DEEP LEARNING FOR COMPUTER VISION

TEAM 2 - Colorization with Conditional GANs



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OVERVIEW

Objectives

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Architecture

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OBJECTIVES

Understanding and testing Conditional GANs

Image colorization using Conditional GANs

Transfer learning on a new dataset

DATASETS

Facades: http://cmp.felk.cvut.cz/~tylecr1/facade/

training samples: 200

testing samples: 178



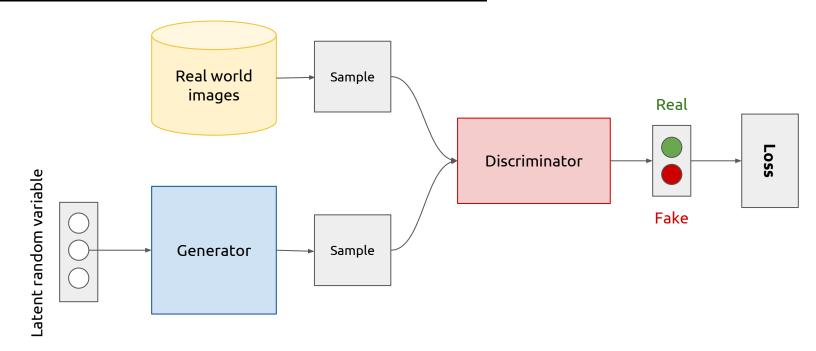
Cat_Dataset: https://www.kagqle.com/crawford/cat-dataset

training samples: 198

testing samples: 104

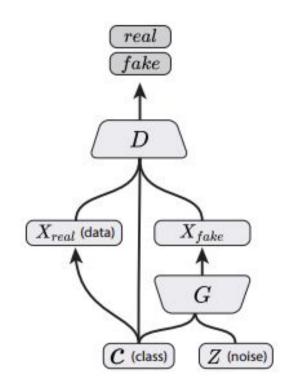


GANs (Generative Adversarial Networks).



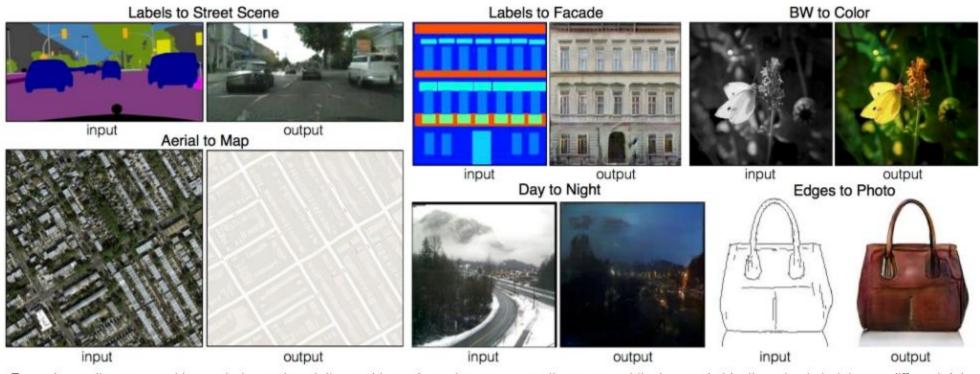
Conditional GAN.

•Unlike an unconditional GAN, the generator has an input condition (image, label, ...)



Conditional GAN (Mirza & Osindero, 2014)

<u>Pix2pix - Image to Image Translation</u>

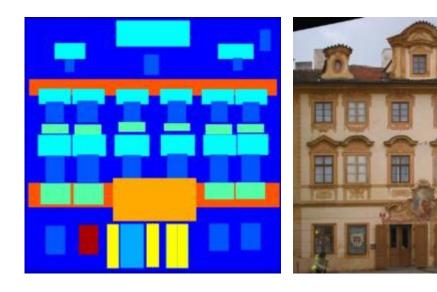


Example results on several image-to-image translation problems. In each case we use the same architecture and objective, simply training on different data.

- Our <u>Generator</u> has 6 blocks, each one composed by:
 - Convolutional Layer
 - Normalization Layer
 - ReLu Layer
- The <u>Discriminator</u> is composed by 3 layers:
 - Convolutional Layer
 - Normalization Layer
 - LeakyReLu Layer

Source code: https://github.com/mrzhu-cool/pix2pix-pytorch.git

Source code example: labels to facades



Label Facade

1. Colorization (from BW to color) train and test with <u>facades dataset</u>

epochs = 200, # learning rate = 0.0002 (default)



Original



Black and white



Generated

https://github.com/telecombcn-dl/2018-dlcv-team2

2. Colorization (from BW to color) test with <u>cat dataset</u>



3. Transfer Learning

epochs = 50, # learning rate = 0.0002 (default)



Original



Transfer

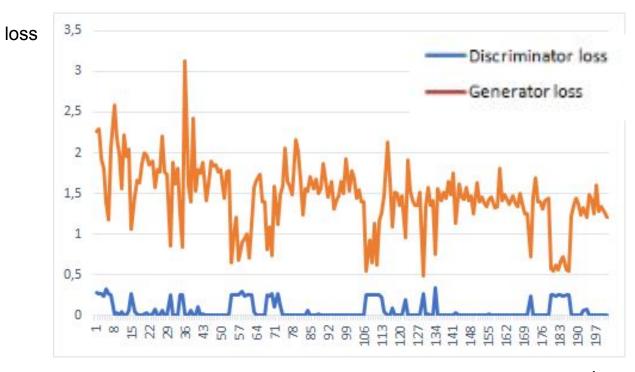
https://github.com/telecombcn-dl/2018-dlcv-team2

RESULTS

Colorization trained for 200 epochs, with 200 facades training

images:

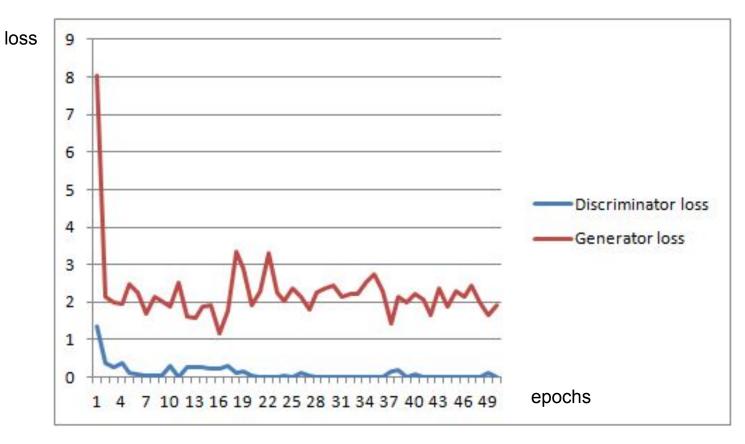
Loss log:



RESULTS

 Transfer learning for cat dataset trained for 50 epochs, with 198 training images:

Loss log:



CONCLUSIONS AND LESSONS LEARNED

- Pix2pix gives overall good results in different image-to-image translation tasks
- Loss is a bit counter intuitive in GANs
- Transfer learning works better with similar datasets
- To improve the results of transfer learning → play with hyperparameters
- Imagenet, tiny imagenet: too long to train even with GPU

THANK YOU FOR YOUR ATTENTION!

Any questions?

https://github.com/telecombcn-dl/2018-dlcv-team2/blob/master/train.py