

# Homework 1

Computation Camp

Due August 9, 2023

Submit Link: <https://classroom.github.com/a/PIR0phPT> (<https://classroom.github.com/a/PIR0phPT>)

## Question 1

Recall that the factorial function  $n!$  returns  $n! = n \times (n - 1) \times \dots \times 2 \times 1$ . Write a function that executes this called `factorial2` that uses a for loop.

## Question 2

Consider the polynomial

$$p(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n = \sum_{i=0}^n a_i x_i$$

Using `enumerate()` in your loop, write a function `p` such that `p(x, coeff)` computes the value of the polynomial given a point  $x$  and an array of coefficients `coeff`.

## Question 3

Compute an approximation for  $\pi$  using Monte Carlo. You may only use `rand()` for random number generation. Hint: If  $U$  is a bivariate uniform random variable on the unit square  $(0, 1)^2$ , then the probability that  $U$  lies in a subset  $B$  of  $(0, 1)^2$  is equal to the area of  $B$ .

## Question 4

Let the data generating process for  $y$  be

$$y = ax_1 + bx_1^2 + cx_2 + d + \sigma w$$

where  $y, x_1, x_2$  are scalars,  $a, b, c, d$  are parameters to estimate, and  $w \sim N(0, 1)$  iid.

- First, draw  $N = 50$  values for  $x_1, x_2$  from iid normal distributions.
- Then, draw a  $w$  vector for the  $N$  values to generate  $y$  for the simulated data. Use  $a = 0.1, b = 0.2, c = 0.5, d = 1.0$ , and  $\sigma = 0.1$ . Repeat draws of  $w$  until you have 200 different simulations of the  $y$  values.
- Finally, calculate ordinary least squares manually for each of the 200 simulations. Plot histograms for each parameter  $a, b, c, d$ .

## Question 5

Take a random walk starting from  $x_0 = 1$  :

$$x_{t+1} = \alpha x_t + \sigma \epsilon_{t+1},$$

where  $t = 0, \dots, t_{\max}$ . Assume that  $x_{t_{\max}} = 0$  with certainty and that  $\{\epsilon_t\}$  is drawn from an iid standard normal. Start with  $\sigma = 0.2$  and  $\alpha = 1.0$ . For a given path  $\{x_t\}$  define a first-passage time as  $T_a = \min \{t \mid x_t \leq a\}$ .

- Calculate the first-passage time  $T_0$  for 100 simulated random walks (up to  $t_{\max} = 200$ ) and plot a histogram.
- Plot the sample mean of  $T_0$  from the simulation for  $\alpha \in \{0.8, 1.0, 1.2\}$ .

## Question 6

Recall that the root of a univariate function  $f(\cdot)$  is an  $x$  such that  $f(x) = 0$ . One solution method to find local roots of smooth functions is called Newton's method. Starting with an  $x_0$  guess, a function  $f(\cdot)$  and a first derivative  $f'(\cdot)$ , the algorithm is to repeat

$$x^{n+1} = x^n - \frac{f(x^n)}{f'(x^n)}$$

until  $|x^{n+1} - x^n|$  is below some tolerance threshold. Code a function that implements Newton's method. The function should accept arguments a function `f`, its derivative `f_prime`, a starting guess `x_0`, a tolerance `tol`, and a maximum number of iterations `maxiter`. Test this function with  $f(x) = (x - 1)^3$  and another function of your choice where you can analytically find the derivative.

## Question 7

We consider the capital investment problem of an infinitely lived household with log preferences over consumption. Production is given by  $Y_t = Z_t K_t^\theta$ , where  $K_t$  is current capital,  $\theta = 0.36$ , and  $Z_t \in \{Z_g, Z_b\}$  is the period's current productivity level. Assume that capital depreciates at rate  $\delta = 0.025$ , and households discount at rate  $\beta = 0.99$ . Assume further that the states of productivity are  $Z_g = 1.25$ ,  $Z_b = 0.2$ , and that transitions between the two states are given by a two-state Markov process:

$$\Pi = \begin{bmatrix} 0.977 & 0.023 \\ 0.074 & 0.926 \end{bmatrix}$$

so, for instance,  $P(Z_{t+1} = Z_g \mid Z_t = Z_g) = 0.977$ . The household's value function  $V(K, Z)$  is given by

$$V(K, Z) = \max_{K'} \left\{ u(c) + \beta \sum_{Z'} \Pi(Z', Z) V(K', Z') \right\}$$

$$c = ZK^\theta + (1 - \delta)K - K'.$$

Set the capital grid to be from 0.01 to 45 with 1000 grid points.

A poorly-written version of this model exists in the Homework 1 repository. Update my code to include the stochastic transitions and to run faster than it currently does.