## **Instructions:**

Complete this assignment within the space on your GitHub repo in a folder called Assignment\_03. In this folder, save your answers to a file called my\_A3\_functions.py, following the sample script in the folder Assignment\_03 in the course repository. When you are finished, submit your files to your repository and upload the link to Webcourses.

- 1. Follow the function design recipe to define functions for all of the following exercises. for each function, create three examples to test your functions. Record the definitions in the sample script my\_A3\_functions.py
  - (a) For Assignment 2, you wrote a function CESutility() that calculated the value of the Constant Elasticity of Substitution utility function  $u(x;y;r) = (x^r + y^r)^{\frac{1}{r}}$ . In this function, the first two arguments are x and y, respectively, and the third is r. Write an augmented version of the function called CESutility\_valid() that returns the same value as CESutility() when x and y are non-negative numbers and r is strictly positive but returns the value None otherwise. For each case of negative numbers, make your function print a message that tells the user what is wrong with the inputs.
  - (b) Extend the above function with another wrapper function CESutility\_in\_budget(x, y, r, p\_x, p\_y, w) that evaluates CESutility\_valid() when the consumer's choice of goods x and y are in budget and returns None otherwise. That is, given prices  $p_x$  and  $p_y$ , the consumer's basket of goods should cost no more than their wealth w:  $w \ge p_x x + p_y y$ . The function should also return None if any of the prices are negative or if r is not positive.
  - (c) Write a python function logit() that will calculate the logit link function

$$\ell(x; \beta_0, \beta_1) = Prob(y = 1|x) = \frac{e^{\beta_0 + x\beta_1}}{1 + e^{\beta_0 + x\beta_1}}$$

The first argument is x and the last two are  $\beta_0$  and  $\beta_1$ . For your examples, you may use the fact that  $e^{\ln a} = a$  to create examples that evaluate to fractions.

(d) The likelihood function of the logistic regression model is used to estimate coefficients in logistic regression. Logistic regression is used to model binary events, i.e. whether or not an event occurred. For each observation i, the observation  $y_i$  equals 1 if the event occurred and 0 if it did not. Build on the function from above and write a python function  $logit\_like()$  that calculates the log-likelihood of observation  $(y_i; x_i)$ . That is, it returns the log of the function  $\ell(x; \beta_0, \beta_1)$  if  $y_i = 1$  or the log of the function  $1 - \ell(x; \beta_0, \beta_1)$  if  $y_i = 0$ . This function will have four arguments  $(y_i, x_i; \beta_0, \beta_1)$  in that order.