

Question 4: GAN + reconstruction loss term

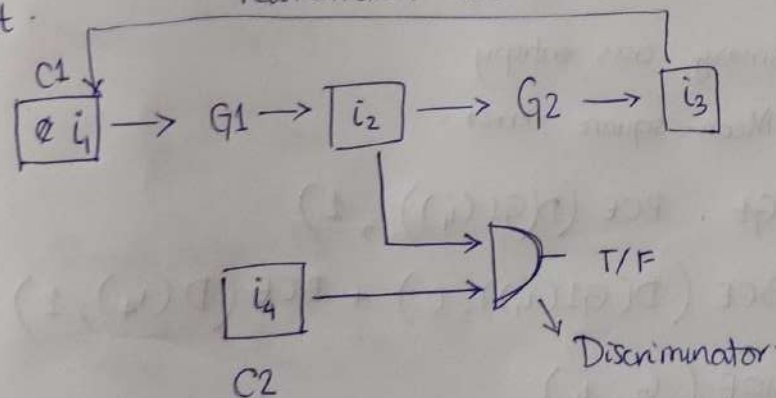
3 neural networks are used.

G1 : takes input from C1 sampled images i_1 & is ^{expected} to output modified image i_2 with color of category C2.

C2 : { generator - 1 }

G2 : { generator - 2 } : takes image (modified) i_2 from output of G1 as input & tries to reconstruct image i_1 by producing i_3 . It is expected that image i_3 is as close as possible to image i_1 .

D : Discriminator : Trains to correctly classify whether image given to it as input is from category C2 or not.



Reconstruction loss \rightarrow Mean squared error between elements of two images.

This loss gradient information updates weights of both $G1$ as well as $G2$.

$G1$ has to create images i_2 such that discriminator fails to reject it. This component also updates weights for $G1$.

Discriminator has to

- correctly classify images i_4 from G_2 .

- correctly reject images i_2 from $G1$.

Both these components update weights for D .

Sequence of training operation is provided via code.

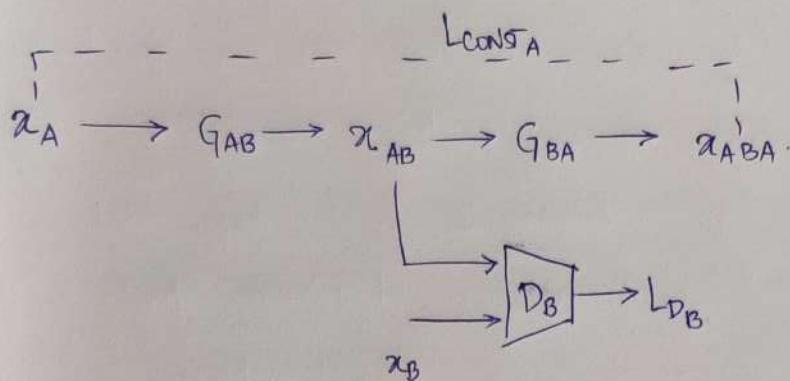
BCE - Binary cross entropy

MSE - Mean-square error.

$$L1 = G1 \text{ BCE}(D(G1(i_1)), 1)$$

$$L2 = \text{BCE}(D(G1(i_1)), 0) + \text{BCE}(D(i_4), 1)$$

$$L3 = \text{MSE}(i_3, i_4)$$



$G_{AB} (G1) :$

$$R_{G_{AB}} = \mathbb{E}_{x_A \sim A} d(x_A, x_{ABA})$$

$$GAN_{G_{AB}} = \mathbb{E}_{x_A \sim A} \log D_B(G_{AB}(x_A))$$

$$L_{G_{AB}} = R_{G_{AB}} + GAN_{G_{AB}}$$

$G_{BA} (G2) :$

$$R_{G_{BA}} = \mathbb{E}_{x_A \sim A} d(x_A, x_{ABA})$$

$$L_{G_{BA}} = R_{G_{BA}}$$

Discriminator $D_B (D) :$

$$L_{D_B}^{real} = \mathbb{E}_{x_B \sim B} \log D_B(x_B)$$

$$L_{D_B}^{gen} = \mathbb{E}_{x_B \sim B} \log (1 - D_B(G_{AB}(x_A)))$$

$$L_{D_B} = L_{D_B}^{real} + L_{D_B}^{gen}$$