Generative Cooperative Net for Image Generation and Data Augmentation

University of Tehran School of Electrical and Computer Engineering

Deep Learning in Computer Vision and Speech Processing

Dr. R. Hosseini

Erfan Mirzaei, Mohammad Mahdi Mehmanchi, Sara Jahedazad

Summer 2021

Content:

- Related Works
- Datasets
- GCN Details
- Results
- Conclusion



Other Generative Models

- Autoregressive
- Generative Adversarial Networks
- Variational Auto-Encoder
- Flow-based Generative Model

Conventional Augmentation Methods

- Random Cropping
- Scale Variation
- Affine Transformation ,....

Limitations and Issues with Other Works and How This Paper Solves the Problems

- Supervised Methods → Generalization
- ► GAN → Limited Controllability Over the Desired Features
- ► Conditional GAN → Large Dataset

Main Contributions:

- 1. Can employ smaller dataset due to absence of need for learning the distribution of visual concept.
- 2. Deploying a classifier facilitates knowledge transformation, to solve supervised generative model problems.

Datasets:

- The Karolinska Directed Emotional Faces (KDEF)
- OMNIST

KDEF Preprocessing



KDEF Dataset Consists of 4900, 562*762 RGB Pictures of Human Expressions.

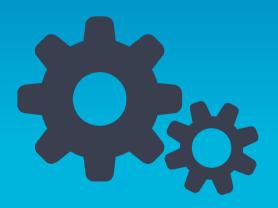
Preprocessing Below was Applied:

- Face Detection and Resizing
- Mirroring
- Rotation (0, 90, 180, and 270 degrees)
- Train & Test Splitting





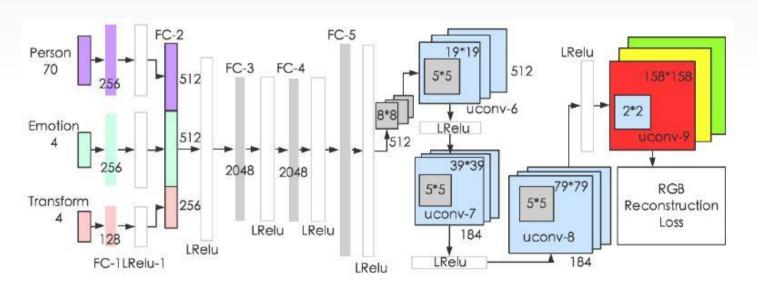
GCN Architecture:



- Generator
- Classifier

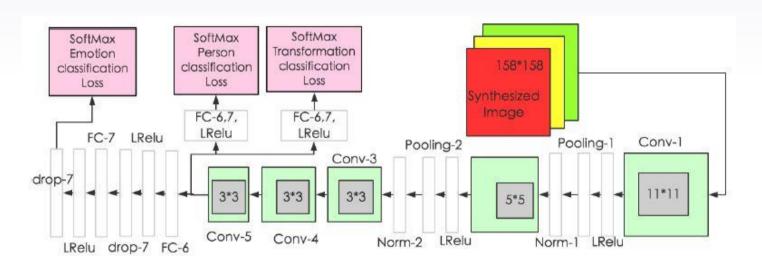
Generator

Inspired by Network Introduced by Dosovitskiy et al.



Classifier

Inspired by AlexNet.



Loss Function

$$\frac{1}{N} \{ \sum_{i=1}^{N} \left\| Pixel_r^i - Pixel_s(f_{i_1}, \dots, f_{i_M})^i \right\|^2 - [\sum_{i=1}^{N} \sum_{f=1}^{M} \sum_{j=1}^{K_f} Weight_f \{ y_f^i = j \} \log \frac{e^{\theta_{f_j}^I x_f^i}}{\sum_{l=1}^{K_f} e^{\theta_{f_l}^T x_f^i}}] \}$$

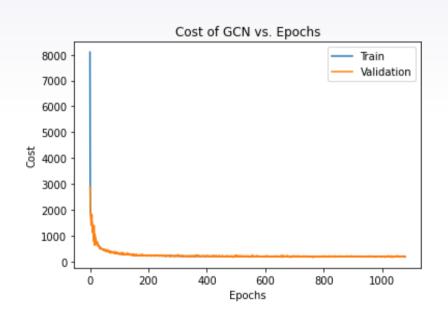
Hyper Parameters:

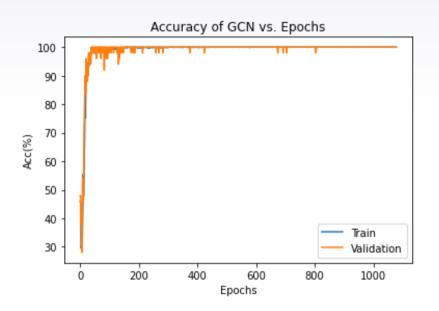
- Optimizer: Adam
- Learning Rate: 0.0002
- $(\beta_1, \beta_2): (0.9, 0.99)$

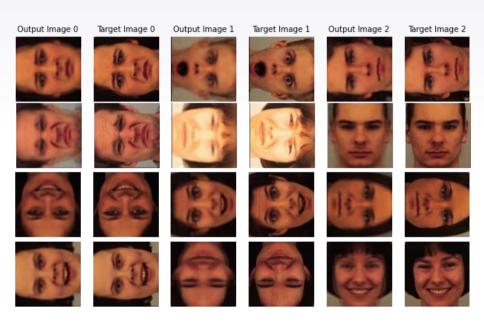
- ϵ : 10⁻⁸
- Batch Size: 64
- Epochs: 1100

Image Generation Results

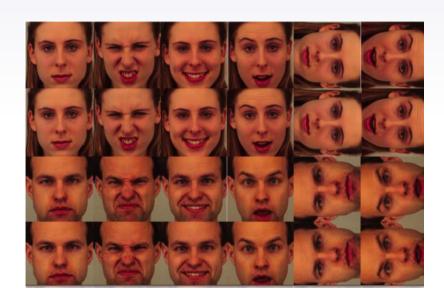












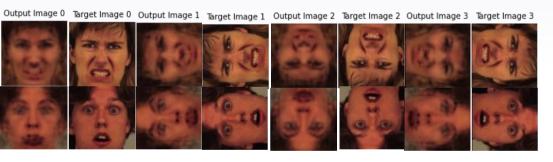
Training set faces reconstructed by the paper



Low level features generation of the test set Our implementation



Low level features generation of the test set Original paper





High level features generation of the test set Our implementation

High level features generation of the test set Original paper

Data Augmentation

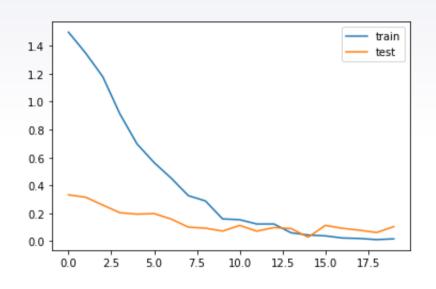


Results

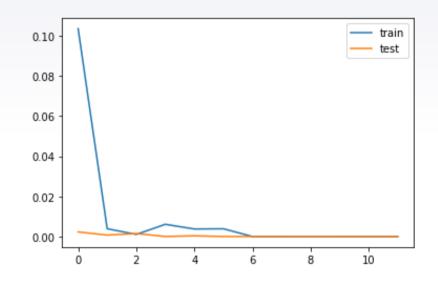


Augmented images (from left: Surprised, Disgusted, Happy, Neutral)

Results



Loss function without augmented data



Loss function with augmented data

Results

Without augmented data	With augmented data
92%	100%

Accuracy on test images

QMNIST Preprocessing

QMNIST Consists of 28*28 Black and White Pictures of Handwritten Digits Whose Writers are Known.

Preprocessing Below was Applied:

- Coloring the Pictures (to R, G and B)
- Rotation (0, 90, 180 and 270 Degrees)

** 10 Writers' Handwritten Digits Were Chosen. Either a Specific Rotation or Color of a Specific Digit was Taken Out for the Test Set.

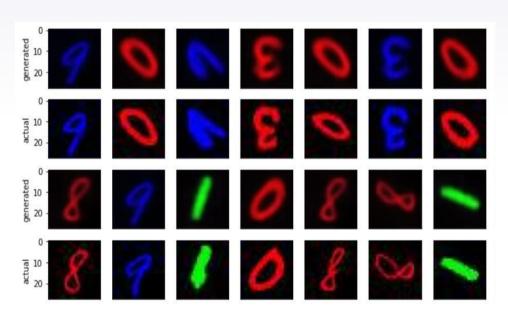








QMNIST



Low level features generation of the test set Our implementation

Low level features generation of the test set Original paper

Conclusion and Suggestions



Conclusion



Critiques

- Information provided for data preprocessing and train/test splitting is vague
 - Unclear pre processing used for (Q)MNIST
 - Unclear test set for face generation
 - Used 5 emotions instead of 4 for the classification!!
- Network structure explanation is not exact:
 - Mismatch between network structure in the figure and text
 - No clear explanation for MNIST dataset network
- Learning rate decay seems unreasonable!!

THANKS! Any questions?

You can contact us:

- erfunmirzaei@gmail.com
- mehmanchimahdi@gmail.com
- Sara.jahedazad@gmail.com



References

- Presentation template by <u>SlidesCarnival</u>
- Illustrations by <u>Sergei Tikhonov</u>
- Xu, Q., Qin, Z., & Wan, T. (2019, March). Generative cooperative net for image generation and data augmentation. In International Symposium on Integrated Uncertainty in Knowledge Modelling and Decision Making (pp. 284-294). Springer, Cham.

References

- Dosovitskiy, A., Tobias Springenberg, J., & Brox, T. (2015). Learning to generate chairs with convolutional neural networks. In Proceedings of the IEEE conference on computer vision and pattern recognition (pp. 1538-1546).
- Krizhevsky, A., Sutskever, I., & Hinton, G. E. (2012). Imagenet classification with deep convolutional neural networks. Advances in neural information processing systems, 25, 1097-1105.