## Program 3: Program to build a neural network with a single hidden layer using TensorFlow.

Aim: The aim of this program to build a neural network with a single hidden layer using TensorFlow. Procedure:import tensorflow as tf import numpy as np # Step 1: Prepare the Data # Generate some example data x train = np.array([[1], [2], [3], [4], [5]], dtype=np.float32) # Input features  $y_{train} = np.array([[2], [4], [6], [8], [10]], dtype=np.float32) # Target outputs (y = 2x)$ # Step 2: Build the Model model = tf.keras.Sequential([ tf.keras.layers.Dense(10, activation='relu', input shape=(1,)), # Single hidden layer with 10 neurons tf.keras.layers.Dense(1) # Output layer with 1 neuron for regression ]) # Step 3: Compile the Model model.compile(optimizer='adam', loss='mean squared error') # Step 4: Train the Model print("Training the model...") model.fit(x train, y train, epochs=100, verbose=0) # Train for 100 epochs print("Model training complete.") # Step 5: Test the Model x test = np.array([[6]], dtype=np.float32) # Test input predicted y = model.predict(x test) print(f"Predicted y for  $x=\{x \text{ test}[0][0]\}$ : {predicted y[0][0]}") output:-Training the model... Model training complete. =] - 0s 22ms/step

Predicted y for x=6.0: 12.0

## Program 4- Build 3 networks, each with atleast 10 hidden layers in deep learning.

**Aim:**-The aim of this program to build 3 networks, each with atleast 10 hidden layers in deep learning

## Procedure:-

# Model 3: Multiclass Classification

```
import tensorflow as tf
import numpy as np
# Step 1: Define helper function to build a model
def build deep network(input shape, num hidden layers=10, units per layer=32,
activation='relu', output units=1, output activation=None):
  model = tf.keras.Sequential()
  # Input Laver
  model.add(tf.keras.layers.Dense(units per layer, activation=activation,
input shape=input shape))
  # Hidden Layers
  for in range(num hidden layers - 1):
    model.add(tf.keras.layers.Dense(units per layer, activation=activation))
  # Output Layer
  model.add(tf.keras.lavers.Dense(output units, activation=output activation))
  return model
# Step 2: Generate synthetic data for demonstration
x train = np.random.rand(1000, 10) \# 1000 samples, 10 features
y train reg = np.sum(x train, axis=1) # Regression target: sum of inputs
y train binary = (y train reg > 5).astype(int) # Binary classification target: sum > 5
y train multiclass = np.random.randint(0, 3, size=(1000,)) # Multiclass classification: 3 classes
# Step 3: Build and train three models
# Model 1: Regression
print("Training Model 1: Regression")
model regression = build deep network(input shape=(10,), output units=1,
output activation=None)
model regression.compile(optimizer='adam', loss='mean squared error', metrics=['mse'])
model_regression.fit(x_train, y_train_reg, epochs=10, batch_size=32, verbose=1)
# Model 2: Binary Classification
print("\nTraining Model 2: Binary Classification")
model binary = build deep network(input shape=(10,), output units=1,
output activation='sigmoid')
model binary.compile(optimizer='adam', loss='binary crossentropy', metrics=['accuracy'])
model binary.fit(x train, y train binary, epochs=10, batch size=32, verbose=1)
```

```
print("\nTraining Model 3: Multiclass Classification")
model_multiclass = build_deep_network(input_shape=(10,), output_units=3,
output_activation='softmax')
model_multiclass.compile(optimizer='adam', loss='sparse_categorical_crossentropy',
metrics=['accuracy'])
model_multiclass.fit(x_train, y_train_multiclass, epochs=10, batch_size=32, verbose=1)
# Step 4: Model Summary
print("\nModel Summaries:")
print("\nModel 1 Summary:")
model_regression.summary()
print("\nModel 2 Summary:")
model_binary.summary()
print("\nModel 3 Summary:")
model multiclass.summary()
```