

# A Generative Adversarial Network for Brain Stroke Image Analysis

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# Summary

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# Motivation

Ischemic stroke is caused by partial or total restriction of blood supply to part of the brain. During an acute stroke, prolonged ischemia results in irreversible tissue death.

2 Million brain cells die every minute.

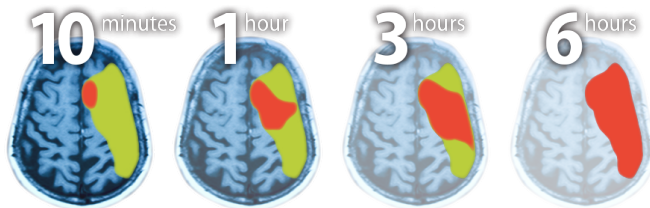


Figure: Source: <https://outlook.wustl.edu/2014/jun/stroke>

# Motivation

Decisions about ischemic stroke therapy are highly time-sensitive and rely on distinguishing between the infarcted core tissue and hypoperfused lesions.

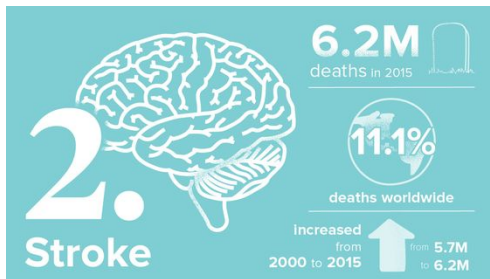


Figure: Source:

<https://www.healthline.com/health/top-10-deadliest-diseases>

# Motivation

Automated methods that can locate and segment ischemic stroke lesions can aid clinician decisions about acute stroke treatment.

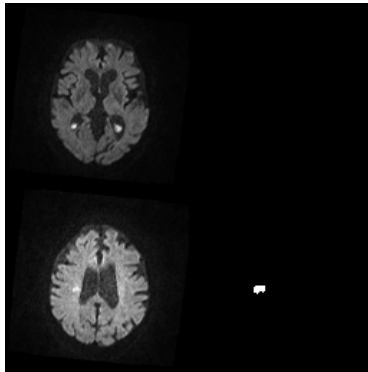


Figure: DWI of brain without and with stroke and the segmentation

# Motivation

Generative Adversarial Networks are one of the hottest topics in Artificial Intelligence right now. These networks, are able to learn how to produce data from a dataset that is indistinguishable from the original data and generate segmentation mask from images.

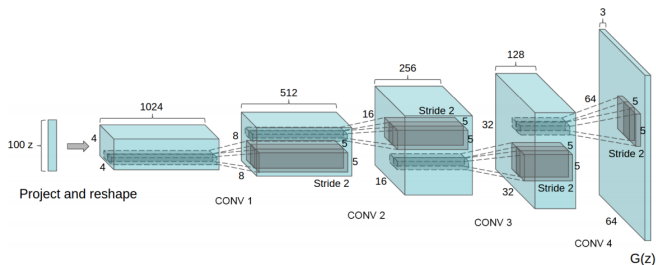


Figure: Source: [3]

# Preliminary Objectives

- 1 Study the use of Generative Adversarial Networks for generating synthetic images of Brain Stroke.
- 2 Study the use of Generative Adversarial Networks for generating segmentation mask of Stroke from Brain Images.

# Expected Results

- 1 Generate synthetic images of Brain Images preserving the condition of stroke using Generative Adversarial Networks.
- 2 Generate segmentation mask of Brain Stroke using Generative Adversarial Networks.



# Dataset

- 1 Total Images: 13200
- 2 75% train & 25% validation.
- 3 82% without Stroke & 18% with Stroke.
- 4 Image Size: 128x128 px.
- 5 Image Type: DWI (Diffusion Weighted Resonance Magnetic Image).
- 6 Image Source: University Hospital Basel - Switzerland

## \*Advance

- 1 Study and implementation of recent models such as vGAN [1], vGAN [1], cGAN [2], DCGAN [3] y cDCGAN [2],[3] or the automatic generation of synthetic images of cerebral infarcts.
- 2 Reproduce and modify the models adding a resnet blocks [4].
- 3 From cDCGAN [2][3] with resnet blocks [4] produce a synthetic images of cerebral infarcts and segmentation.

## \*Current Status

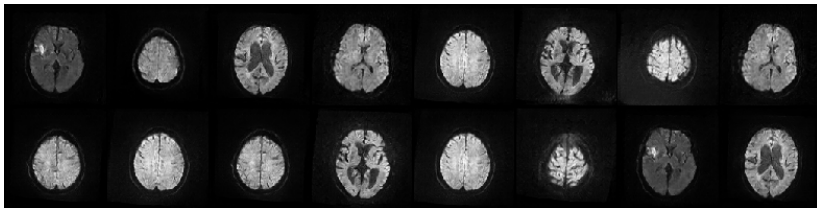


Figure: Original (1st row) and Generated (2nd row) CT Images

# References

- 1 I. Goodfellow, et al., "Generative Adversarial Networks", Advances in Neural Information Processing Systems (NIPS), 2014, pp. 2672-2680.
- 2 M. Mirza, S. Osindero, "Conditional Generative Adversarial Nets", arXiv:1411.1784, 2014.
- 3 Alec Radford, Luke Metz, Soumith Chintala, "Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks", arXiv:1511.06434, 2015.
- 4 K. He, X. Zhang, S. Ren, J. Sun, "Deep Residual Learning for Image Recognition", The IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2016, pp. 770-778.