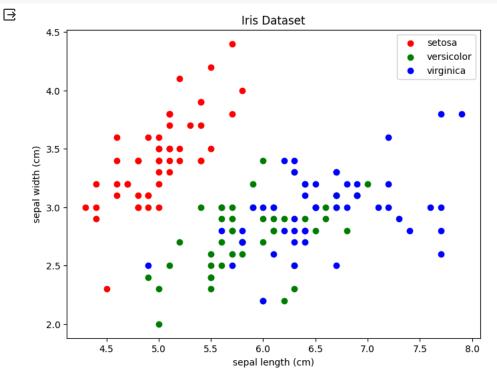
```
iris flower sepration using ANN/FUZZY/MEMBERSHIP/GENETIC ALGO.ipynb - Colab
5/4/24, 6:53 PM
    !pip install -U scikit-fuzzy
    !pip install deap
          Requirement already satisfied: scikit-fuzzy in /usr/local/lib/python3.10/dist-packages (0.4.2)
Requirement already satisfied: numpy>=1.6.0 in /usr/local/lib/python3.10/dist-packages (from scikit-fuzzy) (1.25.2)
           Requirement already satisfied: scipy>=0.9.0 in /usr/local/lib/python3.10/dist-packages (from scikit-fuzzy) (1.11.4)
          Requirement already satisfied: networkx>=1.9.0 in /usr/local/lib/python3.10/dist-packages (from scikit-fuzzy) (3.3) Requirement already satisfied: deap in /usr/local/lib/python3.10/dist-packages (1.4.1)
          Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from deap) (1.25.2)
    import numpy as np
    import matplotlib.pyplot as plt
    import skfuzzy as fuzz
    from deap import base, creator, tools, algorithms
    import tensorflow as tf
    from sklearn.datasets import load iris
     from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import OneHotEncoder
    # Load Iris dataset
```

```
iris = load iris()
X = iris.data
y = iris.target
\verb|plot_iris_data(X, y)| # Visualize the Iris dataset|
```



```
#again reshape v axis for training/testing
iris = load_iris()
X = iris.data
y = iris.target.reshape(-1, 1)
# One-hot encode the target variable
```

```
encoder = OneHotEncoder(sparse output=False)
y_onehot = encoder.fit_transform(y)
# Split data into train and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y_onehot, test_size=0.2, random_state=42)
```

```
#(ANN) model
def create_ann_model():
    model = tf.keras.Sequential([
        {\tt tf.keras.layers.Dense(10, activation='relu', input\_shape=(4,)),}\\
        tf.keras.layers.Dense(10, activation='relu')
        tf.keras.layers.Dense(3, activation='softmax')
    model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

```
# Fuzzy Logic model
def create_fuzzy_logic_model():
    # Define input and output variables
    sepal_length_universe = np.linspace(4, 8, 100)
    sepal_width_universe = np.linspace(2, 4.5, 100)
    petal_length_universe = np.linspace(1, 7, 100)
    petal_width_universe = np.linspace(0, 2.5, 100)
    #fuzzy membership functions
    sepal_length_fuzzy = fuzz.gaussmf(sepal_length_universe, mean=5.5, sigma=1)
    sepal_width_fuzzy = fuzz.gaussmf(sepal_width_universe, mean=3.5, sigma=0.5)
    petal_length_fuzzy = fuzz.gaussmf(petal_length_universe, mean=4, sigma=1)
    petal_width_fuzzy = fuzz.gaussmf(petal_width_universe, mean=1.2, sigma=0.3)
    # Fuzzy rules
    rule1 = np.fmin(sepal_length_fuzzy, sepal_width_fuzzy)
rule2 = np.fmin(petal_length_fuzzy, petal_width_fuzzy)
    rule3 = np.fmin(rule1, rule2) # Combine rules AND operator
    # Aggregation
    aggregated = np.fmax(rule1, rule2, rule3)
    # Defuzzification (weighted average based on model)
    class_weights = np.argmax(y\_train, axis=1) + 1
    predicted_class = np.average(np.arange(3), weights=aggregated, axis=0)
    return predicted class
# Visualization fuzzy membership functions
def plot_fuzzy_memberships():
    sepal_length_universe = np.linspace(4, 8, 100)
sepal_length_fuzzy = fuzz.gaussmf(sepal_length_universe, mean=5.5, sigma=1)
plot_fuzzy_memberships(sepal_length_universe, sepal_length_fuzzy)
    plt.figure(figsize=(8, 6))
    plt.plot(sepal_length_universe, sepal_length_fuzzy, label='Sepal Length')
    plt.xlabel('Sepal Length')
    plt.ylabel('Membership Degree')
plt.title('Fuzzy Membership Functions')
    plt.legend()
# Genetic Algorithm
def evaluate(individual):
    # Evaluate fitness
    ann_weights = []
    ann_biases = []
    for layer in ann_model.layers:
         weights_shape = layer.get_weights()[0].shape
biases_shape = layer.get_weights()[1].shape
         num_weights = np.prod(weights_shape)
         num_biases = np.prod(biases_shape)
         ann_weights.append(individual[start:start + num_weights].reshape(weights_shape))
         ann\_biases.append(individual[start + num\_weights:start + num\_weights + num\_biases].reshape(biases\_shape))
         \verb|start += num_weights + num_biases|\\
    ann model.set weights(ann weights + ann biases)
    y_pred_ann = ann_model.predict(X_train)
    y_pred_
```