Q2 The following data record the amount of electricity output, *y*, generated by a windmill over 25 separate fifteen minute periods, during each of which the average wind speed (in miles per hour) is recorded. The R data frame windmill contains the measured output (output) and average wind speed (velocity). Some output has been deleted.

Two predictors of output are considered:

- x_1 velocity;
- x_2 the reciprocal of velocity, that is, $x_2 = 1/x_1$,

and a simple linear regression model is fit separately using each predictor in turn.

```
1 > x1<-windmill$velocity</pre>
 2 > x2<-1/windmill$velocity
 3 > y<-windmill$output
 4 >
 5 > fit1<-lm(y~x1); summary(fit1)</pre>
 6 Coefficients:
              Estimate Std. Error t value Pr(>|t|)
 7
 8 (Intercept) 0.13796
                                   0.847
                                           0.406
                        0.16287
 9 x1
                0.23764
                           0.02421
10 ---
11 Signif. codes: 0 *** 0.001 ** 0.01 * 0.05
                                                    . 0.1
                                                                1
12
13 Residual standard error: 0.2985 on 23 degrees of freedom
14 Multiple R-squared: 0.8072,
                                  Adjusted R-squared:
                                                        0.7989
15 F-statistic:
                     on 1 and 23 DF, p-value: 1.087e-09
16
17 > fit2 < -lm(y \sim x2); summary(fit2)
18 Coefficients:
              Estimate Std. Error t value Pr(>|t|)
19
20 (Intercept) 3.04008 0.09597 31.68 < 2e-16 ***
21 x2
              -7.41975
                           0.46032
2.2 ---
23 Signif. codes: 0 *** 0.001 ** 0.01 * 0.05
                                                    . 0.1
                                                                1
24
25 Residual standard error: 0.1939 on 23 degrees of freedom
26 Multiple R-squared: 0.9187,
                                  Adjusted R-squared:
                                                        0.9151
27 F-statistic:
                    on 1 and 23 DF, p-value: 5.022e-14
```

(a) What conclusions about the predictive capability of the two predictors can be made on the basis of this output ? Make specific reference to line numbers when citing evidence to support your conclusions.

4 Marks

(b) Predict, using each model in turn, the electricity output produced if the average wind velocity in a given period is 6.5 miles per hour.

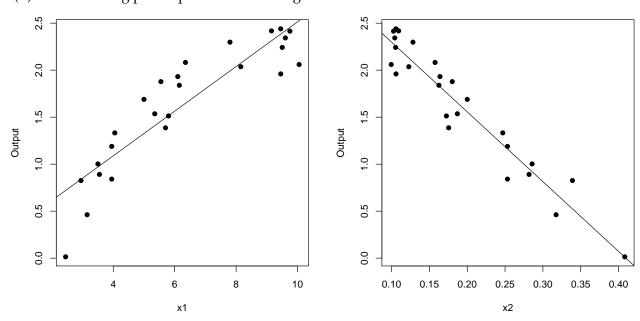
4 Marks

Question continued on the next page.

(c) For any linear regression analysis, the sums of squares decomposition

$$SS_{T} = SS_{Res} + SS_{R} \label{eq:sstars}$$

holds. Compute SS_T for the model based on x_1 , and the model based on x_2 .



(d) The following plot depicts the two straight line fits

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Q2

By considering the residuals implied by these plots, comment on the adequacy of the two straight line models.

2 Marks

4 Marks

(e) The following code computes the hat matrix, \mathbf{H} , corresponding to the analysis based on x_1 , and extracts the diagonal elements, here rounded to four decimal places.

```
28 > X<-cbind(1,x1)
29 > H<-X %*% solve(t(X) %*% X) %*% t(X)
30 > round(diag(H),4)
31 [1] 0.0414 0.0750 0.0950 0.0882 0.0433 0.0636 0.0504 0.0900 0.1135 0.1070
32 [11] 0.0454 0.0401 0.1070 0.0420 0.1202 0.0402 0.0556 0.1120 0.1354 0.0401
33 [21] 0.1091 0.1346 0.0721 0.0750 0.1036
```

What is the sum of the diagonal elements of \mathbf{H} ?

(f) The SS_T quantity from part (c) can be written

$$SS_T = \mathbf{y}^{\top} (\mathbf{I}_n - \mathbf{H}_1) \mathbf{y}$$

Define the matrix \mathbf{H}_1 , and write down its trace.

(g) Show that H_1 is symmetric and idempotent.

2 Marks

2 Marks

2 Marks

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Answer to Q2:

(a) In both models, there is good predictive ability for the response. The first analysis using x_{i1} indicates a significantly positive slope (line (9: 0.23764/0.02421 = 9.815) and an R^2 statistic of 0.8072 (line 14), which is reasonably high. The second analysis using $1/x_{i1}$ indicates a significantly negative slope (line (21: -7.41975/0.46032 = -16.118) and an even higher R^2 statistic of 0.98 (line 26), which is very high. The models indicate that generation increases with increasing wind velocity, as might be predicted. 2 Marks

(b) We have

First model : $\hat{y} = 0.13796 + 0.23764 \times 6.5 = 1.68262$ Second model : $\hat{y} = 3.04008 - 7.41975/6.5 = 1.89858$

(c) We know that

$$R^{2} = \frac{\mathrm{SS}_{\mathrm{R}}}{\mathrm{SS}_{\mathrm{T}}} = 1 - \frac{\mathrm{SS}_{\mathrm{Res}}}{\mathrm{SS}_{\mathrm{T}}} = 1 - \frac{(n-2)\widehat{\sigma}^{2}}{\mathrm{SS}_{\mathrm{T}}}$$

Therefore for the first analysis

$$SS_{Res} = 23 \times 0.2985^2 = 2.049352$$
$$SS_{T} = \frac{SS_{Res}}{(1 - R^2)} = \frac{2.049352}{1 - 0.8072} = 10.62942$$
$$SS_{R} = 10.62942 - 2.049352 = 8.580068$$

and for the second analysis

$$\begin{split} &\mathrm{SS}_{\mathrm{Res}} = 23 \times 0.1939^2 = 0.8647883 \\ &\mathrm{SS}_{\mathrm{T}} = \frac{\mathrm{SS}_{\mathrm{Res}}}{(1-R^2)} = \frac{0.8647883}{1-0.9187} = 10.63356 \\ &\mathrm{SS}_{\mathrm{R}} = 10.62942 - 0.8647883 = 9.7688 \end{split}$$

Of course, SS_T should be identical in the two models, so any difference must be due to rounding.

4 Marks

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To confirm in R: > anova(fit1) Analysis of Variance Table Response: y Df Sum Sq Mean Sq F value Pr(>F) x1 1 8.5839 8.5839 96.324 1.087e-09 *** Residuals 23 2.0496 0.0891 Signif. codes: 0 * * * 0.001 ** 0.01 0.05 0.1 > anova(fit2) Analysis of Variance Table Response: y Df Sum Sq Mean Sq F value Pr(>F) 259.81 5.022e-14 *** x2 1 9.7688 9.7688 Residuals 23 0.8648 0.0376 0.001 0.01 0.05 0.1 Signif. codes: * * * 0 **

```
4 Marks
```

- (d) The left panel implies a quadratic pattern (residuals greater than zero near the mean x as the points are above the line of best fit, below near the ends of the range of x): the model fit is deficient, indicating an incorrect conditional mean model $\mathbb{E}_{Y_i|X_i}[Y_i|\mathbf{x}_i]$ specification. In the right panel, no such pattern would be evident, so the model fit is adequate 4 Marks
- (e) The sum of the diagonal elements of **H** is always equal to p = 2 for the simple linear regression. There is no need to add the 20 numbers !

2 Marks

(f) The matrix \mathbf{H}_1 is the hat matrix from the linear regression model with only the intercept included. It is an $(n \times n)$ matrix with all elements equal to 1/n.

2 Marks

(g) \mathbf{H}_1 is clearly symmetric as all elements are identical. It is idempotent as, in the calculation of \mathbf{H}_1^2 , the (i, j)th entry is the inner product of the *i*th row and *j*th column of \mathbf{H}_1 – this equals

$$[1/n \ 1/n \cdots 1/n] [1/n \ 1/n \cdots 1/n]^{\top} = \sum_{l=1}^{n} \frac{1}{n^2} = \frac{1}{n}$$

so again each element is equal to 1/n.

2 Marks