Problem 1.

In the gradient descent algorithm, $\alpha>0$ is the learning rate. If α is small enough, then the function value guarantees to decrease. In practice, we may anneal α , meaning that we start from a relatively large α , but decrease it gradually.

Show that α cannot be decreased too fast. If α is decreased too fast, even if it is strictly positive, the gradient descent algorithm may not converge to the optimum of a convex function.

Hint: Show a specific loss and an annealing scheduler such that the gradient descent algorithm fails to converge to the optimum.

Suppose we have a function f(x) = wx, and we want to optimize it using L_1 loss: J(w) = |f(x) - t|. If we have only one data point in our dataset $\mathcal{D} = \{(1,0)\}$, and we want to find the value w that fits this dataset. In this case, it is easy to see that the global optimal is $w^* = 0$.

With the gradient descent algorithm, each time step t we update w with $w^{(t)} = w^{(t-1)} - \nabla_w J(w) = w^{(t-1)} - \alpha^{(t-1)} \nabla_w |w^{(t-1)}|$, where $w \neq 0$.

Suppose our gradient descent starts with $w^{(0)}=1$, and it has a small initial learning rate $\alpha^{(0)}=0.1$. Therefore, the gradient descent is converging to $w^*=0$ from the w>0 side. Thus, we have $\nabla_w|w^{(t-1)}|=1$.

Now, let us use an annealed learning rate $\alpha^{(t)}=\frac{1}{2^t}\alpha^{(0)}$. The annealed gradient descent computes $w^{(t)}$ as following

$$w^{(t)} = w^{(0)} - \alpha^{(0)} \left(\frac{1}{2^0} + \frac{1}{2^1} + \dots + \frac{1}{2^{t-1}} \right)$$
 (1)

We know that

$$\lim_{t \to \infty} \left(\frac{1}{2^0} + \frac{1}{2^1} + \dots + \frac{1}{2^{t-1}} \right) = 2 \tag{2}$$

Therefore,

$$\lim_{t \to \infty} w^{(t)} = w^{(0)} - 2\alpha^{(0)} = 0.8 > 0 \tag{3}$$

This example shows that, a decayed learning rate may prevents the gradient descent algorithm from having enough energy for finding better local/global optimums.

Problem 2

Consider the Monty Hall game in a TV show. There are three closed doors, behind which are a car and two goats placed randomly.

a) You are asked to open a door by the host. Say, you would like to open Door 1. What is the probability of getting a car?

- b) The host knows where the car is but he/she does not tell you. Instead, the host will open another door with a goat.
 - i) If the car is behind Door 2, the host can only open Door 3.
 - ii) If the car is behind Door 1, the host can open either Door 2 or Door 3. He/she will do it with equal probability.

Say, the host has opened Door 3. What is the probability of having the car behind Door 1 now? What is the probability of having the car behind Door 2 now?

Following the hint, calculating

$$P(A = 1|B = 1) = \frac{P(A=1)P(B=1|A=1)}{P(A=1)P(B=1|A=1) + P(A=2)P(B=1|A=2) + P(A=3)P(B=1|A=3)} = \frac{1/2 \cdot 1/3}{1/2 \cdot 1/3 \cdot 1 + 1/3 \cdot 0} = 1/3.$$

c) If my goal is to get the car, should I change my first choice (i.e., open Door 1 or Door 2)? Yes. The probability of winning a car if sticks to one's choice, as calculated above, is 1/3 while the probability of winning a car if switches, is 1 - 1/3 = 2/3. One can also calculate the value of P(A = 2|B = 1) following the similar calculations detailed in the hint:

$$P(A = 2|B = 1) = \frac{P(A=2)P(B=1|A=2)}{P(A=1)P(B=1|A=1) + P(A=2)P(B=1|A=2) + P(A=3)P(B=1|A=3)} = \frac{1/3 \cdot 1}{1/2 \cdot 1/3 + 1/3 \cdot 1} = 2/3.$$

Hint: Suppose Door 1 is chosen in the first stage. Consider the random variables

- $A \in \{1, 2, 3\}$ is the door that has a car
- $B \in \{0, 1\}$ is whether the host opens Door 3 with a goat

Based on Bayes' rule, we have

$$P(A = 1|B = 1) = \frac{P(A=1)P(B=1|A=1)}{P(A=1)P(B=1|A=1) + P(A=2)P(B=1|A=2) + P(A=3)P(B=1|A=3)}$$

where

- The numerator is essentially P(A = 1, B = 1).
 - \circ P(A = 1) = 1/3, the probability that the car is behind Door 1 before the game starts.
 - P(B = 1|A = 1) = 1/2, the probability that the host will open Door 3 if the car is behind Door 1.
- The denominator is essentially P(B), the probability that the host opens Door 3 with a goat.
 - The first term P(A = 1)P(B = 1|A = 1) is the same as numerator
 - For the second term P(A = 2)P(B = 1|A = 2), we have
 - P(A = 2) = 1/3
 - P(B = 1|A = 2) = 1, because if the car is behind Door 2, the host can only open Door 3.
 - For the second term P(A = 3)P(B = 1|A = 3), we have
 - P(A = 3) = 1/3
 - But P(B = 1|A = 3) = 0, because if the car is behind Door 3, there is no chance that the host can open the Door 3 with a goat, i.e., B = 1 cannot happen.

Now you can compute the probability of P(A = 1|B = 1).







Reference:

 $\frac{\text{https://en.wikipedia.org/wiki/Monty Hall problem\#:}\sim:\text{text=Before}\%20\text{the}\%20\text{host}\%20\text{opens}\%20\text{a},\%2\text{F2}\%20\%}{3\text{D}\%201\%2\text{F6}}.$

Problem 3.

Prove that the expectation is a linear system.

$$\mathbb{E}_{X \sim P(X)}[a \, f(X) \, + \, b \, g(X)] \, = \, a \, \mathbb{E}_{X \sim P(X)}[f(X)] \, + \, b \, \mathbb{E}_{X \sim P(X)}[g(X)]$$

$$\mathbb{E}_{X \sim P(X)}[a \, f(X) \, + \, b \, g(X)] \, = \, \sum_{x} P(x)[a f(x) \, + \, b g(x)] \, = \, a \, \sum_{x} P(x) f(x) \, + \, b \, \sum_{x} P(x) g(x) \, = \, a \, \mathbb{E}_{X \sim P(X)}[f(X)] \, + \, b \, \mathbb{E}_{X \sim P(X)}[g(X)]$$

END