| Marwadi University Marwadi Chandarana Group | Marwadi University Faculty of Engineering and Technology Department of Information and Communication Technology | |
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| Subject: Gen AI | Write a code for VANILLA GAN. USE DATASET MNIST FROM KERAS. | |
| Experiment | Date: | Enrolment No:92200133020 |

CODE:

```
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential, Model
from tensorflow.keras.layers import Dense, LeakyReLU, BatchNormalization, Reshape, Flatten, Input
from tensorflow.keras.optimizers import Adam
                                                                                            Python
(X_{train}, _), (_, _) = mnist.load_data()
X_{train} = (X_{train.astype}(np.float32) - 127.5) / 127.5
X_train = np.expand_dims(X_train, axis=-1) # Add channel dimension
                                                                                            Python
# 1. Parameters
img_rows, img_cols, channels = 28, 28, 1
img_shape = (img_rows, img_cols, channels)
z_dim = 100
                                                                                            Python
                                                                 Activate Windows
def build_generator(z_dim):
   model = Sequential()
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model.add(Dense(256, input_dim=z_dim))
   model.add(LeakyReLU(alpha=0.2))
   model.add(BatchNormalization(momentum=0.8))
   model.add(Dense(512))
   model.add(LeakyReLU(alpha=0.2))
   model.add(BatchNormalization(momentum=0.8))
   model.add(Dense(1024))
   model.add(LeakyReLU(alpha=0.2))
   model.add(BatchNormalization(momentum=0.8))
   model.add(Dense(np.prod(img_shape), activation='tanh'))
   model.add(Reshape(img_shape))
    return model
                                                                                             Python
def build_discriminator(img_shape):
   model = Sequential()
   model.add(Flatten(input_shape=img_shape))
   model.add(Dense(512))
   model.add(LeakyReLU(alpha=0.2))
   model.add(Dense(256))
   model.add(LeakyReLU(alpha=0.2))
   model.add(Dense(1, activation='sigmoid'))
    return model
                                                                 Activate Windows
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```

```
optimizer = Adam(0.0002, 0.5)
                                                                                               Python
(X_{train}, _), (_, _) = mnist.load_data()
X_{\text{train}} = (X_{\text{train.astype}}(\text{'float32'}) - 127.5) / 127.5
X_train = np.expand_dims(X_train, axis=-1)
                                                                                               Python
generator = build_generator(z_dim)
discriminator = build_discriminator(img_shape)
# Compile the discriminator
discriminator.compile(loss='binary crossentropy', optimizer=optimizer, metrics=['accuracy'])
discriminator.trainable = False
# Build and compile the combined model
z = Input(shape=(z dim,))
img = generator(z)
valid = discriminator(img)
                                                                   Activate Windows
combined = Model(z, valid)
combined.compile(loss='binary_crossentropy', optimizer=optimizer) Go to Settings to activate Windows.
Click to add a breakpoint
patcn_size = 128
save_interval = 1000
half_batch = batch_size // 2
for epoch in range(1, epochs + 1):
    idx = np.random.randint(0, X_train.shape[0], half_batch)
    real_imgs = X_train[idx]
    noise = np.random.normal(0, 1, (half_batch, z_dim))
    fake_imgs = generator.predict(noise)
    d_loss_real = discriminator.train_on_batch(real_imgs, np.ones((half_batch, 1)))
    d_loss_fake = discriminator.train_on_batch(fake_imgs, np.zeros((half_batch, 1)))
    d_loss = 0.5 * np.add(d_loss_real, d_loss_fake)
    # Train Generator
    noise = np.random.normal(0, 1, (batch_size, z_dim))
    valid_y = np.ones((batch_size, 1))
    g_loss = combined.train_on_batch(noise, valid_y)
    if epoch % save interval == 0 or epoch == 1:
         print(f"{epoch} [D loss: {d_loss[0]:.4f}, acc.: {100*d_loss[1]:.2f}%] [G loss: {g_loss:.
```

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| 2/2 | Os 30ms/step |
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| 2/2 | Os 49ms/step |
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