1. Consider a sample of people who were asked to report their sex (Everyone answers, and everyone answers either male or female). From their answers, we construct the following matrix:

\[ X = \begin{bmatrix} 1 & M & F \end{bmatrix} \]  \hspace{1cm} (1)

In the matrix, each row corresponds to one person's response; 1 is a column of 1s, \( M \) is a column with a 1 in the \( i \)th row if the \( i \)th person answered male and a zero otherwise, and \( F \) is a column with a 1 in the \( i \)th row if person \( i \) answered female and a zero otherwise.
Is \( X'X \) invertible. Why or why not?

2. We discussed in class that variance matrixes are always positive semi-definite but not always positive definite. In the case of scalar random variables, a singular variance matrix means a zero variance — that is, the random variable is not really random at all, but fixed.

Now, let's think about what a positive semi-definite but not positive definite (hereafter PSDNPD) variance matrix means for the random vector it corresponds to. Let \( x \) be a random vector with PSDNPD variance matrix \( V \).

(a) Must there be a fixed non-zero vector \( \alpha \), such that \( \alpha'V\alpha=0 \)? Why or why not?
(b) Suppose there is such an \( \alpha \), what would the variance of \( \alpha'x \) be?
(c) What does this tell us about \( \alpha'x \)?
(d) Does this mean that \( \alpha'x = 0 \)?
(e) So, what does it mean for a random vector to have a PSDNPD variance matrix?

3. You are interested in the effect of going to the emergency room on a person's life expectancy. There is a sample of people collected on a particular day, some of whom went to an emergency room and some of whom did not. They are followed over time until they die. From each person there are two items in the dataset: \( T_i \) the length of time from the day of the survey until death and \( ER_i \), a variable which equals one if they went to the emergency room and zero if they did not. You wish to run a regression,
but the people who conducted the study will not let you have the data, only summary statistics. What should you ask for and how will you use it?

4. Considering the example in problem 3, consider two linear models:

(a) \[ T_i = \beta_1 + \beta_2 E R_i + \epsilon_i \]
(b) \[ E R_i = \beta_0 + \beta_4 T_i + \epsilon_i \]

Let’s call the OLS estimates you would get from regressions run on these equations \( \hat{\beta}_{1,OLS}, \hat{\beta}_{2,OLS} \), … What relationship will there be between \( \hat{\beta}_{2,OLS}, \hat{\beta}_{4,OLS} \)? What sign (±) do you expect for \( \hat{\beta}_{2,OLS}, \hat{\beta}_{4,OLS} \)?

5. We usually interpret regression models as embodying some causal story (the RHS causing the LHS); however, OLS does not “know” this: it produces estimates regardless of whether our implicit causal story is right. Discuss this fact in the context of the models in question 4 and your expectations about the signs of the relevant coefficients.

6. You may recall that, in the bivariate regression model,

\[ Y = \beta_1 + \beta_2 X_2 + \epsilon \] (2)

the OLS estimate of \( \beta_2 \) is \( \hat{\beta}_{2,OLS} = \frac{\hat{\text{Cov}}(Y,X_2)}{\text{V}(X_2)} \). This means that if the correlation (or covariance) between \( Y \) and \( X_2 \) is positive, then \( \hat{\beta}_2 \) is positive. Consider now a slightly more complex model:

\[ Y = \beta_1 X_1 + \beta_2 X_2 + \epsilon \] (3)

Let’s suppose that \( \overline{Y} = \overline{X_1} = \overline{X_2} = 0 \). Please calculate an expression for \( \hat{\beta}_{1,OLS} \) and \( \hat{\beta}_{2,OLS} \) in terms of the variances and covariances of \( Y, X_1, X_2 \). Using your formula, discuss whether or not it can happen that \( \hat{\text{Cov}}(Y, X_1) > 0 \) while \( \hat{\beta}_{1,OLS} < 0 \). If it cannot, why not? If it can, then how? Your goal is both to construct a correct argument and to explain in intuitive English why your result holds.