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# sketch assistant : Human in the loop application to drawing sketches

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## Abstract

Sketch Assistant to convert text to image using Generative adversarial network.

## 1 Introduction

Aim of the project is to generate image of sketch by using text as input. These types of applications can be used in application which involve modeling or where user is not able to draw sketch. As said in the proposal, we'll be creating application which is Human in the loop based which takes in description of the image as input and generate an image, If the user isn't satisfied with image generated, he/she provides feedback via natural language instruction. This feedback is fed into network to improvise on the image generated. We are concentrating on particular set of images involving birds obtained from sketchy dataset[1] and handcraft corpus involving limited amount of topics.

### 1.1 Dataset Preparation

This project seemed to be intensive **interms** of preparing data needed for engineering. Following are the initial steps taken to clean up the data. 1) Obtained data in the svg format and generated strokes out the images. 2) generates images with incremental strokes 3) resize it to fit into the model 4) dataset given was in RGBA format which needed to be converted to grayscale.

We tried couple of approaches to annotate the data, following are the findings

- **Unsupervised learning**

We tried to cluster strokes[2] by length/pixel density and find the group of strokes.Used Kmeans clustering to group the strokes and annotate groups. As there was no symmetry in the images drawn, the clusters generated were not of any use. when K was low the stroke were more generic and when K was high smaller number of outlier strokes didn't contribute to drawings.

- **Supervised learning**

Hand annotate images into 8 set of categories and use them as input to GAN, we annotated around 1000 images into 8 categories.

- **Semi-supervised learning**

Use Handannotated images to identify classes of unannotated images. Implemented convolutional neural network to classify images into attributes. Accuracy of 0.48 was low as the unannotated images were diverse interms of strokes.

## 2 Modeling

We are using Generative Adversarial Text to Image Synthesis paper as reference to device GAN model with incorporating feedback network. Current network consists of following layers:

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generator:

Layer (type)	Output Shape	Param #	Connected to
input_4 (InputLayer)	(None, 200, 200, 1)	0	
conv2d_3 (Conv2D)	(None, 200, 200, 64)	6464	input_4[0][0]
activation_5 (Activation)	(None, 200, 200, 64)	0	conv2d_3[0][0]
max_pooling2d_1 (MaxPooling2D)	(None, 100, 100, 64)	0	activation_5[0][0]
conv2d_4 (Conv2D)	(None, 96, 96, 128)	204928	max_pooling2d_1[0][0]
activation_6 (Activation)	(None, 96, 96, 128)	0	conv2d_4[0][0]
max_pooling2d_2 (MaxPooling2D)	(None, 48, 48, 128)	0	activation_6[0][0]
flatten_1 (Flatten)	(None, 294912)	0	max_pooling2d_2[0][0]
input_3 (InputLayer)	(None, 9)	0	
dense_5 (Dense)	(None, 200)	58982600	flatten_1[0][0]
dense_4 (Dense)	(None, 200)	2000	input_3[0][0]
concatenate_2 (Concatenate)	(None, 400)	0	dense_5[0][0] dense_4[0][0]
activation_7 (Activation)	(None, 400)	0	concatenate_2[0][0]
dense_6 (Dense)	(None, 1)	401	activation_7[0][0]
activation_8 (Activation)	(None, 1)	0	dense_6[0][0]

Total params: 59,196,393

Trainable params: 59,196,393

Non-trainable params: 0

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discriminator

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 9)	0	
input_2 (InputLayer)	(None, 9)	0	
dense_1 (Dense)	(None, 400)	4000	input_1[0][0]
dense_2 (Dense)	(None, 400)	4000	input_2[0][0]
concatenate_1 (Concatenate)	(None, 800)	0	dense_1[0][0] dense_2[0][0]

activation_1 (Activation)	(None, 800)	0	concatenate_1[0][0]
dense_3 (Dense)	(None, 80000)	64080000	activation_1[0][0]
batch_normalization_1 (BatchNorm	(None, 80000)	320000	dense_3[0][0]
activation_2 (Activation)	(None, 80000)	0	batch_normalization_1[0][0]
reshape_1 (Reshape)	(None, 50, 50, 32)	0	activation_2[0][0]
up_sampling2d_1 (UpSampling2D)	(None, 100, 100, 32)	0	reshape_1[0][0]
conv2d_1 (Conv2D)	(None, 100, 100, 32)	102432	up_sampling2d_1[0][0]
activation_3 (Activation)	(None, 100, 100, 32)	0	conv2d_1[0][0]
up_sampling2d_2 (UpSampling2D)	(None, 200, 200, 32)	0	activation_3[0][0]
conv2d_2 (Conv2D)	(None, 200, 200, 1)	801	up_sampling2d_2[0][0]
activation_4 (Activation)	(None, 200, 200, 1)	0	conv2d_2[0][0]
model_2 (Model)	(None, 1)	59196393	activation_4[0][0] input_2[0][0]

## 2.1 Evaluation

Evaluating the quality of these systems is a hard problem. Salimans propose to use an image classifier to evaluate the quality of generated images called Inception score. The intuition behind this that we want out model to generate meaningful objects. We could use top-r metric for selecting r relevant images out of all the images generated by the network.

## 3 Dataset

Sketch Dataset , which contains pairs of images and sketches for 125 categories and acquires 75,471 sketches.

Annotations generated for birds includes file head beak wing tail body leg full eyes and image\_id

We took category bird from the dataset for our experiment.

## 4 References

- [1] <http://sketchy.eye.gatech.edu/>
- [2] Clustering Hand-Drawn Sketches via Analogical Generalization, Maria D. Chang, Kenneth D. Forbus ,Qualitative Reasoning Group, Northwestern University
- [3] Generative Adversarial Text to Image Synthesis, Scott Reed, Zeynep Akata, Xinchun Yan, Lajanugen Logeswaran , Proceedings of the 33 rd International Conference on Machine Learning, New York, NY, USA, 2016