Simple Linear Models

Lecture 04.1: General Linear Models

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Module: Linear, Non-linear, and Mixed Effects Models

Readings

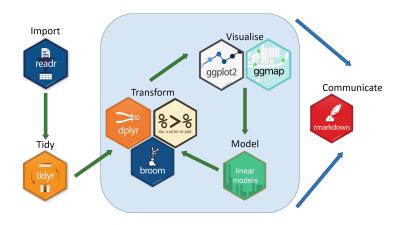
Required for class:

► NA

Optional:

- ► Crawley, M. Statistics: An Introduction Using R
- ▶ Bolker, B. Ecological Models and Data in R Ebook version

On to statistical analysis.



Data

A dataset looking at salmon residence time in streams and how that varies with sex, age, year and precipitation.



Data

This dataset has multiple types of X variables (both categorical and continuous) and a single Y variable so we can look at different types of linear models.

##	# A t	ibble:	68 x	5			
##	Re	sidenc	eTime	Sex	Age	Year	${\tt Precip}$
##			<dbl></dbl>	<chr>></chr>	<chr>></chr>	<dbl></dbl>	<dbl></dbl>
##	1		1	f	a	2001	16.1
##	2		2.5	m	a	2000	9.10
##	3		3	f	a	2001	21.6
##	4		3	m	a	2001	17.9
##	5		3	f	a	2002	17.7
##	6		3	f	a	2002	8.10
##	7		3	f	j	2002	4.48
##	8		3	f	j	2002	9.74
##	9		3	f	j	2002	7.48
##	10		3	f	j	2002	17.7
##	#	with	58 moi	re rows	3		

General Linear Models

General Linear Models refer to linear regression models that have a continuous dependent variable (Y) and a single or series of independent variables (X's) that can be either continuous or categorical. They all assume normal distributions. These models can have specific names depending on their type, but are all linear models.

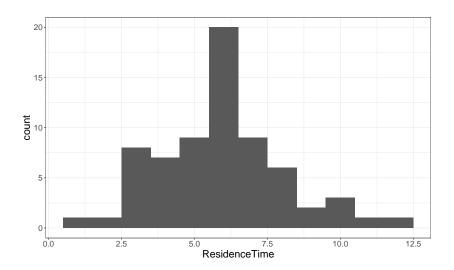
- ▶ Regression continuous Y, 1 continuous X. lm()
- ► ANOVA continuous Y, categorical X. aov()
- ► Multiple Linear Regression continuous Y, multiple continuous X. lm()
- ► **ANCOVA** continuous Y, at least 1 continuous X and at least 1 categorical X. lm()

General Linear Models - Assumptions

- 1. Relationships are (all) linear
 - ► For regressions only
- 2. (Multivariate) Normal distributions of error variance ϵ
- 3. Equal variance (aka Homoscedasticity)
 - ► ANOVA's are pretty robust to this
- 4. Independence of observed samples

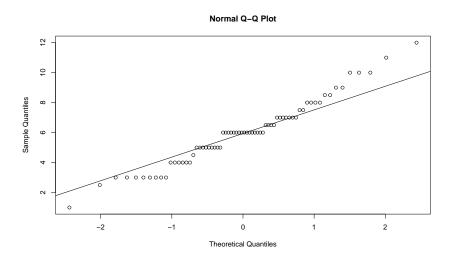
Normality Assumption

How are the data distributed?



Normality Assumption

Try a normal Q-Q plot



Normality Assumption

Try a **Shapiro-Wilk Test**

- \blacktriangleright H_0 : Data are not different from a normal distribution
- \blacktriangleright H_a : Data are different from a normal distribution

```
shapiro.test(salmon$ResidenceTime)
```

```
##
## Shapiro-Wilk normality test
##
## data: salmon$ResidenceTime
## W = 0.97122, p-value = 0.1169
```

Regression - Experimental Design

Dependent variable (Y) is continuous, independent variable (X) is continuous.

 $Y_i = \beta_0 + \beta_i X_i + \epsilon$ (β_0 is the intercept, β_i is the slope coefficient and ϵ is the error)

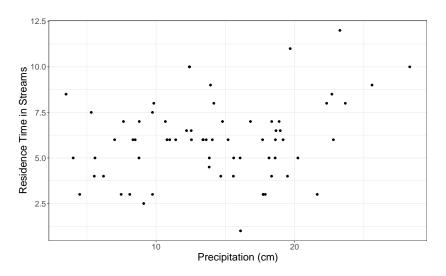
 H_0 : no relationship between X and Y

Some Questions...

- ▶ How does elevation (X) alter a plant's seed production (Y)?
- ► How does temperature (X) alter a lizard's metabolic rate (Y)?
- ► How does the year (X) influence the average global temperature (Y)?

Regression

Does residence time within a stream depend on the amount of precipitation?



Regression

Does residence time within a stream depend on the amount of precipitation?

test_reg <- lm(ResidenceTime ~ Precip, data = salmon)</pre>

```
summary(test_reg)
##
## Call:
## lm(formula = ResidenceTime ~ Precip, data = salmon)
##
## Residuals:
##
      Min 1Q Median
                             30
                                    Max
## -5.1676 -1.3786 0.0781 1.1288 5.1342
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 4.60153 0.69182 6.651 6.72e-09 ***
## Precip 0.09729 0.04522 2.152 0.0351 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.103 on 66 degrees of freedom
## Multiple R-squared: 0.06555, Adjusted R-squared: 0.0514
## F-statistic: 4.63 on 1 and 66 DF, p-value: 0.03508
```

Regression

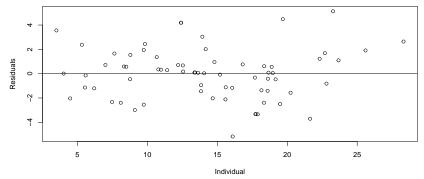
$$Y_i = \beta_0 + \beta_i X_i + \epsilon$$
$$Y = 4.6015 + 0.0973X + \epsilon$$

```
##
## Call:
## lm(formula = ResidenceTime ~ Precip, data = salmon)
##
## Residuals:
             10 Median 30
##
      Min
                                     Max
## -5.1676 -1.3786 0.0781 1.1288
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 4.60153 0.69182 6.651 6.72e-09 ***
## Precip
               0.09729
                        0.04522 2.152 0.0351 *
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.103 on 66 degrees of freedom
## Multiple R-squared: 0.06555, Adjusted R-squared: 0.0514
## F-statistic: 4.63 on 1 and 66 DF, p-value: 0.03508
```

Residuals

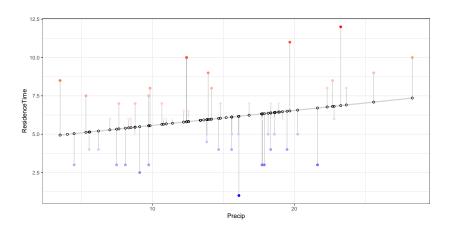
Residual plots show you the difference between your observed data (y), and the expected, or fitted value (\hat{y}) .

 $Residual = y - \hat{y}$ $\texttt{test_reg_resid} \leftarrow \texttt{resid(test_reg)}$



Residuals

 $Residual = y - \hat{y}$



ANOVA - Experimental Design

Dependent variable (Y) is continuous, independent variable (X) is categorical.

 $Y_{ij} = \mu + \alpha_i + \epsilon_{ij}$ (μ is the grand mean, α_i is the i^{th} group mean and ϵ_{ij} is the error)

 H_0 : no difference among groups

Some Questions...

- ► How does a diet treatment (X) alter an animal's growth rate (Y)?
- ► How do nutrient additions (X) alter plant species diversity (Y)?
- ► How does sex of an organism (X) alter it's feeding behavior (Y)?
- ► How does plant family (X) alter a plant's SLA (Y)?

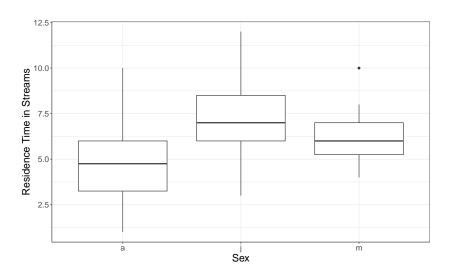
ANOVA vs t-test

For 2 groups of equal size, ANOVA's and t-tests give you the same result.

```
v \leftarrow c(5,4,4,3,3,7,5,7,6,6)
t.test(v ~ x)
##
## Welch Two Sample t-test
##
## data: y by x
## t = 4.5356, df = 8, p-value = 0.00191
## alternative hypothesis: true difference in means between group f and group m is not equal to 0
## 95 percent confidence interval:
## 1.179777 3.620223
## sample estimates:
## mean in group f mean in group m
##
             6.2
anova(lm(v ~ x))
```

ANOVA

Is residence time within a stream a function of the age of the fish?



ANOVA

Is residence time within a stream a function of the age of the fish?

ANOVA - Post Hoc Test

But this only tells you that age significantly predicts residence time. Which groups are different from each other?

```
TukeyHSD(test_aov)
```

```
##
    Tukey multiple comparisons of means
       95% family-wise confidence level
##
##
## Fit: aov(formula = ResidenceTime ~ Age, data = salmon)
##
## $Age
##
            diff
                        lwr
                                  upr
                                          p adi
## i-a 2.2765152 0.8850487 3.6679816 0.0006159
## m-a 1.5454545 0.1240606 2.9668485 0.0299676
## m-j -0.7310606 -2.1225271 0.6604059 0.4226850
```

Multiple Linear Regression - Experimental Design

Dependent variable (Y) is continuous, multiple independent variables (X) that are all continuous.

$$Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \dots + \epsilon$$

 \triangleright β_i are **partial regression coefficients** - the effect of X_i while holding all other X constant

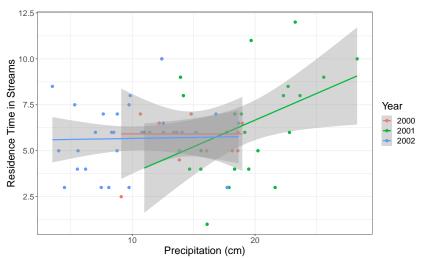
 H_0 : no difference among slopes

Some Questions...

- ► How does elevation (X1) and temperature (X2) alter a plant's seed production (Y)?
- ► How does temperature (X1) and humidity (X2) alter a lizard's metabolic rate (Y)?
- ► How does the year (X1) and atmospheric CO2 level (X2) influence the average global temperature (Y)?

Multiple Linear Regression

Does residence time within a stream depend on the amount of precipitation and the year?



Multiple Linear Regression

Does residence time within a stream depend on the amount of precipitation and the year?

```
test mreg <- lm(ResidenceTime ~ Year + Precip, data = salmon)
summary(test_mreg)
##
## Call:
## lm(formula = ResidenceTime ~ Year + Precip. data = salmon)
##
## Residuals:
      Min
              10 Median
                                    Max
## -5 1459 -1 2522 0 0064 1 1341 5 1031
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -236.96285 773.21091 -0.306
                                           0.7602
## Year
               0.12065 0.38619 0.312 0.7557
## Precip 0.10465 0.05125 2.042 0.0452 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.118 on 65 degrees of freedom
## Multiple R-squared: 0.06695, Adjusted R-squared: 0.03825
## F-statistic: 2.332 on 2 and 65 DF, p-value: 0.1052
```

ANCOVA - Experimental Design

Dependent variable (Y) is continuous, multiple independent variables (X), where at least one is continuous and one is categorical.

$$Y_{ij} = \mu + \alpha_i + \beta_{within}(X_{ij} - \overline{X_i}) + \epsilon_{ij}$$

 H_0 : no difference among slopes, no difference among groups.

▶ First compares slopes, then compares groups while holding effects of covariates constant.

Some Questions...

- ► How does elevation (X1) and nutrient addition (X2) alter a plant's seed production (Y)?
- ► How does temperature (X1) and the sex of an individual (X2) alter lizard's metabolic rate (Y)?
- ► How does the year (X1) and atmospheric CO2 level (X2) and habitat type (X3) influence plant biomass (Y)?

ANCOVA

Does residence time within a stream depend on the amount of precipitation, the year, and the sex of the fish?

```
test mreg <- lm(ResidenceTime ~ Precip + Year + Sex , data = salmon)
summary(test_mreg)
##
## Call:
## lm(formula = ResidenceTime ~ Precip + Year + Sex, data = salmon)
##
## Residuals:
      Min
               10 Median
                                    Max
## -4.2393 -0.8995 -0.1261 0.8873 4.0080
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -317.30460 589.84225 -0.538 0.59248
## Precip
           0.13975 0.03942 3.545 0.00074 ***
             0.15936 0.29460 0.541 0.59042
## Year
              3.15523 0.45666 6.909 2.68e-09 ***
## Sexm
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.615 on 64 degrees of freedom
## Multiple R-squared: 0.4656, Adjusted R-squared: 0.4405
## F-statistic: 18.59 on 3 and 64 DF, p-value: 8.78e-09
```

ANCOVA

Does residence time within a stream depend on the amount of precipitation, the year, and the sex of the fish?

