

# Deep Learning – Summer Semester 2024 Exercise for Lecture – Mathematical Foundation

## Task 1 (Differentiation)

Calculate the first-order derivative f'(x) in each case:

- a)  $f(x) = 5x^4$
- b)  $g(x) = 3\sin(x)^2$
- c)  $z(x) = \log(\sqrt{e^{2x}})$

#### Task 2 (Tangent Hyperbolicus)

The hyperbolic tangent

$$tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

is an example of a (non-linear) activation function frequently used in deep learning applications.

- a) Draw the graph of the function
- b) Calculate the first-order derivative tanh'(x)
- c) Show that  $tanh'(x) = 1 (tanh(x))^2$

### Task 3 (Sigmoid Activation Function)

The sigmoid function

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

is an example of a (non-linear) activation function frequently used in deep learning applications.

- a) Draw the graph of the function
- b) Calculate the first-order derivative  $\sigma'(x)$
- c) Show that  $\sigma'(x) = \sigma(x)(1 \sigma(x))$

#### Task 4 (Convex Functions)

a) Decide whether the following functions are convex or not:

$$f(x) = |x|$$
,  $g(x) = \sin(x)$ ,  $h(x) = e^x$ 

- b) Let f and g be convex functions. Show that  $h(x) := \max(f(x), g(x))$  is also convex
- c) Conclude from subtask a) that the ReLU activation function ReLU(x) = max(0, x) is convex
- d) Let f and g be convex functions. Show that f + g is also convex
- e) Let f be a convex function and h a linear function. Show that  $f \circ h$  is convex.

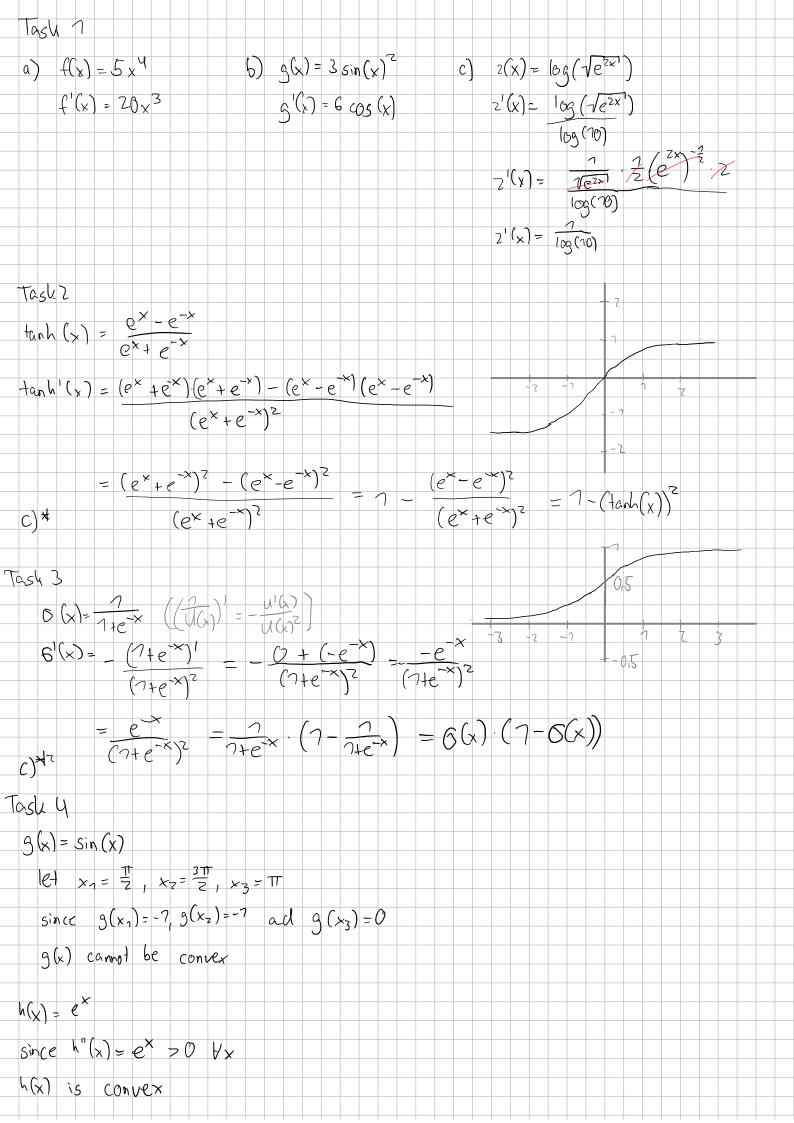
#### **Task 5 (Logistic Regression)**

The loss function in logistic regression has the form

$$L(\boldsymbol{\theta}) = -\sum_{i=1}^{m} y^{(i)} \log(\sigma(\boldsymbol{\theta}^{T} \boldsymbol{x}^{(i)})) + (1 - y^{(i)}) \log(1 - \sigma(\boldsymbol{\theta}^{T} \boldsymbol{x}^{(i)}))$$

where  $\sigma(x) = \frac{1}{1+e^{-x}}$  is the sigmoid function and  $\theta = (\theta_0, \dots, \theta_p)^T$  denotes the weights of the model. The convention  $x_0^{(i)} := 1$  is used.

- a) Show that  $f_1(x) = -\log(\sigma(x))$  and  $f_2(x) = -\log(1 \sigma(x))$  are convex. Calculate their second derivatives
- b) Conclude from a) that L is a convex function in the weights  $\theta$
- c) Calculate the partial derivatives of L according to the weights  $\theta_0, \dots, \theta_p$  and describe the gradient  $\nabla L_{\theta}$  in vectorized form. Use the representation from task 3c



\* Tash 2c tanh(x) = Stanh(x)  $\tanh(x) = \int 1 - (\tanh(x))^2 dx$ tanh (x) 2 + sech (x) 2 > 7 =  $\int sech(x)^2 dx$ = tanh(x) + C A2 Task 3c  $\sigma(x) = \int \sigma(x) (1 - \sigma(x)) dx$  $= \int_{0}^{\infty} \sigma(x) - \sigma(x)^{2} dx$ = 1 7+ex - (7+e-x) dx  $= \int \frac{1 - \frac{1}{1 + e^{x}}}{(1 + e^{x})} dx$  $= \int \frac{e^{x}}{(e^{x}+1)^{2}} dx \qquad |u=e^{x}+7| \rightarrow du=e^{x} dx$ = 1 1/2 du  $= -\frac{1}{e^{x}+1} = \frac{1}{e^{-x}+1}$ Task 5 a) Show fr=-log(o(x)) and fz(x)=-log(7-o(x)) are convex with 0(x)= 1+0-x  $f_1(x) - \log(\frac{1}{1+e^{-x}}) = -(\log(1) - \log(1+e^{-x})) = \log(1+e^{-x})$  $f_{1}(x) = -\left(\frac{1}{1+e^{-x}}\right) \cdot -\left(1+e^{-x}\right)^{-2} \cdot \left(-e^{-x}\right) = -\frac{e^{-x}}{1+e^{-x}}$  $f_{1}(x) = \left(\frac{1}{1+e^{-x}} \cdot \left(-e^{-x}\right)\right) = -\left(\frac{1}{1+e^{-x}} \cdot \left(-e^{-x}\right) \cdot \left(-e^{-x}\right) + \frac{1}{1+e^{-x}} \cdot e^{-x}$  $= \frac{(-e^{x}) \cdot (-e^{-x})}{(7+e^{-x})^2} + \frac{e^{-x}}{7+e^{-x}}$ 

