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## **Homework-7 Report**

### 1. Generative Models for Text:

- a. Building a generative model to mimic the writing style of prominent British Mathematician, Philosopher, prolific writer, and political activist, Bertrand Russell.
- b. Downloaded the following 4 textbooks in text format:
  - i. The Problems of Philosophy
  - ii. The Analysis of Mind
  - iii. Mysticism and Logic and Other Essays
  - iv. Our Knowledge of the External World as a Field for Scientic Method in Philosophy

## c. LSTM:

- i. Concatenated the textbooks downloaded in 1.b.
- ii. Total Characters: 1611845 Total Vocabulary: 100
- iii. Window size selected as W=100.
- iv. --
- v. Total Patterns: 1611745
- vi. --
- vii. Used Softmax output layer to yield a probability prediction for each of the characters between 0 and 1 and used log loss (cross entropy) as the objective function for the network.
- viii. Used whole dataset as Training dataset.
- ix. Number of Epochs chosen=15.
- x. Model checkpoint is used to keep the network weights to determine each time an improvement in loss is observed at the end of the epoch.

Epoch 1/15

1611745/1611745 [===========] - 387s 240us/step - loss: 3.0493

Epoch 00001: loss improved from inf to 3.04928, saving model to weights-improvement-01-3.0493.hdf5

Epoch 2/15

loss: 2.8945

Epoch 00002: loss improved from 3.04928 to 2.89454, saving model to weights-improvement-02-2.8945.hdf5

Epoch 3/15

1611745/1611745 [============] - 387s 240us/step -

loss: 2.8276

Epoch 00003: loss improved from 2.89454 to 2.82755, saving model to weightsimprovement-03-2.8276.hdf5 Epoch 4/15 loss: 2.7838 Epoch 00004: loss improved from 2.82755 to 2.78377, saving model to weightsimprovement-04-2.7838.hdf5 Epoch 5/15 loss: 2.7462 Epoch 00005: loss improved from 2.78377 to 2.74624, saving model to weightsimprovement-05-2.7462.hdf5 Epoch 6/15 1611745/1611745 [============] - 388s 241us/step loss: 2.7128 Epoch 00006: loss improved from 2.74624 to 2.71278, saving model to weightsimprovement-06-2.7128.hdf5 Epoch 7/15 loss: 2.6830 Epoch 00007: loss improved from 2.71278 to 2.68295, saving model to weightsimprovement-07-2.6830.hdf5 Epoch 8/15 loss: 2.6541 Epoch 00008: loss improved from 2.68295 to 2.65409, saving model to weightsimprovement-08-2.6541.hdf5 Epoch 9/15 loss: 2.6260 Epoch 00009: loss improved from 2.65409 to 2.62601, saving model to weightsimprovement-09-2.6260.hdf5 Epoch 10/15 1611745/1611745 [============] - 388s 241us/step loss: 2.5979 Epoch 00010: loss improved from 2.62601 to 2.59792, saving model to weightsimprovement-10-2.5979.hdf5 Epoch 11/15 loss: 2.5702

Epoch 00011: loss improved from 2.59792 to 2.57024, saving model to weights-improvement-11-2.5702.hdf5

Epoch 12/15

1611745/1611745 [===========] - 388s 241us/step -

loss: 2.5438

Epoch 00012: loss improved from 2.57024 to 2.54382, saving model to weights-improvement-12-2.5438.hdf5

Epoch 13/15

loss: 2.5181

Epoch 00013: loss improved from 2.54382 to 2.51808, saving model to weights-improvement-13-2.5181.hdf5

Epoch 14/15

1611745/1611745 [===========] - 388s 241us/step -

loss: 2.4926

Epoch 00014: loss improved from 2.51808 to 2.49257, saving model to weights-improvement-14-2.4926.hdf5

Epoch 15/15

1611745/1611745 [===============] - 390s 242us/step -

loss: 2.4691

Epoch 00015: loss improved from 2.49257 to 2.46909, saving model to weights-improvement-15-2.4691.hdf5

<keras.callbacks.History at 0x7fe3209b2e10>

The best set of weights in terms of loss is in Epoch 15 with a loss of 2.4691

xi. Sentence used for prediction: There are those who take mental phenomena naively, just as they would physical phenomena. This school of psychologists tends not to emphasize the object.

Output: is the sere th the the sore of the sere th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sore and the sare "and the sored of the sere th the the soene th the sored of the sere th the sored of

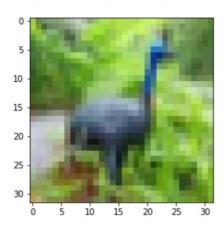
We can see that the above predicted output isn't accurate. We can improve the performance of the model by increasing the number of Epoch's or by inputting more data (textbooks).

## 2. (Deep) CNNs for Image Colorization:

- a. Dataset downloaded
- b. Extracted the class birds from train and test datasets.

Shape of the data: (6000, 32, 32, 3)

One of the images:



#### c. Pixel's selected:

[[164 206 84]

[105 140 61]

[118 148 101]

...

[156 179 95]

[158 180 98]

[157 179 98]]

d. K-means clustering on the P vectors using k = 4:

KMeans(algorithm='auto', copy\_x=True, init='k-means++', max\_iter=300, n\_clusters=4, n\_init=10, n\_jobs=None, precompute\_distances='auto', random\_state=None, tol=0.0001, verbose=0)

Cluster centers:

array([[156.37287845, 155.66372502, 135.67160076],

[206.58262579, 211.86674913, 211.57118882],

[49.07657012, 50.05161286, 38.58521485],

[109.02727327, 108.6914443, 82.80830299]])

e. Original to Grayscale conversion is done using skimage.color rgb2grey function.

# f. Deep Convolutional Neural Network: Model summary:

Layer (type)	Output Shape P	aram #			
conv2d_1 (Conv2D)	(None, 32, 32, 64)	1664	 		
max_pooling2d_1 (N	MaxPooling2 (None, 32,	32, 64)	0		
conv2d_2 (Conv2D)	(None, 32, 32, 64)	1024	164		
max_pooling2d_2 (N	NaxPooling2 (None, 32,	32, 64)	0		
dense_1 (Dense)	(None, 32, 32, 32)	2080			
dense_2 (Dense)	(None, 32, 32, 4)	132			
Total parame: 106.2	======================================	======	=====	=======	:====:

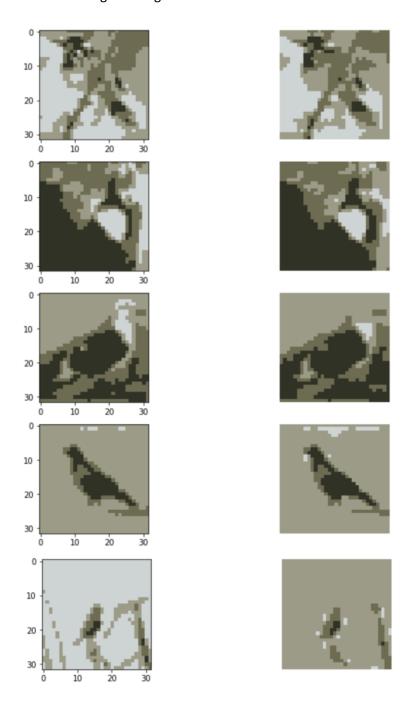
Total params: 106,340 Trainable params: 106,340 Non-trainable params: 0

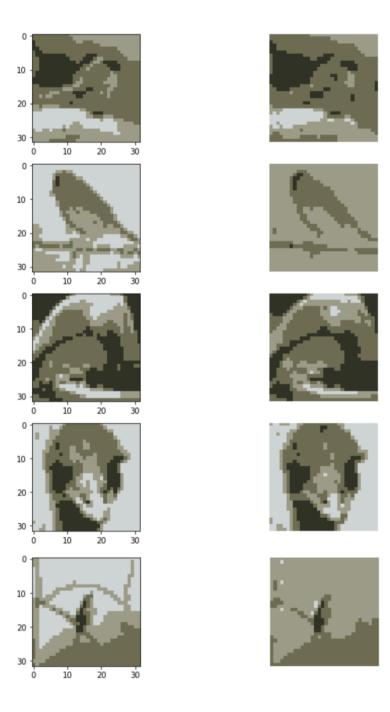
#### Output of each Epoch:

Train on 5400 samples, validate on 600 samples Epoch 1/20

- 2s loss: 0.8433 acc: 0.7128 val\_loss: 0.8026 val\_acc: 0.7208 Epoch 2/20
- 2s loss: 0.7774 acc: 0.7242 val\_loss: 0.7565 val\_acc: 0.7165 Epoch 3/20
- 2s loss: 0.7259 acc: 0.7322 val\_loss: 0.7038 val\_acc: 0.7364 Epoch 4/20
- 2s loss: 0.6895 acc: 0.7356 val\_loss: 0.6700 val\_acc: 0.7340 Epoch 5/20
- 2s loss: 0.6563 acc: 0.7404 val\_loss: 0.6412 val\_acc: 0.7450 Epoch 6/20
- 2s loss: 0.6358 acc: 0.7408 val\_loss: 0.6166 val\_acc: 0.7448 Epoch 7/20
- 2s loss: 0.6150 acc: 0.7421 val\_loss: 0.6206 val\_acc: 0.7218 Epoch 8/20
- 2s loss: 0.5951 acc: 0.7451 val\_loss: 0.5807 val\_acc: 0.7479 Epoch 9/20
- 2s loss: 0.5750 acc: 0.7489 val\_loss: 0.6060 val\_acc: 0.7083 Epoch 10/20
- 2s loss: 0.5649 acc: 0.7468 val\_loss: 0.5714 val\_acc: 0.7258 Epoch 11/20
- 2s loss: 0.5519 acc: 0.7477 val\_loss: 0.5429 val\_acc: 0.7456 Epoch 12/20
- 2s loss: 0.5386 acc: 0.7488 val\_loss: 0.5317 val\_acc: 0.7519 Epoch 13/20
- 2s loss: 0.5274 acc: 0.7488 val\_loss: 0.5197 val\_acc: 0.7518 Epoch 14/20
- 2s loss: 0.5163 acc: 0.7491 val\_loss: 0.5115 val\_acc: 0.7513 Epoch 15/20
- 2s loss: 0.5072 acc: 0.7485 val\_loss: 0.4973 val\_acc: 0.7493 Epoch 16/20
- 2s loss: 0.4940 acc: 0.7499 val\_loss: 0.4893 val\_acc: 0.7551 Epoch 17/20
- 2s loss: 0.4870 acc: 0.7485 val\_loss: 0.4854 val\_acc: 0.7573 Epoch 18/20
- 2s loss: 0.4814 acc: 0.8098 val\_loss: 0.4805 val\_acc: 0.8496 Epoch 19/20
- 2s loss: 0.4669 acc: 0.8575 val\_loss: 0.4659 val\_acc: 0.8668 Epoch 20/20
- 2s loss: 0.4611 acc: 0.8587 val\_loss: 0.4594 val\_acc: 0.8469

Comparing the artificially colored versions of the first 10 images in the test set with the original images:





Plotting training and test errors:

