

CX-12 Implement a Deep convolutional GAN to generate complex color images

Aim:- To implement a Deep convolution Generative

Adversarial Network (DCGAN) that can learn to generate realistic color images from random noise using the CIFAR-10 dataset

Objectives :-

1. To understand and apply the Generative Adversarial Network (GAN) framework.
2. To design and train a deep convolutional Generator that creates synthetic color images
3. To design and train a convolutional Discriminator that distinguishes between real and fake images
4. To evaluate the visual quality of generated images as training progresses.

Pseudocode :-

1. Import required libraries
2. Load CIFAR-10 dataset and normalize image pixel b/w  
[-1, 1]
3. Define the generator network
4. Define the Discriminator network.
5. Combine both networks
6. Training loop.
7. Save trained generator model and generated samples

Result :-

A Deep convolutional GAN was successfully implemented to generate realistic color images. After sufficient epochs ( $\approx 50+$ ) the generator produces images that visually resemble real CIFAR-10 samples, demonstrating the power of DCGANs in unsupervised image generation.

S. J. DDS

outPut:-

- During training, the generator loss decreases gradually - Meaning, it's improving at producing realistic images.
- The discriminator loss fluctuates - indicating the adversarial balance b/w the two networks.
- Generated images start has random noise and progress resemble CIFAR-10-like objects.

The CNN model to ~~different~~ training

Epoch over Training & validation.

Epoch 1/50 Generator loss :- 0.9890  
Disriminator loss :- 1.3835

10/50 Generator loss :- 2.3227  
Disriminator loss :- 0.9163

20/50 Generator loss :- 2.1335  
Disriminator loss :- 0.8985

30/50 Generator loss :- 1.3841  
Disriminator loss :- 1.0203

40/50 Generator loss :- 1.4775  
Disriminator loss :- 0.9890

50/50 Generator loss :- 1.1928  
Disriminator loss :- 1.1307

The screenshot shows a Google Colab notebook interface. On the left, there's a sidebar with a 'Table of contents' section containing links to 'Welcome to Colab!', 'Getting started', 'Data science', 'Machine learning', 'More resources', and 'Featured examples'. Below the sidebar, there's a '+ Section' button. The main area has a code editor with a single cell containing Python code for generating images using TensorFlow and Keras. The code imports necessary libraries, loads CIFAR-10 data, normalizes it, defines buffer sizes, batch sizes, and latent dimensions, and builds a generator model using Sequential layers like Dense, BatchNormalization, and Conv2DTranspose. At the bottom of the code cell is a blue run button with a white play icon. The status bar at the bottom right shows the time as 01:46 and the runtime as T4 (Python 3).

```
[1]: import tensorflow as tf
from tensorflow.keras import layers
import matplotlib.pyplot as plt
import numpy as np

# -----
# Load and preprocess CIFAR-10
# -----
(x_train, _, _, _) = tf.keras.datasets.cifar10.load_data()
x_train = x_train.astype("float32")
x_train = (x_train - 127.5) / 127.5 # Normalize to [-1, 1]

BUFFER_SIZE = 50000
BATCH_SIZE = 128
LATENT_DIM = 100
EPOCHS = 50

train_dataset = tf.data.Dataset.from_tensor_slices(x_train).shuffle(BUFFER_SIZE).batch(BATCH_SIZE)

# -----
# Generator Model
# -----
def build_generator():
    model = tf.keras.Sequential([
        layers.Dense(8 * 8 * 256, use_bias=False, input_shape=(LATENT_DIM,)),
        layers.BatchNormalization(),
        layers.ReLU(),
        layers.Reshape((8, 8, 256)),
        layers.Conv2DTranspose(128, (5,5), strides=(2,2), padding='same', use_bias=False),
        layers.BatchNormalization(),
        layers.ReLU(),
        layers.Conv2DTranspose(1, (5,5), strides=(2,2), padding='same', use_bias=False),
        layers.Activation('tanh')
    ])
    return model
```

The screenshot shows a Google Colab notebook titled "Welcome to Colab". The left sidebar contains a "Table of contents" with sections like "Welcome to Colab!", "Getting started", "Data science", "Machine learning", and "More resources". A "Featured examples" section is also present. The main workspace displays Python code for building a Generative Adversarial Network (GAN). The code defines two functions: "build\_generator()" and "build\_discriminator()", both utilizing TensorFlow's Keras API. The generator uses Conv2DTranspose layers with various configurations (e.g., 64 units, 5x5 kernel, stride 2) and activation functions like ReLU and tanh. The discriminator uses Conv2D layers with similar configurations and LeakyReLU activation. Both models include dropout layers. The final part of the code shows how to build the GAN using these components.

```
[1]: layers.Conv2DTranspose(64, (5,5), strides=(2,2), padding='same', use_bias=False),
      layers.BatchNormalization(),
      layers.ReLU(),
      layers.Conv2DTranspose(3, (5,5), strides=(1,1), padding='same', use_bias=False, activation='tanh')
    ])
    return model

# -----
# Discriminator Model
# -----
def build_discriminator():
    model = tf.keras.Sequential([
        layers.Conv2D(64, (5,5), strides=(2,2), padding='same', input_shape=[32, 32, 3]),
        layers.LeakyReLU(alpha=0.2),
        layers.Dropout(0.3),

        layers.Conv2D(128, (5,5), strides=(2,2), padding='same'),
        layers.LeakyReLU(alpha=0.2),
        layers.Dropout(0.3),

        layers.Flatten(),
        layers.Dense(1, activation='sigmoid')
    ])
    return model

# -----
# Build GAN
# -----
generator = build_generator()
discriminator = build_discriminator()

cross_entropy = tf.keras.losses.BinaryCrossentropy(from_logits=False)
```

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```
[1]
# -----
# Training functions
# -----
@tf.function
def train_step(images):
    noise = tf.random.normal([BATCH_SIZE, LATENT_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 = cross_entropy(tf.ones_like(fake_output), fake_output)
        disc_loss_real = cross_entropy(tf.ones_like(real_output), real_output)
        disc_loss_fake = cross_entropy(tf.zeros_like(fake_output), fake_output)
        disc_loss = disc_loss_real + disc_loss_fake

        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

# -----
# Generate and display images
# -----
def generate_and_save_images(model, epoch):
    noise = tf.random.normal([16, LATENT_DIM])
    predictions = model(noise, training=False)
```

Variables Terminal ✓ 01:46 T4 (Python 3)

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```
[1] # -----
# Train the DCGAN
# -----
for epoch in range(1, EPOCHS + 1):
    for image_batch in train_dataset:
        g_loss, d_loss = train_step(image_batch)

    print(f"Epoch {epoch}/{EPOCHS} | Generator Loss: {g_loss:.4f} | Discriminator Loss: {d_loss:.4f}")
    if epoch % 10 == 0:
        generate_and_save_images(generator, epoch)
```

Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>

170498071/170498071 14s 0us/step

```
/usr/local/lib/python3.12/dist-packages/keras/src/layers/core/dense.py:93: UserWarning: Do not pass an `input_shape`/`input_dim` argument to super().__init__(activity_regularizer, **kwargs)
/usr/local/lib/python3.12/dist-packages/keras/src/layers/convolutional/base_conv.py:113: UserWarning: Do not pass an `input_shape`/`input_d
super().__init__(activity_regularizer, **kwargs)
/usr/local/lib/python3.12/dist-packages/keras/src/layers/activations/leaky_relu.py:41: UserWarning: Argument `alpha` is deprecated. Use `neg
warnings.warn(
Epoch 1/50 | Generator Loss: 0.9840 | Discriminator Loss: 1.3835
Epoch 2/50 | Generator Loss: 3.1005 | Discriminator Loss: 0.7256
Epoch 3/50 | Generator Loss: 2.4983 | Discriminator Loss: 0.3565
Epoch 4/50 | Generator Loss: 2.6726 | Discriminator Loss: 0.3086
Epoch 5/50 | Generator Loss: 2.6250 | Discriminator Loss: 0.2491
Epoch 6/50 | Generator Loss: 2.4817 | Discriminator Loss: 0.7983
Epoch 7/50 | Generator Loss: 2.7962 | Discriminator Loss: 0.2988
Epoch 8/50 | Generator Loss: 5.6742 | Discriminator Loss: 0.2018
Epoch 9/50 | Generator Loss: 3.3226 | Discriminator Loss: 0.4563
Epoch 10/50 | Generator Loss: 2.3227 | Discriminator Loss: 0.9163
```

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Epoch 10

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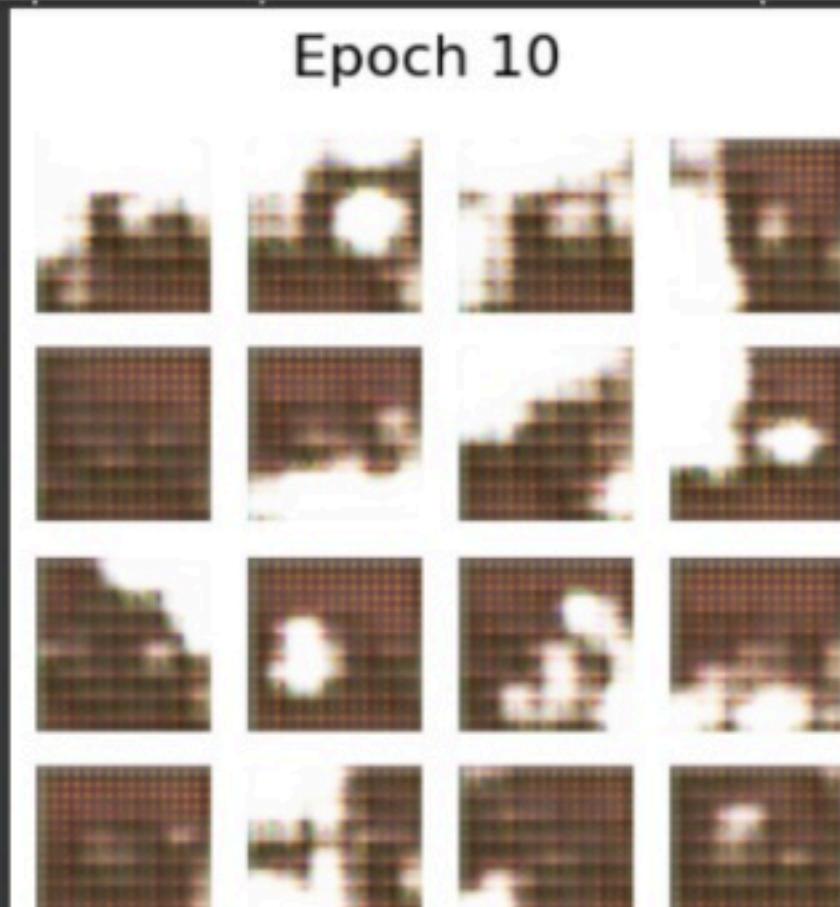
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Epoch 10



Epoch 11/50 | Generator Loss: 2.7002 | Discriminator Loss: 0.5032  
Epoch 12/50 | Generator Loss: 2.1884 | Discriminator Loss: 0.5766  
Epoch 13/50 | Generator Loss: 1.7097 | Discriminator Loss: 1.3700  
Epoch 14/50 | Generator Loss: 1.7213 | Discriminator Loss: 0.6607  
Epoch 15/50 | Generator Loss: 1.5948 | Discriminator Loss: 0.7198  
Epoch 16/50 | Generator Loss: 1.8186 | Discriminator Loss: 0.8346  
Epoch 17/50 | Generator Loss: 1.6664 | Discriminator Loss: 0.8812  
Epoch 18/50 | Generator Loss: 1.3570 | Discriminator Loss: 1.1007  
Epoch 19/50 | Generator Loss: 2.0084 | Discriminator Loss: 0.4675  
Epoch 20/50 | Generator Loss: 2.1335 | Discriminator Loss: 0.8985

Epoch 20

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Epoch 20

Epoch 21/50 | Generator Loss: 1.8843 | Discriminator Loss: 0.8181  
Epoch 22/50 | Generator Loss: 1.8564 | Discriminator Loss: 1.4031  
Epoch 23/50 | Generator Loss: 1.5022 | Discriminator Loss: 0.8977  
Epoch 24/50 | Generator Loss: 2.4385 | Discriminator Loss: 0.5607  
Epoch 25/50 | Generator Loss: 1.5738 | Discriminator Loss: 0.7621  
Epoch 26/50 | Generator Loss: 1.3153 | Discriminator Loss: 1.1137  
Epoch 27/50 | Generator Loss: 1.9380 | Discriminator Loss: 0.6921  
Epoch 28/50 | Generator Loss: 1.0993 | Discriminator Loss: 1.0375  
Epoch 29/50 | Generator Loss: 1.1910 | Discriminator Loss: 1.2646  
Epoch 30/50 | Generator Loss: 1.3841 | Discriminator Loss: 1.0203

Epoch 30

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Epoch 30

Epoch 31/50 | Generator Loss: 1.2532 | Discriminator Loss: 1.0982  
Epoch 32/50 | Generator Loss: 1.3170 | Discriminator Loss: 1.7191  
Epoch 33/50 | Generator Loss: 1.2660 | Discriminator Loss: 1.3758  
Epoch 34/50 | Generator Loss: 1.3181 | Discriminator Loss: 0.9501  
Epoch 35/50 | Generator Loss: 1.3896 | Discriminator Loss: 1.0550  
Epoch 36/50 | Generator Loss: 1.0913 | Discriminator Loss: 1.2456  
Epoch 37/50 | Generator Loss: 1.1644 | Discriminator Loss: 1.3073  
Epoch 38/50 | Generator Loss: 1.1355 | Discriminator Loss: 1.1686  
Epoch 39/50 | Generator Loss: 1.1943 | Discriminator Loss: 1.2741  
Epoch 40/50 | Generator Loss: 1.4775 | Discriminator Loss: 0.9890

Epoch 40

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Epoch 40

Epoch 41/50 | Generator Loss: 1.0110 | Discriminator Loss: 1.6527  
Epoch 42/50 | Generator Loss: 1.3656 | Discriminator Loss: 1.1725  
Epoch 43/50 | Generator Loss: 1.2712 | Discriminator Loss: 1.0234  
Epoch 44/50 | Generator Loss: 1.1048 | Discriminator Loss: 1.0794  
Epoch 45/50 | Generator Loss: 0.9776 | Discriminator Loss: 1.3757  
Epoch 46/50 | Generator Loss: 1.1483 | Discriminator Loss: 1.4964  
Epoch 47/50 | Generator Loss: 0.9877 | Discriminator Loss: 1.1750  
Epoch 48/50 | Generator Loss: 1.1319 | Discriminator Loss: 1.0681  
Epoch 49/50 | Generator Loss: 1.0524 | Discriminator Loss: 1.7417  
Epoch 50/50 | Generator Loss: 1.1928 | Discriminator Loss: 1.1307

Epoch 50

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Epoch 48/50 | Generator Loss: 1.1319 | Discriminator Loss: 1.0081  
Epoch 49/50 | Generator Loss: 1.0524 | Discriminator Loss: 1.7417  
Epoch 50/50 | Generator Loss: 1.1928 | Discriminator Loss: 1.1307

Epoch 50

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