$$\frac{\partial G_{1}(x)}{\partial x} = \frac{-(-e^{2x})}{(1+e^{2x})^{2}} = \frac{1}{1+e^{2x}} \left(1 - \frac{1}{1+e^{2x}}\right) = G_{1}(x) \left(1 - S_{1}(x)\right)$$

$$\frac{\partial 6_{2}(\pi)}{\partial \pi} = \frac{(e^{\pi} + e^{\pi})^{2} - (e^{\pi} - e^{\pi})^{2}}{(e^{\pi} + e^{\pi})^{2}} = 1 - \left(\frac{e^{\pi} - e^{\pi}}{e^{\pi} + e^{\pi}}\right)^{2} = 1 - 8_{2}(\pi)$$

The forward pass values are written in the graph

loss= 
$$(f(x)-g)^2 = (-0.73-1)^2 = 2.99$$
  
 $\partial L = [-2] = (f(x)-y) = 2[f(x)-y]$   
 $\partial L = [-2] = 2[f(x)-y]$   
 $\partial S = [-8(2)]$ 

The backward pass is also denoted on the graph.

The upstream gradient \* local gradient = down Stream gredient

is used for calculations base on the slides.

The final gredient with respect to each variable is highlighter

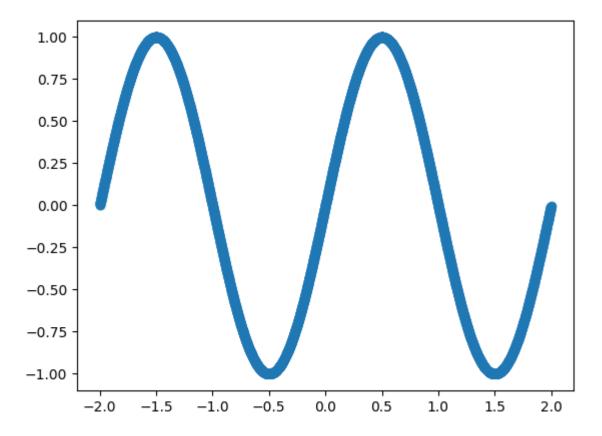
$$J=1 \implies \begin{cases} 011 - 1011 = -1 - (-0.23) = -0.77 \\ 011 = 01 - 1001 = 1 - (0.23) = 0.77 \\ 012 = 02 - 1001 = -2 - (-1.62) = -0.38 \\ 012 = 02 - 1001 = 1 - (-1.62) = 2.69 \end{cases}$$

### Task 3

```
In []: # First we import some necessary libraries
    import tensorflow as tf
    import numpy as np
    import matplotlib.pyplot as plt
    from sklearn.model_selection import train_test_split

In []: # The following function creates the sine wave data
    def sine_wave(i):
        X = np.arange(-2,2,0.001)
        y = np.sin(X* i * np.pi/4)
        return X,y

In []: # Plot a sample of the sine wave function
        X, y = sine_wave(4)
        plt.scatter(X, y)
        plt.show()
```

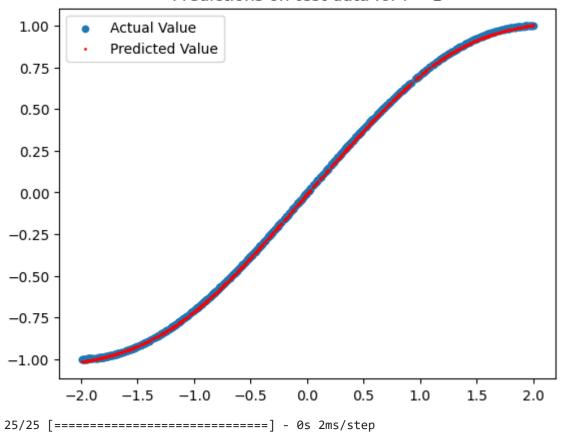


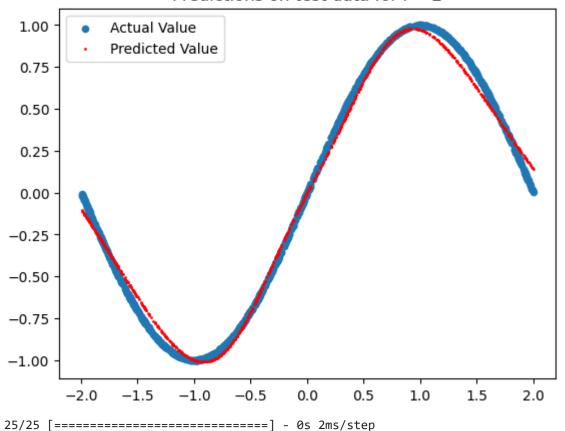
```
In []: # The following function builds a multilayer perceptron with one hidden layer

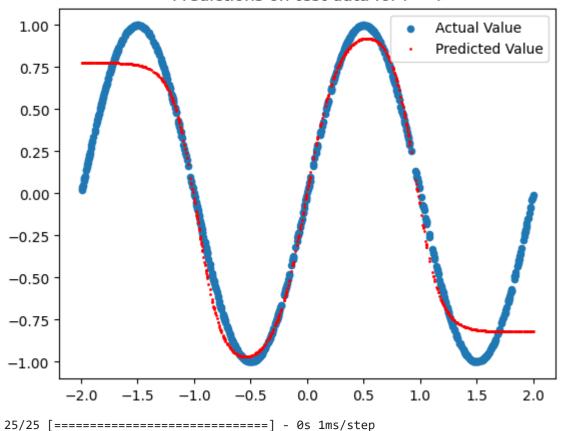
def build_network(num_hidden):
    inp = tf.keras.Input(shape=(1,))
    hidden = tf.keras.layers.Dense(num_hidden, activation='sigmoid')(inp)
    outp = tf.keras.layers.Dense(1)(hidden)
    model = tf.keras.Model(inputs=inp, outputs=outp, name='sine_model')
    return model
```

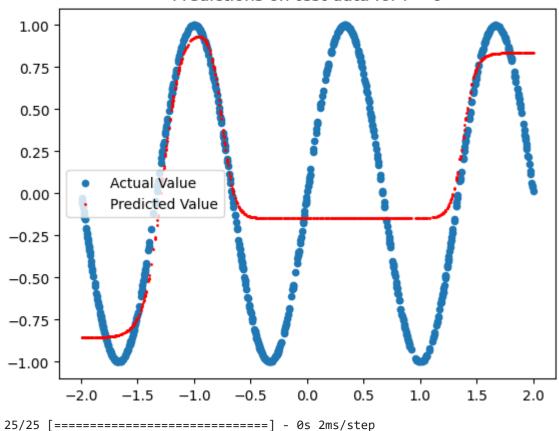
```
In []: # The following function runs training for a given set of arguments.

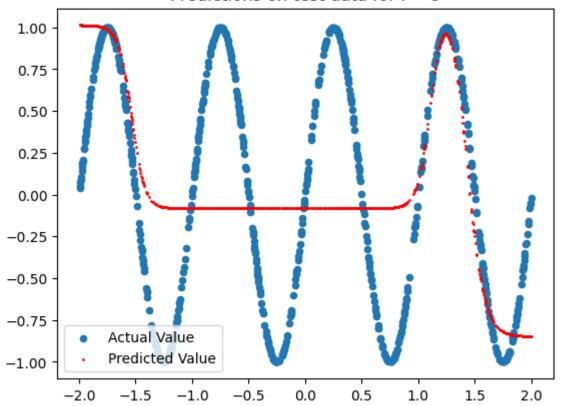
def train(num_hidden,i):
    model = build_network(num_hidden)
    X,y = sine_wave(i)
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
    model.compile(
        optimizer=tf.keras.optimizers.Adam(learning_rate=0.1),
        loss=tf.keras.losses.MeanSquaredError()
```





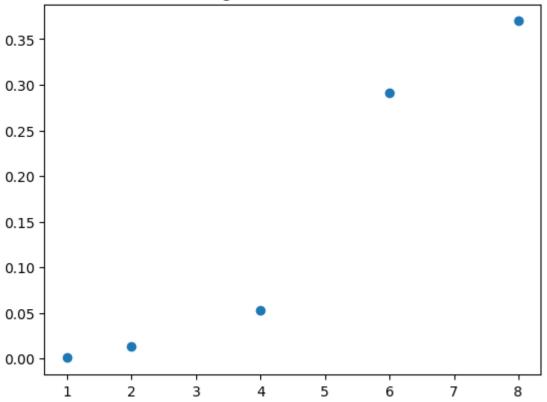






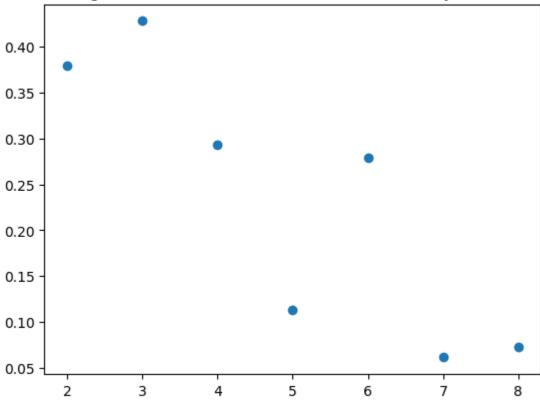
```
In [ ]: fig, ax = plt.subplots()
    ax.scatter([1, 2, 4, 6, 8,], avgs)
    ax.set_title(f'Average error based on the i')
    plt.show()
```

### Average error based on the i



```
In [ ]: ranges = range(2, 9)
    avgs = []
    for i in ranges:
        average_training_loss, X_test, predictions, y_test = train(i, 6)
        avgs.append(average_training_loss)
    fig, ax = plt.subplots()
    ax.scatter(ranges, avgs)
    ax.set_title(f'Average error based on the number of hidden layer neurons')
    plt.show()
```

### Average error based on the number of hidden layer neurons



In [ ]: