14.8 Appendix: R Output for the Samara Example by EV Nordheim, MK Clayton & BS Yandell, December 11, 2003

In this appendix we will briefly illustrate some of the regression commands available in R by using the samara data and the lm command. Note that lm allows for the possibility of having several predictors. This is important in multiple regression, a topic we will not pursue in this chapter.

We have entered the data with x in column V1 and y in column V2.

> samara = read.table("http://www.stat.wisc.edu/~st571-1/data/samara.dat")
> x = samara\$V1
> y = samara\$V2

Alternatively, you can enter data as we have sometimes done:

> x = c(1.72, 1.72, 1.77, 1.78, 1.82, 1.85, 1.88, 1.93, 1.96, 1.96, + 2, 2, 2.03, 2.06) > y = c(0.85, 0.86, 0.72, 0.79, 0.82, 0.8, 0.99, 0.94, 0.82, 0.89, + 0.95, 1, 0.98, 0.99)

The R command plot produces a scatterplot of y versus x. The line

> plot(x, y)



To regress y on x, we proceed as follows:

```
> samara.lm = lm(y ~ x)
> summary(samara.lm)
Call:
lm(formula = y ~ x)
Residuals:
                    Median
    Min
               1Q
                                 ЗQ
                                         Max
-0.10345 -0.03416 0.00803
                            0.04917
                                    0.11057
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                         0.2992
(Intercept) -0.1551
                                 -0.518
                                         0.61355
              0.5503
                         0.1579
                                  3.485
                                         0.00451 **
х
___
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
Residual standard error: 0.06605 on 12 degrees of freedom
Multiple R-Squared: 0.503,
                                  Adjusted R-squared: 0.4616
F-statistic: 12.14 on 1 and 12 DF, p-value: 0.004506
> anova(samara.lm)
Analysis of Variance Table
Response: y
          Df
               Sum Sq Mean Sq F value
                                         Pr(>F)
           1 0.052985 0.052985
                               12.144 0.004506 **
х
Residuals 12 0.052357 0.004363
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The object samara.lm contains all the information about the regression fit. The summary command provides the form of the regression equation, the estimates of the intercept "(Intercept)" and slope ("x"). Also provided are the estimated standard deviations ("Std. Error") of these quantities, T values corresponding to $H_0: b_0 = 0$ and $H_0: b_1 = 0$, and p-values for these tests. R then prints an estimate of σ_e and R^2 . The notation Adjusted R-squared refers to an adjusted version of R^2 important in multiple regression. Finally, the anova produces the ANOVA table for the regression, including the F value and p-value for $H_0: b_1 = 0$.

You can add the regression line to the data plot using the following command. However, we will not show the plot here.

> lines(x, predict(samara.lm))

R has a number of definitions of residual suitable for different purposes. To produce a residual plot as we have defined it, we use the **plot** command.

```
> plot(samara.lm, which = 1)
```



If we want to see all the predicted values and residuals, we can use the commands predict(samara.lm) and resid(samara.lm), respectively. Actually, there are four possible plots for lm objects. For instance, the second plot is the Q-Q plot:

> plot(samara.lm, which = 2)



R can also be used to obtain \hat{Y}_{est} and its estimated standard error and to obtain confidence intervals for \hat{Y}_{est} and \hat{Y}_{pred} . We do this below for the value $x_* = 1.80$ by using the predict command. We use it twice to get confidence and prediction intervals.

> predict(samara.lm, data.frame(x = 1.8), se.fit = TRUE, interval = "confidence") \$fit fit lwr upr [1,] 0.8354017 0.7857125 0.885091 \$se.fit [1] 0.02280564 \$df [1] 12 \$residual.scale [1] 0.0660539 > predict(samara.lm, data.frame(x = 1.8), se.fit = TRUE, interval = "prediction") \$fit fit lwr upr [1,] 0.8354017 0.6831462 0.9876571 \$se.fit [1] 0.02280564 \$df [1] 12 \$residual.scale [1] 0.0660539