

# PHYS 3142 HW 9

Due date: 11:59 PM 17<sup>th</sup> Apr. 2022

- Submit a report that includes your results and your python scripts
- Make sure your code can run
- Write comments in your code
- If you submit the assignment after the deadline or the report is missing, you can only get at most 80% of the full marks.
- If there is any kind of plagiarism, all students involved will get zero marks.

## 1 Ferromagnetism (50 points)

In the mean-field theory of ferromagnetism, the strength of magnetization  $M$  of a ferromagnetic material like iron depends on temperature  $T$  according to the formula

$$M = \mu \tanh \frac{JM}{k_B T}, \quad (1)$$

where  $\mu$  is a magnetic moment,  $J$  is a coupling constant, and  $k_B$  is Boltzmann's constant. To simplify the expression, define  $x = M/\mu$  and  $C = \frac{\mu J}{k_B}$  so that

$$x = \tanh \frac{Cx}{T} \quad (2)$$

The magnetization  $x$  (which is in units of  $\mu$ ) can be solved using the relaxation method. Denote  $x$  after  $i$  iterations as  $x_i$ , the initial guess as  $x_0$ .

**1. a** Let  $C = 1$ . Use  $x_0 = 1$  as your initial guess. Using 50 iterations in the relaxation method at each temperature, plot the temperature dependence of  $x$  from  $T = 0.1$  to  $T = 2$ . The plot shows that equation 2 describes a phase transition.

**1. b** Following p. 10 of lecture 16, denote the accurate solution as  $x^*$ , the error for  $x_i$  is  $\epsilon_i = |x_i - x^*|$ .

Start from  $x_i - x^* = f(x_{i-1}) - x^*$  and Taylor expand  $f(x_{i-1})$  at  $x^*$ , derive the estimation of the error

$$\epsilon_i \approx \left| \frac{(x_{i-1} - x_i)^2}{2x_{i-1} - x_{i-2} - x_i} \right| \quad (3)$$

Assume that after a few iterations  $|x_i - x_{i-1}|$  is small, and the derivative at  $x^*$  can be approximated using backward difference at  $x_{i-1}$ .

$$f'(x^*) \approx f'(x_{i-1}) \approx \frac{f(x_{i-1}) - f(x_{i-2})}{x_{i-1} - x_{i-2}} \quad (4)$$

**1. c** For  $C = 1$ ,  $T = 0.9$  and  $x_0 = 1$ , first estimate  $x^*$  using 50 iterations. Plot the actual error  $|x_i - x^*|$  and the estimation from eq. 3 in the same figure. Start with  $i = 5$  up to  $i = 30$ . This shows the validity of the estimation. Also, plot the graph using log scale for the error (use `plt.semilogy()`) to show that the error decreases exponentially.

**1. d** Plot the actual error  $|x - x^*|$  for relaxation method and for over-relaxation method (with  $\omega = 1$ ) in the same figure. Estimate  $x^*$  using 50 iterations in the over-relaxation method. Other parameters are the same as in 1. c.

## 2 Wien's displacement law (50 points)

Planck's radiation law tells us that the intensity of radiation per unit area and per unit wavelength  $\lambda$  from a black body at temperature  $T$  is

$$I(\lambda) = \frac{2\pi hc^2 \lambda^{-5}}{e^{hc/(\lambda k_B T)} - 1} \quad (5)$$

Denote the  $\lambda$  that gives the maximum  $I(\lambda)$  as  $\lambda_{peak}$ . Wien's displacement law states that

$$\lambda_{peak} = \frac{b}{T} \quad (6)$$

For example, this law can be used to estimate the temperature of the sun.

Define  $x = \frac{hc}{\lambda k_B T}$  ( $x > 0$ ), omitting the proportionality constants,  $I(x) \propto f(x)$ , with

$$f(x) = \frac{x^5}{e^x - 1} \quad (7)$$

Denote the  $x$  with maximum  $I(x)$  (for  $x > 0$ ) as  $x_{peak}$ , the Wien's displacement constant is  $b = \frac{hc}{k_B x_{peak}}$ .

**2. a** Write down  $\frac{df(x)}{dx}$  analytically. Plot  $f(x)$  and  $\frac{df(x)}{dx}$  on the same figure. Use a range of  $x$  from  $x = 0.1$  to  $x = 20$ . Also plot a horizontal dashed line at  $y = 0$ .  $x_{peak}$  is the root of  $\frac{df(x)}{dx}$  near  $x = 5$ .

**2. b** Write down  $\frac{d^2f(x)}{dx^2}$  analytically. Use Newton's method to find the root of  $\frac{df(x)}{dx}$ . Estimate the error from the  $i^{\text{th}}$  iteration as  $\epsilon_i \approx |x_{i+1} - x_i|$ . Try some values of  $x_0$ , if the correct  $x_{peak}$  cannot be reached, what  $x$  do you have instead? Why the update fails for  $x = 0$ ? Using  $x_0 = 5$ , how many iterations do you need to have  $\epsilon_i < 10^{-6}$ ? Write down that  $x_i$ .

**2. c** Use Golden's ratio search to find the maximum of  $f(x)$ . Denote  $i$  as how many times you redefine  $x_1$  or  $x_4$ , and  $\epsilon_i = x_4 - x_1$ . Using  $x_1 = 0.1$  and  $x_4 = 20$  to start the search, how many

iterations do you need to have  $\epsilon_i < 10^{-6}$ ? Write down  $(x_1 + x_4)/2$  after that accuracy is reached. Make sure the result agrees with parts **a** and **b**.

**2. d** Give at least one advantage and one disadvantage for each of the two methods. You can comment on the accuracy, the stability or other aspects.

## Optional

### 3 Gradient descent method (10 points)

The numerical gradient descent method is

$$x_i = x_{i-1} - \gamma \frac{f(x_{i-1}) - f(x_{i-2})}{x_{i-1} - x_{i-2}} \quad (8)$$

You have 3 parameters,  $\gamma, x_0$  and  $x_1$ . Use the  $f(x)$  in question 2. Try different  $\gamma, x_0, x_1$  and plot all the  $x_i$  (for  $i \leq 30$ ) to determine if  $x_i$  converges or not. When you find suitable parameters, attach the plot in your report.

Plot  $\frac{d^2 f(x)}{dx^2}$ . Comment on how to choose a suitable  $\gamma$ . How to choose the sign and the magnitude of  $\gamma$ ?