# Statistical Modeling in R

One big part of statistics is fitting a *model* to data. R has many functions for doing this, but I'll mention only lm, which fits a linear model.

Models in R are often specified using *formulas*, that say how one thing is modelled in terms of other things.

For lm, we want to specify that some *response* variable is modelled as a linear combination (plus noise) of some *explanatory* variables. This is done using a formula such as

growth ~ ave\_temp + fertilizer + variety

This might express that the amount by which some plant grows is linearly related to the average temperature, the amount of fertilizer used, and a set of indicator variables indicating the variety of the plant.

#### A Simple Example of a Linear Model

Here, I'll show the results of a very simple linear model, relating the volume of wood in a cherry tree to its girth (diameter of trunk). The data is in the data frame **trees** that comes with R.

Here's a plot of the data:



trees\$Girth

## Fitting the Model with lm

We can fit a linear model for volume given girth as follows:

```
> lm (trees$Volume ~ trees$Girth)
```

Call: lm(formula = trees\$Volume ~ trees\$Girth)

Coefficients: (Intercept) trees\$Girth

-36.943 5.066

The result says that best fit model for the volume is

Volume = -36.943 + 5.066 Girth + noise

We can get the same result with an abbreviated formula by saying the data comes from the data frame **trees**:

lm (Volume ~ Girth, data=trees)

## Using the Result of lm

The value returned by lm is an object of class "lm", which has special methods for printing and other operations.

We can save the result, and then get the regression coefficients with coef.

> m <- lm (Volume ~ Girth, data=trees)
> coef(m)
(Intercept) Girth
-36.943459 5.065856

We could use these coefficients to predict the volume for a new tree, with girth of 11.6:

```
> coef(m) %*% c(1,11.6) # %*% will compute the dot product
      [,1]
[1,] 21.82048
```

#### Getting More Details on the Model Fitted

We can also ask for more statistical details with summary:

> summary(m) Call: lm(formula = Volume ~ Girth, data = trees) Residuals: Min 1Q Median 3Q Max -8.065 -3.107 0.152 3.495 9.587 Coefficients: Estimate Std. Error t value Pr(>|t|) (Intercept) -36.9435 3.3651 -10.98 7.62e-12 \*\*\* 5.0659 0.2474 20.48 < 2e-16 \*\*\* Girth Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1 Residual standard error: 4.252 on 29 degrees of freedom Multiple R-squared: 0.9353, Adjusted R-squared: 0.9331

F-statistic: 419.4 on 1 and 29 DF, p-value: < 2.2e-16

#### Plotting the Regression Line

We can also plot the regression line from the fitted model on top of a scatterplot of the data, using abline(m):



The plot shows some indication that the relationship is actually curved.

## Some Useful Functions of Vectors

The unique function returns a vector of unique values:

```
> colours <- c("red","blue","red","red","green","blue")
> unique(colours)
[1] "red" "blue" "green"
```

The **sort** function sorts a vector in increasing order (or decreasing order if you use **decreasing=TRUE**):

> ages <- c(4,9,12,2,4,9,10)
> sort(ages)
[1] 2 4 4 9 9 10 12
> sort(unique(ages),decreasing=TRUE)
[1] 12 10 9 4 2

The which.min and which.max functions give the index of the smallest and largest elements in a vector (first occurrence if they occur more than once):

```
> which.min(ages)
[1] 4
> which.max(ages)
[1] 3
```