# Indicator Variable Regression

# R Handout

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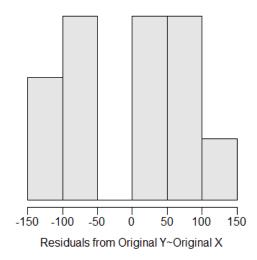
# Laysan Teal Monitoring

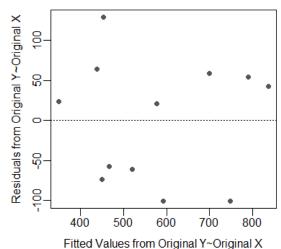
Monitoring the abundance of animal populations can be expensive. Researchers often attempt to find methods for accurately predicting abundance from simple and inexpensive metrics. In one study, Reynolds et al. (2017) compared the abundance of Laysan Teal (Anas laysanensis) as estimated from the time-intensive Lincoln-Petersen mark-resight method to the maximum annual observed count of Laysan Teal made bi-monthly on Midway Atoll. They wanted to determine if there was a significantly positive relationship between the mark-resight abundance estimate and the maximum annual count, and if that relationship differed between the breeding and non-breeding seasons.

#### Fitting the Linear Model

```
> lm1 <- lm(mr.estimate~max.count*season,data=df)</pre>
```

> transChooser(lm1)





# Lecture Support - ANOVA Demonstration

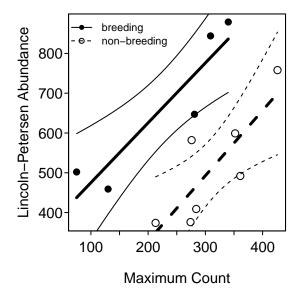
```
> lm1 <- lm(mr.estimate~max.count*season,data=df)</pre>
> anova(lm1)
                 Df Sum Sq Mean Sq F value
max.count
                  1 83421
                              83421 10.6370 0.011503
                  1 190750 190750 24.3225 0.001147
season
                                359 0.0457 0.835991
                       359
max.count:season
Residuals
                     62740
                               7843
> lm2 <- lm(mr.estimate~max.count+season,data=df)</pre>
> anova(lm2)
          Df Sum Sq Mean Sq F value
                                        Pr(>F)
              83421
                      83421 11.899 0.0072813
max.count
                     190750
                             27.207 0.0005521
season
           1 190750
Residuals
              63099
                       7011
> lm3 <- lm(mr.estimate~max.count,data=df)</pre>
> anova(lm3)
          Df Sum Sq Mean Sq F value Pr(>F)
max.count 1 83421
                      83421 3.2862 0.09994
```

# Model Exploration and Summarization

25385

Residuals 10 253849

> fitPlot(lm1,interval="confidence",xlab=xlb1,ylab=ylb1,legend="topleft",cex.leg=0.7)



### > summary(lm1)

#### Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
                              324.6130
                                          94.4805
                                                    3.436 0.00888
max.count
                                1.5048
                                           0.3779
                                                    3.982 0.00405
seasonnon-breeding
                             -323.9986
                                         187.8052 -1.725 0.12278
max.count:seasonnon-breeding
                                0.1355
                                           0.6336
                                                    0.214 0.83599
```

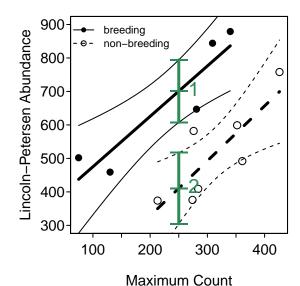
Residual standard error: 88.56 on 8 degrees of freedom Multiple R-squared: 0.814, Adjusted R-squared: 0.7442 F-statistic: 11.67 on 3 and 8 DF, p-value: 0.002718

#### > confint(lm1)

```
2.5 % 97.5 % (Intercept) 106.7406493 542.485322 max.count 0.6333927 2.376185 seasonnon-breeding -757.0781733 109.081017 max.count:seasonnon-breeding -1.3255575 1.596583
```

# > ( nd <- data.frame(max.count=c(250,250),season=c("breeding","non-breeding")) )</pre>

```
max.count season
1 250 breeding
2 250 non-breeding
```



 obs max.count
 season
 fit
 lwr
 upr

 1
 1
 250
 breeding
 700.8101
 607.3089
 794.3114

 2
 2
 250
 non-breeding
 410.6898
 304.4186
 516.9610

# Fish Energy Density

Hartman and Brandt (1995) examined the relationship between energy density and percent dry weight for four species of fish – Bay Ancovy (*Anchoa mitchilli*), Bluefish (*Pomatomus saltatrix*), Striped Bass (*Morone saxatilis*), and Weakfish (*Cynoscion regalis*). They wanted to describe the relationship between these two variables and determine if there were any differences among species.

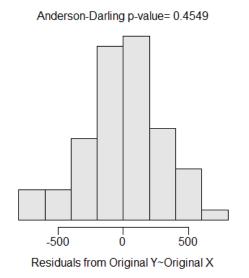
```
> FED <- read.csv("FishEnergyDensity.csv")
> str(FED)

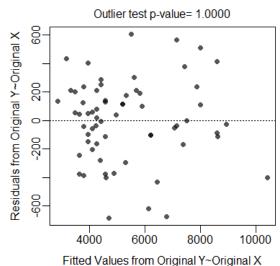
'data.frame': 64 obs. of 3 variables:
$ species: Factor w/ 4 levels "bayanchovy","bluefish",..: 2 2 2 2 2 2 2 2 2 2 2 ...
$ dw : int 39 34 34 32 31 30 30 29 26 25 ...
$ ed : int 10000 9000 8500 8100 7500 7100 7700 6100 5900 5500 ...
```

# Model Fitting and Analysis

```
> lm1 <- lm(ed~dw*species,data=FED)
```

### > transChooser(lm1)





#### > anova(lm1)

```
Pr(>F)
                 Sum Sq
                          Mean Sq F value
            1 170693154 170693154 1858.966 < 2.2e-16
dw
species
            3
               10592036
                           3530679
                                     38.452 1.258e-13
dw:species
                           1368539
                                     14.904 3.002e-07
            3
                4105617
Residuals 56
                5142008
                            91822
```

### > compSlopes(lm1)

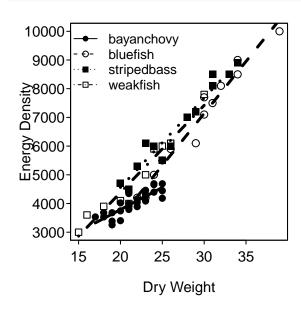
```
Multiple Slope Comparisons (using the 'holm' adjustment)
```

```
comparison diff 95% LCI 95% UCI p.unadj p.adj bluefish-bayanchovy 208.30952 145.62788 270.99117 0.00000 0.00000 2 stripedbass-bayanchovy 157.62055 94.29774 220.94336 0.00001 0.00005 weakfish-bayanchovy 149.60379 83.20906 215.99852 0.00003 0.00012 stripedbass-bluefish -50.68898 -101.08569 -0.29227 0.04873 0.10290 weakfish-bluefish -58.70573 -112.91192 -4.49955 0.03430 0.10290 weakfish-stripedbass -8.01676 -62.96310 46.92958 0.77116 0.77116
```

# Slope Information (using the 'holm' adjustment)

```
level slopes 95% LCI 95% UCI p.unadj p.adj
bayanchovy 154.1905 102.2341 206.1469 0 0
weakfish 303.7943 262.4567 345.1318 0 0
stripedbass 311.8110 275.6128 348.0092 0 0
bluefish 362.5000 327.4355 397.5645 0 0
```

### > fitPlot(lm1,xlab="Dry Weight",ylab="Energy Density",legend="topleft",cex.leg=0.8)



## Follow-Up Model Fitting and Analysis

```
> FED1 <- filterD(FED, species!="bayanchovy")
> lm2 <- lm(ed~dw*species, data=FED1)
> anova(lm2)
```

```
Sum Sq
                          Mean Sq F value
                                               Pr(>F)
            1 104962559 104962559 838.6208 < 2.2e-16
dw
            2
                2583571
                          1291785
                                   10.3210 0.0003476
species
dw:species
           2
                 556352
                           278176
                                    2.2225 0.1247887
                4005150
                           125161
Residuals
          32
```

> compIntercepts(lm2)

Warning: Removed an interaction term from 'mdl' (i.e., assumed parallel lines) to test intercepts.

Tukey HSD on means adjusted assuming parallel lines

```
comparison diff 95% LCI 95% UCI p.adj
1 stripedbass-bluefish 631.3980 291.1082 971.6878 0.0001843341
2 weakfish-bluefish 506.4749 144.5318 868.4180 0.0044160914
3 weakfish-stripedbass -124.9231 -480.8922 231.0461 0.6693967026
```

Mean ed when dw=26.10526

```
bluefish stripedbass weakfish 5828.663 6460.061 6335.138
```

### > compIntercepts(lm2,common.cov=0)

Warning: Removed an interaction term from 'mdl' (i.e., assumed parallel lines) to test intercepts.

Tukey HSD on means adjusted assuming parallel lines

```
comparison diff 95% LCI 95% UCI p.adj
1 stripedbass-bluefish 631.3980 291.1082 971.6878 0.0001843341
2 weakfish-bluefish 506.4749 144.5318 868.4180 0.0044160914
3 weakfish-stripedbass -124.9231 -480.8922 231.0461 0.6693967026
```

Mean ed when dw=0

```
bluefish stripedbass weakfish -2752.451 -2121.053 -2245.976
```

> fitPlot(lm2,xlab="Dry Weight",ylab="Energy Density",legend="topleft",cex.leg=0.8)

