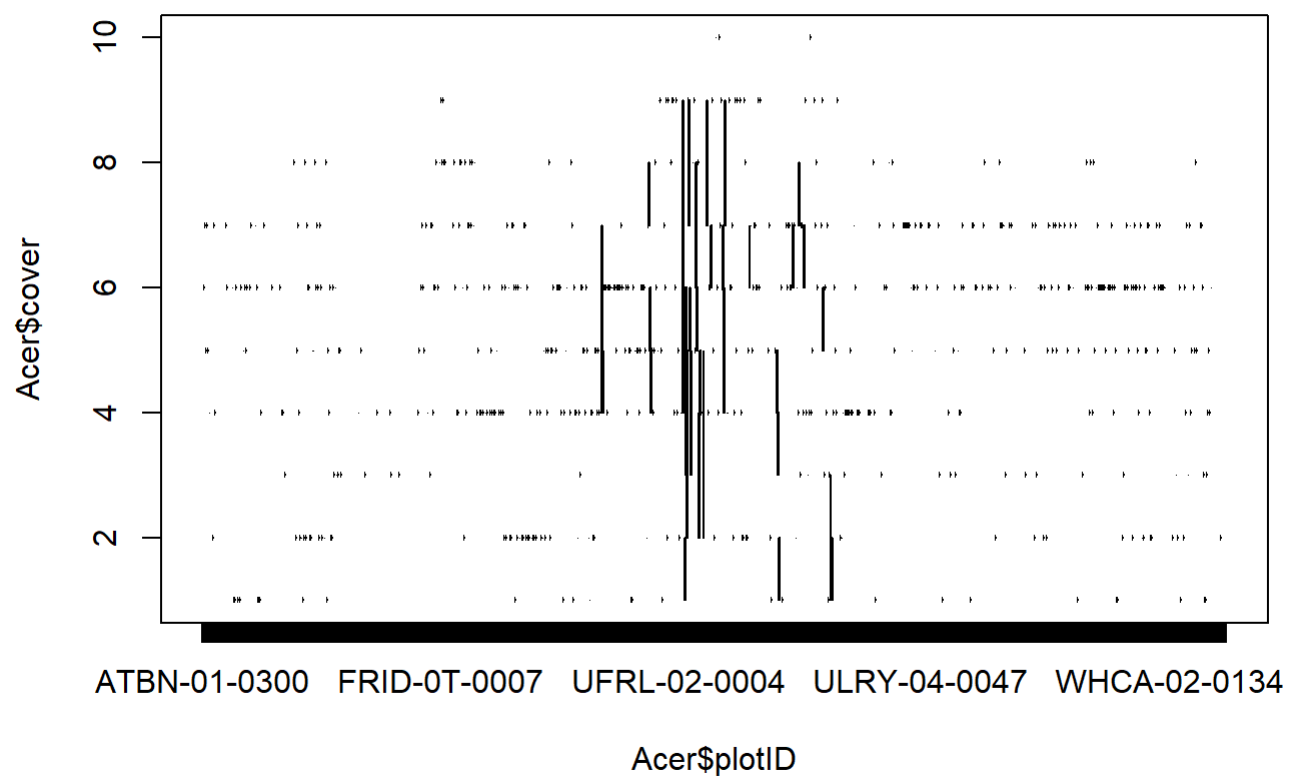
```
pairs(Acer)
```



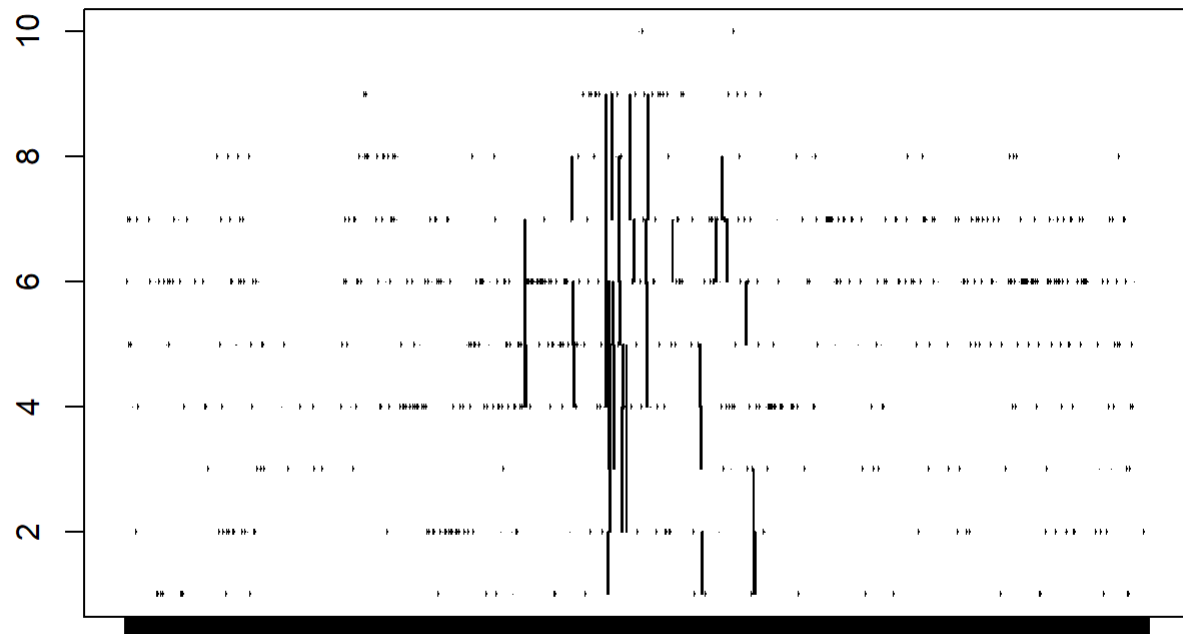
```
pairs(Abies)
```



```
plot(Acer$cover~Acer$plotID)
```

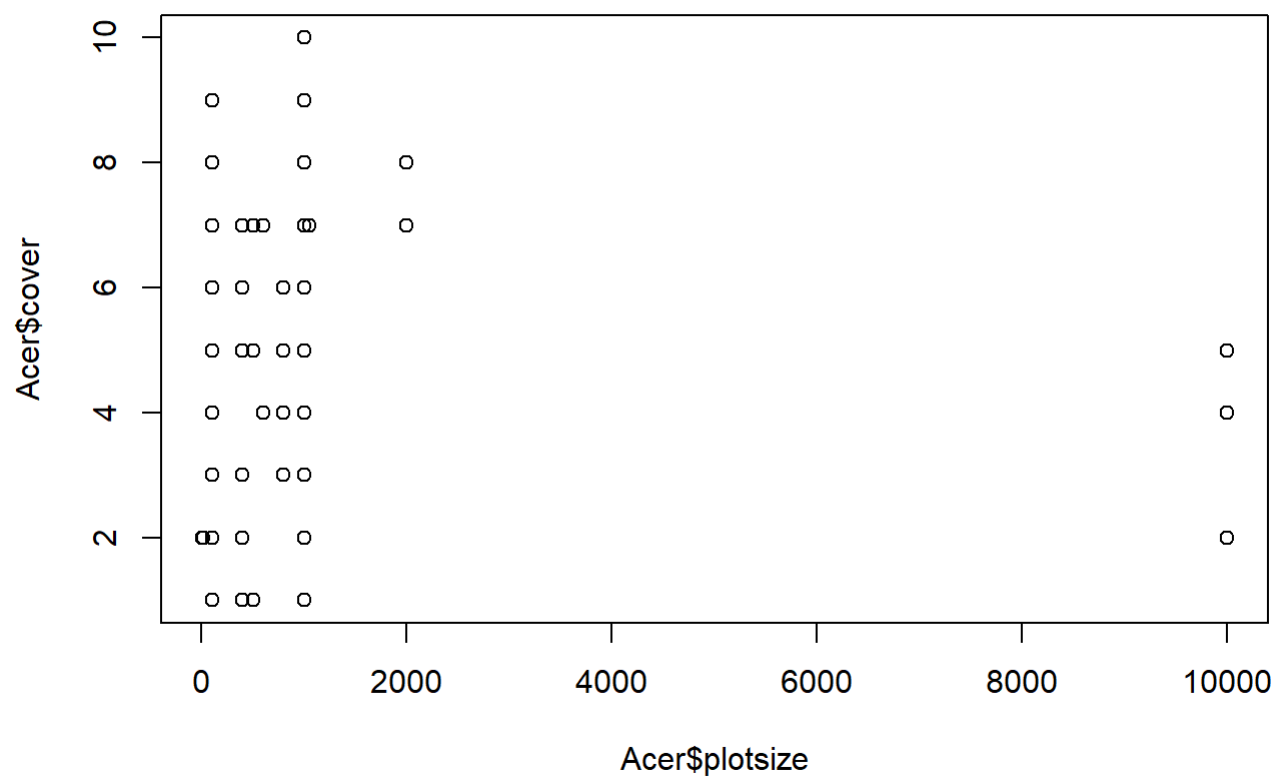


```
boxplot(Acer$cover~Acer$plotID)
```

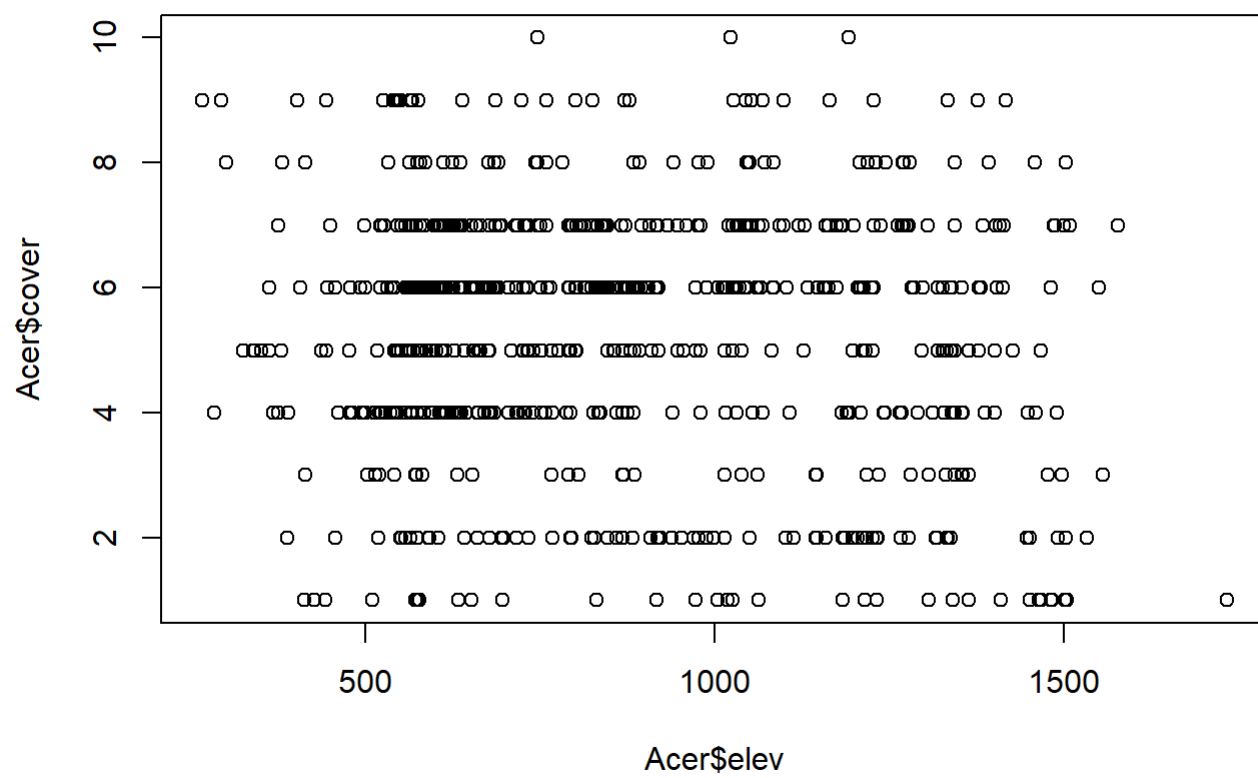


ATBN-01-0300 FRID-0T-0007 UFRL-02-0004 ULRY-04-0047 WHCA-02-0134

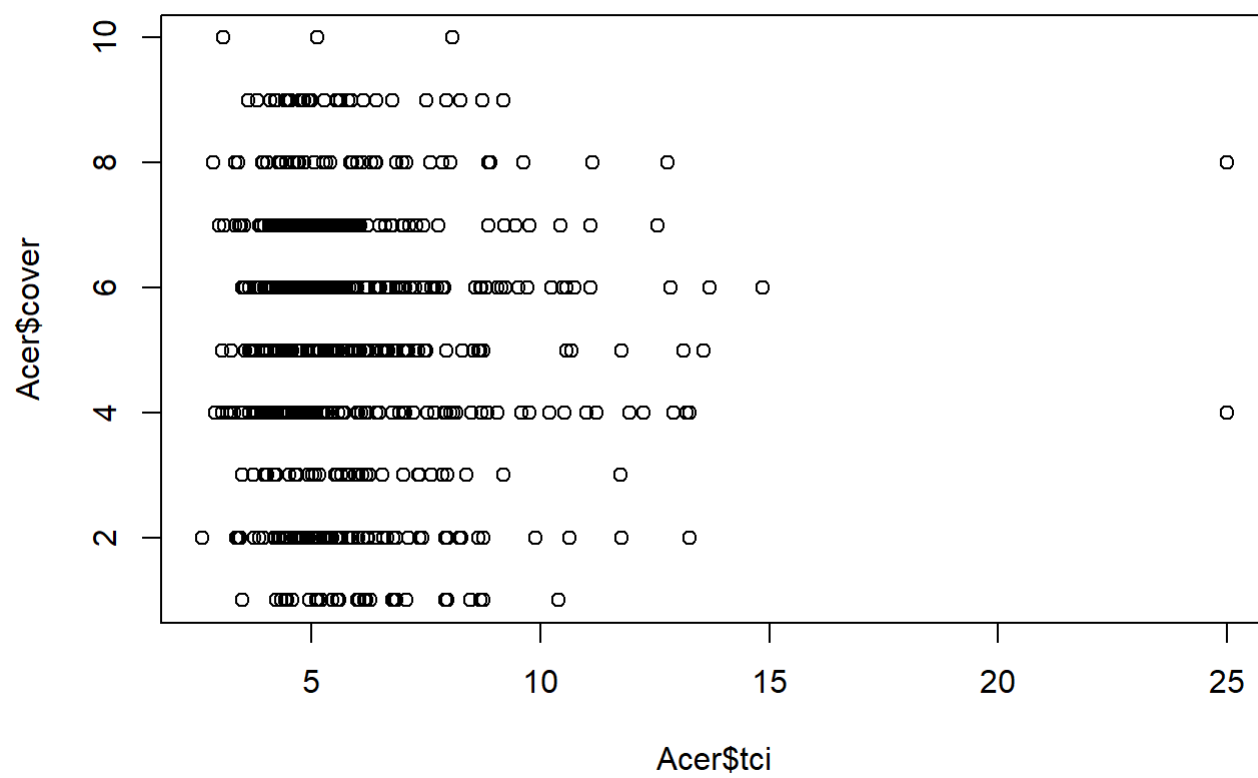
```
plot(Acer$cover~Acer$plotsize)
```



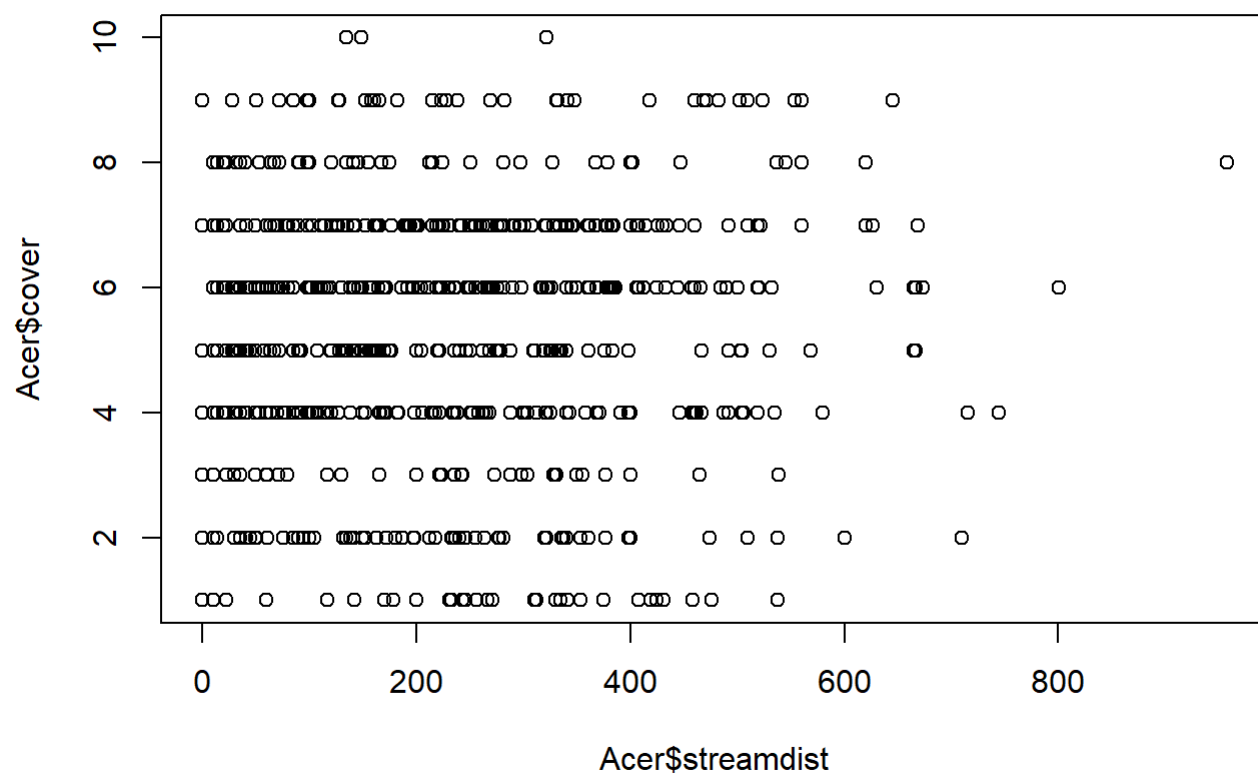
```
plot(Acer$cover~Acer$elev)
```



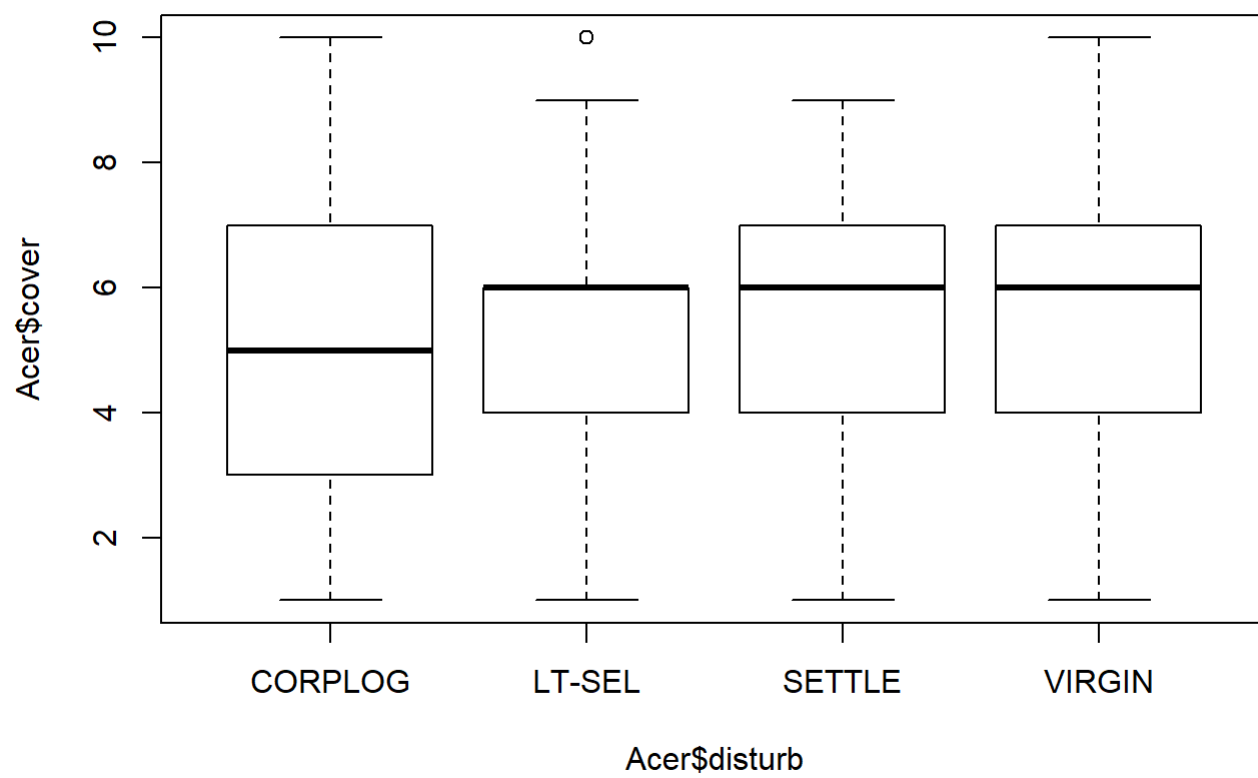
```
plot(Acer$cover~Acer$elev)
```



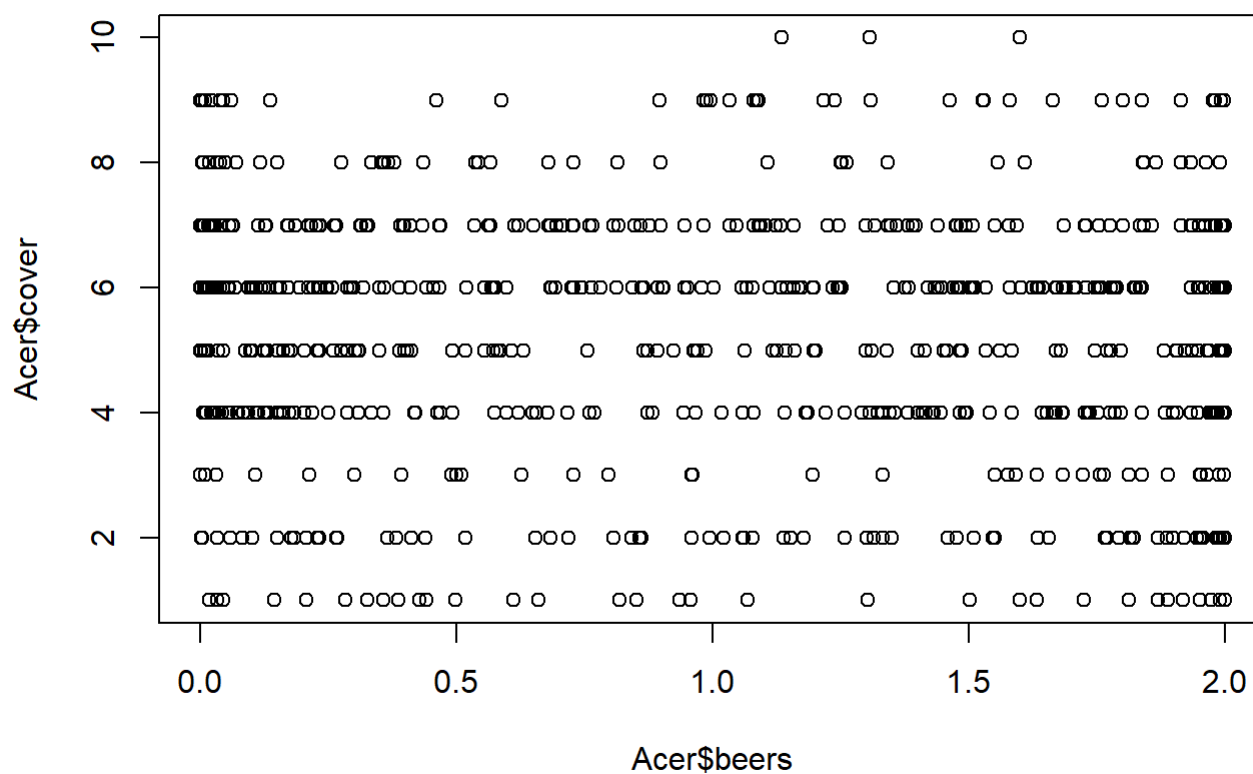
```
plot(Acer$cover~Acer$streamdist)
```

```
plot(Acer$cover~Acer$disturb)
```



```
plot(Acer$cover~Acer$beers)
```



```
# take a look at plotID, elevation, tci, streamdist -- Lesser important beers, disturb, plotsize
```

After looking at the plots of each variable as it relates to cover I thought the most important variables to consider were plotID, elevation, tci, and streamdist.

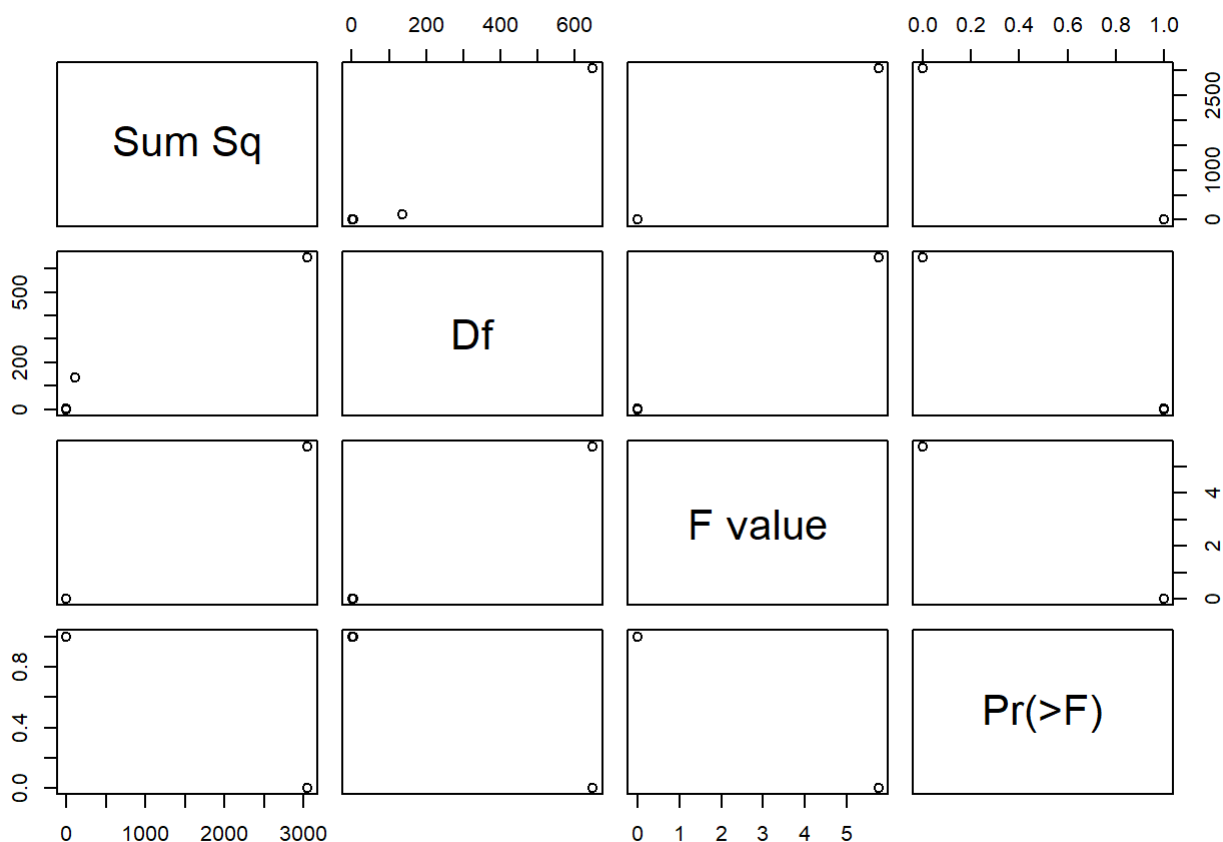
```
AcerFull<-Anova(aov(cover~as.factor(plotID)+elev+tci+streamdist+beers+disturb+plotsize, data = Acer))
```

```
## Note: model has aliased coefficients
##      sums of squares computed by model comparison
```

```
summary(AcerFull)
```

##	Sum Sq	Df	F value	Pr(>F)
## Min. :	0.0	Min. : 0.0	Min. :0.0000	Min. :0.0000
## 1st Qu.:	0.0	1st Qu.: 1.0	1st Qu.:0.0000	1st Qu.:1.0000
## Median :	0.0	Median : 1.0	Median :0.0000	Median :1.0000
## Mean :	450.8	Mean : 99.0	Mean :0.9576	Mean :0.8333
## 3rd Qu.:	56.0	3rd Qu.: 36.5	3rd Qu.:0.0000	3rd Qu.:1.0000
## Max. :	3043.8	Max. :648.0	Max. :5.7458	Max. :1.0000
## NA's :	1		NA's :2	NA's :2

```
plot(AcerFull)
```



```
AcerAOV<- aov(cover~as.factor(plotID)+elev+tci+streamdist+beers+disturb+plotsize, data = Acer)
AcerLM<- lm(cover~plotID+elev+tci+streamdist+beers+disturb+plotsize, data = Acer)
summary(AcerAOV)
```

```
##              Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(plotID) 649   3197    4.927   6.026 <2e-16 ***
## elev              1      0    0.000   0.000     1
## tci               1      0    0.000   0.000     1
## streamdist        1      0    0.000   0.000     1
## beers             1      0    0.000   0.000     1
## disturb           3      0    0.000   0.000     1
## Residuals        137    112    0.818
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM)$adj.r.squared
```

```
## [1] 0.804105
```

```
Acer1<-aov(cover~tci+elev+streamdist+beers+disturb+plotsize, data = Acer)
AcerLM1<-lm(cover~tci+elev+streamdist+beers+disturb+plotsize, data = Acer)
summary(Acer1)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## tci           1   12.1    12.05     2.997 0.08379 .
## elev          1   34.3    34.32     8.538 0.00358 **
## streamdist    1   30.2    30.17     7.504 0.00629 **
## beers         1   40.8    40.84    10.159 0.00149 **
## disturb       3    8.4     2.82     0.700 0.55206
## plotsize      1   27.7    27.73     6.898 0.00880 **
## Residuals    785 3155.8     4.02
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM1)
```

```
##
## Call:
## lm(formula = cover ~ tci + elev + streamdist + beers + disturb +
##      plotsize, data = Acer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7731 -1.2704  0.3525  1.4185  5.2049
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   6.3754823   0.4477404   14.239 < 2e-16 ***
## tci           -0.0438862   0.0334663   -1.311  0.19012
## elev          -0.0008589   0.0002970   -2.892  0.00394 **
## streamdist     0.0012370   0.0004590    2.695  0.00720 **
## beers         -0.2988236   0.1044015   -2.862  0.00432 **
## disturbLT-SEL  0.1457197   0.2094514    0.696  0.48681
## disturbSETTLE  0.1222650   0.2610710    0.468  0.63969
## disturbVIRGIN  0.3550208   0.2410740    1.473  0.14124
## plotsize      -0.0003217   0.0001225   -2.626  0.00880 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.005 on 785 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.0464, Adjusted R-squared:  0.03668
## F-statistic: 4.775 on 8 and 785 DF, p-value: 9.557e-06
```

```
Acer2<-aov(cover~tci+elev+streamdist+beers+plotsize, data = Acer)
AcerLM2<-lm(cover~tci+elev+streamdist+beers+plotsize, data = Acer)
summary(Acer2)
```

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## tci         1     12   12.05     3.001 0.08363 .
## elev        1     34   34.32     8.547 0.00356 **
## streamdist  1     30   30.17     7.512 0.00627 **
## beers       1     41   40.84    10.169 0.00148 **
## plotsize    1     27   27.43     6.829 0.00914 **
## Residuals  788   3165     4.02
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM2)
```

```
##
## Call:
## lm(formula = cover ~ tci + elev + streamdist + beers + plotsize,
##     data = Acer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7987 -1.3037  0.3438  1.4254  5.1959
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.5024516   0.3529416   18.424 < 2e-16 ***
## tci          -0.0467835   0.0333089    -1.405 0.160554
## elev         -0.0008253   0.0002443    -3.379 0.000764 ***
## streamdist   0.0012763   0.0004522     2.822 0.004887 **
## beers        -0.3056302   0.1027921    -2.973 0.003036 **
## plotsize     -0.0003173   0.0001214    -2.613 0.009138 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.004 on 788 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.04376,    Adjusted R-squared:  0.03769
## F-statistic: 7.212 on 5 and 788 DF,  p-value: 1.296e-06
```

```
Acer3<-aov(cover~elev+streamdist+beers+plotsize, data = Acer)
AcerLM3<-lm(cover~elev+streamdist+beers+plotsize, data = Acer)
summary(Acer3)
```

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## elev       1     25   25.09    6.241 0.01269 *
## streamdist  1     42   41.50   10.321 0.00137 **
## beers       1     37   36.79    9.150 0.00257 **
## plotsize    1     33   33.50    8.331 0.00400 **
## Residuals  789   3172    4.02
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM3)
```

```
##
## Call:
## lm(formula = cover ~ elev + streamdist + beers + plotsize, data = Acer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.9265 -1.3419  0.3182  1.4210  5.2248
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.1725065   0.2635689   23.419 < 2e-16 ***
## elev        -0.0007846   0.0002427   -3.233  0.00128 **
## streamdist   0.0014348   0.0004382    3.275  0.00110 **
## beers        -0.2903855   0.1022804   -2.839  0.00464 **
## plotsize     -0.0003458   0.0001198   -2.886  0.00400 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.005 on 789 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.04136,    Adjusted R-squared:  0.0365
## F-statistic: 8.511 on 4 and 789 DF,  p-value: 1.002e-06
```

```
Acer4<-aov(cover~elev*streamdist*beers*plotsize, data = Acer)
AcerLM4<-lm(cover~elev+streamdist+beers+plotsize, data = Acer)
summary(Acer4)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## elev          1   25.1    25.09   6.347 0.01196 *
## streamdist    1   41.5    41.50  10.497 0.00125 **
## beers         1   36.8    36.79   9.306 0.00236 **
## plotsize      1   33.5    33.50   8.474 0.00371 **
## elev:streamdist 1    0.8     0.81   0.204 0.65187
## elev:beers     1   77.3    77.28  19.548 1.12e-05 ***
## streamdist:beers 1    3.3     3.31   0.837 0.36041
## elev:plotsize  1    0.4     0.37   0.092 0.76115
## streamdist:plotsize 1    2.5     2.46   0.623 0.43008
## beers:plotsize  1    1.4     1.45   0.366 0.54537
## elev:streamdist:beers 1    3.1     3.13   0.793 0.37356
## elev:streamdist:plotsize 1    1.8     1.79   0.453 0.50119
## elev:beers:plotsize 1    3.0     3.02   0.765 0.38205
## streamdist:beers:plotsize 1    0.6     0.62   0.157 0.69158
## elev:streamdist:beers:plotsize 1    2.5     2.48   0.626 0.42905
## Residuals      778 3075.8    3.95
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM4)
```

```
##
## Call:
## lm(formula = cover ~ elev + streamdist + beers + plotsize, data = Acer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.9265 -1.3419  0.3182  1.4210  5.2248
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  6.1725065   0.2635689   23.419 < 2e-16 ***
## elev        -0.0007846   0.0002427   -3.233 0.00128 **
## streamdist   0.0014348   0.0004382    3.275 0.00110 **
## beers       -0.2903855   0.1022804   -2.839 0.00464 **
## plotsize    -0.0003458   0.0001198   -2.886 0.00400 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.005 on 789 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.04136,    Adjusted R-squared:  0.0365
## F-statistic: 8.511 on 4 and 789 DF,  p-value: 1.002e-06
```

```
Acer5<-aov(cover~elev+streamdist+plotsize, data = Acer)
AcerLM5<-lm(cover~elev+streamdist+plotsize, data = Acer)
summary(Acer5)
```



```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## elev           1      25    25.09    6.186 0.01309 *
## streamdist     1      42    41.50   10.230 0.00144 **
## plotsize       1      38    37.88    9.337 0.00232 **
## Residuals    790    3205     4.06
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1 observation deleted due to missingness
```

```
summary(AcerLM5)
```

```
##
## Call:
## lm(formula = cover ~ elev + streamdist + plotsize, data = Acer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7572 -1.3318  0.3265  1.4854  5.2108
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.9553499  0.2533512   23.506 < 2e-16 ***
## elev        -0.0008504  0.0002427   -3.505 0.000483 ***
## streamdist   0.0014425  0.0004401    3.278 0.001093 **
## plotsize    -0.0003669  0.0001201   -3.056 0.002321 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.014 on 790 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.03157,    Adjusted R-squared:  0.02789
## F-statistic: 8.584 on 3 and 790 DF,  p-value: 1.3e-05
```

Plot ID is dominating the explanation of the variance. However, when each variable is ran on its own they show significance. This is most likely due to plotID being a composite of the other variables. In other words, the values of the other variables are highly dependent on the plotID marker or visa versa because it is a marker of location. For this reason, I removed plot ID. In this case elevation and distance from stream are the most important variables to consider when trying to predict cover for *Acer rubrum*.

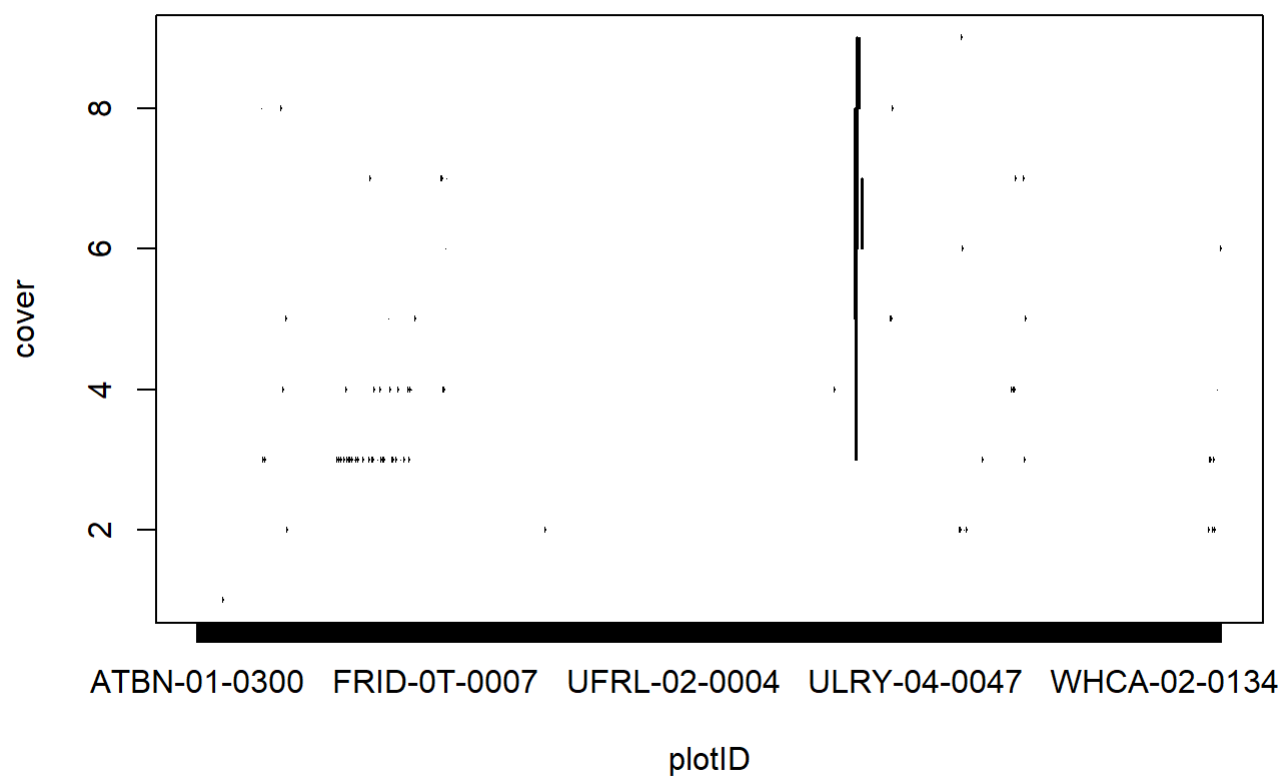
```
AcerGLM<-glm(cover~streamdist+elev+plotsize, data = Acer, family = 'poisson')
summary(AcerGLM)
```

```
##
## Call:
## glm(formula = cover ~ streamdist + elev + plotsize, family = "poisson",
##      data = Acer)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.4562  -0.5994   0.1375   0.6138   2.0766
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  1.821e+00  6.079e-02  29.958  < 2e-16 ***
## streamdist   2.758e-04  9.386e-05   2.938  0.00330 **
## elev         -1.659e-04  5.309e-05  -3.124  0.00178 **
## plotsize     -9.907e-05  3.715e-05  -2.667  0.00765 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 721.46  on 793  degrees of freedom
## Residual deviance: 699.69  on 790  degrees of freedom
##   (1 observation deleted due to missingness)
## AIC: 3416.8
##
## Number of Fisher Scoring iterations: 4
```

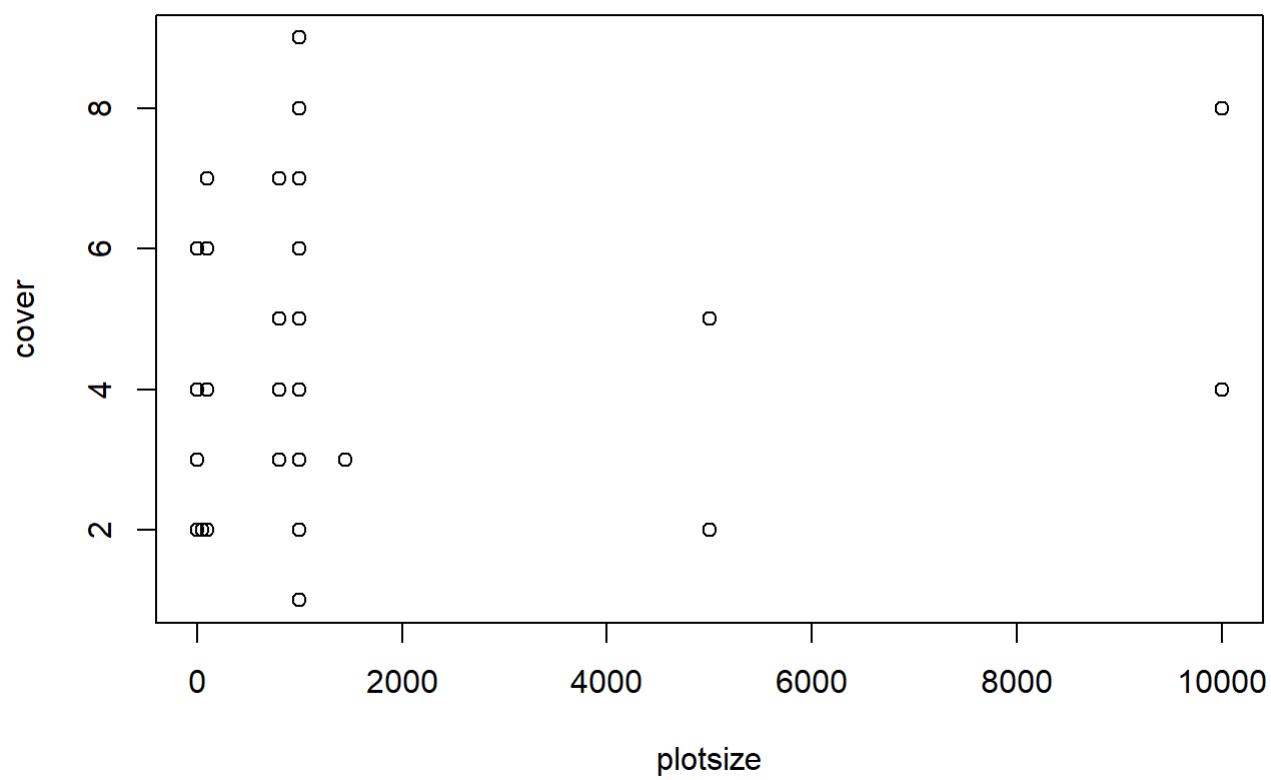
```
pseudo_r2<-1 - AcerGLM$deviance / AcerGLM$null.deviance
pseudo_r2
```

```
## [1] 0.03017477
```

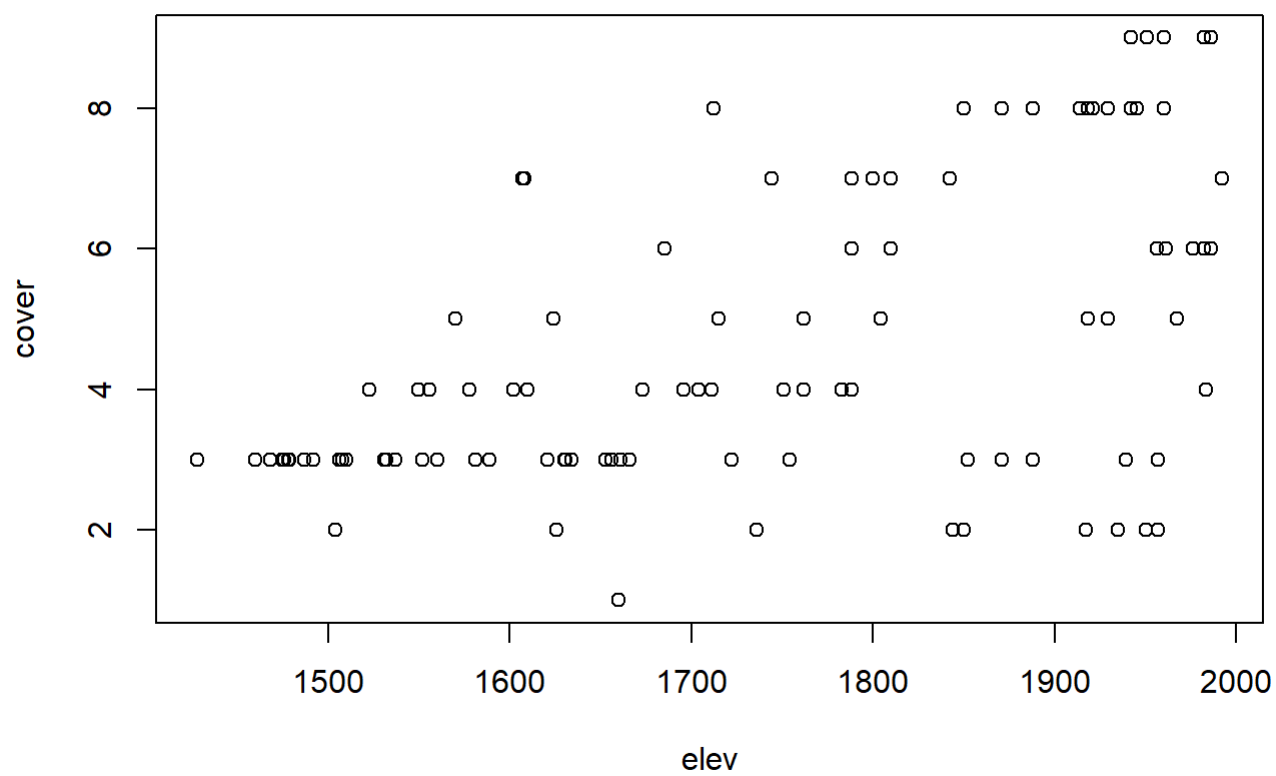
```
plot(cover~plotID, data = Abies)
```



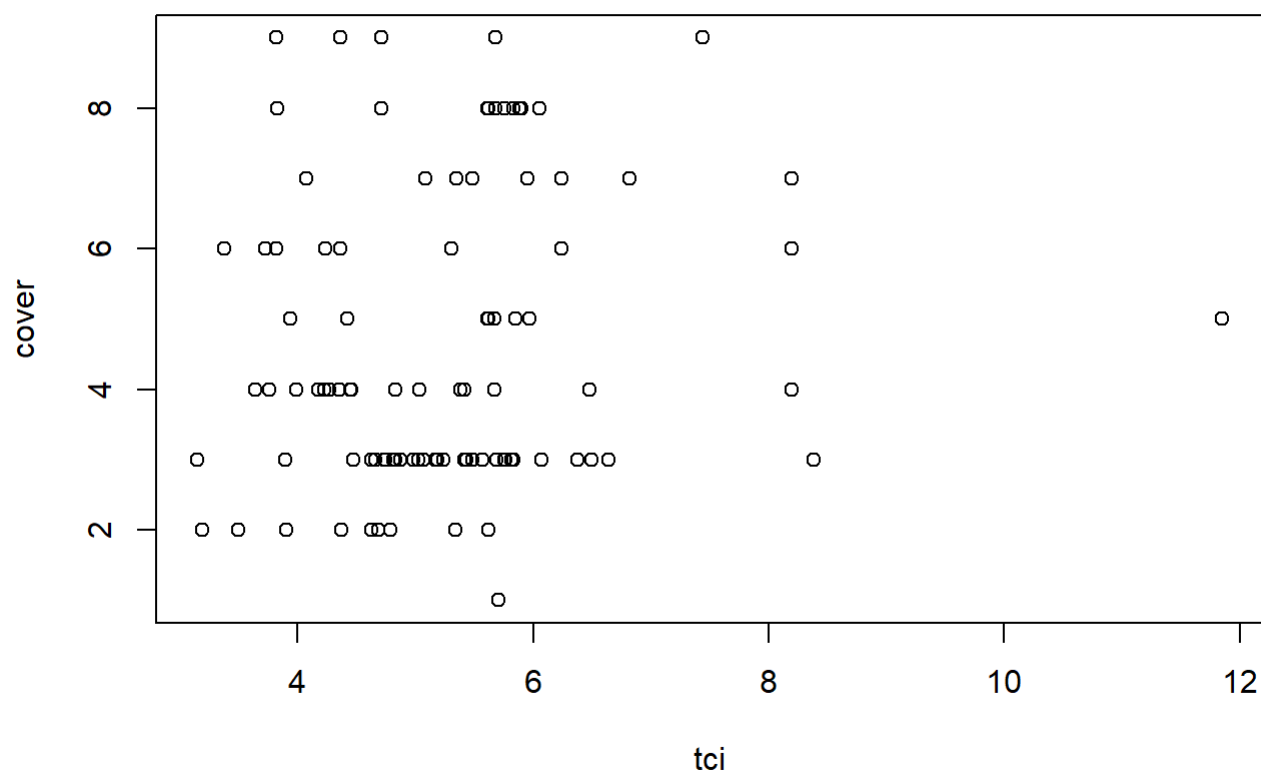
```
plot(cover~plotsize, data = Abies)
```



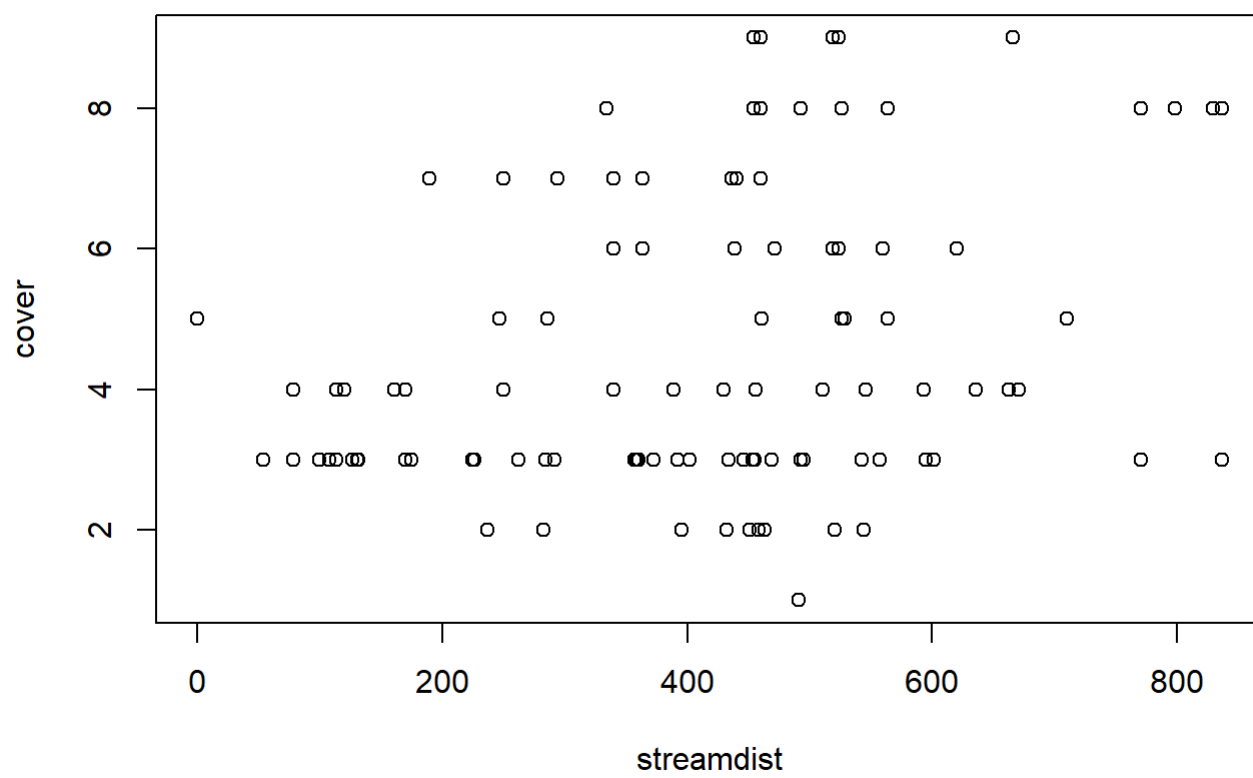
```
plot(cover~elev, data = Abies)
```



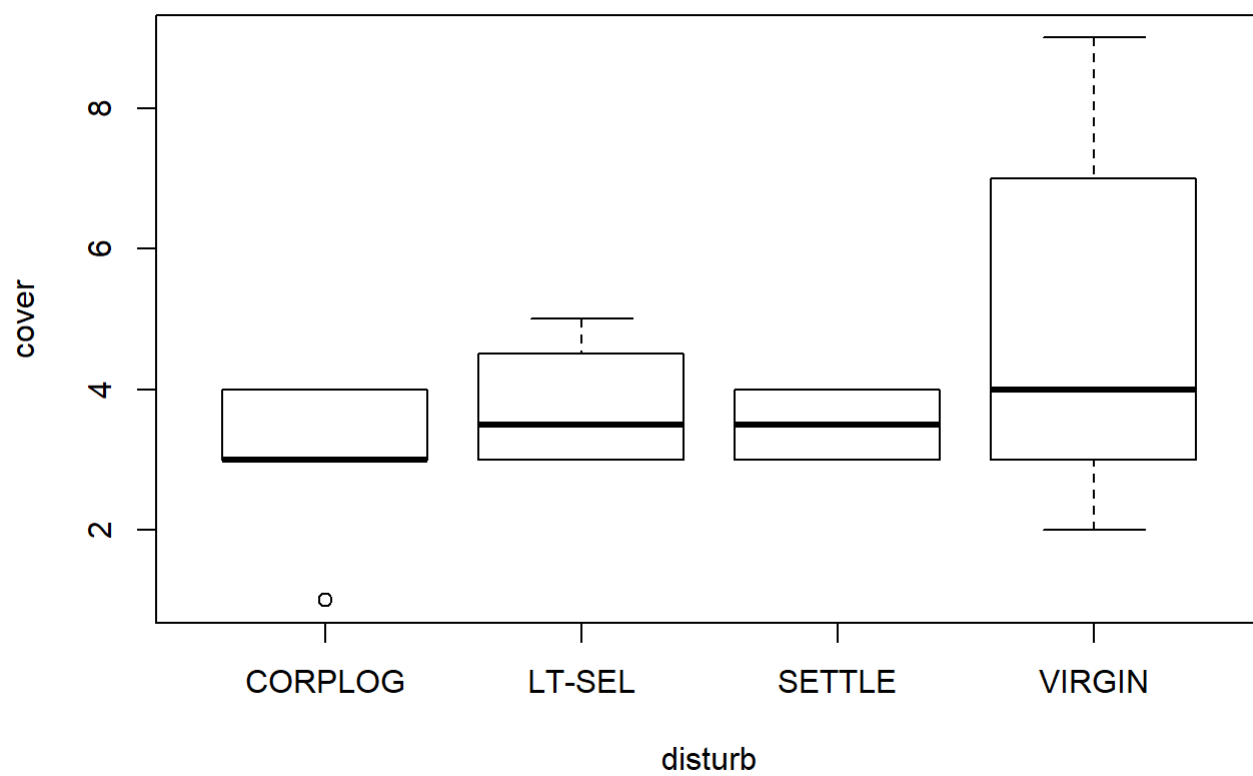
```
plot(cover~tci, data = Abies)
```



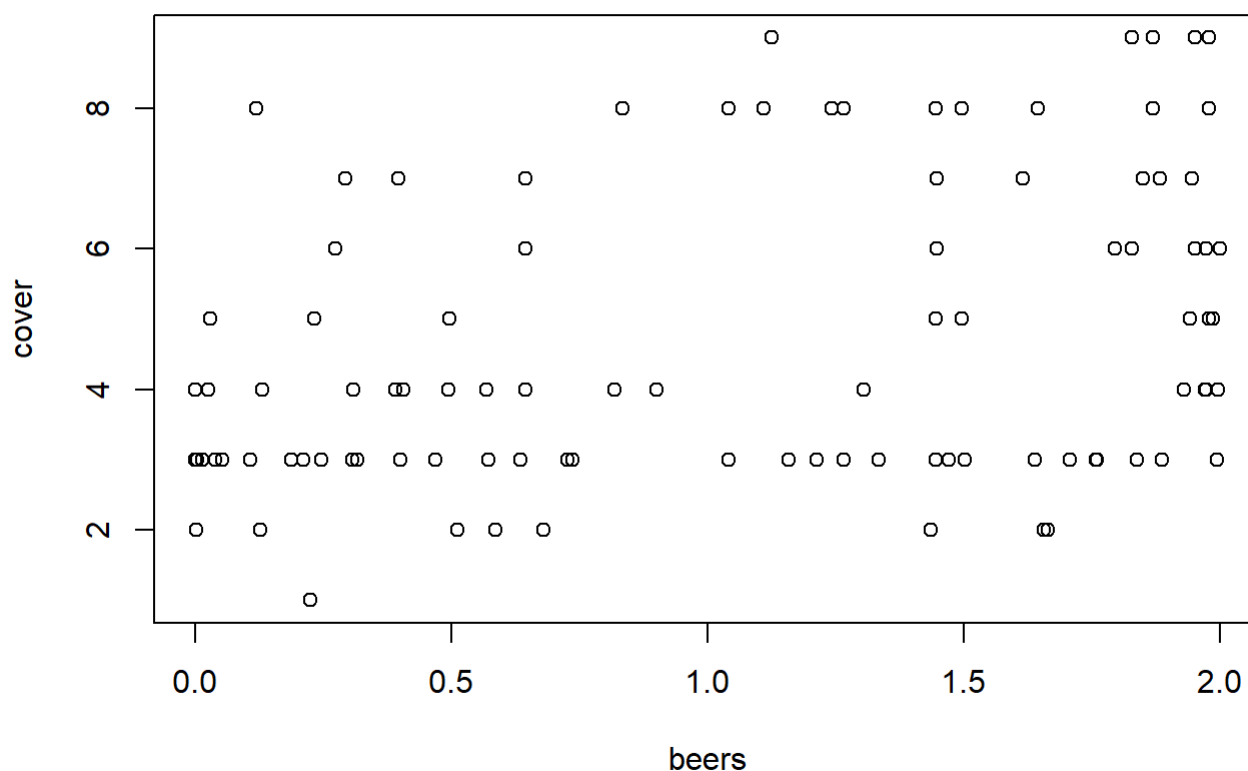
```
plot(cover~streamdist, data = Abies)
```



```
plot(cover~disturb, data=Abies)
```



```
plot(cover~beers, data = Abies)
```

```
#take a look at plot ID, elev, tci, streamdist-- lesser important disturb, beers, plotsize
```

Again, after looking at each relationship individually the variables of the most interest were plotID, elev, tci and streamdist.

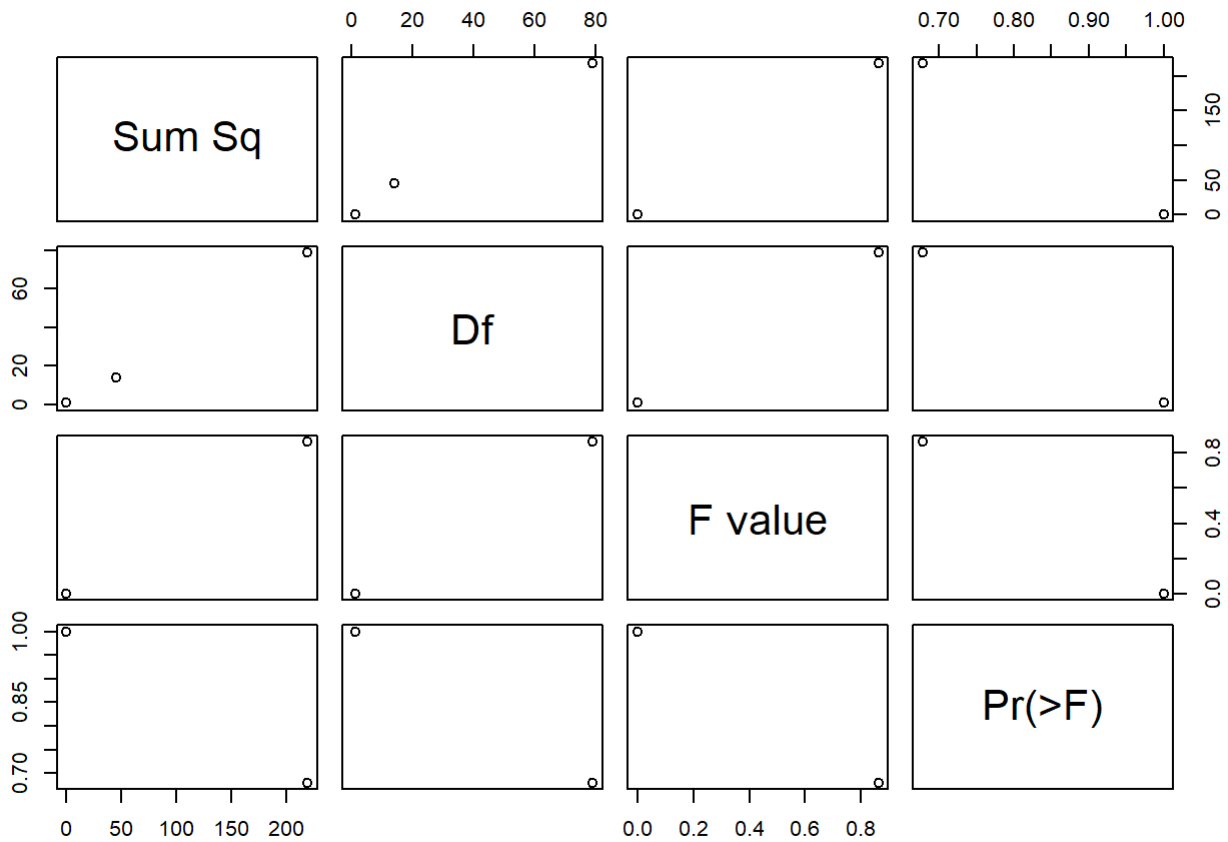
```
AbiesFull<-Anova(lm(cover~plotID+elev+tci+streamdist+beers+disturb+plotsize, data = Abies))
```

```
## Note: model has aliased coefficients
##      sums of squares computed by model comparison
```

```
summary(AbiesFull)
```

```
##      Sum Sq      Df    F value    Pr(>F)
## Min.   : 0.00  Min.   : 0.00  Min.   :0.0000  Min.   :0.6786
## 1st Qu.: 0.00  1st Qu.: 0.75  1st Qu.:0.0000  1st Qu.:1.0000
## Median : 0.00  Median : 1.00  Median :0.0000  Median :1.0000
## Mean   : 43.95  Mean   :12.12  Mean   :0.1723  Mean   :0.9357
## 3rd Qu.: 33.75  3rd Qu.: 4.25  3rd Qu.:0.0000  3rd Qu.:1.0000
## Max.   :218.70  Max.   :79.00  Max.   :0.8613  Max.   :1.0000
## NA's   :2      NA's   :3      NA's   :3
```

```
plot(AbiesFull)
```



```
AbiesAOV<- aov(cover~as.factor(plotID)+elev+tc+streamdist+beers+disturb+plotsize, data = Abies)
AbiesLM<- lm(cover~plotID+elev+tc+streamdist+beers+disturb+plotsize, data = Abies)
summary(AbiesAOV)
```

```
##               Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(plotID) 83    431   5.192   1.615  0.158
## elev              1      0   0.000   0.000  1.000
## tc                1      0   0.000   0.000  1.000
## streamdist        1      0   0.000   0.000  1.000
## beers             1      0   0.000   0.000  1.000
## Residuals         14     45   3.214
```

```
summary(AbiesLM)$adj.r.squared
```

```
## [1] 0.317921
```

```
Abies1<-aov(cover~tc+elev+streamdist+beers+disturb+plotsize, data = Abies)
AbiesLM1<-lm(cover~tc+elev+streamdist+beers+disturb+plotsize, data = Abies)
summary(Abies1)
```

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## tci         1  10.88   10.88    3.837 0.053134 .
## elev        1 158.50  158.50   55.897 4.14e-11 ***
## streamdist  1   1.11    1.11    0.392 0.532542
## beers       1  35.83   35.83   12.635 0.000597 ***
## disturb     3   5.73    1.91    0.673 0.570504
## plotsize    1   0.21    0.21    0.075 0.784530
## Residuals   93 263.70    2.84
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(AbiesLM1)$adj.r.squared
```

```
## [1] 0.3982977
```

```
Abies2<-aov(cover~tci+elev+beers, data = Abies)
AbiesLM2<-lm(cover~tci+elev+beers, data = Abies)
summary(Abies2)
```

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## tci         1  10.88   10.88    3.90 0.051091 .
## elev        1 158.50  158.50   56.82 2.42e-11 ***
## beers       1  33.22   33.22   11.91 0.000825 ***
## Residuals   98 273.36    2.79
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(AbiesLM2)
```

```
##
## Call:
## lm(formula = cover ~ tci + elev + beers, data = Abies)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9474 -1.2389  0.0673  1.1258  4.4466
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -10.426571   1.965465  -5.305 7.00e-07 ***
## tci           0.505556   0.137752   3.670 0.000395 ***
## elev          0.006623   0.001005   6.587 2.25e-09 ***
## beers         0.848679   0.245905   3.451 0.000825 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.67 on 98 degrees of freedom
## Multiple R-squared:  0.4257, Adjusted R-squared:  0.4081
## F-statistic: 24.21 on 3 and 98 DF, p-value: 8.319e-12
```

```
Abies3<-aov(cover~elev*tci*beers, data = Abies)
AbiesLM3<-lm(cover~elev+beers+tci, data = Abies)
summary(Abies3)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## elev           1 141.49  141.49   58.452 1.76e-11 ***
## tci            1  27.89   27.89   11.523 0.001008 **
## beers          1  33.22   33.22   13.726 0.000357 ***
## elev:tci        1  10.64   10.64    4.397 0.038685 *
## elev:beers      1  33.00   33.00   13.632 0.000373 ***
## tci:beers       1   0.54    0.54    0.223 0.637704
## elev:tci:beers  1   1.65    1.65    0.681 0.411502
## Residuals      94 227.53    2.42
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(AbiesLM3)
```

```
##
## Call:
## lm(formula = cover ~ elev + beers + tci, data = Abies)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9474 -1.2389  0.0673  1.1258  4.4466
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -10.426571   1.965465  -5.305 7.00e-07 ***
## elev          0.006623   0.001005   6.587 2.25e-09 ***
## beers         0.848679   0.245905   3.451 0.000825 ***
## tci           0.505556   0.137752   3.670 0.000395 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.67 on 98 degrees of freedom
## Multiple R-squared:  0.4257, Adjusted R-squared:  0.4081
## F-statistic: 24.21 on 3 and 98 DF, p-value: 8.319e-12
```

```
Abies5<-aov(cover~elev+tci, data = Abies)
AbiesLM5<-lm(cover~elev+tci, data = Abies)
summary(Abies5)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## elev           1 141.49   141.49  45.688 9.71e-10 ***
## tci            1  27.89    27.89   9.007  0.0034 **
## Residuals     99 306.58     3.10
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(AbiesLM5)
```

```
##
## Call:
## lm(formula = cover ~ elev + tci, data = Abies)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9957 -0.8333  0.1053  0.9229  4.1766
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -10.475197   2.070892  -5.058 1.95e-06 ***
## elev         0.007391    0.001033   7.154 1.48e-10 ***
## tci          0.430070    0.143303   3.001  0.0034 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.76 on 99 degrees of freedom
## Multiple R-squared:  0.3559, Adjusted R-squared:  0.3429
## F-statistic: 27.35 on 2 and 99 DF,  p-value: 3.503e-10
```

Unlike *Acer rubrum*, plotID was not significantly important in predicting cover in *Acer fraseri*. Instead, the three most important parameters were elevation, tci and beers. However, beers and elevation are not independent and have a significant interaction. Another consideration is the highest adjusted R squared is much lower than *Acer rubrum* at about 0.4.

```
AbiesGLM<-glm(cover~elev+tci+beers, data = Abies, family = 'poisson')
summary(AbiesGLM)
```

```
##
## Call:
## glm(formula = cover ~ elev + tci + beers, family = "poisson",
##      data = Abies)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.87039  -0.56446   0.04234   0.47595   2.04892
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.7933522  0.5710240  -3.141  0.00169 **
## elev         0.0014528  0.0002897   5.015  5.3e-07 ***
## tci          0.1060297  0.0361364   2.934  0.00334 **
## beers        0.1879386  0.0711588   2.641  0.00826 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 98.822  on 101  degrees of freedom
## Residual deviance: 54.988  on  98  degrees of freedom
## AIC: 401.51
##
## Number of Fisher Scoring iterations: 4
```

```
pseudo_r2Abies<-1 - AbiesGLM$deviance / AbiesGLM$null.deviance
pseudo_r2Abies
```

```
## [1] 0.4435648
```

The GLM did assist in the pseudo R squared which was about 0.44. Overall the GLM did not greatly change the outcomes of the models.

3. Take away messages *Acer rubrum* cover is best explained by the plotID which is very dependent on the other parameters being measured. *Acer fraseri* cover was less well explained by parameters measured, but the best choices were elevation, beers, and tci. This makes sense when accounting for the habitat flexibility for each of these species. *Acer rubrum*, a habitat generalist's most important determinant is the plotID. This alludes to clumping of individuals because the greatest indicator of cover is the location where similar cover values are. Additionally, when plotID was removed the adjusted R squared decreased dramatically meaning that the cover of *Acer rubrum* was hard to predict without knowing which plot it is. Adversely, a habitat specialist in *Abies fraseri* is more concerned with specific values of elevation, tci, and beers because they require a narrower range of ideal habitat characteristics.