Resembling Better Lungs

Using Generative Adversarial Networks (GANs) to create CT-Scans of COVID-19 infected lungs

Dieter Hoogestraat, @opencampus GANs WiSe2021



Idea

Resemble CT lung scans of **Covid 19** infected persons as done in 2020 during the OpenCampus course "Machine Learning with **Tensorflow**", using new techniques to get (perhaps) better results, i.e. getting better insights in the possibilities and practical issues of using GANs for reproducing detailed imaging products.



Agenda

- Idea (already done)
- Starting point: ML, TF and AF
- Pathsways
- Results
- Entropy
- Conclusion & Possible Future Work



Start

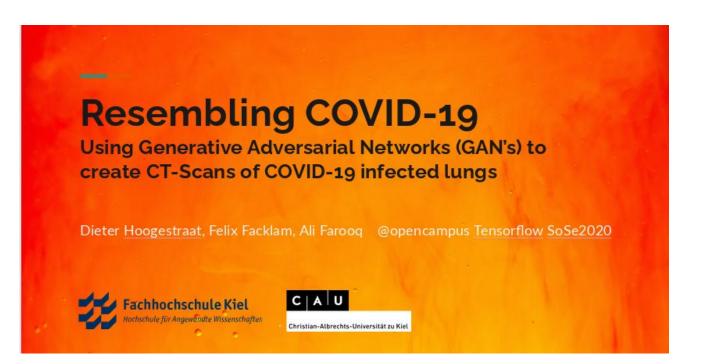
OC SoSe202

Tensorflow

Ali Farooq

GAN

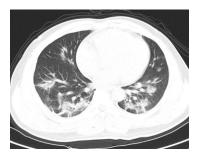
Covid19 CT





Start - former results

Real



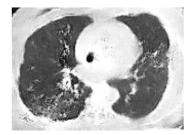


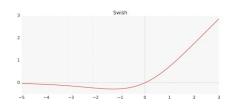


GAN





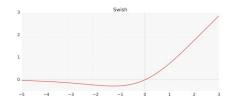


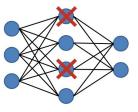


- Use another activation function
- Transfer the job from TF to PyTorch
- Implement a Wasserstein-GAN-GP
- Add new CT-Scans, i.e. new data
- (added: examine the metalcover-results in relation to the initial Covid-GAN-results)



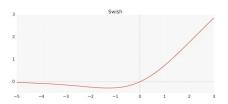
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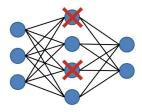






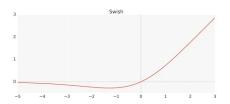
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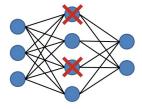






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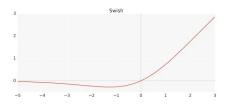


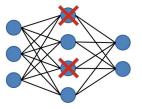




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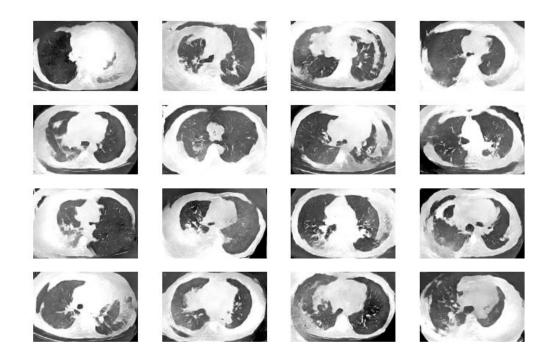




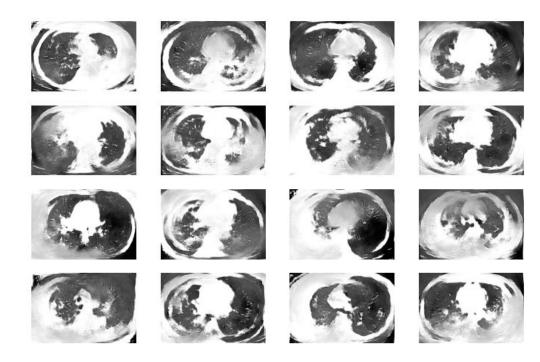
(some) Results



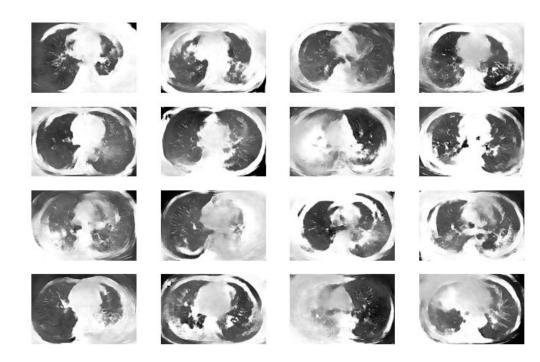
06-2020
LeakyRelu
no dropout
1.250 epochs



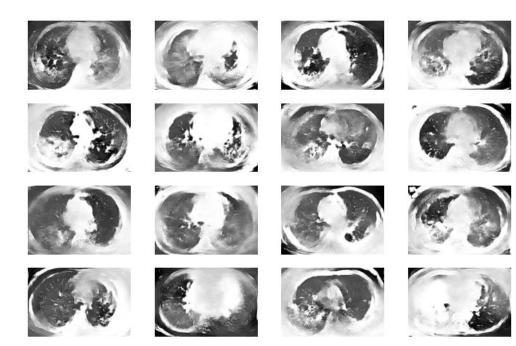
Swish no dropout 1.250 epochs



Swish 30% dropout 1.250 epochs

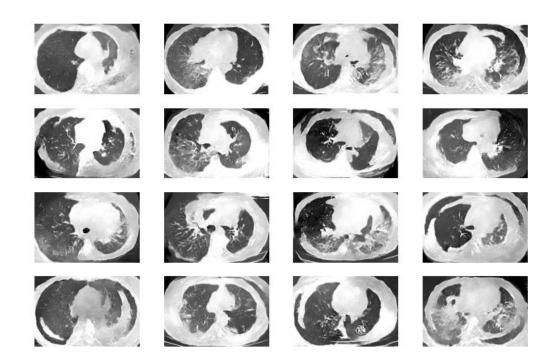


Swish 50% dropout 1.250 epochs



LeakyRelu no dropout 14.720 epochs

with lots of instability



From TF-DCGAN to PyTorch-WGAN+GP

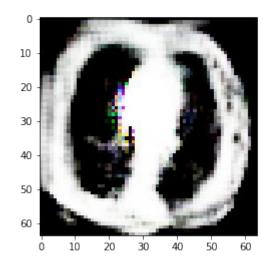
- Wasserstein-GAN with Gradient Penalty from week 3
- running WGAN-GP on old data: some images caused problems
- new data (-)
- working on images took a lot of time => using Google-Colab Pro+

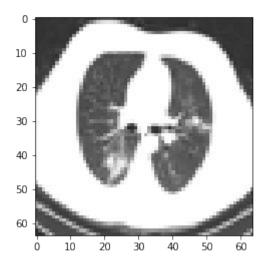
 Additionally: after metalcover-talk, I decided to use exactly that model, what took even more time

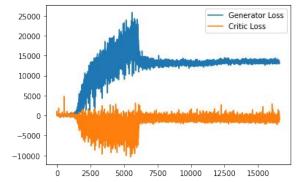


- PyTorch
- LeakyRelu
- no dropout
- Wasserstein
- GP

- LeakyRelu
- no dropout
- Wasserstein
- GP









Resembling Better Lungs Christian-Albrechts-Universitätzu Kiel

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"Batch Normalization in the discriminator breaks Wasserstein GANs with gradient penalty. The authors themselves advocate the usage of layer normalization instead, but this is clearly written in bold in their paper (https://papers.nips.cc/paper/7159-improved-training-of-wasserstein-gans.pdf). [...]

By the way transposed convolutions produce stairway artifacts in your output images. Use image resizing instead. For an indepth explanation of that phenomenon I can recommend the following resource (https://distill.pub/2016/deconv-checkerboard/)."

Thomas Peinitz

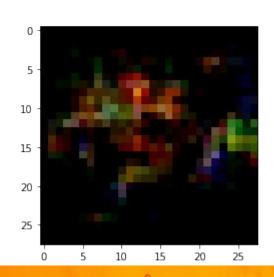
https://stackoverflow.com/users/3994824/thomas-pinetz

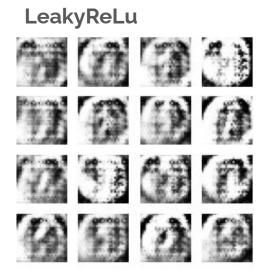


Results from 2020 - Déjà-vu

500 epochs, learning rate 1E-4, resolution 28x28

ReLu





Conclusion and Future Work

- Wasserstein-GANs don't work better just because they are Wasserstein GANs
- Batch Normalization could be a problem for Wasserstein GANs
- Easily extensible code is available, easy data handling not
- Generated images can be used for future covid-19 projects

