Cover Sheet

Faculty name: Computers and Artificial Intelligence - Helwan

university

Course name: Selected Topics CS-2

Team number 18

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Paper details

Paper name: UNSUPERVISED REPRESENTATION LEARNING WITH DEEP CONVOLUTIONAL GENERATIVE ADVERSARIAL NETWORKS

Authors name: Kaiming He, Xiangyu Zhang, Shaoqing Ren, and

Jian Sun

Publisher name: Soumith Chintala

Content: DC GAN

Year of publish: 2016

Dataset : The ImageNet

About dataset : The ImageNet dataset is a large-scale image classification dataset that contains over 1.2 million images across 1,000 categories. The dataset is widely used in computer vision research and is often used as a benchmark for evaluating new image classification models.

The implemented algorithms: the ResNet architecture

Results: the authors trained ResNet models with up to 152 layers on the ImageNet dataset and achieved a top-5 error rate of 3.57%, which was significantly better than the previous state-of-the-art result of 3.57%. They also showed that ResNet models can be trained faster and with fewer parameters than previous deep neural network architectures.

General Information on the selected dataset

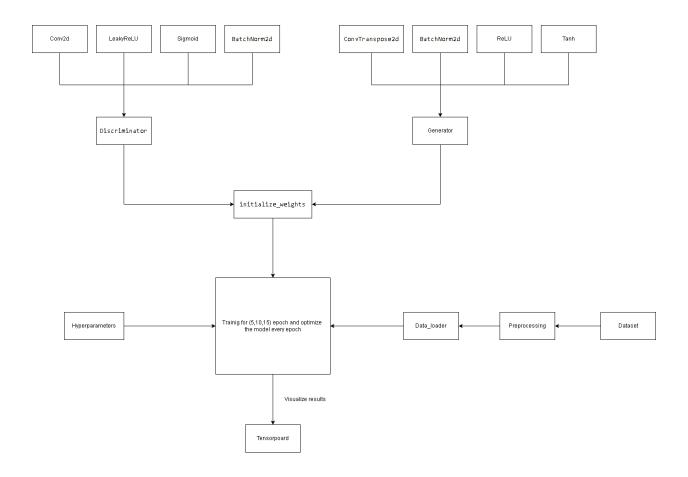
Name: Human Faces

Desciption : A web scraped dataset of human faces suggested for image processing models

Link: https://www.kaggle.com/datasets/ashwingupta3012/human-faces

Total number of samples : 7219

Block diagram for the model



Hyperparameters

Learning rate : 2e-4 (0.0002)

Batch size : {64, 128, 256}

Image size: 64

Noise Dimensions: 100

Number of epochs : {5, 10, 15}

Discriminator features: 64

Generator features: 64

Optimizer: Adam

Results details

First run

Hyperparameters:

```
Learning_Rate = 2e-4
Batch_Size = 128
Image_size = 64
Channels_img = 3
Z_dim = 100
Num Epochs = 5
```

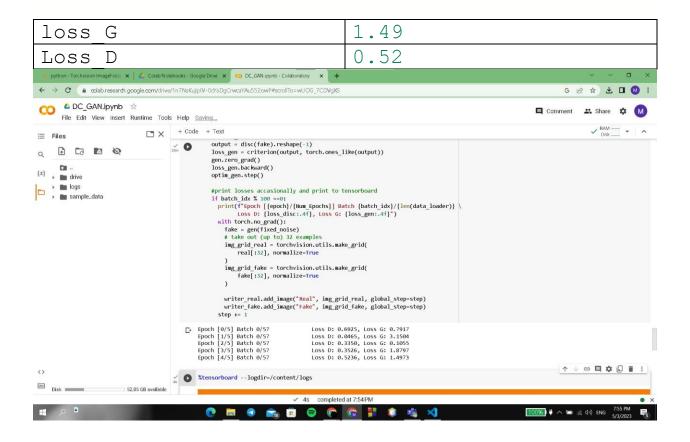
CPU

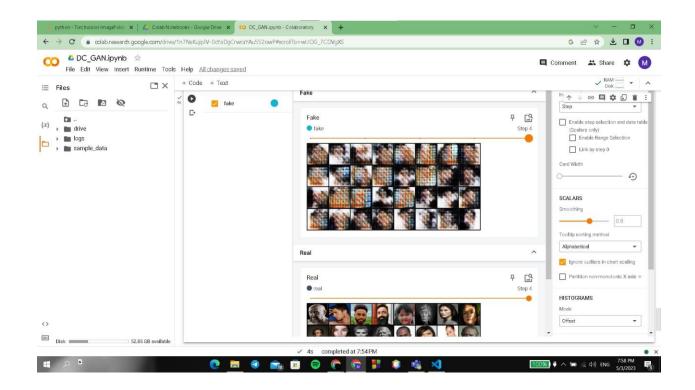
```
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Loss
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                                                                                                                  Comment A Share $ M
     File Edit View Insert Runtime Tools Help All changes saved
                                                                                                                              ✓ RAM → ^
           loss_gen = criterion(output, torch.ones_like(output))
gen.zero_grad()
loss_gen.backward()
optim_gen.step()
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\{x\}
                     accasionally and print to tensorboard
           img_grid_fake = torchvision.utils.make_grid(
    fake[:32], normalize=True
              <>
```

Second run

Hyperparameters:

```
Learning_Rate = 2e-4
Batch_Size = 128
Image_size = 64
Channels_img = 3
Z_dim = 100
Num_Epochs = 5
```

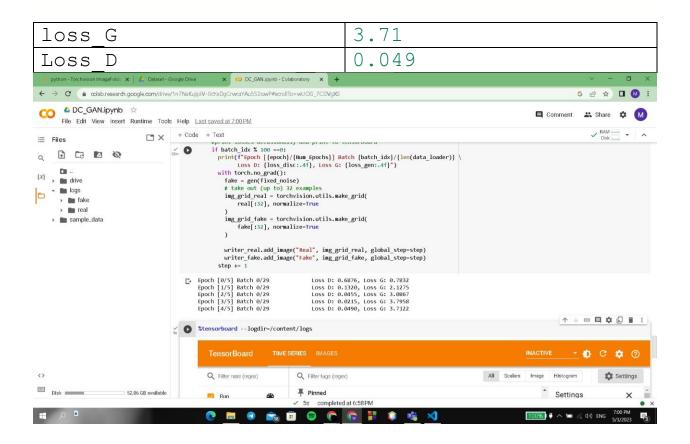


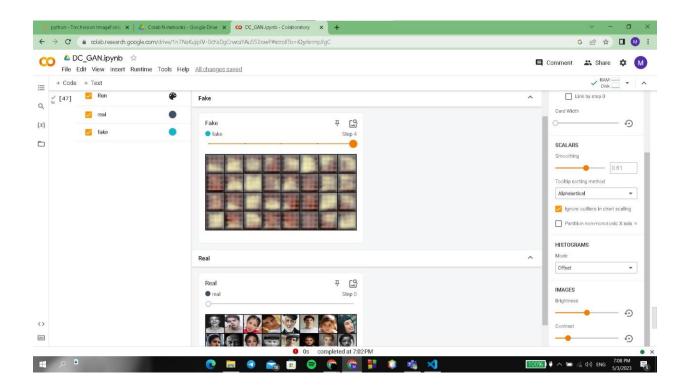


Third run

Hyperparameters:

```
Learning_Rate = 2e-4
Batch_Size = 256
Image_size = 64
Channels_img = 3
Z_dim = 100
Num Epochs = 5
```

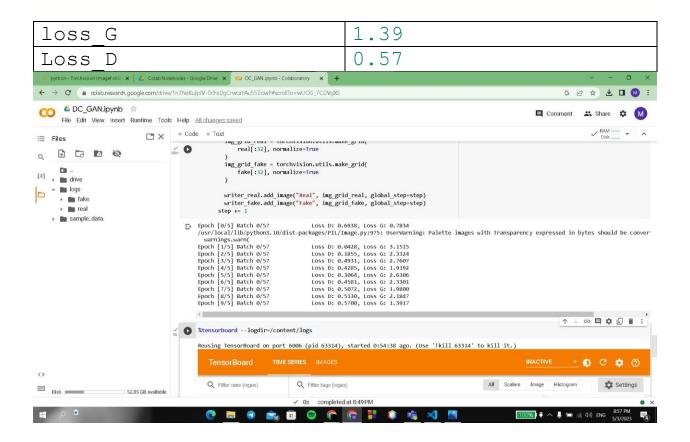


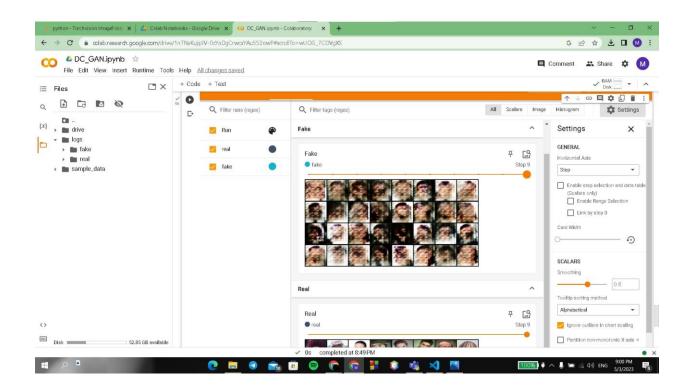


Fourth run

Hyperparameters:

```
Learning_Rate = 2e-4
Batch_Size = 128
Image_size = 64
Channels_img = 3
Z_dim = 100
Num_Epochs = 10
```





Fifth run

Hyperparameters:

```
Learning_Rate = 2e-4
Batch_Size = 64
Image_size = 64
Channels_img = 3
Z_dim = 100
Num_Epochs = 10
```

