```
import tensorflow as tf
from tensorflow.keras import layers
import matplotlib.pyplot as plt
import numpy as np
import time
# Load and preprocess the CIFAR-10 dataset
(train images, train labels), (_, _) =
tf.keras.datasets.cifar10.load data()
train images = train images.astype('float32')
train images = (train_images - 127.5) / 127.5 # Normalize the images
to [-1, 1]
def build generator():
    model = tf.keras.Sequential()
    model.add(layers.Dense(8*8*256, use bias=False,
input shape=(100,))
    model.add(layers.BatchNormalization())
    model.add(layers.LeakyReLU())
    model.add(layers.Reshape((8, 8, 256)))
    model.add(layers.Conv2DTranspose(128, (5, 5), strides=(1, 1),
padding='same', use bias=False))
    model.add(layers.BatchNormalization())
    model.add(layers.LeakyReLU())
    model.add(layers.Conv2DTranspose(64, (5, 5), strides=(2, 2),
padding='same', use bias=False))
    model.add(layers.BatchNormalization())
    model.add(layers.LeakyReLU())
    model.add(layers.Conv2DTranspose(3, (5, 5), strides=(2, 2),
padding='same', use bias=False, activation='tanh'))
    return model
def build discriminator():
    model = tf.keras.Sequential()
    model.add(layers.Conv2D(64, (5, 5), strides=(2, 2),
padding='same', input shape=[32, 32, 3]))
    model.add(layers.LeakyReLU())
    model.add(layers.Dropout(0.3))
    model.add(layers.Conv2D(128, (5, 5), strides=(2, 2),
padding='same'))
    model.add(layers.LeakyReLU())
    model.add(layers.Dropout(0.3))
    model.add(layers.Flatten())
    model.add(layers.Dense(1, activation='sigmoid'))
    return model
```

```
# Define loss function and optimizers
cross entropy = tf.keras.losses.BinaryCrossentropy(from logits=True)
def discriminator loss(real output, fake output):
    real loss = cross entropy(tf.ones like(real output), real output)
    fake loss = cross entropy(tf.zeros like(fake output), fake output)
    total_loss = real_loss + fake_loss
    return total loss
def generator loss(fake output):
    return cross entropy(tf.ones like(fake output), fake output)
generator optimizer = tf.keras.optimizers.Adam(1e-4)
discriminator optimizer = tf.keras.optimizers.Adam(le-4)
EPOCHS = 100
noise dim = 100
num examples to generate = 16
seed = tf.random.normal([num examples to generate, noise dim])
@tf.function
def train step(images):
    noise = tf.random.normal([BATCH SIZE, noise dim])
    with tf.GradientTape() as gen tape, tf.GradientTape() as
disc tape:
        generated images = generator(noise, training=True)
        real output = discriminator(images, training=True)
        fake output = discriminator(generated images, training=True)
        gen loss = generator loss(fake output)
        disc loss = discriminator loss(real output, fake output)
    gradients_of_generator = gen_tape.gradient(gen loss,
generator.trainable variables)
    gradients of discriminator = disc tape.gradient(disc loss,
discriminator.trainable variables)
    generator_optimizer.apply_gradients(zip(gradients_of_generator,
generator.trainable variables))
discriminator optimizer.apply gradients(zip(gradients of discriminator
, discriminator.trainable variables))
    return gen loss, disc loss
# Create and compile models
generator = build generator()
discriminator = build discriminator()
```

generator.summary()

Model: "sequential_4"

Layer (type)	Output Shape	Param #
dense_4 (Dense)	(None, 16384)	1638400
<pre>batch_normalization_6 (Bat chNormalization)</pre>	(None, 16384)	65536
leaky_re_lu_10 (LeakyReLU)	(None, 16384)	0
reshape_2 (Reshape)	(None, 8, 8, 256)	0
<pre>conv2d_transpose_6 (Conv2D Transpose)</pre>	(None, 8, 8, 128)	819200
<pre>batch_normalization_7 (Bat chNormalization)</pre>	(None, 8, 8, 128)	512
leaky_re_lu_11 (LeakyReLU)	(None, 8, 8, 128)	0
<pre>conv2d_transpose_7 (Conv2D Transpose)</pre>	(None, 16, 16, 64)	204800
<pre>batch_normalization_8 (Bat chNormalization)</pre>	(None, 16, 16, 64)	256
leaky_re_lu_12 (LeakyReLU)	(None, 16, 16, 64)	0
<pre>conv2d_transpose_8 (Conv2D Transpose)</pre>	(None, 32, 32, 3)	4800

Total params: 2733504 (10.43 MB) Trainable params: 2700352 (10.30 MB) Non-trainable params: 33152 (129.50 KB)

discriminator.summary()

Model: "sequential_5"

Layer (type)	Output Shape	Param #
conv2d_4 (Conv2D)	(None, 16, 16, 64)	4864
<pre>leaky_re_lu_13 (LeakyReLU)</pre>	(None, 16, 16, 64)	0

```
dropout 4 (Dropout)
                             (None, 16, 16, 64)
 conv2d 5 (Conv2D)
                             (None, 8, 8, 128)
                                                       204928
 leaky re lu 14 (LeakyReLU)
                            (None, 8, 8, 128)
 dropout 5 (Dropout)
                             (None, 8, 8, 128)
                                                       0
 flatten 2 (Flatten)
                                                       0
                             (None, 8192)
 dense 5 (Dense)
                             (None, 1)
                                                       8193
Total params: 217985 (851.50 KB)
Trainable params: 217985 (851.50 KB)
Non-trainable params: 0 (0.00 Byte)
# Initialize the models
generator = build generator()
discriminator = build discriminator()
# Build the models to ensure all layers are initialized
generator.build(input shape=(None, 100)) # Adjust the input shape
based on your noise dimension
discriminator.build(input shape=(None, 32, 32, 3)) # Adjust input
shape to match CIFAR-10 image dimensions
# Create and compile models after building
generator optimizer = tf.keras.optimizers.Adam(1e-4)
discriminator optimizer = tf.keras.optimizers.Adam(1e-4)
# Configure the optimizers with the variables
generator optimizer.build(generator.trainable variables)
discriminator optimizer.build(discriminator.trainable variables)
@tf.function
def train step(images):
    noise = tf.random.normal([BATCH_SIZE, noise_dim])
    with tf.GradientTape() as gen_tape, tf.GradientTape() as
disc tape:
        generated images = generator(noise, training=True)
        real output = discriminator(images, training=True)
        fake output = discriminator(generated images, training=True)
        gen loss = generator loss(fake output)
        disc loss = discriminator loss(real output, fake output)
    gradients of generator = gen tape.gradient(gen loss,
generator.trainable variables)
    gradients of discriminator = disc tape.gradient(disc loss,
discriminator.trainable variables)
```

```
generator_optimizer.apply_gradients(zip(gradients_of_generator,
generator.trainable variables))
discriminator_optimizer.apply_gradients(zip(gradients_of_discriminator)
, discriminator.trainable variables))
    return gen loss, disc loss
# Prepare the dataset for training
BUFFER SIZE = 20000
BATCH \overline{S}IZE = 32
train dataset =
tf.data.Dataset.from tensor slices(train images).shuffle(BUFFER SIZE).
batch(BATCH SIZE)
def generate and save images(model, epoch, test input):
    predictions = model(test_input, training=False)
    fig = plt.figure(figsize=(4, 4))
    for i in range(predictions.shape[0]):
        plt.subplot(4, 4, i+1)
        plt.imshow((predictions[i, :, :, :] + 1) / 2) # Adjusting
normalization for display
        plt.axis('off')
    plt.suptitle(f"Generated Images at Epoch {epoch}")
    plt.show() # Ensures the image is displayed in Jupyter Notebook
    plt.savefig(f'image at epoch {epoch:04d}.png') # Save each
epoch's generated images
def train(dataset, epochs):
    for epoch in range(epochs):
        start time = time.time() # Start timing the epoch
        for image batch in dataset:
            gen loss, disc loss = train step(image batch) # Capture
losses from train step
        # Generate and display images
        generate and save images (generator, epoch + 1, seed)
        end time = time.time() # End timing the epoch
        print(f"Epoch {epoch+1}/{EPOCHS} - Generator Loss:
{gen loss.numpy():.4f}, Discriminator Loss: {disc loss.numpy():.4f} -
Time Taken: {end time - start time:.2f} sec")
# Start the training process
train(train dataset, EPOCHS)
```



Epoch 1/100 - Generator Loss: 0.9341, Discriminator Loss: 1.2777 - Time Taken: 169.87 sec

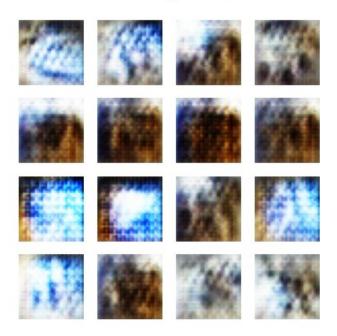
<Figure size 640x480 with 0 Axes>



Epoch 2/100 - Generator Loss: 1.0615, Discriminator Loss: 1.2185 - Time Taken: 164.02~sec

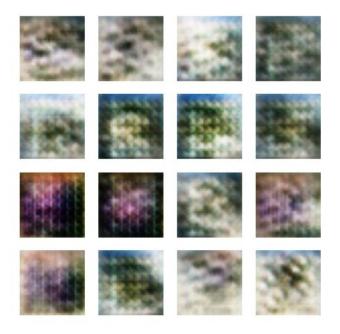
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 3



Epoch 3/100 - Generator Loss: 0.8092, Discriminator Loss: 1.6920 -

Time Taken: 158.22 sec



Epoch 4/100 - Generator Loss: 0.7361, Discriminator Loss: 1.2987 - Time Taken: 157.09 sec

<Figure size 640x480 with 0 Axes>



Epoch 5/100 - Generator Loss: 1.2502, Discriminator Loss: 1.1228 - Time Taken: 156.99 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 6



Epoch 6/100 - Generator Loss: 0.9812, Discriminator Loss: 1.4240 - Time Taken: 156.94 sec



Epoch 7/100 - Generator Loss: 0.7597, Discriminator Loss: 1.5265 - Time Taken: 159.94 sec

<Figure size 640x480 with 0 Axes>



Epoch 8/100 - Generator Loss: 1.0404, Discriminator Loss: 1.1523 - Time Taken: 159.51 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 9

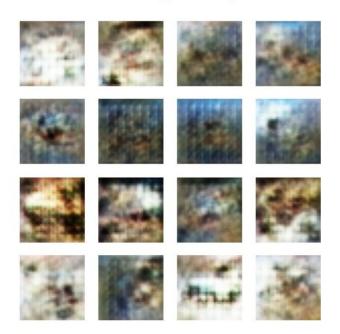


Epoch 9/100 - Generator Loss: 1.2566, Discriminator Loss: 1.0486 - Time Taken: 156.18 sec



Epoch 10/100 - Generator Loss: 1.1700, Discriminator Loss: 1.0708 - Time Taken: 156.57 sec

<Figure size 640x480 with 0 Axes>



Epoch 11/100 - Generator Loss: 0.8928, Discriminator Loss: 1.6335 - Time Taken: 155.94 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 12



Epoch 12/100 - Generator Loss: 1.2045, Discriminator Loss: 1.0142 - Time Taken: 156.27 sec



Epoch 13/100 - Generator Loss: 1.0434, Discriminator Loss: 1.0453 - Time Taken: 157.41 sec

<Figure size 640x480 with 0 Axes>



Epoch 14/100 - Generator Loss: 0.9640, Discriminator Loss: 1.0896 -

Time Taken: 155.61 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 15



Epoch 15/100 - Generator Loss: 0.9160, Discriminator Loss: 1.2906 - Time Taken: 160.00~sec



Epoch 16/100 - Generator Loss: 0.8248, Discriminator Loss: 1.6522 - Time Taken: 156.40 sec

<Figure size 640x480 with 0 Axes>



Epoch 17/100 - Generator Loss: 1.2123, Discriminator Loss: 0.8915 - Time Taken: 156.01 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 18



Epoch 18/100 - Generator Loss: 1.0547, Discriminator Loss: 1.1490 - Time Taken: 156.06 sec



Epoch 19/100 - Generator Loss: 0.9116, Discriminator Loss: 1.3088 - Time Taken: 158.26 sec

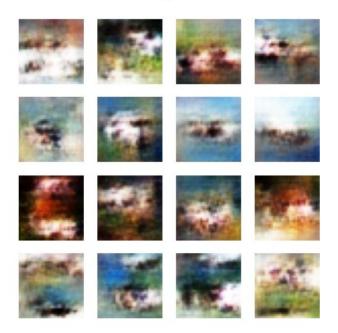
<Figure size 640x480 with 0 Axes>



Epoch 20/100 - Generator Loss: 1.1773, Discriminator Loss: 1.0139 - Time Taken: 160.95~sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 21



Epoch 21/100 - Generator Loss: 0.9831, Discriminator Loss: 1.0983 - Time Taken: 156.52 sec



Epoch 22/100 - Generator Loss: 1.0132, Discriminator Loss: 0.9665 - Time Taken: 156.92 sec

<Figure size 640x480 with 0 Axes>



Epoch 23/100 - Generator Loss: 1.1542, Discriminator Loss: 1.1903 - Time Taken: 156.20 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 24



Epoch 24/100 - Generator Loss: 1.1038, Discriminator Loss: 1.1550 - Time Taken: 157.13 sec



Epoch 25/100 - Generator Loss: 1.0746, Discriminator Loss: 0.9275 - Time Taken: 156.16 sec

<Figure size 640x480 with 0 Axes>



Epoch 26/100 - Generator Loss: 0.9935, Discriminator Loss: 1.1247 - Time Taken: 156.03 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 27

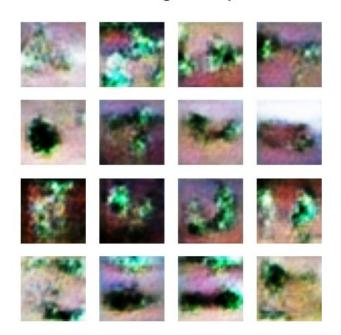


Epoch 27/100 - Generator Loss: 1.0260, Discriminator Loss: 1.1715 - Time Taken: 156.72 sec



Epoch 28/100 - Generator Loss: 1.1032, Discriminator Loss: 1.2488 - Time Taken: 156.56 sec

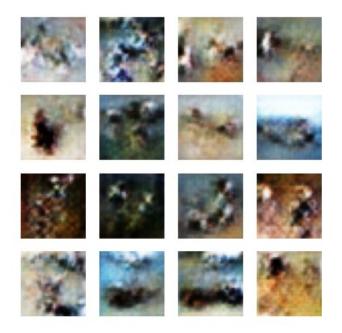
<Figure size 640x480 with 0 Axes>



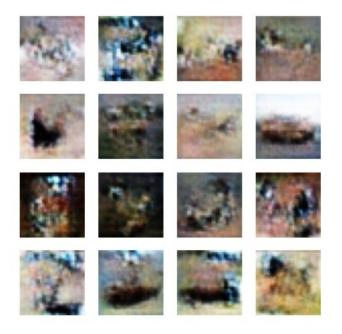
Epoch 29/100 - Generator Loss: 1.1730, Discriminator Loss: 0.7714 - Time Taken: 156.32 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 30

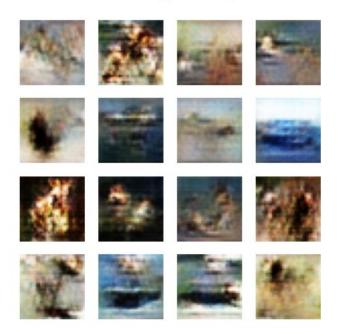


Epoch 30/100 - Generator Loss: 1.0729, Discriminator Loss: 1.3286 - Time Taken: 157.97 sec



Epoch 31/100 - Generator Loss: 1.3947, Discriminator Loss: 1.1169 - Time Taken: 156.96 sec

<Figure size 640x480 with 0 Axes>



Epoch 32/100 - Generator Loss: 1.0527, Discriminator Loss: 1.3176 - Time Taken: 158.68 sec

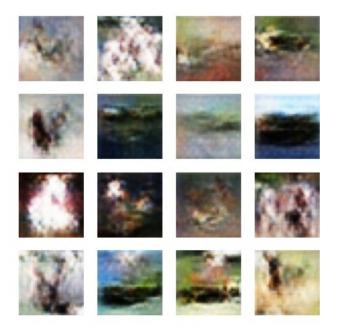
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 33



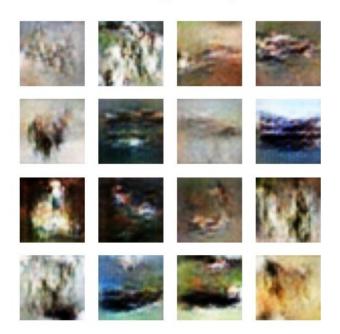
Epoch 33/100 - Generator Loss: 1.5937, Discriminator Loss: 0.7611 -

Time Taken: 162.52 sec



Epoch 34/100 - Generator Loss: 1.2580, Discriminator Loss: 0.9335 - Time Taken: 159.49 sec

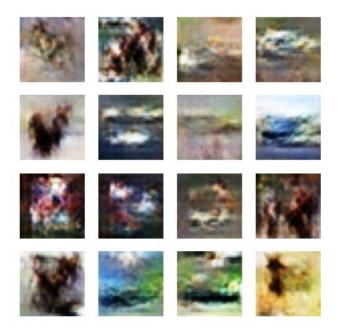
<Figure size 640x480 with 0 Axes>



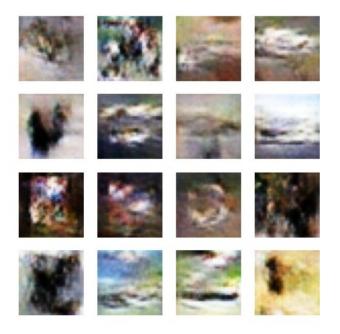
Epoch 35/100 - Generator Loss: 1.1185, Discriminator Loss: 1.3051 - Time Taken: 157.73 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 36

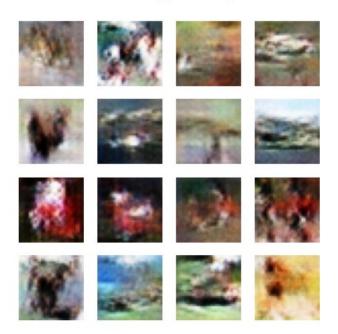


Epoch 36/100 - Generator Loss: 1.2999, Discriminator Loss: 0.9704 - Time Taken: 159.24 sec



Epoch 37/100 - Generator Loss: 1.2421, Discriminator Loss: 0.9770 - Time Taken: 155.95 sec

<Figure size 640x480 with 0 Axes>



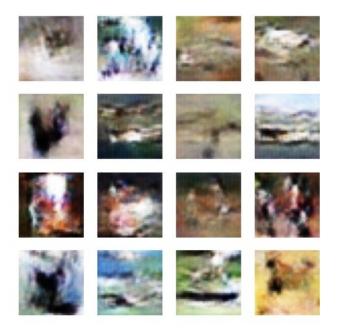
Epoch 38/100 - Generator Loss: 1.0985, Discriminator Loss: 1.3586 - Time Taken: 160.45 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 39

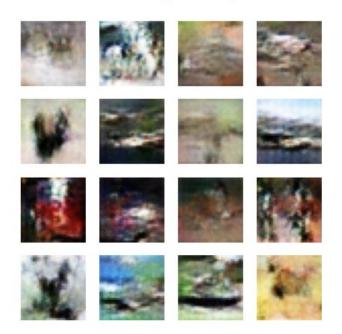


Epoch 39/100 - Generator Loss: 1.4627, Discriminator Loss: 0.8453 - Time Taken: 156.66 sec



Epoch 40/100 - Generator Loss: 1.2762, Discriminator Loss: 1.1688 - Time Taken: 156.91 sec

<Figure size 640x480 with 0 Axes>



Epoch 41/100 - Generator Loss: 1.1450, Discriminator Loss: 1.1354 -

Time Taken: 159.32 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 42

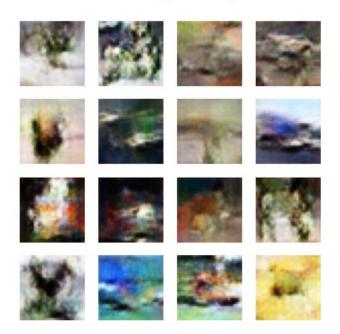


Epoch 42/100 - Generator Loss: 1.1264, Discriminator Loss: 1.2262 - Time Taken: 157.60 sec



Epoch 43/100 - Generator Loss: 1.2602, Discriminator Loss: 1.2943 - Time Taken: 156.18 sec

<Figure size 640x480 with 0 Axes>

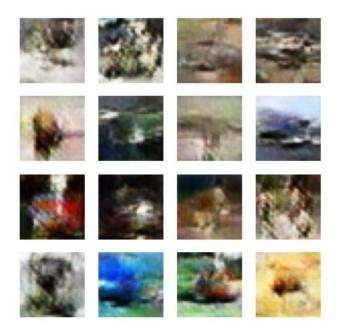


Epoch 44/100 - Generator Loss: 0.8526, Discriminator Loss: 1.1841 -

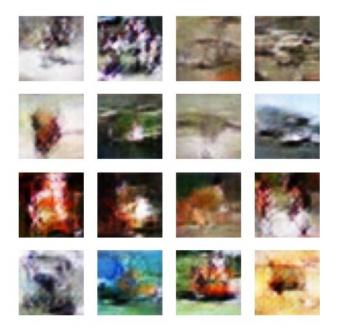
Time Taken: 156.62 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 45

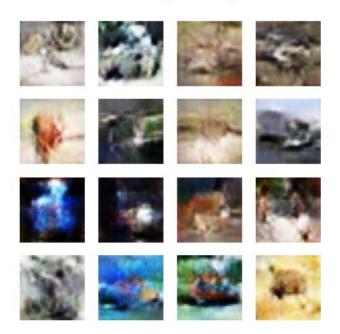


Epoch 45/100 - Generator Loss: 1.1130, Discriminator Loss: 1.0342 - Time Taken: 156.47 sec



Epoch 46/100 - Generator Loss: 1.0734, Discriminator Loss: 1.1260 - Time Taken: 156.39 sec

<Figure size 640x480 with 0 Axes>

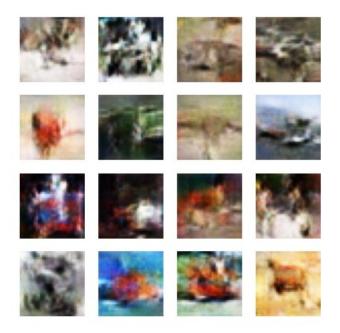


Epoch 47/100 - Generator Loss: 0.9555, Discriminator Loss: 1.3291 -

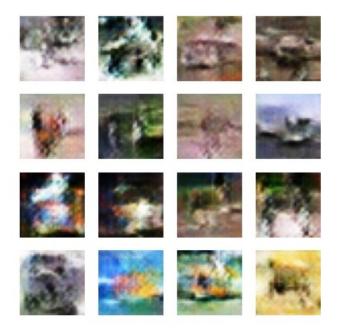
Time Taken: 156.82 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 48

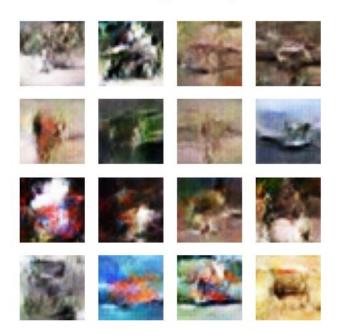


Epoch 48/100 - Generator Loss: 1.1265, Discriminator Loss: 1.4634 - Time Taken: 156.17 sec



Epoch 49/100 - Generator Loss: 1.2326, Discriminator Loss: 1.0481 - Time Taken: 157.89 sec

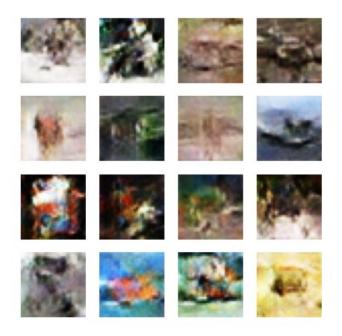
<Figure size 640x480 with 0 Axes>



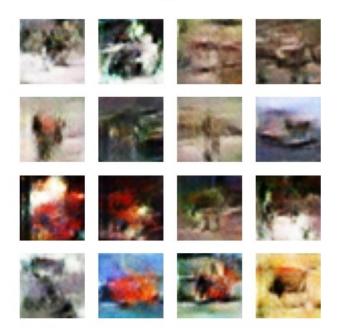
Epoch 50/100 - Generator Loss: 0.8791, Discriminator Loss: 1.3052 - Time Taken: 156.59 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 51

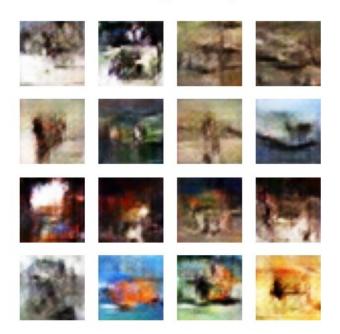


Epoch 51/100 - Generator Loss: 1.0660, Discriminator Loss: 0.9225 - Time Taken: 155.79 sec



Epoch 52/100 - Generator Loss: 1.2222, Discriminator Loss: 1.2645 - Time Taken: 155.69 sec

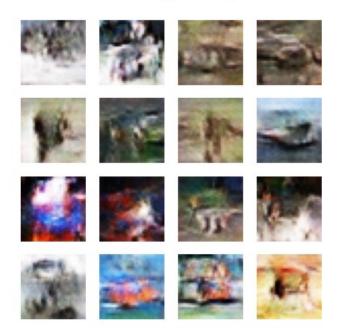
<Figure size 640x480 with 0 Axes>



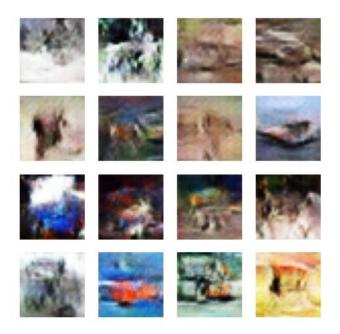
Epoch 53/100 - Generator Loss: 1.4689, Discriminator Loss: 1.1544 - Time Taken: 157.59 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 54

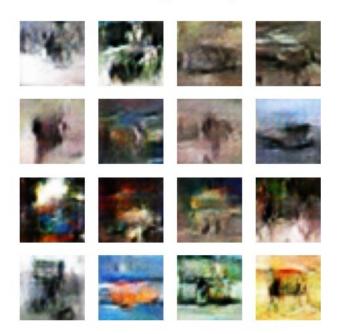


Epoch 54/100 - Generator Loss: 0.9999, Discriminator Loss: 1.1987 - Time Taken: 156.31 sec



Epoch 55/100 - Generator Loss: 1.1720, Discriminator Loss: 1.1516 - Time Taken: 156.93 sec

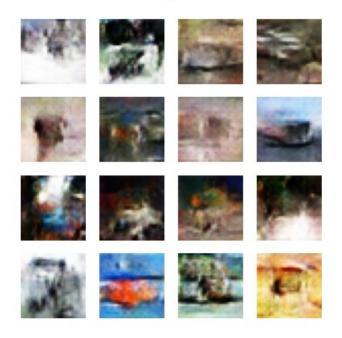
<Figure size 640x480 with 0 Axes>



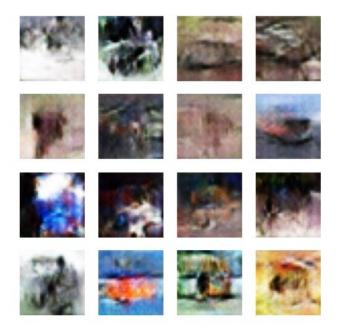
Epoch 56/100 - Generator Loss: 1.0397, Discriminator Loss: 1.3021 - Time Taken: 155.96 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 57

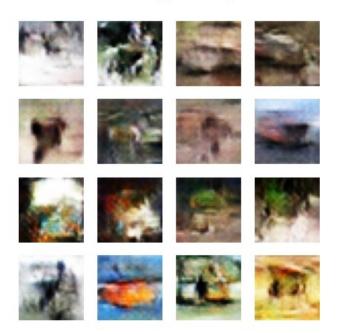


Epoch 57/100 - Generator Loss: 1.1892, Discriminator Loss: 0.9922 - Time Taken: 156.39 sec



Epoch 58/100 - Generator Loss: 1.1008, Discriminator Loss: 1.1881 - Time Taken: 155.72 sec

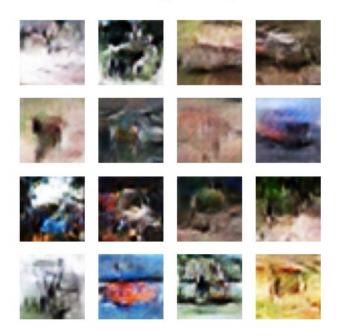
<Figure size 640x480 with 0 Axes>



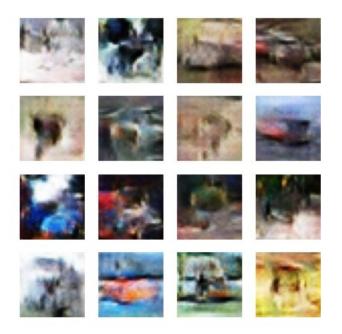
Epoch 59/100 - Generator Loss: 0.9843, Discriminator Loss: 0.9189 - Time Taken: 156.44 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 60

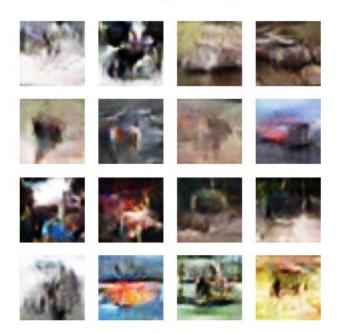


Epoch 60/100 - Generator Loss: 1.1277, Discriminator Loss: 0.8943 - Time Taken: 157.23 sec



Epoch 61/100 - Generator Loss: 1.2222, Discriminator Loss: 1.1953 - Time Taken: 157.34 sec

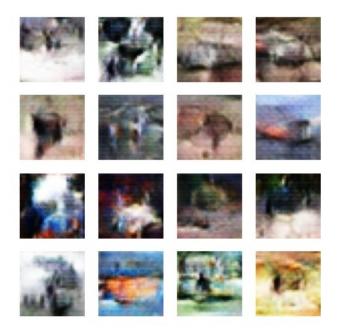
<Figure size 640x480 with 0 Axes>



Epoch 62/100 - Generator Loss: 1.1388, Discriminator Loss: 1.2270 - Time Taken: 158.14 sec

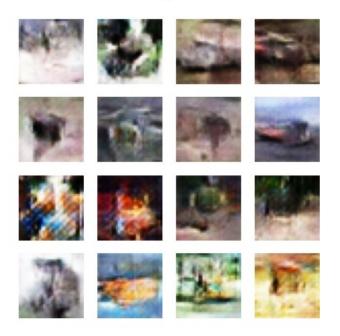
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 63



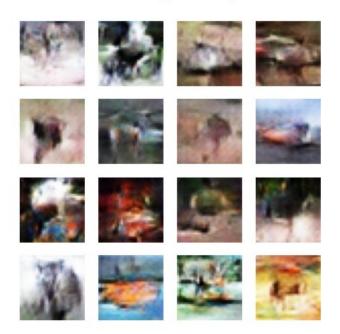
Epoch 63/100 - Generator Loss: 1.3563, Discriminator Loss: 0.8103 -

Time Taken: 161.55 sec



Epoch 64/100 - Generator Loss: 1.1755, Discriminator Loss: 1.2944 - Time Taken: 162.24~sec

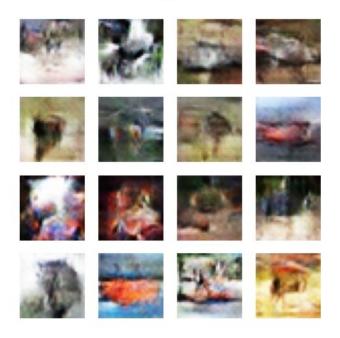
<Figure size 640x480 with 0 Axes>



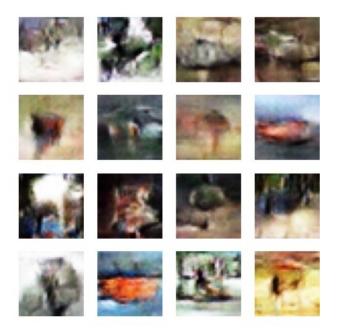
Epoch 65/100 - Generator Loss: 1.0022, Discriminator Loss: 1.1964 - Time Taken: 163.97 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 66

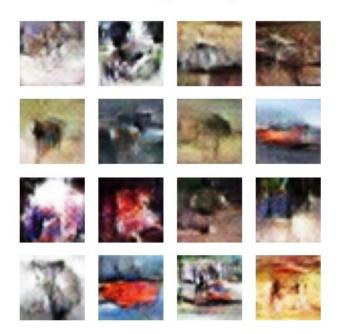


Epoch 66/100 - Generator Loss: 1.2827, Discriminator Loss: 1.5088 - Time Taken: 156.49 sec



Epoch 67/100 - Generator Loss: 1.4232, Discriminator Loss: 0.8453 - Time Taken: 156.14 sec

<Figure size 640x480 with 0 Axes>

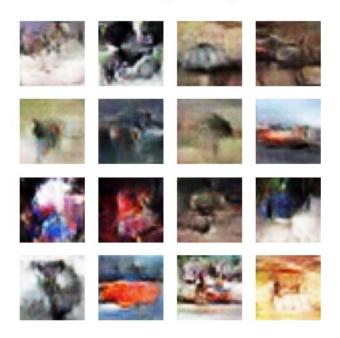


Epoch 68/100 - Generator Loss: 1.1634, Discriminator Loss: 1.0500 -

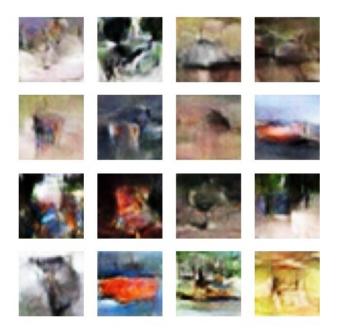
Time Taken: 156.45 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 69

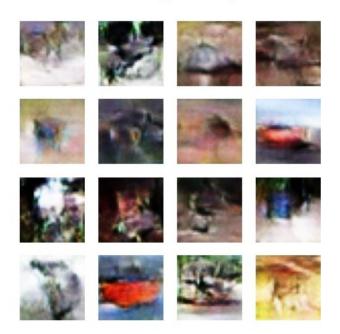


Epoch 69/100 - Generator Loss: 1.0675, Discriminator Loss: 1.0085 - Time Taken: 156.18 sec



Epoch 70/100 - Generator Loss: 1.1919, Discriminator Loss: 0.9675 - Time Taken: 156.95 sec

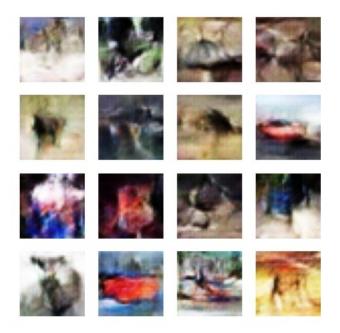
<Figure size 640x480 with 0 Axes>



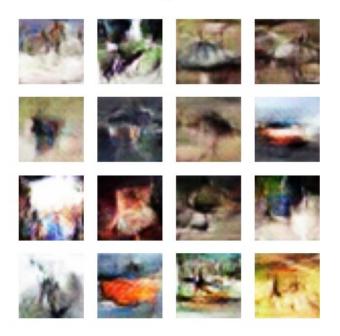
Epoch 71/100 - Generator Loss: 1.1848, Discriminator Loss: 1.5320 - Time Taken: 156.00 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 72

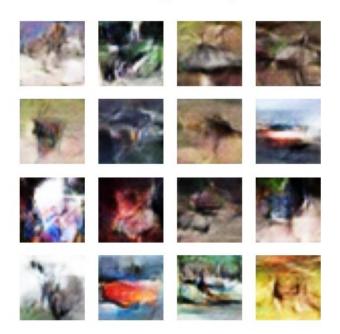


Epoch 72/100 - Generator Loss: 1.3209, Discriminator Loss: 0.9803 - Time Taken: 155.16 sec



Epoch 73/100 - Generator Loss: 0.8987, Discriminator Loss: 1.3290 - Time Taken: 156.59 sec

<Figure size 640x480 with 0 Axes>

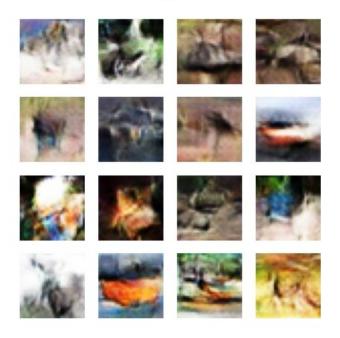


Epoch 74/100 - Generator Loss: 1.3787, Discriminator Loss: 0.9477 -

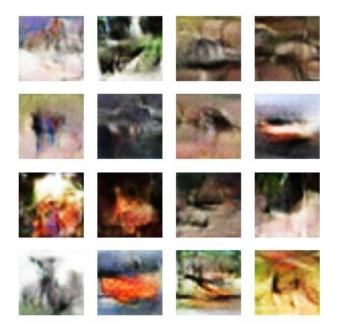
Time Taken: 155.80 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 75

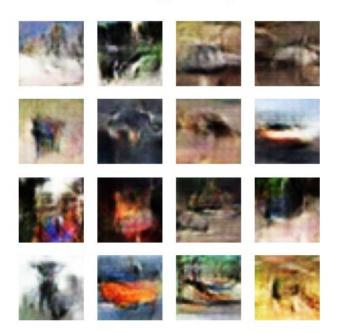


Epoch 75/100 - Generator Loss: 1.0859, Discriminator Loss: 1.0391 - Time Taken: 155.40 sec



Epoch 76/100 - Generator Loss: 1.2058, Discriminator Loss: 1.1969 - Time Taken: 157.58 sec

<Figure size 640x480 with 0 Axes>

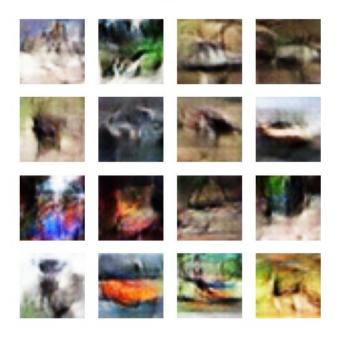


Epoch 77/100 - Generator Loss: 1.4552, Discriminator Loss: 1.1564 -

Time Taken: 160.29 sec

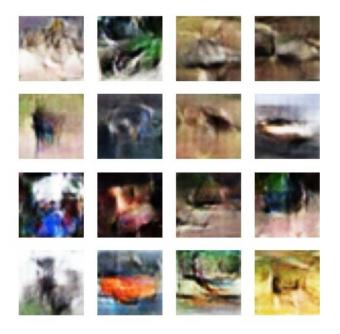
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 78



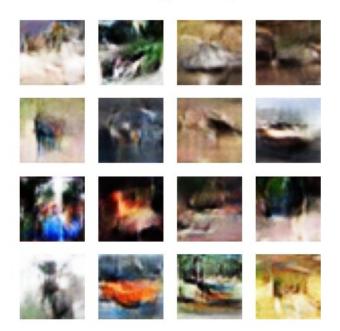
Epoch 78/100 - Generator Loss: 1.4965, Discriminator Loss: 0.7879 -

Time Taken: 159.86 sec



Epoch 79/100 - Generator Loss: 1.4413, Discriminator Loss: 1.0765 - Time Taken: 156.42 sec

<Figure size 640x480 with 0 Axes>

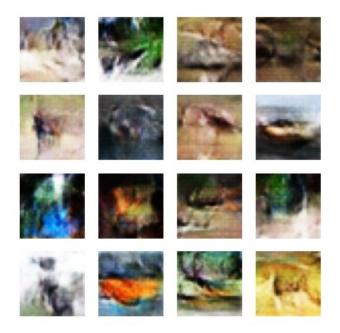


Epoch 80/100 - Generator Loss: 0.8938, Discriminator Loss: 1.4187 -

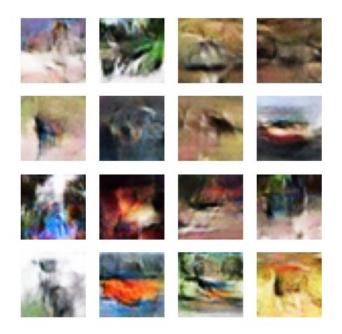
Time Taken: 157.91 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 81

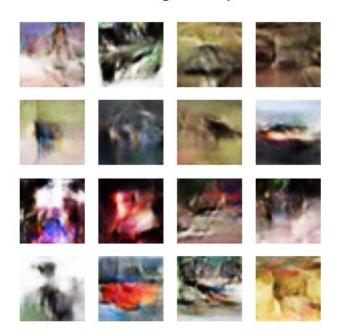


Epoch 81/100 - Generator Loss: 1.0738, Discriminator Loss: 1.2642 - Time Taken: 157.31 sec



Epoch 82/100 - Generator Loss: 1.0687, Discriminator Loss: 1.2624 - Time Taken: 157.98 sec

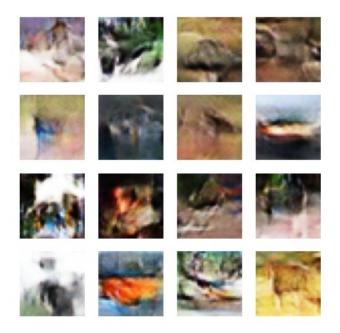
<Figure size 640x480 with 0 Axes>



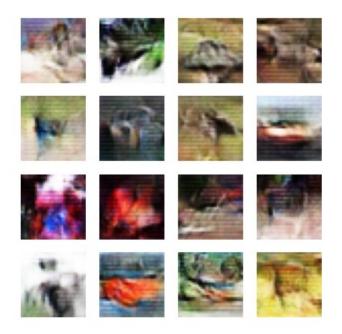
Epoch 83/100 - Generator Loss: 1.1962, Discriminator Loss: 1.1078 - Time Taken: 156.44 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 84

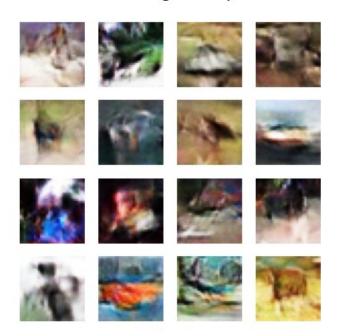


Epoch 84/100 - Generator Loss: 1.1126, Discriminator Loss: 1.2522 - Time Taken: 156.61 sec



Epoch 85/100 - Generator Loss: 1.2485, Discriminator Loss: 0.8965 - Time Taken: 156.30 sec

<Figure size 640x480 with 0 Axes>

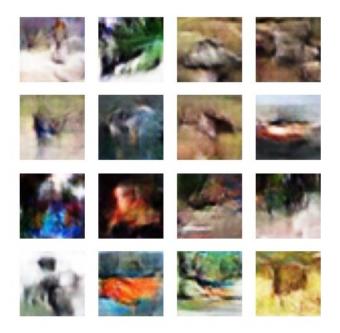


Epoch 86/100 - Generator Loss: 1.1019, Discriminator Loss: 1.3628 -

Time Taken: 162.77 sec

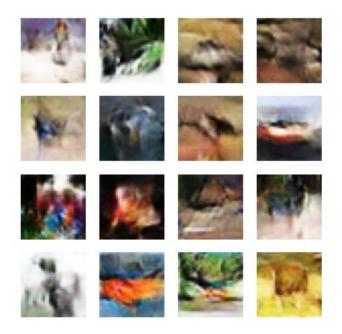
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 87



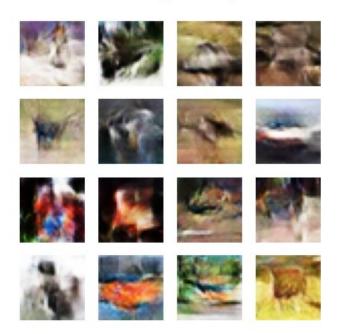
Epoch 87/100 - Generator Loss: 1.0935, Discriminator Loss: 1.3950 -

Time Taken: 168.68 sec



Epoch 88/100 - Generator Loss: 1.2551, Discriminator Loss: 0.8256 - Time Taken: 161.82 sec

<Figure size 640x480 with 0 Axes>

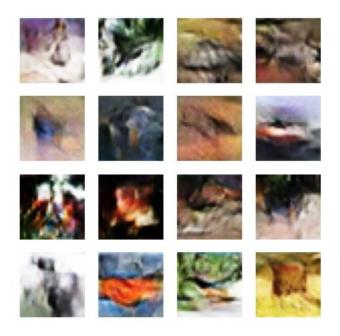


Epoch 89/100 - Generator Loss: 1.0529, Discriminator Loss: 1.2448 -

Time Taken: 160.93 sec

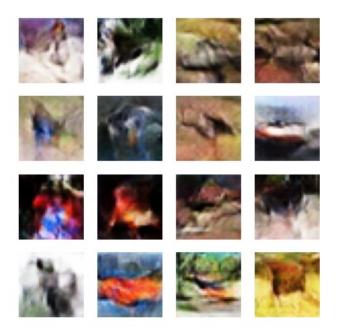
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 90



Epoch 90/100 - Generator Loss: 1.1724, Discriminator Loss: 1.3229 -

Time Taken: 161.28 sec



Epoch 91/100 - Generator Loss: 1.1050, Discriminator Loss: 1.0499 - Time Taken: 161.49 sec

<Figure size 640x480 with 0 Axes>

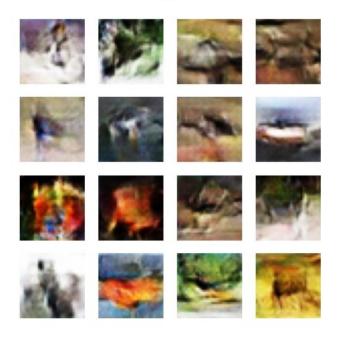


Epoch 92/100 - Generator Loss: 1.2284, Discriminator Loss: 1.0910 -

Time Taken: 160.66 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 93



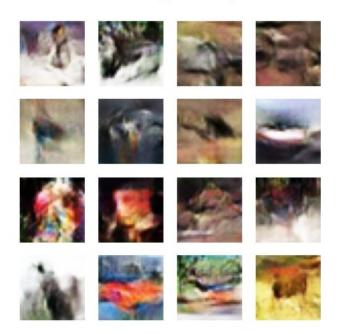
Epoch 93/100 - Generator Loss: 1.1708, Discriminator Loss: 0.9951 -

Time Taken: 161.95 sec



Epoch 94/100 - Generator Loss: 1.2667, Discriminator Loss: 1.2792 - Time Taken: 160.46~sec

<Figure size 640x480 with 0 Axes>



Epoch 95/100 - Generator Loss: 0.8980, Discriminator Loss: 1.1151 - Time Taken: 160.98~sec

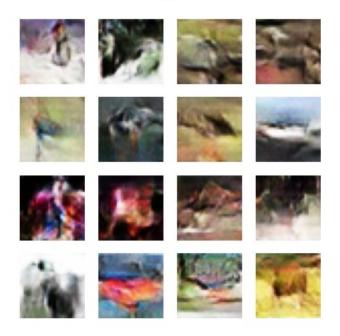
<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 96



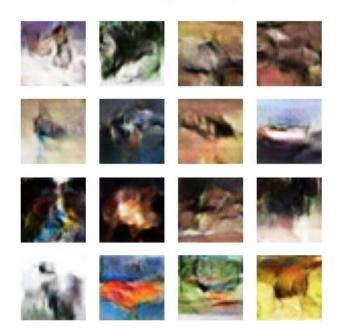
Epoch 96/100 - Generator Loss: 1.1504, Discriminator Loss: 1.2770 -

Time Taken: 161.73 sec



Epoch 97/100 - Generator Loss: 1.0856, Discriminator Loss: 1.1220 - Time Taken: 162.22 sec

<Figure size 640x480 with 0 Axes>

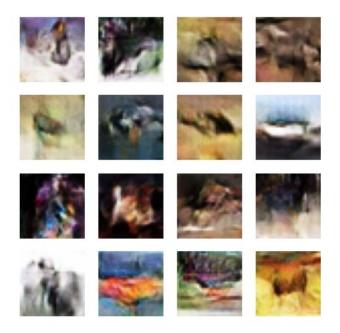


Epoch 98/100 - Generator Loss: 1.1443, Discriminator Loss: 0.9166 -

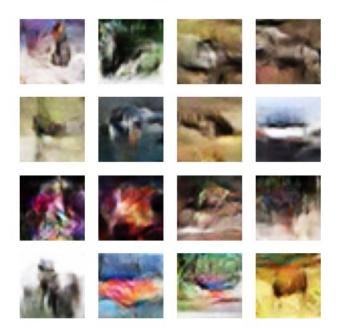
Time Taken: 162.24 sec

<Figure size 640x480 with 0 Axes>

Generated Images at Epoch 99



Epoch 99/100 - Generator Loss: 1.3180, Discriminator Loss: 1.0746 - Time Taken: 162.00~sec



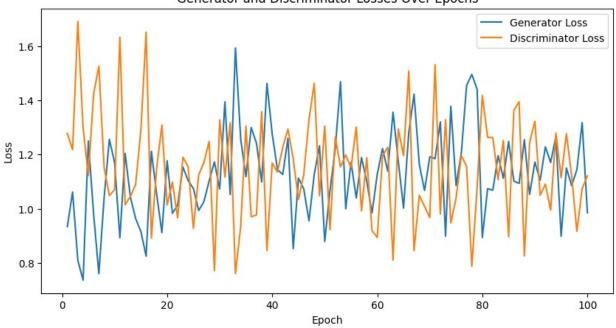
```
Epoch 100/100 - Generator Loss: 0.9847, Discriminator Loss: 1.1212 -
Time Taken: 161.69 sec
<Figure size 640x480 with 0 Axes>
# Save Models After Training
generator.save("cifar10 generator 2.h5")
discriminator.save("cifar10 discriminator 2.h5")
print("Training Complete and Models Saved!")
WARNING:tensorflow:Compiled the loaded model, but the compiled metrics
have yet to be built. `model.compile metrics` will be empty until you
train or evaluate the model.
WARNING: tensorflow: Compiled the loaded model, but the compiled metrics
have yet to be built. `model.compile_metrics` will be empty until you
train or evaluate the model.
Training Complete and Models Saved!
c:\Python311\Lib\site-packages\keras\src\engine\training.py:3103:
UserWarning: You are saving your model as an HDF5 file via
`model.save()`. This file format is considered legacy. We recommend
using instead the native Keras format, e.g.
`model.save('my model.keras')`.
  saving api.save model(
def plot training losses(gen losses, disc losses, epochs):
    plt.figure(figsize=(12, 6))
    plt.plot(epochs, gen_losses, label='Generator Loss', color='blue')
    plt.plot(epochs, disc losses, label='Discriminator Loss',
```

```
color='red')
    plt.title('Training and Discriminator Losses Over Epochs')
    plt.xlabel('Epoch')
    plt.ylabel('Loss')
    plt.legend()
    plt.grid(True)
    plt.show()
# Generator and discriminator losses for each epoch
gen losses = [
    0.9341, 1.0615, 0.8092, 0.7361, 1.2502, 0.9812, 0.7597, 1.0404,
1.2566, 1.1700,
    0.8928, 1.2045, 1.0434, 0.9640, 0.9160, 0.8248, 1.2123, 1.0547,
0.9116, 1.1773,
    0.9831, 1.0132, 1.1542, 1.1038, 1.0746, 0.9935, 1.0260, 1.1032,
1.1730, 1.0729,
    1.3947, 1.0527, 1.5937, 1.2580, 1.1185, 1.2999, 1.2421, 1.0985,
1.4627. 1.2762.
    1.1450, 1.1264, 1.2602, 0.8526, 1.1130, 1.0734, 0.9555, 1.1265,
1.2326, 0.8791,
    1.0660, 1.2222, 1.4689, 0.9999, 1.1720, 1.0397, 1.1892, 1.1008,
0.9843, 1.1277,
    1.2222, 1.1388, 1.3563, 1.1755, 1.0022, 1.2827, 1.4232, 1.1634,
1.0675, 1.1919,
    1.1848, 1.3209, 0.8987, 1.3787, 1.0859, 1.2058, 1.4552, 1.4965,
1.4413, 0.8938,
    1.0738, 1.0687, 1.1962, 1.1126, 1.2485, 1.1019, 1.0935, 1.2551,
1.0529, 1.1724,
    1.1050, 1.2284, 1.1708, 1.2667, 0.8980, 1.1504, 1.0856, 1.1443,
1.3180, 0.9847
]
disc losses = [
    1.2777, 1.2185, 1.6920, 1.2987, 1.1228, 1.4240, 1.5265, 1.1523,
1.0486, 1.0708,
    1.6335, 1.0142, 1.0453, 1.0896, 1.2906, 1.6522, 0.8915, 1.1490,
1.3088, 1.0139,
    1.0983, 0.9665, 1.1903, 1.1550, 0.9275, 1.1247, 1.1715, 1.2488,
0.7714, 1.3286,
    1.1169, 1.3176, 0.7611, 0.9335, 1.3051, 0.9704, 0.9770, 1.3586,
0.8453, 1.1688,
    1.1354, 1.2262, 1.2943, 1.1841, 1.0342, 1.1260, 1.3291, 1.4634,
1.0481, 1.3052,
    0.9225, 1.2645, 1.1544, 1.1987, 1.1516, 1.3021, 0.9922, 1.1881,
0.9189, 0.8943,
    1.1953, 1.2270, 0.8103, 1.2944, 1.1964, 1.5088, 0.8453, 1.0500,
1.0085, 0.9675,
    1.5320, 0.9803, 1.3290, 0.9477, 1.0391, 1.1969, 1.1564, 0.7879,
1.0765, 1.4187,
    1.2642, 1.2624, 1.1078, 1.2522, 0.8965, 1.3628, 1.3950, 0.8256,
```

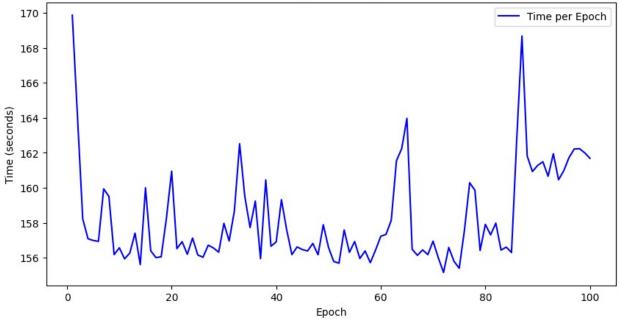
```
1.2448, 1.3229,
    1.0499, 1.0910, 0.9951, 1.2792, 1.1151, 1.2770, 1.1220, 0.9166,
1.0746, 1.1212
# Time taken for each epoch
times = [
    169.87, 164.02, 158.22, 157.09, 156.99, 156.94, 159.94, 159.51,
156.18, 156.57,
    155.94, 156.27, 157.41, 155.61, 160.00, 156.40, 156.01, 156.06,
158.26, 160.95,
    156.52, 156.92, 156.20, 157.13, 156.16, 156.03, 156.72, 156.56,
156.32, 157.97,
    156.96, 158.68, 162.52, 159.49, 157.73, 159.24, 155.95, 160.45,
156.66, 156.91,
    159.32, 157.60, 156.18, 156.62, 156.47, 156.39, 156.82, 156.17,
157.89. 156.59.
    155.79, 155.69, 157.59, 156.31, 156.93, 155.96, 156.39, 155.72,
156.44, 157.23,
    157.34, 158.14, 161.55, 162.24, 163.97, 156.49, 156.14, 156.45,
156.18, 156.95,
    156.00, 155.16, 156.59, 155.80, 155.40, 157.58, 160.29, 159.86,
156.42, 157.91,
    157.31, 157.98, 156.44, 156.61, 156.30, 162.77, 168.68, 161.82,
160.93, 161.28,
    161.49, 160.66, 161.95, 160.46, 160.98, 161.73, 162.22, 162.24,
162.00, 161.69
import matplotlib.pyplot as plt
epochs = range(1, 101)
# Plotting the losses
plt.figure(figsize=(10, 5))
plt.plot(epochs, gen losses, label='Generator Loss')
plt.plot(epochs, disc_losses, label='Discriminator Loss')
plt.title('Generator and Discriminator Losses Over Epochs')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()
# Plotting the training times
plt.figure(figsize=(10, 5))
plt.plot(epochs, times, color='blue', label='Time per Epoch')
plt.title('Training Time Per Epoch')
plt.xlabel('Epoch')
plt.ylabel('Time (seconds)')
```

plt.legend() plt.show()









```
epochs = list(range(1, 101)) # Epochs from 1 to 100
plot_training_losses(gen_losses, disc_losses, epochs)
```



```
import tensorflow as tf
# Load trained models
generator = tf.keras.models.load model("cifar10 generator 2.h5")
discriminator =
tf.keras.models.load model("cifar10 discriminator 2.h5")
print("□ Models loaded successfully!")
WARNING:tensorflow:From c:\Python311\Lib\site-packages\keras\src\
losses.py:2976: The name tf.losses.sparse softmax cross entropy is
deprecated. Please use
tf.compat.v1.losses.sparse softmax cross entropy instead.
WARNING:tensorflow:From c:\Python311\Lib\site-packages\keras\src\
backend.py:1398: The name tf.executing eagerly outside functions is
deprecated. Please use
tf.compat.vl.executing eagerly outside functions instead.
WARNING:tensorflow:From c:\Python311\Lib\site-packages\keras\src\
layers\normalization\batch normalization.py:979: The name
tf.nn.fused batch norm is deprecated. Please use
tf.compat.vl.nn.fused batch norm instead.
WARNING: tensorflow: No training configuration found in the save file,
so the model was *not* compiled. Compile it manually.
WARNING: tensorflow: No training configuration found in the save file,
so the model was *not* compiled. Compile it manually.

☐ Models loaded successfully!
```

```
import numpy as np
# Generate 1000 fake images using the trained generator
noise = tf.random.normal([1000, 100]) # 1000 noise vectors
generated images = generator(noise, training=False).numpy()
print("
    Generated images for evaluation!")
☐ Generated images for evaluation!
import matplotlib.pyplot as plt
import tensorflow as tf
import numpy as np
# Load trained generator model
generator = tf.keras.models.load model("cifar10 generator 2.h5")
# Load real CIFAR-10 dataset
(train_images, _), (_, _) = tf.keras.datasets.cifar10.load_data()
train_images = (train_images.astype("float32") - 127.5) / 127.5 #
Normalize to [-1,1]
# Generate fake images using the trained generator
num images = 10 # Number of images to compare
noise = tf.random.normal([num images, 100]) # Noise for generator
generated images = generator(noise, training=False).numpy()
# Select random real images
real images = train images[np.random.choice(len(train images),
num images, replace=False)]
# Plot real vs. generated images side by side
fig, axes = plt.subplots(\frac{2}{2}, num images, figsize=(\frac{15}{4}))
for i in range(num images):
    # Real Images
    axes[0, i].imshow((real images[i] + 1) / 2) # Normalize to [0, 1]
for display
    axes[0, i].axis("off")
    # Generated Images
    axes[1, i].imshow((generated images[i] + 1) / 2) # Normalize to
[0,1] for display
    axes[1. il.axis("off")
# Titles for rows
axes[0, 0].set_ylabel("Real Images", fontsize=12, fontweight="bold")
axes[1, 0].set ylabel("Generated Images", fontsize=12,
fontweight="bold")
plt.suptitle("Comparison of Real vs. Generated CIFAR-10 Images",
```

```
fontsize=14, fontweight="bold")
plt.show()
```

WARNING:tensorflow:No training configuration found in the save file, so the model was *not* compiled. Compile it manually.

Comparison of Real vs. Generated CIFAR-10 Images



```
import matplotlib.pyplot as plt
import tensorflow as tf
import numpy as np
# Load trained generator model
generator = tf.keras.models.load model("cifar10 generator 2.h5")
# Generate a large batch of images
num samples = 64  # Generate 64 images to analyze diversity
noise = tf.random.normal([num samples, 100])
generated images = generator(noise, training=False).numpy()
# Plot generated images in a grid
fig, axes = plt.subplots(8, 8, figsize=(8, 8))
for i, ax in enumerate(axes.flat):
    ax.imshow((generated images[i] + \frac{1}{2}) # Normalize to [0,1] for
display
    ax.axis("off")
plt.suptitle("Diversity in Generated CIFAR-10 Images", fontsize=14,
fontweight="bold")
plt.show()
WARNING:tensorflow:No training configuration found in the save file,
so the model was *not* compiled. Compile it manually.
```

Diversity in Generated CIFAR-10 Images

