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Introduction to Intelligent Computing : Final Project Report

Methodology

The first thing we did about our classifier was to fix the labels after having imported the data. So we replaced spaces by “_”. The model we choose for our classifier is a CNN classifier. The different parameters we used are : 30 classes, a width and height of 64, 8000 images per class, 800 valid images per class and 1 test image per class. The next step was to code a drawing function. We also used another database to have more data so we can train more our model. We reshaped, normalized and split the data into X_train, Y_train and validation data set. Then, we tried to improve our model using a more complex one.

As for the generator, we used bot DCGAN and ACGAN but the methodology between them is almost the same. We import the data like we did for the classifier and normalize the images. Then, we defined the generator model and also the discriminator model. The discriminator evaluates the drawings giving them a score : a positive one is a “good” image and a negative one a “bad” image. We added checkpoints in order to be able to backtrack if needed.

How to train the model

The classifier uses a categorical cross entropy function to evaluate the loss and the Adam optimizer. We use two callbacks : an early stopping function and the reduceLROnPlat one. We use a batch size of 32, 50 epochs and a validation split of 5% (**Illustration 1**). We printed two graphs measuring the accuracy and the loss based on the epochs number. Those results were decent but they can be improved (**Illustration 2**), so we tried to improve our model. It has more layers and we only use 20 epochs this time (**Illustration 3**). We printed the graphs once more to notice that it improved our

results (*Illustration 4*). We tried to improve it once more using data augmentation and then training it on 100 epochs.

We trained the generator using 100 epochs and a DC/ACGAN (*Illustration 6*). We also used a loss measurement function and the Adam optimizer. Then, we printed some of the drawings so we were able to see how accurate our model is. Our discriminator function also helped to remove automatically not significant drawings (*Illustration 5*).

Test Result

The test results of our classifier evolved since we improved our model twice. First, the testing accuracy was around 38% (*Illustration 2*). But with our next improvement, our score improved a lot to the value of approximately 70% (*Illustration 4*). However, the last model improvement did not change much our previous results.

To measure the test results of the generator, it was more subjective because of the fact that only us can tell if the drawing is accurate or not, but with the discriminator function, we ended up having a really good precision among test drawings (*Illustrations 7&8*).

Demo Result

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Discussion

This project was a first approach of AI recognition and generation for us so we all learned a lot doing it. We met some issues doing it such as the GPU limitation. None of us had one, so we used GPU mode on Google Collab. However, since the training and prediction processes can be very long, we sometimes got disconnected so we have to start over again. In addition, our misunderstanding of the demo part may decrease our final grade a lot which disappoints us since we were confident about our models.

Conclusion

The main goal of this project was to be able to create a model that can classify but mainly generate accurate drawings. Using a CNN classifier and GANs, we think that we managed to develop models that will meet this project expectations.

As a last word, we can say that since the AI recognition and generation domains are expanding a lot nowadays, this project gave us some tools to know more about them. Thus, it may help us for the following of our studies and eventually our future jobs.

Illustrations

Model: "sequential_8"

Layer (type)	Output Shape	Param #
conv2d_32 (Conv2D)	(None, 64, 64, 32)	320
max_pooling2d_20 (MaxPooling)	(None, 32, 32, 32)	0
conv2d_33 (Conv2D)	(None, 32, 32, 64)	18496
max_pooling2d_21 (MaxPooling)	(None, 16, 16, 64)	0
dropout_12 (Dropout)	(None, 16, 16, 64)	0
flatten_8 (Flatten)	(None, 16384)	0
dense_16 (Dense)	(None, 680)	11141800
dropout_13 (Dropout)	(None, 680)	0
dense_17 (Dense)	(None, 30)	20430

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Total params: 11,181,046
Trainable params: 11,181,046
Non-trainable params: 0

Illustration 1: 1st classifier model

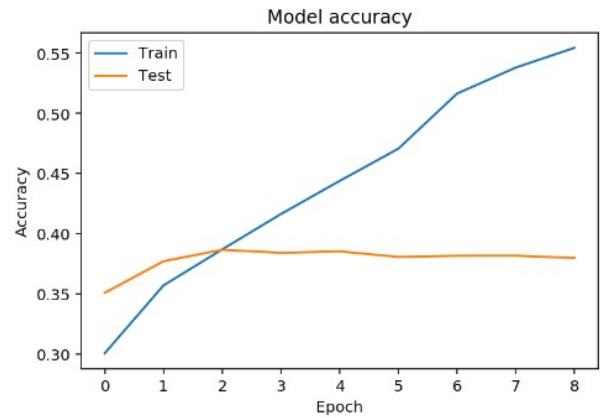


Illustration 2: 1st classifier results

Model: "sequential_10"

Layer (type)	Output Shape	Param #
conv2d_40 (Conv2D)	(None, 64, 64, 16)	144
batch_normalization_18 (Batch Normalization)	(None, 64, 64, 16)	64
activation_18 (Activation)	(None, 64, 64, 16)	0
conv2d_41 (Conv2D)	(None, 64, 64, 16)	2304
batch_normalization_19 (Batch Normalization)	(None, 64, 64, 16)	64
activation_19 (Activation)	(None, 64, 64, 16)	0
max_pooling2d_25 (MaxPooling)	(None, 32, 32, 16)	0
conv2d_42 (Conv2D)	(None, 32, 32, 32)	4608
batch_normalization_20 (Batch Normalization)	(None, 32, 32, 32)	128
activation_20 (Activation)	(None, 32, 32, 32)	0
conv2d_43 (Conv2D)	(None, 32, 32, 32)	9216
batch_normalization_21 (Batch Normalization)	(None, 32, 32, 32)	128
activation_21 (Activation)	(None, 32, 32, 32)	0
max_pooling2d_26 (MaxPooling)	(None, 16, 16, 32)	0
conv2d_44 (Conv2D)	(None, 16, 16, 64)	18432
batch_normalization_22 (Batch Normalization)	(None, 16, 16, 64)	256
activation_22 (Activation)	(None, 16, 16, 64)	0
conv2d_45 (Conv2D)	(None, 16, 16, 64)	36864
batch_normalization_23 (Batch Normalization)	(None, 16, 16, 64)	256
activation_23 (Activation)	(None, 16, 16, 64)	0
max_pooling2d_27 (MaxPooling)	(None, 8, 8, 64)	0
dropout_15 (Dropout)	(None, 8, 8, 64)	0
flatten_10 (Flatten)	(None, 4096)	0
dense_20 (Dense)	(None, 512)	2097664
dense_21 (Dense)	(None, 30)	15390

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Total params: 3,436,632
Trainable params: 2,659,321
Non-trainable params: 777,311

Illustration 3: 2nd classifier model

Discriminator model:
Model: "model_9"

Layer (type)	Output Shape	Param #	Connected to
input_16 (InputLayer)	[(None, 28, 28, 1)]	0	
sequential_6 (Sequential)	(None, 12544)	387840	input_16[0][0]
generation_1 (Dense)	(None, 1)	12545	sequential_6[1][0]
auxiliary_1 (Dense)	(None, 30)	376350	sequential_6[1][0]

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Total params: 776,735
Trainable params: 776,735
Non-trainable params: 0

Illustration 5: Discriminator model

Combined model:
Model: "model_11"

Layer (type)	Output Shape	Param #	Connected to
input_19 (InputLayer)	[(None, 100)]	0	
input_20 (InputLayer)	[(None, 1)]	0	
model_10 (Model)	(None, 28, 28, 1)	2659897	input_19[0][0] input_20[0][0]
model_9 (Model)	[(None, 1), (None, 3)]	776735	model_10[1][0]

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Total params: 3,436,632
Trainable params: 2,659,321
Non-trainable params: 777,311

Illustration 6: Generator model

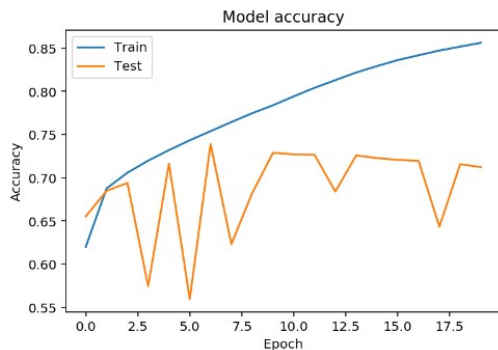


Illustration 4: 2nd classifier results

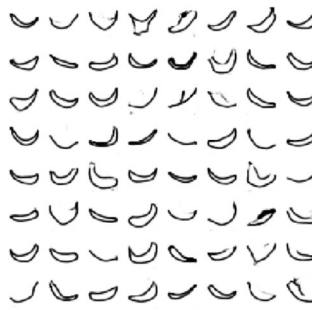


Illustration 7: 64 generated bananas

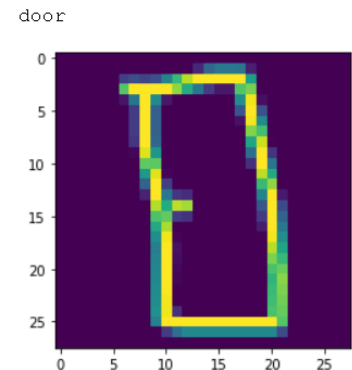


Illustration 8: Generated drawing and its label