

## Congratulations! You passed!

 Grade  
 received 100%

 Latest Submission  
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 To pass 80% or  
 higher

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1. Using Newton's method, find an approximation recursive formula for  $\sqrt{2}$ .

1 / 1 point

To help you, remember that  $\sqrt{2}$  is the positive solution for  $x^2 - 2$ , so you can use  $f(x) = x^2 - 2$ .

- ☐  $x_{k+1} = x_k - \frac{2x_k}{x_k^2 - 2}$   
☐  $x_{k+1} = \frac{x_k^2 - 2}{2x_k}$   
☐  $x_{k+1} = \frac{2x_k}{x_k^2 - 2}$   
☒  $x_{k+1} = x_k - \frac{x_k^2 - 2}{2x_k}$

 Correct

Correct! By applying the formula  $x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$  with  $f(x) = x^2 - 2$  and  $f'(x) = 2x$ , you got the right result!

2. Regarding the previous question, suppose you don't know any approximation for  $\sqrt{2}$  and only that it is a positive real number such that  $x^2 = 2$ . Which value from the list below will result in the fastest convergence?

1 / 1 point

- ☐ 4  
☐ 3  
☒ 2  
☐ The initial value does not impact in the Newton's method convergence.

 Correct

Correct! We know that  $\sqrt{2}$  is a number between 1 and 2, so 2 is the closest value in this list of options, therefore is the value that will converge faster!

3. Let's continue investigating the method we are developing to compute the  $\sqrt{2}$ . Remember that we used the fact that  $\sqrt{2}$  is one of the roots of  $x^2 - 2$ . What would happen if we have chosen a negative value as initial point?

1 / 1 point

- ☐ The algorithm would not converge.  
☐ The algorithm would converge to  $\sqrt{2}$ .  
☒ The algorithm would converge to the negative root of  $x^2 - 2$ .  
☐ The algorithm would converge to 0.

 Correct

Correct! Any negative number will be closer to  $-\sqrt{2}$  instead of  $\sqrt{2}$ !

4. Did you know that it is possible to calculate the *reciprocal* of any number *without performing division*? (The reciprocal of a non-zero real number  $a$  is  $\frac{1}{a}$ ).

1 / 1 point

Setting a non-zero real number  $a$ , use the function  $f(x) = a - \frac{1}{x} = a - x^{-1}$  to find such formula.

This method was in fact used in older IBM computers to implement division in hardware!

So, the iteration formula to find the reciprocal of  $a$ , in this case, is:

- ☒  $x_{k+1} = 2x_k - ax_k^2$   
☐  $x_{k+1} = 2x_k + ax_k^2$   
☐  $x_{k+1} = 2x_k - x_k^2$   
☐  $x_{k+1} = x_k - ax_k^2$

 Correct

Correct! By applying the Newton's method formula with function  $f(x) = a - \frac{1}{x} = a - x^{-1}$  and  $f'(x) = \frac{1}{x^2}$  and some manipulations, you got the result!

5. Suppose we want to find the minimum value (suppose we already know that the minimum exists and is unique) of  $x \log(x)$  where  $x \in (0, +\infty)$ . Using Newton's method, what recursion formula we must use?

1 / 1 point

Hint:  $f(x) = x \log(x)$ ,  $f'(x) = \log(x) + 1$  and  $f''(x) = \frac{1}{x}$

- ☐  $x_{k+1} = x_k - \frac{x_k \log(x_k)}{\log(x_k) + 1}$
- ☐  $x_{k+1} = x_k - x_k^2 \log(x_k)$
- ☐  $x_{k+1} = x_k - \log(x_k)$
- ☒  $x_{k+1} = x_k - x_k (\log(x_k) + 1)$

✓ Correct

Correct! By applying the formula  $x_{k+1} = x_k - \frac{f'(x_k)}{f''(x_k)}$  you got the result!

6. Regarding the *Second Derivative Test* to decide whether a point with  $f'(x) = 0$  is a local minimum or local maximum, check all that apply.

1 / 1 point

- ☐ If  $f''(x) < 0$  then  $x$  is a local minimum.
- ☒ If  $f''(x) > 0$  then  $x$  is a local minimum.

✓ Correct

Correct! If  $f'(x) = 0$  and  $f''(x) < 0$  then  $x$  is a local maximum!

- ☐ If  $f''(x) = 0$  then  $x$  is an inflection point.
- ☒ If  $f''(x) = 0$  then the test is inconclusive.

✓ Correct

Correct! If  $f'(x) = f''(x) = 0$ , then the test is inconclusive!

7. Let  $f(x, y) = x^2 + y^3$ , then the Hessian matrix,  $H(x, y)$  is:

1 / 1 point

- ☐  $H(x, y) = \begin{bmatrix} 2x & 3y^2 \\ 3y^2 & 2x \end{bmatrix}$
- ☒  $H(x, y) = \begin{bmatrix} 2 & 0 \\ 0 & 6y \end{bmatrix}$
- ☐  $H(x, y) = \begin{bmatrix} 0 & 2 \\ 6y & 0 \end{bmatrix}$
- ☐  $H(x, y) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

✓ Correct

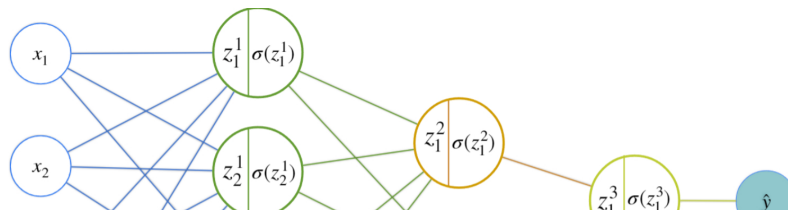
Correct! Using the formula  $H(x, y) = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix}$  it is straightforward to obtain the result!

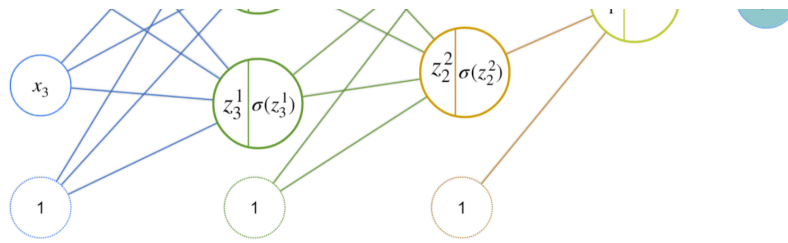
8. How many parameters has a Neural Network with:

1 / 1 point

- Input layer of size 3
- One hidden layer with 3 neurons
- One hidden layer with 2 neurons
- Output layer with size 1

An image is provided below:





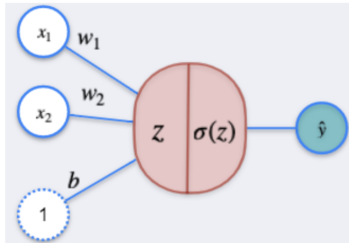
- ☐ 11  
☐ 8  
☒ 23  
☐ 3

✓ Correct

Correct! There are  $3 \cdot 3 + 3 = 12$  parameters in the first hidden layer,  $3 \cdot 2 + 2 = 8$  parameters in the second hidden layer and  $2 + 1 = 3$  parameters in the output layer!

9. Given the following Single Layer Perceptron with Sigmoid function as activation function, and log-loss as Loss Function ( $L$ ), the value for  $\frac{\partial L}{\partial w_1}$  is:

1 / 1 point



- ☐  $-(y - \hat{y})$   
☒  $-(y - \hat{y})x_1$   
☐  $-(y - \hat{y})x_2$   
☐ 1

✓ Correct

Correct! As you saw in the lecture [Classification with Perceptron](#), the value is  $-(y - \hat{y})x_1$

10. Suppose you have a function  $f(x, y)$  with  $\nabla f(x_0, y_0) = (0, 0)$  and such that

1 / 1 point

$$H(x_0, y_0) = \begin{bmatrix} 2 & 0 \\ 0 & 10 \end{bmatrix}$$

Then the point  $(x_0, y_0)$  is a:

- ☐ Local maximum.  
☒ Local minimum.  
☐ Saddle point.  
☐ We can't infer anything with the given information.

✓ Correct

Correct! The matrix in that point has two positive eigenvalues, therefore it is a local minimum!