```
In [1]: # Import Libraries
         import numpy as np
         import pandas as pd
         from sklearn.datasets import load iris
         from sklearn.model_selection import train_test_split
         import matplotlib.pyplot as plt
In [2]: # Load dataset
         data = pd.read_csv('iris.csv')
         data.head()
Out[2]:
            sepal.length sepal.width petal.length petal.width variety
                                                   0.2 Setosa
         0
                   5.1
                              3.5
                                         1.4
         1
                   4.9
                                                   0.2 Setosa
                              3.0
                                         1.4
         2
                   4.7
                              3.2
                                         1.3
                                                   0.2 Setosa
         3
                   4.6
                              3.1
                                         1.5
                                                   0.2 Setosa
                   5.0
                              3.6
                                                   0.2 Setosa
                                         1.4
In [3]: | data.shape
Out[3]: (150, 5)
In [4]:
        X = data[['sepal.length', 'sepal.width', 'petal.length', 'petal.width']].values
         #label_to_int = {label: idx for idx, label in enumerate(df['variety'].unique())}
         # Create the target vector 'target'
        y = data['variety'].values
In [5]: # Get dummy variable
        y = pd.get_dummies(y).values
        y[:3]
Out[5]: array([[1, 0, 0],
                [1, 0, 0],
                [1, 0, 0]], dtype=uint8)
In [6]: #Split data into train and test data
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=20, random_state=
```

```
In [7]: # Initialize variables
        learning_rate = 0.1
        iterations = 5000
        N = y_train.size
        # number of input features
        input_size = 4
        # number of hidden layers neurons
        hidden_size = 2
        # number of neurons at the output layer
        output_size = 3
        results = pd.DataFrame(columns=["mse", "accuracy"])
In [8]: # Initialize weights
        np.random.seed(10)
        # initializing weight for the hidden layer
        W1 = np.random.normal(scale=0.5, size=(input_size, hidden_size))
        # initializing weight for the output layer
        W2 = np.random.normal(scale=0.5, size=(hidden_size , output_size))
In [9]: def sigmoid(x):
            return 1 / (1 + np.exp(-x))
        def mean_squared_error(y_pred, y_true):
            return ((y_pred - y_true)**2).sum() / (2*y_pred.size)
        def accuracy(y_pred, y_true):
            acc = y_pred.argmax(axis=1) == y_true.argmax(axis=1)
            return acc.mean()
```

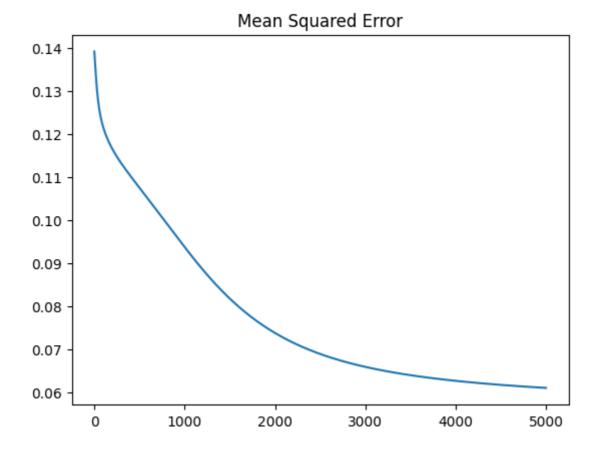
```
In [10]: | for itr in range(iterations):
             # feedforward propagation
             # on hidden layer
             Z1 = np.dot(X_train, W1)
             A1 = sigmoid(Z1)
             # on output layer
             Z2 = np.dot(A1, W2)
             A2 = sigmoid(Z2)
             # Calculating error
             mse = mean_squared_error(A2, y_train)
             acc = accuracy(A2, y_train)
             results=results.append({"mse":mse, "accuracy":acc},ignore_index=True )
             # backpropagation
             E1 = A2 - y_train
             dW1 = E1 * A2 * (1 - A2)
             E2 = np.dot(dW1, W2.T)
             dW2 = E2 * A1 * (1 - A1)
             # weight updates
             W2\_update = np.dot(A1.T, dW1) / N
             W1_update = np.dot(X_train.T, dW2) / N
             W2 = W2 - learning_rate * W2_update
             W1 = W1 - learning rate * W1 update
           results=results.append({"mse":mse, "accuracy":acc},ignore_index=True )
         C:\Users\Dell\AppData\Local\Temp\ipykernel_10888\548591417.py:16: FutureWarning:
         The frame.append method is deprecated and will be removed from pandas in a future
         version. Use pandas.concat instead.
           results=results.append({"mse":mse, "accuracy":acc},ignore_index=True )
```

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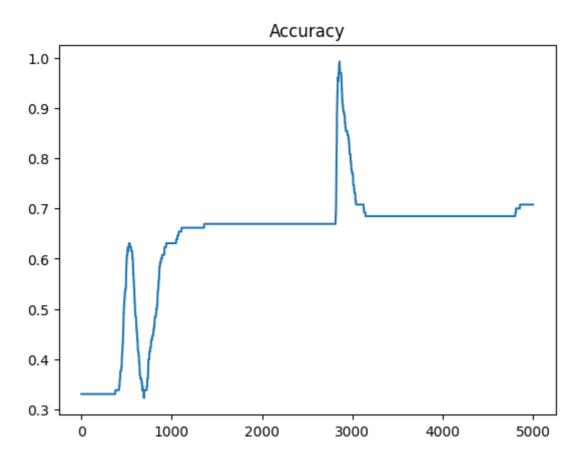
```
In [11]: results.mse.plot(title="Mean Squared Error")
```

Out[11]: <AxesSubplot: title={'center': 'Mean Squared Error'}>



In [12]: results.accuracy.plot(title="Accuracy")

Out[12]: <AxesSubplot: title={'center': 'Accuracy'}>



```
In [13]: # feedforward
    Z1 = np.dot(X_test, W1)
    A1 = sigmoid(Z1)

    Z2 = np.dot(A1, W2)
    A2 = sigmoid(Z2)

acc = accuracy(A2, y_test)
    print("Accuracy: {}".format(acc))
```

Accuracy: 0.8