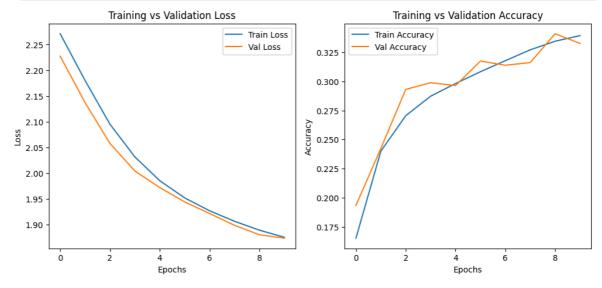
```
In [1]: import tensorflow as tf
        from tensorflow import keras
        from keras.models import Sequential
        from keras.layers import Dense, Flatten
        from keras.optimizers import SGD
        from keras.utils import to_categorical
        import matplotlib.pyplot as plt
In [2]: # =========
        # 1. Load Dataset (CIFAR-10)
        # =========
         (x_train, y_train), (x_test, y_test) = keras.datasets.cifar10.load_data()
        # Normalize pixel values (0-1)
        x_train = x_train.astype("float32") / 255.0
        x_test = x_test.astype("float32") / 255.0
        # One-hot encode labels
        y_train = to_categorical(y_train, 10)
        y_test = to_categorical(y_test, 10)
In [3]: # ==========
        # 2. Define Model Architecture
        # ========
        model = Sequential([
            Flatten(input_shape=(32, 32, 3)), # Flatten 32x32x3 image
Dense(512, activation='sigmoid'), # Hidden Layer
Dense(256, activation='sigmoid'), # Hidden Layer
Dense(10, activation='softmax') # Output Layer (10 class)
                                                     # Output layer (10 classes)
        ])
       C:\Users\aryan\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\sr
       c\layers\reshaping\flatten.py:37: UserWarning: Do not pass an `input shape`/`inpu
       t_dim` argument to a layer. When using Sequential models, prefer using an `Input
       (shape)` object as the first layer in the model instead.
         super().__init__(**kwargs)
In [4]: # =========
        # 3. Compile Model (SGD optimizer)
        # =========
        model.compile(optimizer=SGD(learning rate=0.01),
                       loss='categorical_crossentropy',
                       metrics=['accuracy'])
In [5]: # =========
        # 4. Train Model
        # =========
        history = model.fit(x_train, y_train,
                             validation_data=(x_test, y_test),
                             epochs=10,
                             batch_size=64)
```

```
Epoch 1/10
                  16s 19ms/step - accuracy: 0.1341 - loss: 2.3053 - va
      782/782 ----
      l_accuracy: 0.1933 - val_loss: 2.2271
      Epoch 2/10
      782/782 -----
                           14s 18ms/step - accuracy: 0.2291 - loss: 2.2023 - va
      l_accuracy: 0.2423 - val_loss: 2.1370
      Epoch 3/10
                                - 20s 18ms/step - accuracy: 0.2619 - loss: 2.1154 - va
      782/782 -
      l_accuracy: 0.2931 - val_loss: 2.0577
      Epoch 4/10
                           20s 18ms/step - accuracy: 0.2859 - loss: 2.0416 - va
      782/782 -
      l accuracy: 0.2989 - val loss: 2.0043
      Epoch 5/10
      782/782 — 21s 18ms/step - accuracy: 0.2923 - loss: 1.9999 - va
      l_accuracy: 0.2964 - val_loss: 1.9721
      Epoch 6/10
                            ----- 15s 19ms/step - accuracy: 0.3045 - loss: 1.9616 - va
      l_accuracy: 0.3175 - val_loss: 1.9443
      Epoch 7/10
      782/782 -
                            15s 19ms/step - accuracy: 0.3172 - loss: 1.9353 - va
      1_accuracy: 0.3139 - val_loss: 1.9219
      Epoch 8/10
                           ----- 23s 21ms/step - accuracy: 0.3262 - loss: 1.9099 - va
      782/782 -
      l_accuracy: 0.3162 - val_loss: 1.8989
      Epoch 9/10
                          15s 19ms/step - accuracy: 0.3266 - loss: 1.8989 - va
      l_accuracy: 0.3410 - val_loss: 1.8806
      Epoch 10/10
                           14s 18ms/step - accuracy: 0.3386 - loss: 1.8781 - va
      782/782 -
      l accuracy: 0.3327 - val loss: 1.8739
In [6]: # ===========
        # 5. Evaluate Model
        # =========
        loss, acc = model.evaluate(x_test, y_test, verbose=0)
        print(f"Test Accuracy on CIFAR-10: {acc*100:.2f}%")
      Test Accuracy on CIFAR-10: 33.27%
In [7]: # ============
       # 6. Plot Loss & Accuracy
        # =========
        plt.figure(figsize=(12,5))
        # Training vs Validation Loss
        plt.subplot(1,2,1)
        plt.plot(history.history['loss'], label="Train Loss")
        plt.plot(history.history['val_loss'], label="Val Loss")
        plt.xlabel("Epochs")
        plt.ylabel("Loss")
        plt.legend()
        plt.title("Training vs Validation Loss")
        # Training vs Validation Accuracy
        plt.subplot(1,2,2)
        plt.plot(history.history['accuracy'], label="Train Accuracy")
        plt.plot(history.history['val_accuracy'], label="Val Accuracy")
        plt.xlabel("Epochs")
        plt.ylabel("Accuracy")
        plt.legend()
```

```
plt.title("Training vs Validation Accuracy")
plt.show()
```



In []: