1. Consider the following data set on $Y$ and two independent variables $x_{1}$ and $x_{2}$ :

| $Y$ | $x_{1}$ | $x_{2}$ |
| ---: | ---: | ---: |
| 5 | 1 | 1 |
| 5 | 1 | -1 |
| 6 | -1 | 1 |
| 8 | -1 | -1 |

I want you to do the following parts by hand, and show all of your work. You can use R to check your work.
(a) Consider the multiple linear regression model

$$
Y_{i}=\beta_{0}+\beta_{1} x_{i 1}+\beta_{2} x_{i 2}+\epsilon_{i}
$$

for $i=1,2,3,4$. Write this model as $\mathbf{Y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\epsilon}$.
(b) Calculate the least squares estimate $\widehat{\boldsymbol{\beta}}=\left(\mathbf{X}^{\prime} \mathbf{X}\right)^{-1} \mathbf{X}^{\prime} \mathbf{Y}$. Hint: $\mathbf{X}^{\prime} \mathbf{X}$ should be a diagonal matrix so inverting it is easy.
(c) Calculate the hat matrix $\mathbf{H}$. Calculate the fitted value vector $\widehat{\mathbf{Y}}$ and the residual vector $\mathbf{e}$. Show $\widehat{\mathbf{Y}}$ and $\mathbf{e}$ are orthogonal vectors in $\mathbb{R}^{4}$. Also, show the residuals sum to zero.
(d) Find the (estimated) variance-covariance matrix of $\widehat{\boldsymbol{\beta}}$. What is the estimated standard error of $\widehat{\beta}_{1}$ ?
2. The brake horsepower (HORSE, $Y$ ) developed by an automobile engine is thought to be a function of the engine speed in revolutions per minute (RPM, $x_{1}$ ), the road octane number of the fuel (OCT, $x_{2}$ ), and the engine compression (COM, $x_{3}$ ). An experiment is run in a laboratory at twelve different times; on each run, the temperature (TEMP, $x_{4}$ ) is also recorded. The data from the experiment are below.

| $Y$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ |
| :---: | :---: | :---: | :---: | :---: |
| 225 | 2000 | 90 | 100 | 71.2 |
| 212 | 1800 | 94 | 95 | 70.3 |
| 229 | 2400 | 88 | 110 | 72.3 |
| 222 | 1900 | 91 | 96 | 69.9 |
| 219 | 1600 | 86 | 100 | 73.2 |
| 278 | 2500 | 96 | 110 | 70.0 |
| 246 | 3000 | 94 | 98 | 70.7 |
| 237 | 3200 | 90 | 100 | 70.8 |
| 233 | 2800 | 88 | 105 | 72.1 |
| 224 | 3400 | 86 | 97 | 71.8 |
| 223 | 1800 | 90 | 100 | 71.1 |
| 230 | 2500 | 89 | 104 | 70.6 |

Consider the multiple linear regression model

$$
Y_{i}=\beta_{0}+\beta_{1} x_{i 1}+\beta_{2} x_{i 2}+\beta_{3} x_{i 3}+\beta_{4} x_{i 4}+\epsilon_{i}
$$

for $i=1,2, \ldots, 12$, or, in matrix notation,

$$
\mathbf{Y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\epsilon}
$$

(a) Use R to calculate $\mathbf{X}^{\prime} \mathbf{X},\left(\mathbf{X}^{\prime} \mathbf{X}\right)^{-1}, \mathbf{X}^{\prime} \mathbf{Y}$, and finally $\widehat{\boldsymbol{\beta}}$ as we did in the notes for the waste data in Example 12.2.
(b) Use R to calculate the hat matrix $\mathbf{H}$.
3. Consider the multiple linear regression model

$$
Y_{i}=\beta_{0}+\beta_{1} x_{i}+\beta_{2}\left(3 x_{i}^{2}-2\right)+\epsilon_{i}
$$

for $i=1,2,3$, where $x_{1}=-1, x_{2}=0$, and $x_{3}=1$.
(a) Put this model into $\mathbf{Y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\epsilon}$ form.
(b) Find the least squares estimates of $\beta_{0}, \beta_{1}$, and $\beta_{2}$. Hint: $\mathbf{X}^{\prime} \mathbf{X}$ is diagonal, so inverting this matrix is easy.
(c) Show that the least squares estimates of $\beta_{0}$ and $\beta_{1}$ are unchanged if $\beta_{2}=0$. Why do you think this happens? Hint: What do you note about the 3 column vectors of $\mathbf{X}$ in part (a)?
4. Hat matrix fun. Consider the multiple linear regression model

$$
\mathbf{Y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\epsilon}
$$

where $\mathbf{X}$ is $n \times p$. Let $\mathbf{H}=\mathbf{X}\left(\mathbf{X}^{\prime} \mathbf{X}\right)^{-1} \mathbf{X}^{\prime}$ denote the hat matrix. Let $\mathbf{I}$ denote the identity matrix that has the same dimensions as $\mathbf{H}$.
(a) What are the dimensions of $\mathbf{H}$ ?
(b) Show that $\mathbf{H}$ is symmetric and idempotent.
(c) Show that $\mathbf{I}-\mathbf{H}$ is symmetric and idempotent.
(d) Show that $\mathbf{H X}=\mathbf{X}$ and $(\mathbf{I}-\mathbf{H}) \mathbf{X}=\mathbf{0}$.
(e) Show that $\mathbf{H Y}=\mathbf{X} \widehat{\boldsymbol{\beta}}$, where $\widehat{\boldsymbol{\beta}}$ is the least-squares estimator of $\boldsymbol{\beta}$. Note that $\widehat{\mathbf{Y}}=\mathbf{X} \widehat{\boldsymbol{\beta}}$ is the vector of least-squares fitted values.
(f) Show that $(\mathbf{I}-\mathbf{H}) \mathbf{Y}=\mathbf{e}$, where $\mathbf{e}=\mathbf{Y}-\widehat{\mathbf{Y}}$ is the vector of least-squares residuals.
$(\mathrm{g})$ Show that $(\mathbf{H Y})^{\prime}(\mathbf{I}-\mathbf{H}) \mathbf{Y}=0$. What does this imply about the residuals and fitted values from a least-squares fit?
(h) Show that $(\mathbf{Y}-\mathbf{X} \widehat{\boldsymbol{\beta}})^{\prime}(\mathbf{Y}-\mathbf{X} \widehat{\boldsymbol{\beta}})=\mathbf{Y}^{\prime}(\mathbf{I}-\mathbf{H}) \mathbf{Y}$, the error (residual) sum of squares.

