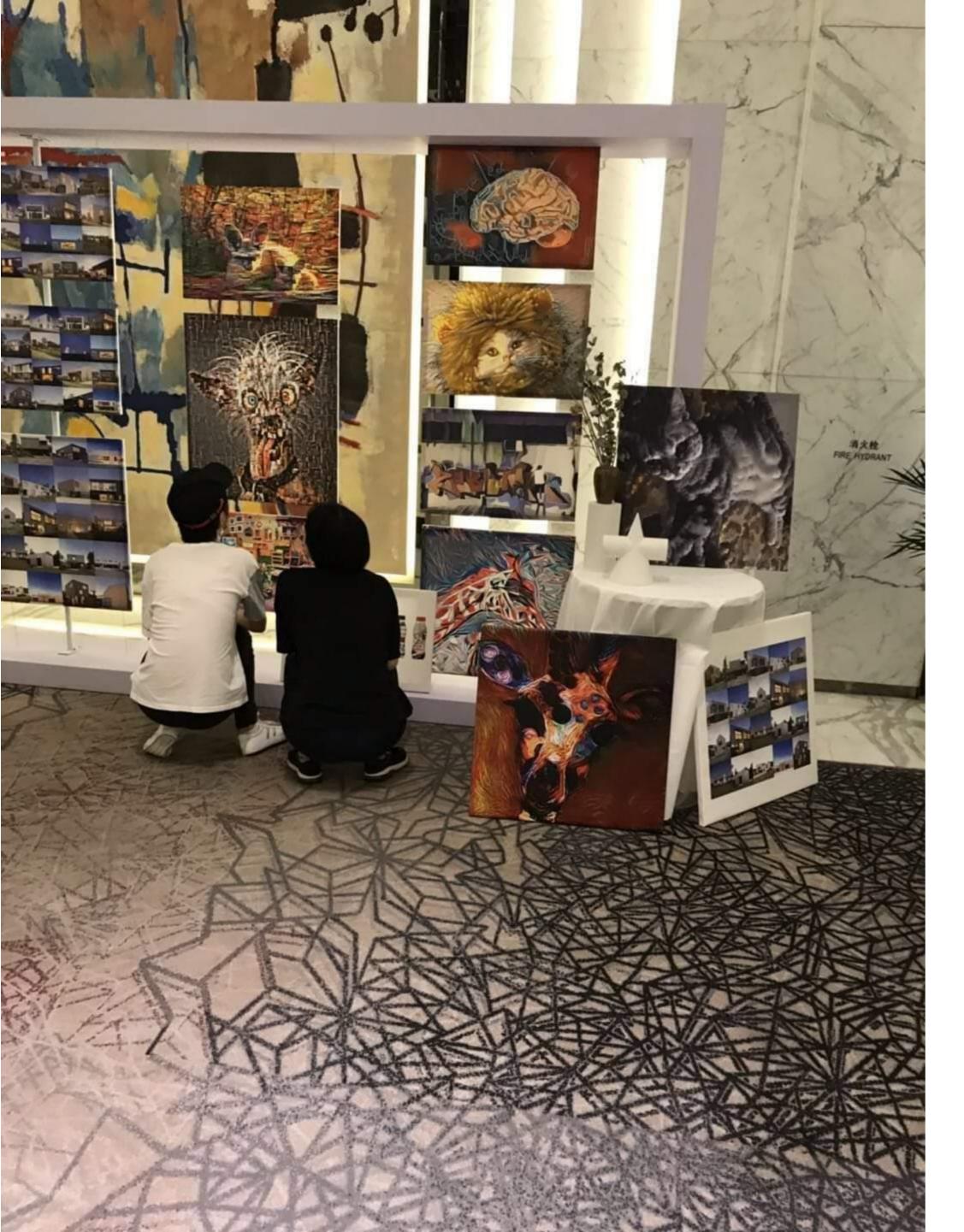


實做案例位置 github

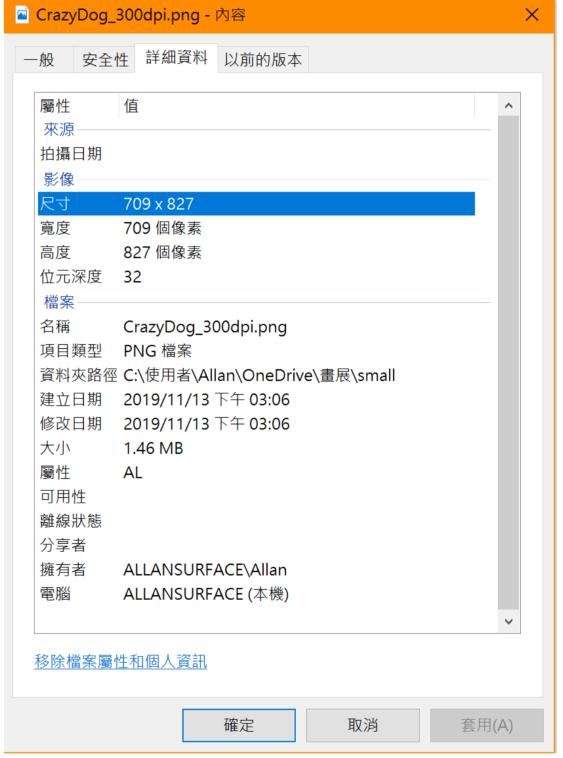
https://github.com/AllanYiin/DeepBelief_Course5_Examples/tree/master/epoch402_越放大越清晰

colab

https://drive.google.com/open?id=1ppB5QOt4bE7ZprMZAZ37rSWZCpJ4DSLG







LR **放大策略**SR



線性插值

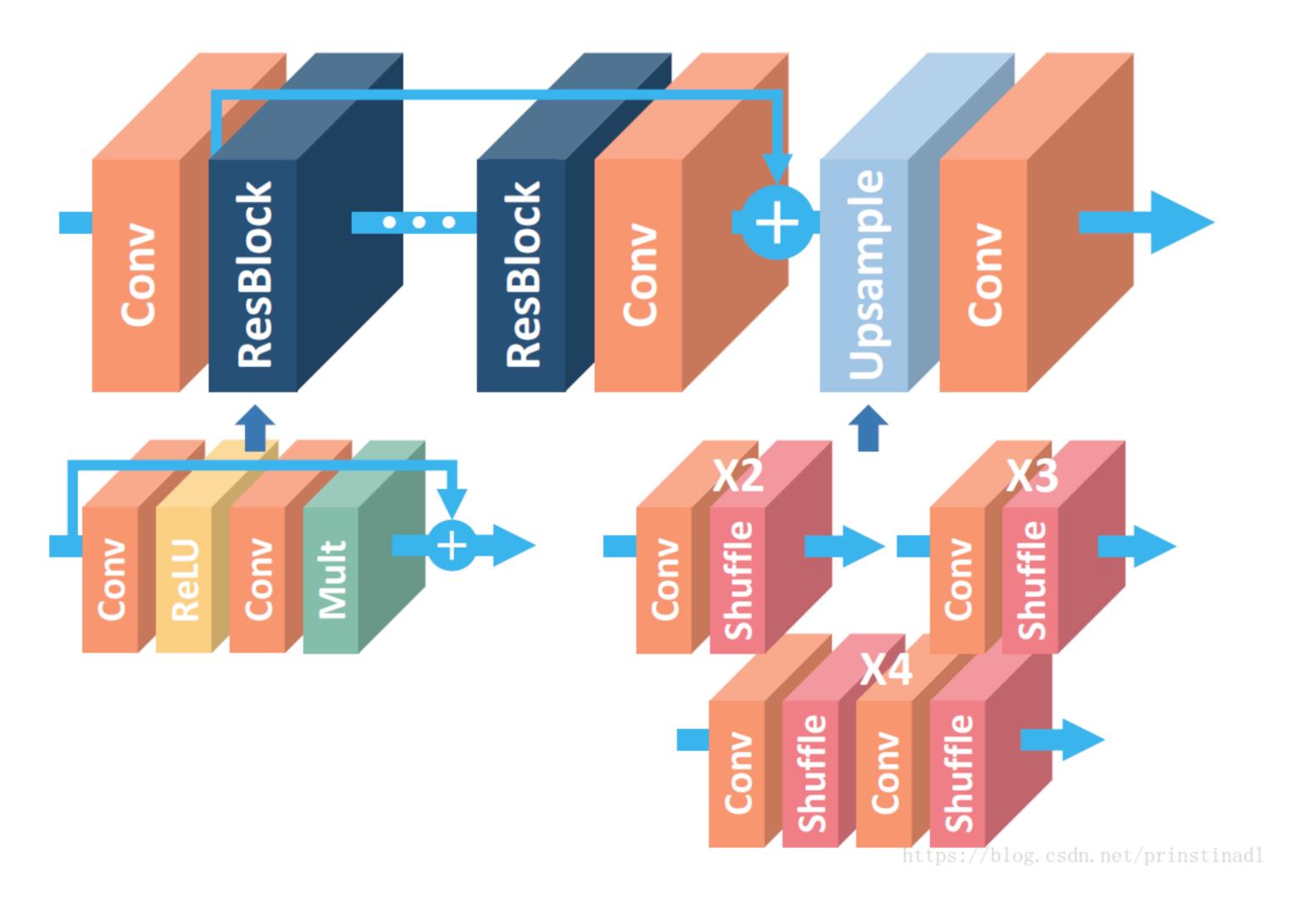






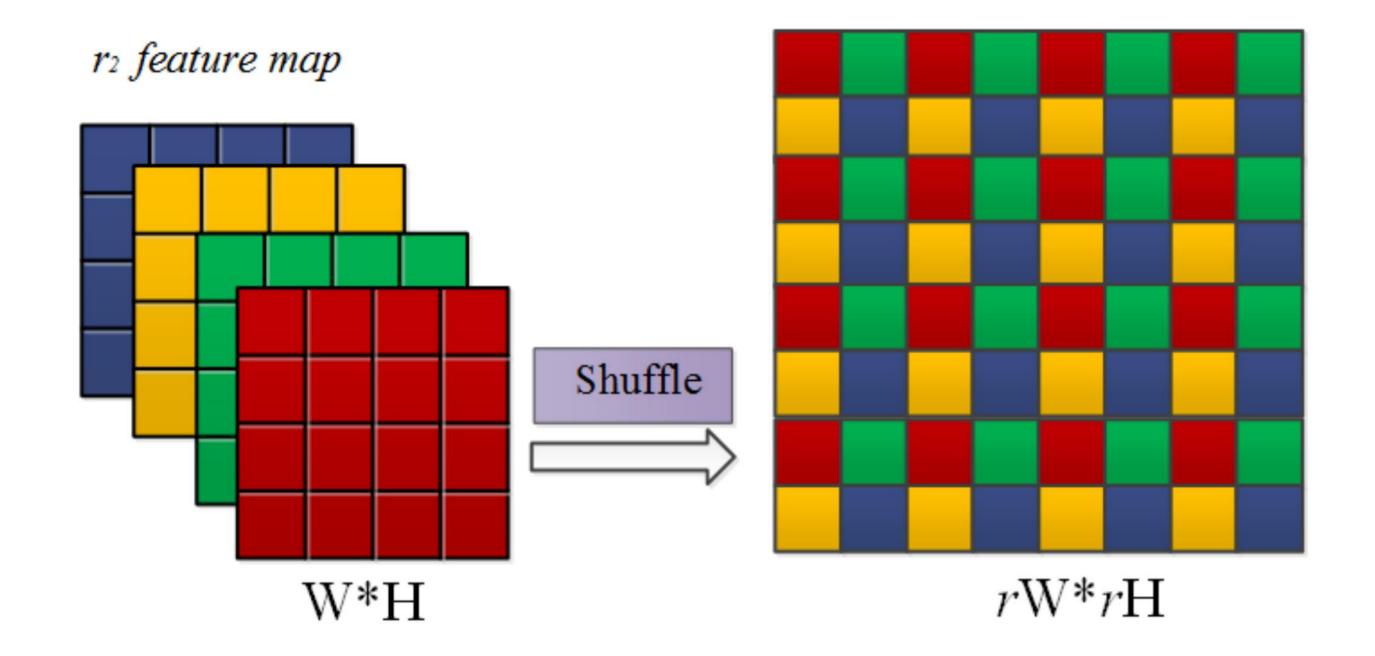
HR





原論文改良

- 把relu換成parameter relu
- 使用pixel shffule



Pixel Shuffle

計算流程

假設入入張量如下,要長寬放大兩倍 (64,128,128)

從通道除4,擴充兩個新維度 (16,2,2,128,128)

轉置將新維度置於最後 (16, 128,128,2,2)

將高度(axis=1)切片再根據axis=3疊合 [(16, m,128,2,2) for m in h]

完成將高度放大2倍 (16, 128,256,2)

將寬度(axis=1)切片再根據axis=3疊合 [(16, n,256,2) for n in w]

完成將寬度放大2倍 (16,256,256)

本次實作改良

- 換掉Batch Normalization
- 將Upscale2x<mark>區域的卷積換成深度可分離卷積</mark> (DepthwiseConv2d)
- 把ResBlock的第一個通道數降為32
- 把16個ResBlock的dilatio分別設置為1,2,4,8
- 最後一層活化函數修改為tanh
- 開頭的9x9卷積換成5x5
- 結尾的9x9卷積換成5x5

原始論文

Total params: 1,546,754

Trainable params: 1,546,754

Non-trainable params: 0

Total MACC: 9,094,381,568.0

Total FLOPs: 18.19697 GFLOPs

模型大小:6.3 MB

本次實作改良

Total params: 642,050

Trainable params: 642,050

Non-trainable params: 0

Total MACC: 2,758,885,376.0 Total FLOPs: 5.52910 GFLOPs

模型大小: 2.7 MB

L1 Loss L2 Loss Smooth L1 Loss....

L2 loss

$$L_2 = \left| f(x) - Y
ight|^2 \ L_2' = 2 f^{'}(x) (f(x) - Y)$$
 (1)

L1 loss

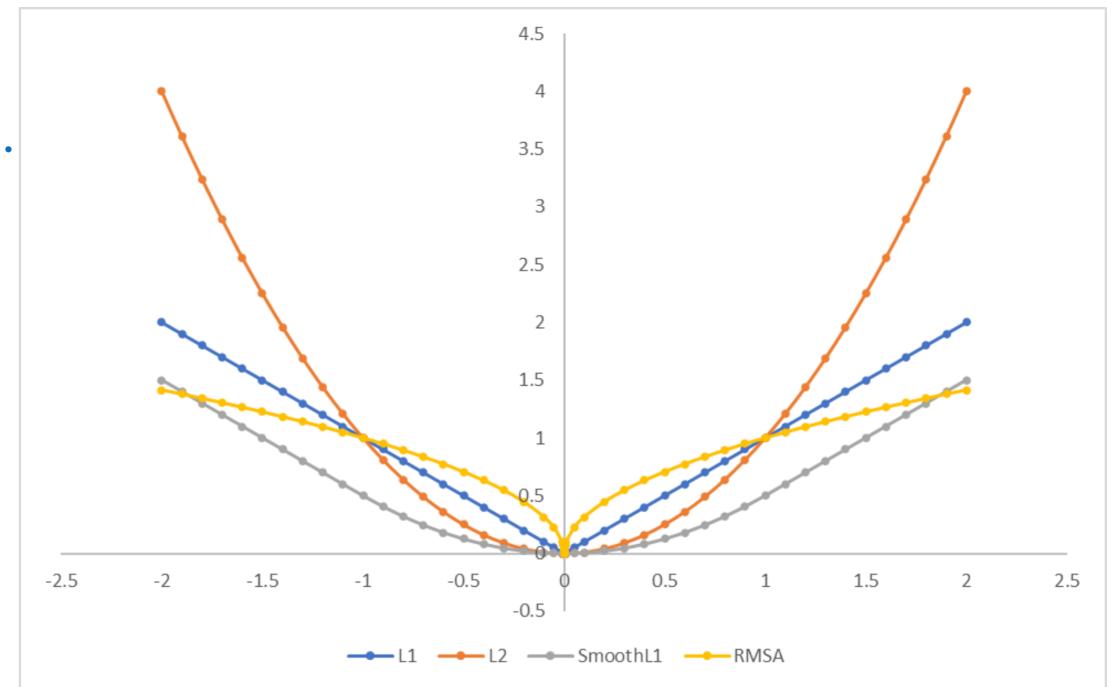
$$L_1 = |f(x) - Y| \ L_1' = \pm f^{'}(x)$$
 (2)

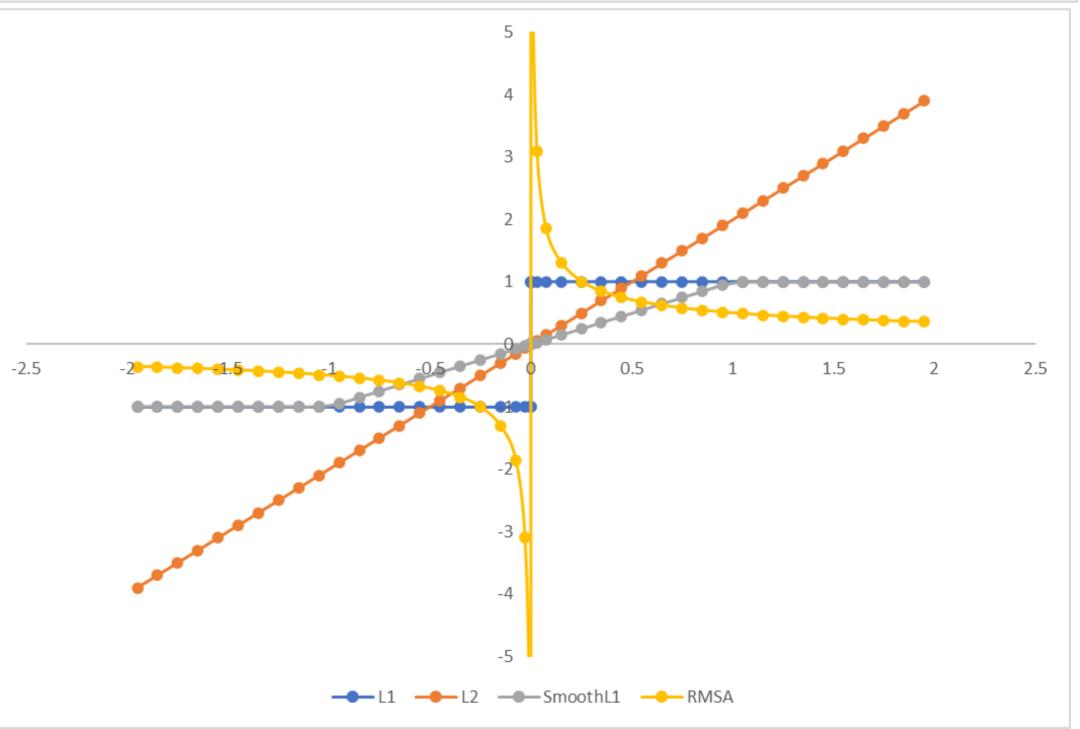
Smooth L1 loss

$$Smooth \quad L_1 = rac{0.5x^2, \;\; |x| < 1}{|x| - 0.5, \;\; x < -1 \, or \, x > 1}$$

......(3

$$x, \ |x| < 1 \ Smooth \ L_{1}^{'} = -1, \ x < -1 \ 1, \ x > 1$$





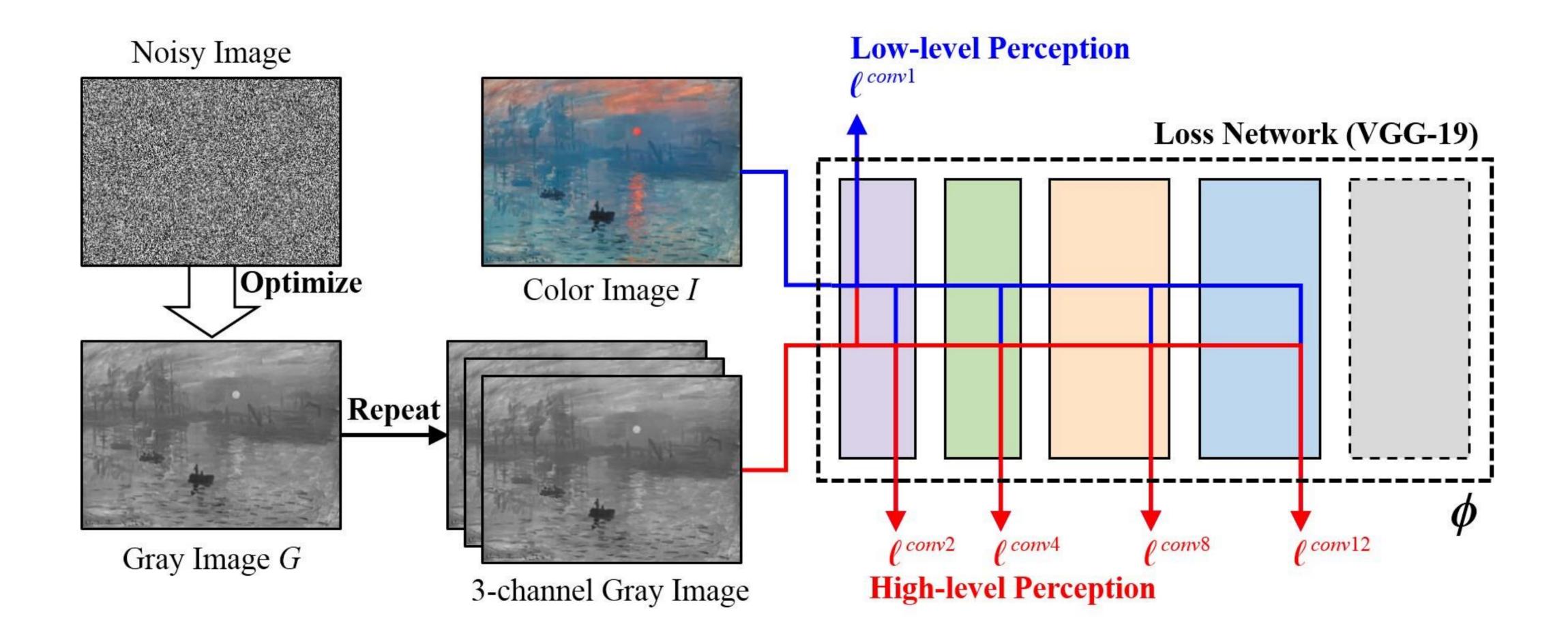


rmse: 19.689% ,psnr: 14.13

rmse: 24.457% ,psnr:12.14

機器只管對應精確位置的像素值是否正確...

Perception Loss



SSIM Structure Similarity Index 結構相似性指標

$$l(x,y) = rac{2\mu_x \mu_y + c_1}{\mu_x^2 + \mu_y^2 + c_1}$$
 照明度(平均值)

$$c(x,y) = rac{2\sigma_x\sigma_y + c_2}{\sigma_x^2 + \sigma_y^2 + c_2}$$
 對比度(標準差/變異數)

$$s(x,y) = rac{\sigma_{xy} + c_3}{\sigma_x \sigma_y + c_3}$$
 結構(共變數)

峰值信號雜訊比 (Peak Signal to Noise Ratio) $PSNR = 10 \times \log \left(\frac{255^2}{MSE} \right)$

$$PSNR = 10 \times \log \left(\frac{255^2}{MSE} \right)$$

$$MSE = \frac{\sum_{n=1}^{FrameSize} (I_n - P_n)^2}{FrameSize}$$

