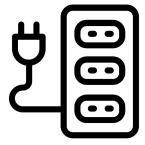


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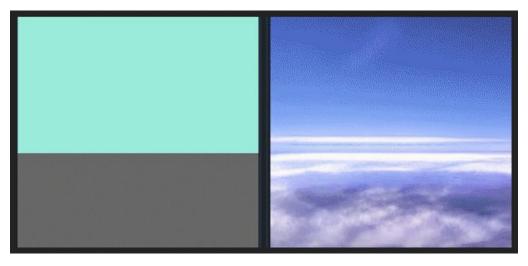
# Overview of GAN Applications

#### **Outline**

- GAN applications
  - Image-to-image translation and extensions to other modalities such as text, audio, and video
  - Image editing, art, and media
  - Medicine and climate change
- GAN adversarial concept use in other research areas



# Image-to-Image



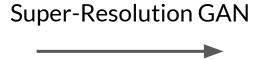
GauGAN

Available from: https://arxiv.org/abs/1903.07291

## Image-to-Image



Original image

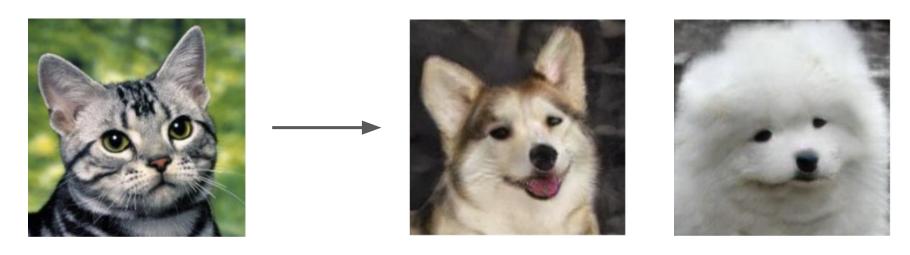




Sharpened image

Available from: https://arxiv.org/abs/1609.04802

# Image-to-Image



Multimodal image-to-image translation

Available from: https://github.com/NVIabs/MUNIT

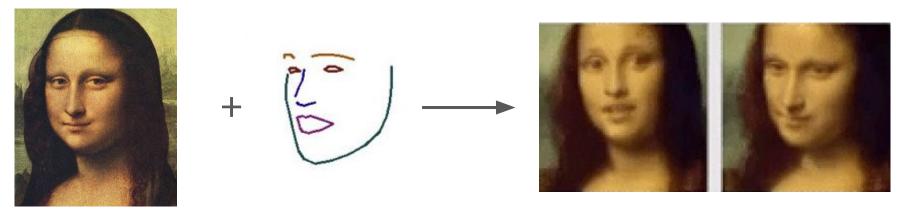
# Text-to-Image

"The bird is black with green and has a very short beak."



Available from: https://arxiv.org/abs/1612.03242

# Image-and-Landmark-to-Video



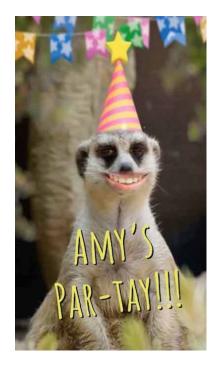
Image

Face landmark

Talking heads

Available from: https://arxiv.org/abs/1905.08233

#### **Application Areas: Image Filters**







Available from: https://www.snapchat.com

## **Application Areas: Image Editing**

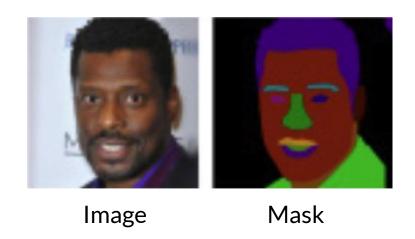


Image editing software

Available from: https://arxiv.org/abs/1907.11922

# Application Areas: Image Editing

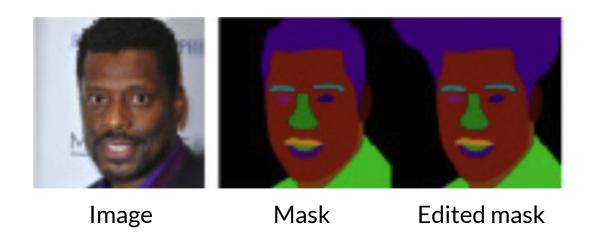


Image editing software

Available from: https://arxiv.org/abs/1907.11922

# Application Areas: Image Editing

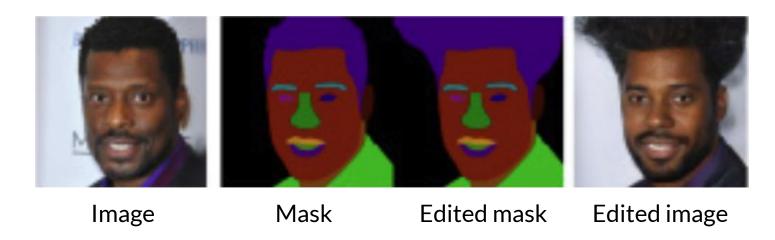


Image editing software

Available from: https://arxiv.org/abs/1907.11922

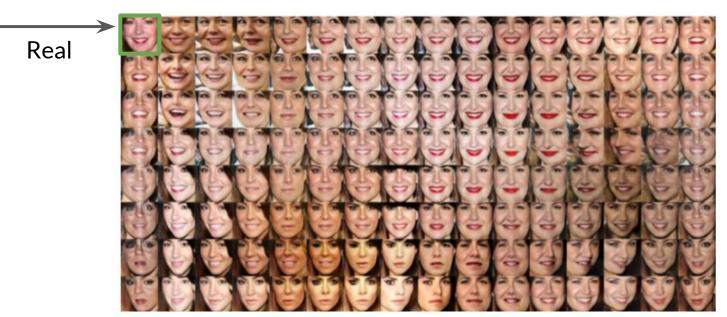
#### Application Areas: Stylized Images



Democratized art

Available from: https://www.youtube.com/watch?v=85l961MmY8Y

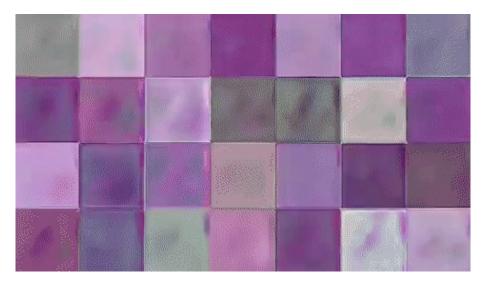
#### Application Areas: Data Augmentation



Increasing dataset size and diversity

Available from: https://arxiv.org/abs/1711.04340

#### Application Areas: Medicine



Simulating tissues

Available from: https://twitter.com/realSharonZhou/status/1182877446690852867

# **Application Areas: Climate Change**

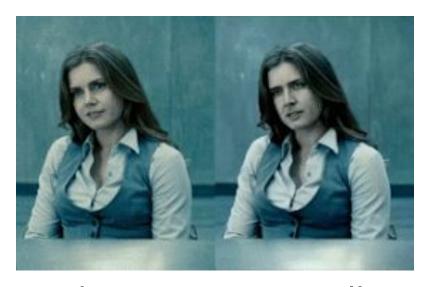




Real input

Generated output

#### Application Areas: Media



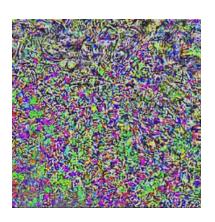
Both pretty cute, actually

Available from: https://en.wikipedia.org/wiki/File:Deepfake\_example.gif

#### Adversarial Research Areas







Predicted class: airliner Predicted probability: 96.8%

Adversarial examples & robustness

Available from: https://adversarial-ml-tutorial.org/introduction/



#### Summary

- Image translation generalizes to many tasks
- Many immediate application areas, including data augmentation
- Other fields use adversarial techniques for realism and robustness





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# Data Augmentation Methods and Uses

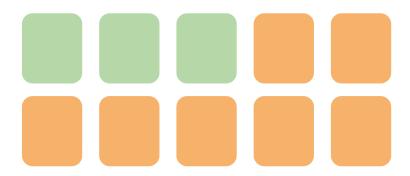
#### Outline

- Data augmentation and use cases
- Implementation of data augmentation



#### Data Augmentation

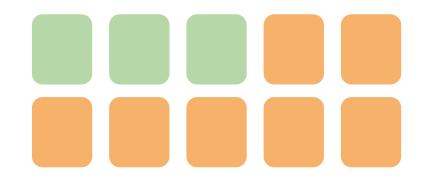
- Supplement data when real data is...
  - Too expensive
  - Too rare



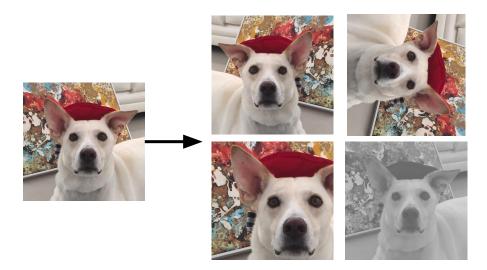
#### Data Augmentation

- Supplement data when real data is...
  - Too expensive
  - Too rare

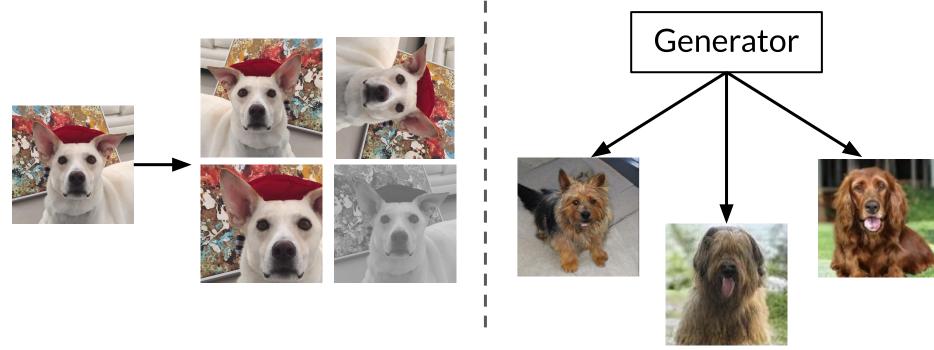
GANs are well suited for this



# How to Augment Data



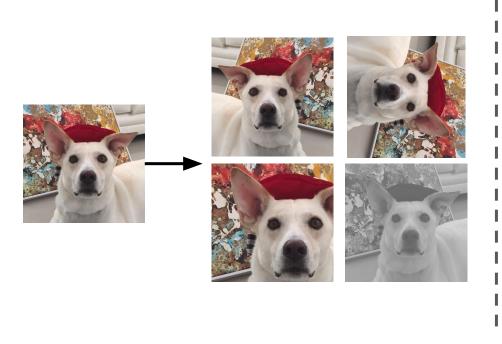
# How to Augment Data

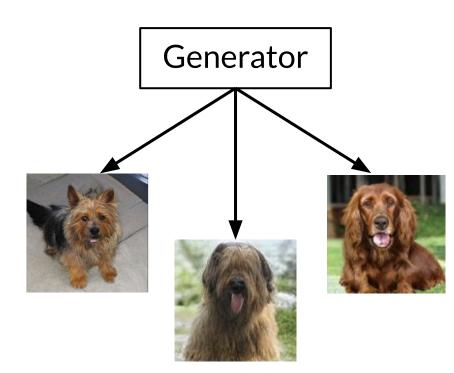


(Right) Available from: https://arxiv.org/abs/1809.11096

#### How to Augment Data

Can mix data augmentation techniques!





(Right) Available from: https://arxiv.org/abs/1809.11096

#### **Use Cases**

Synthetic

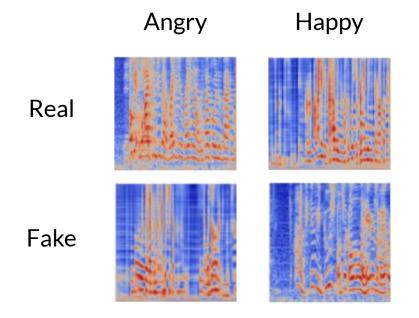


Generated

Gaze detection

Available from: https://arxiv.org/abs/1711.09767

#### **Use Cases**

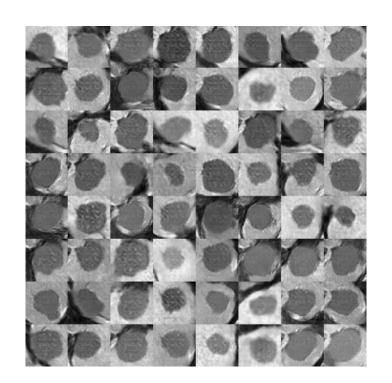


Speech emotion recognition

Available from: https://pdfs.semanticscholar.org/395b/ea6f025e599db710893acb6321e2a1898a1f.pdf

#### **Use Cases**

Synthetic liver lesions



Available from: https://arxiv.org/abs/1803.01229

#### Summary

- Use GANs to generate fake data when real data is too scarce
- GANs have various use cases in data augmentation and beyond!



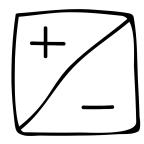


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# Data Augmentation: Pros & Cons

#### Outline

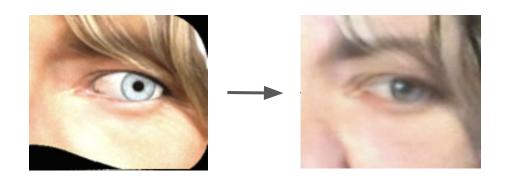
- Pros and cons of data augmentation
- Various use cases



#### Pros of GAN Data Augmentation

Better than hand-crafted synthetic examples

**Synthetic** 

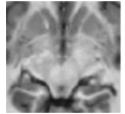


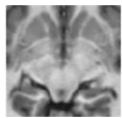
**GAN** refined

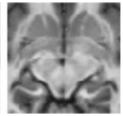
Available from: https://arxiv.org/abs/1711.09767

## Pros of GAN Data Augmentation

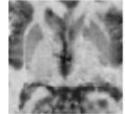
#### Generate more labeled examples

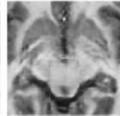


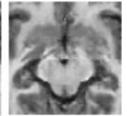




Training set (reals)







Labeled output (fakes)

Available from: https://arxiv.org/abs/1811.10669

#### Pros of GAN Data Augmentation

#### Improve downstream model generalization



Available from: https://www.nature.com/articles/s41598-019-52737-x/figures/3

#### Cons of GAN Data Augmentation

Diversity is limited to the data available



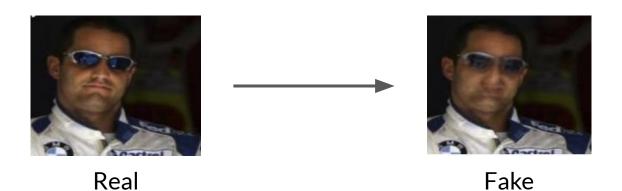
Training set



Generated outputs

#### Cons of GAN Data Augmentation

#### Not useful when overfit to real data



Available from: https://arxiv.org/abs/1902.04202

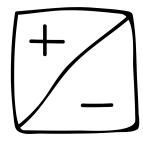
#### Summary

#### Pros:

- Can be better than hand-crafted synthetic examples
- Can generate more labeled examples
- Can improve a downstream model's generalization

#### Cons:

- Can be limited by the available data in diversity
- Can overfit to the real training data





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# GANs for Privacy

#### Outline

- GANs for privacy preservation
- Medical privacy as a motivating example



# Motivations for Medical Privacy

Protects real patient data



### Motivations for Medical Privacy

- Protects real patient data
- Can encourage data-sharing between institutions





### Motivations for Medical Privacy

- Protects real patient data
- Can encourage data-sharing between institutions
- Less expensive and more abundant than real data

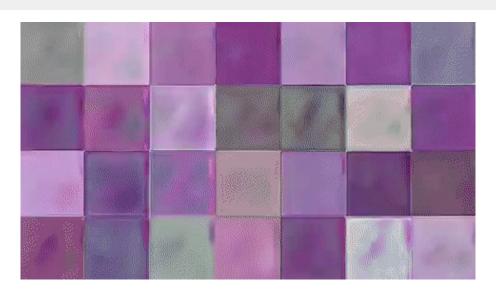






## **Privacy Preservation**

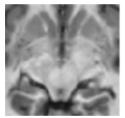
#### GAN tissue patches look real to pathologists

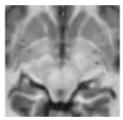


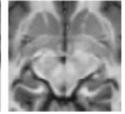
Available from: https://twitter.com/realSharonZhou/status/1182877446690852867

### **Privacy Preservation**

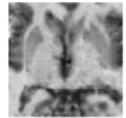
#### **GAN MRIs look realistic**

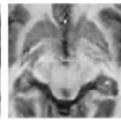


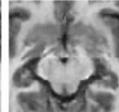




Training set (reals)



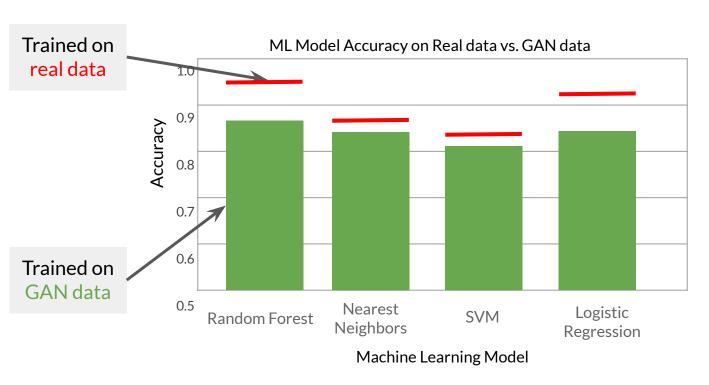




Labeled output (fakes)

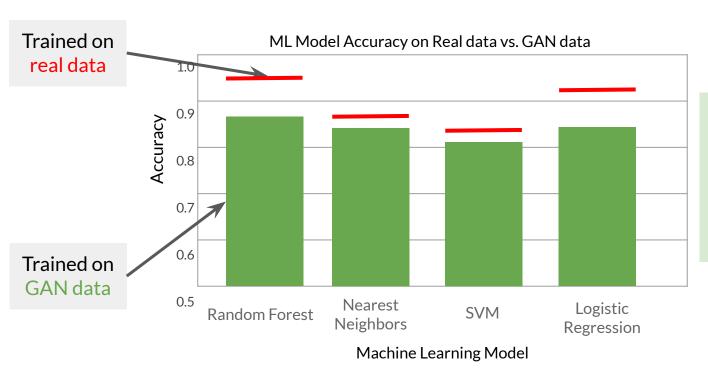
Available from: https://arxiv.org/abs/1811.10669

#### Pro of GANs for Privacy



Available from: https://www.ahajournals.org/doi/epub/10.1161/CIRCOUTCOMES.118.005122

#### Pro of GANs for Privacy



Training with GAN data approaches real data accuracy

Available from: https://www.ahajournals.org/doi/epub/10.1161/CIRCOUTCOMES.118.005122

## Con of GANs for Privacy

#### GAN sample is nearly identical to a real sample



Available from: https://arxiv.org/abs/1902.04202

#### Summary

- GANs can be useful for preserving privacy
  - Sensitive medical data serves as one example
- Caveat: generated samples may mimic the reals too closely
  - Post-processing may help avoid this data leakage





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# GANs for Anonymity

#### Outline

- GANs for anonymity
  - Concealing identity
  - Stealing identity
  - DeepFakes



# Anonymity

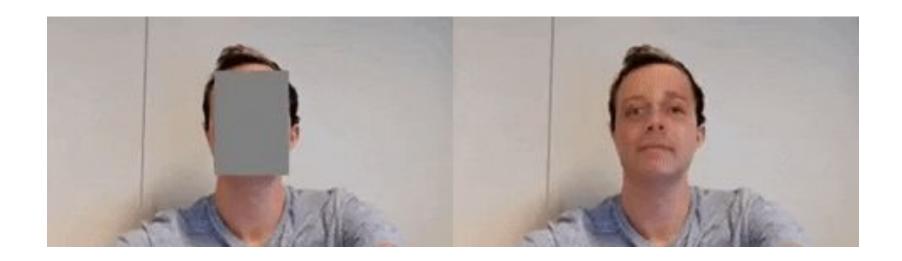


Original image

De-identified image

Available from: https://arxiv.org/abs/1902.04202

# Anonymity



Available from: https://arxiv.org/abs/1909.04538

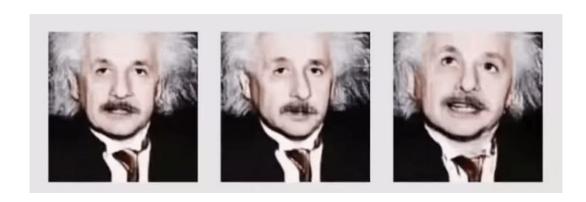
#### Pro of GANs for Anonymity

- Provide safe environment for expression to:
  - Stigmatized groups
  - Assault victims
  - Witnesses
  - Activists



## Con of GANs for Anonymity

#### Deepfakes put words into people's mouths



Available from: https://arxiv.org/abs/1905.08233

#### Summary

- GANs can enable healthy anonymous expression for stigmatized groups
- GANs for anonymization can be used for good or evil
  - Identity theft is not good
  - Use your powers for good

