Accuracy Based:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.datasets import imdb
from tensorflow.keras.preprocessing.sequence import pad sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Embedding, SimpleRNN
# Setting parameters
max features = 10000 # Number of words to consider as features
maxlen = 500 # Cuts off texts after this number of words (among the max features most
common words)
batch size = 32
# Loading data
print('Loading data...')
(x train, y train), (x test, y test) = imdb.load data(num words=max features)
print(len(x train), 'train sequences')
print(len(x test), 'test sequences')
# Padding sequences to ensure uniform length
print('Pad sequences (samples x time)')
x_train = pad_sequences(x_train, maxlen=maxlen)
x \text{ test} = pad \text{ sequences}(x \text{ test, maxlen}=maxlen)
print('x_train shape:', x_train.shape)
print('x_test shape:', x_test.shape)
# Building the model
model = Sequential()
model.add(Embedding(max features, 32))
model.add(SimpleRNN(32))
model.add(Dense(1, activation='sigmoid'))
```

```
model.compile(optimizer='rmsprop', loss='binary_crossentropy', metrics=['acc'])

print(model.summary())

# Training the model

print('Training...')

history = model.fit(x_train, y_train, epochs=10, batch_size=batch_size, validation_split=0.2)

# Evaluating the model

print('Evaluating...')

loss, accuracy = model.evaluate(x_test, y_test)

print('Test Loss:', loss)

print('Test Accuracy:', accuracy)
```

OR

Input Based:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.datasets import imdb
from tensorflow.keras.preprocessing.sequence import pad sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Embedding, SimpleRNN
# Setting parameters
max features = 10000 # Number of words to consider as features
maxlen = 500 # Cuts off texts after this number of words (among the max features most
common words)
batch size = 32
# Loading data
print('Loading data...')
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
print(len(x train), 'train sequences')
print(len(x test), 'test sequences')
```

```
# Padding sequences to ensure uniform length
print('Pad sequences (samples x time)')
x train = pad sequences(x train, maxlen=maxlen)
x \text{ test} = pad \text{ sequences}(x \text{ test, maxlen}=maxlen)
print('x train shape:', x train.shape)
print('x test shape:', x test.shape)
# Building the model
model = Sequential()
model.add(Embedding(max features, 32))
model.add(SimpleRNN(32))
model.add(Dense(1, activation='sigmoid'))
model.compile(optimizer='rmsprop', loss='binary crossentropy', metrics=['acc'])
print(model.summary())
# Training the model
print('Training...')
history = model.fit(x train, y train, epochs=10, batch size=batch size, validation split=0.2)
# Function to preprocess user input
def preprocess input(text):
  word to index = imdb.get word index()
  words = text.lower().split()
  filtered words = [word to index[word] if word in word to index and
word to index[word] < max features else 0 for word in words]
  padded_sequence = pad_sequences([filtered_words], maxlen=maxlen)
  return padded_sequence
# Function to predict sentiment for user input
def predict sentiment(text):
  preprocessed text = preprocess input(text)
  prediction = model.predict(preprocessed text)
  return prediction[0][0]
# Allow user input for sentiment analysis
while True:
  user input = input("Enter a movie review (type 'exit' to quit): ")
```

```
if user_input.lower() == 'exit':
    break
else:
    sentiment = predict_sentiment(user_input)
    if sentiment > 0.5:
        print("Positive Sentiment")
    else:
        print("Negative Sentiment")
```

Simple:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, LSTM, RepeatVector
from tensorflow.keras.callbacks import ModelCheckpoint
# Generate random data for demonstration
data = np.random.rand(1000, 10, 1) # Example data: 1000 sequences of length 10 with 1
feature
# Define model architecture
latent dim = 2 # Dimensionality of the latent space
inputs = Input(shape=(10, 1))
encoded = LSTM(4)(inputs)
encoded = RepeatVector(10)(encoded) # Repeat the encoded representation 10 times
decoded = LSTM(4, return sequences=True)(encoded)
decoded = tf.keras.layers.TimeDistributed(tf.keras.layers.Dense(1))(decoded)
# Build the autoencoder model
autoencoder = Model(inputs, decoded)
# Compile the model
autoencoder.compile(optimizer='adam', loss='mse')
# Print model summary
autoencoder.summary()
# Train the model
autoencoder.fit(data, data, epochs=50, batch size=32, validation split=0.2)
# After training, you can use the encoder and decoder separately if needed
encoder = Model(inputs, encoded)
encoded input = Input(shape=(latent dim, 4))
decoder layer = autoencoder.layers[-2](encoded input)
decoder layer = autoencoder.layers[-1](decoder layer)
decoder = Model(encoded input, decoder layer)
```

Input Based:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, LSTM, RepeatVector
from tensorflow.keras.callbacks import ModelCheckpoint
# Define model architecture
latent dim = 2 # Dimensionality of the latent space
inputs = Input(shape=(10, 1))
encoded = LSTM(4)(inputs)
encoded = RepeatVector(10)(encoded) # Repeat the encoded representation 10 times
decoded = LSTM(4, return sequences=True)(encoded)
decoded = tf.keras.layers.TimeDistributed(tf.keras.layers.Dense(1))(decoded)
# Build the autoencoder model
autoencoder = Model(inputs, decoded)
# Compile the model
autoencoder.compile(optimizer='adam', loss='mse')
# Print model summary
autoencoder.summary()
# Allow user input for sequences
while True:
  user input = input("Enter a sequence of 10 numbers separated by spaces (type 'exit' to
quit): ")
  if user input.lower() == 'exit':
    break
  else:
    # Convert user input to a sequence of floats
    sequence = [float(x) for x in user input.split()]
```

```
if len(sequence) != 10:
    print("Please enter exactly 10 numbers.")
    continue
sequence = np.array(sequence).reshape(1, 10, 1) # Reshape to match model input shape
# Encode the sequence
encoded_sequence = autoencoder.predict(sequence)
# Decode the encoded sequence
decoded_sequence = autoencoder.predict(encoded_sequence)
print("Original Sequence:", sequence)
print("Encoded Sequence:", encoded_sequence)
print("Decoded Sequence:", decoded_sequence)
```

```
#!pip install tensorflow==2.8
import tensorflow as tf
from tensorflow import keras
import numpy as np
# Load MNIST dataset
(X train, \_), (\_, \_) = keras.datasets.mnist.load_data()
X_{train} = X_{train} / 127.5 - 1.0 \# Rescale images to [-1, 1]
X train = np.expand dims(X train, axis=-1)
# Generator model
generator = keras.Sequential([
  keras.layers.Dense(7 * 7 * 128, input_shape=(100,)),
  keras.layers.Reshape((7, 7, 128)),
  keras.layers.Conv2DTranspose(64, kernel size=3, strides=2, padding='same'),
  keras.layers.LeakyReLU(alpha=0.2),
  keras.layers.Conv2DTranspose(1, kernel size=3, strides=2, padding='same',
activation='tanh')
1)
# Discriminator model
discriminator = keras.Sequential([
  keras.layers.Conv2D(64, kernel_size=3, strides=2, padding='same', input_shape=(28, 28,
1)),
  keras.layers.LeakyReLU(alpha=0.2),
  keras.layers.Conv2D(128, kernel size=3, strides=2, padding='same'),
  keras.layers.LeakyReLU(alpha=0.2),
  keras.layers.Flatten(),
  keras.layers.Dense(1, activation='sigmoid')
])
# Compile discriminator
discriminator.compile(loss='binary crossentropy',
optimizer=keras.optimizers.Adam(learning rate=0.0002), metrics=['accuracy'])
# Freeze discriminator's weights during GAN training
```

```
discriminator.trainable = False
# GAN model
gan input = keras.Input(shape=(100,))
generated image = generator(gan input)
gan output = discriminator(generated image)
gan = keras.Model(gan input, gan output)
# Compile GAN
gan.compile(loss='binary crossentropy',
optimizer=keras.optimizers.Adam(learning rate=0.0002))
# Training parameters
batch size = 64
epochs = 10
sample interval = 1000
# Training loop
for epoch in range(epochs):
  # Train discriminator
  idx = np.random.randint(0, X train.shape[0], batch size)
  real images = X train[idx]
  noise = np.random.normal(0, 1, (batch size, 100))
  fake images = generator.predict(noise)
  # Label real and fake images
  real labels = np.ones((batch size, 1))
  fake_labels = np.zeros((batch_size, 1))
  # Train discriminator
  d_loss_real = discriminator.train_on_batch(real_images, real_labels)
  d loss fake = discriminator.train on batch(fake images, fake labels)
  d loss = 0.5 * np.add(d loss real, d loss fake)
  # Train generator
  noise = np.random.normal(0, 1, (batch size, 100))
  g loss = gan.train on batch(noise, real labels)
  # Print progress
  if epoch % sample interval == 0:
```

```
print(f'Epoch {epoch}, D Loss: {d_loss[0]}, G Loss: {g_loss}')

# Print discriminator accuracy

metrics_names = discriminator.metrics_names

accuracy_index = metrics_names.index('accuracy')

_, accuracy = discriminator.evaluate(np.concatenate([real_images, fake_images]),
np.concatenate([real_labels, fake_labels]), verbose=0)

print(f'Discriminator Accuracy: {accuracy:.4f}")
```

```
import tensorflow as tf #!pip install tensorflow==2.8
from tensorflow.keras.applications import VGG16
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten, Dropout
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os # Import the os module !pip install os
from google.colab import drive
drive.mount('/content/drive')
data dir = '/content/drive/MyDrive/Collab'
# Load the pre-trained VGG16 model (without the fully connected layers)
vgg model = VGG16(weights='imagenet', include top=False, input shape=(224, 224, 3))
# Freeze the weights of the pre-trained layers so they are not updated during training
for layer in vgg model.layers:
  layer.trainable = False
# Create a new model
model = Sequential()
# Add the pre-trained VGG16 model
model.add(vgg model)
# Flatten the output of VGG16
model.add(Flatten())
# Add fully connected layers for classification
model.add(Dense(512, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(256, activation='relu'))
model.add(Dropout(0.5))
# Number of classes in your dataset
num classes = 2
# Output layer for multi-class classification
model.add(Dense(num classes, activation='softmax'))
```

```
# Compile the model
model.compile(optimizer=Adam(lr=1e-4), loss='categorical crossentropy',
metrics=['accuracy'])
# Load and preprocess the data using ImageDataGenerator
train data dir = os.path.join(data dir, 'train')
validation data dir = os.path.join(data dir, 'validation')
train datagen = ImageDataGenerator(rescale=1./255)
test datagen = ImageDataGenerator(rescale=1./255)
train generator = train datagen.flow from directory(
  train data dir,
  target size=(224, 224),
  batch size=32,
  class mode='categorical', #Use 'categorical' for multi-class classification
  shuffle=True
)
validation generator = test datagen.flow from directory(
  validation data dir,
  target size=(224, 224),
  batch size=32,
  class mode='categorical', #Use 'categorical' for multi-class classification
  shuffle=False
)
# Class labels
class_labels = train_generator.class_indices
print("Class labels:", class_labels)
# Train the model
model.fit(
  train generator,
  steps per epoch=train generator.samples // train generator.batch size,
  epochs=10, # Adjust the number of epochs as needed
  validation data=validation generator,
  validation_steps=validation_generator.samples // validation_generator.batch_size
```

```
# Evaluate the model on the validation data
validation_loss, validation_accuracy = model.evaluate(validation_generator)
print("Validation Accuracy:", validation_accuracy)
#https://www.kaggle.com/code/samarthsoni106/cat-and-dog-classification-tensorflow/input
```