

Generative Adversarial Networks

Deep Learning — Unit 9

Dr. Jon Krohn

jon@untapt.com

December 16th, 2017

Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

Outline

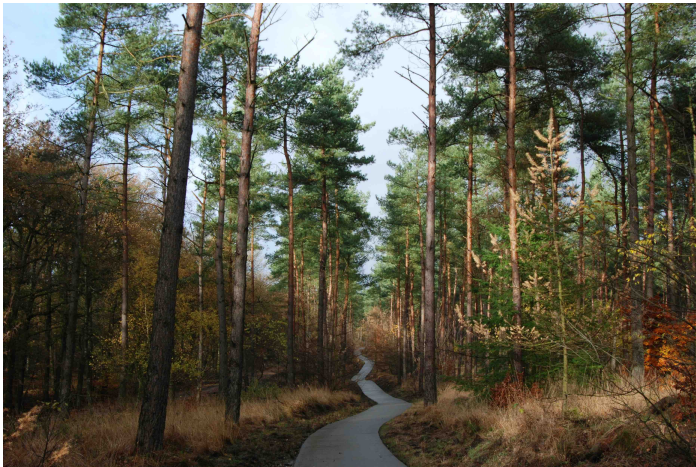
- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

Progress Check

Your Deep Learning Project V



Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Progress Check

Your Deep Learning Project V

Where are you at with respect to the following?

1 Splitting your data

- training set (80% — for optimizing parameters)
- validation set (10% — for hyperparameters)
- test set (10% — don't touch yet!)

2 Building and assessing architecture

- get above chance (simplifying problem, if necessary)
- do existing performance benchmarks exist?
- if not, use a simple architecture as benchmark

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

Recommended Projects

Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

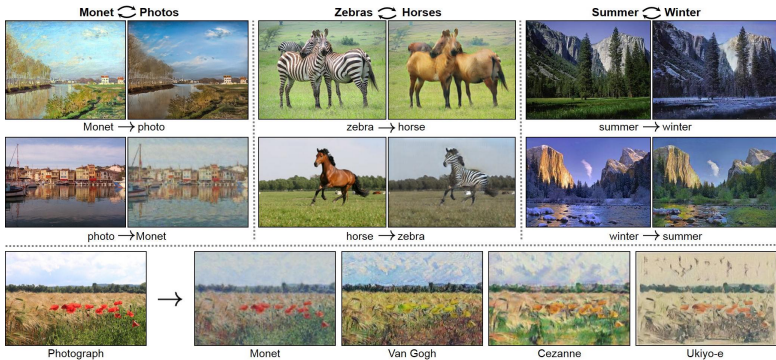
Recommended Projects

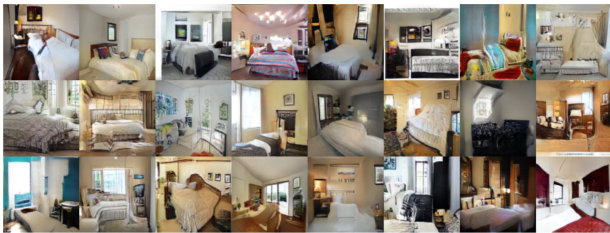
Your Deep Learning Project V

- 1 build a machine-vision architecture to classify images, e.g.:
 - [Fashion MNIST]
 - a subset from the [Cdiscount Kaggle challenge]
 - one of dozens of “image” data sets from [CrowdFlower]
 - one of the *Computer Vision* data sets from [Luke de Oliveira’s post]
- 2 build a natural language processing architecture to classify text, e.g.:
 - Yelp or Amazon sentiment [datasets] from [Zhang et al.]
 - the Yahoo! answers categories data set from Zhang et al.
 - one of dozens of “sentiment” or “text” data sets from CrowdFlower
 - one of the *Natural Language* data sets from Luke de Oliveira’s post

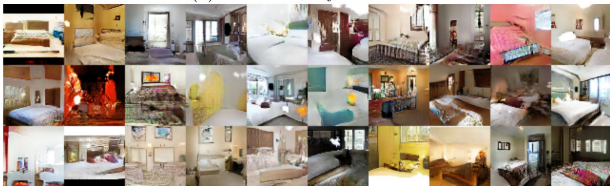
Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation



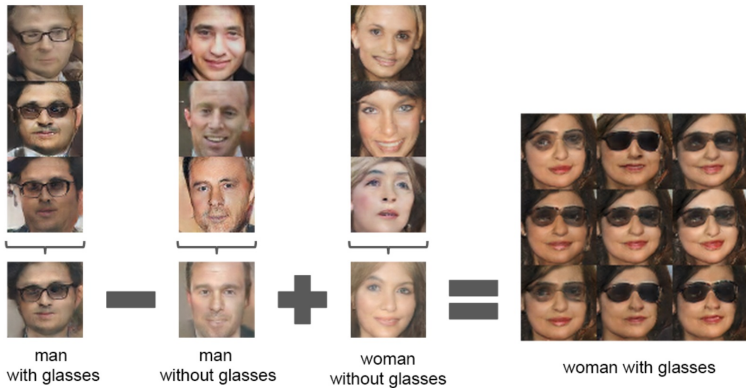


(a) Generated by LSGANs.



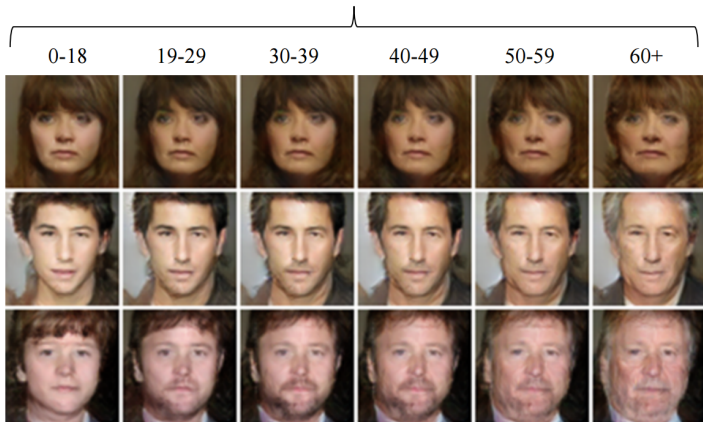
(b) Generated by DCGANs (Reported in [13]).

Figure 5: Generated images on LSUN-bedroom.



[“celebrity” latent-space interpolation]

Face Aging



YOLO



INPUT



undo

clear

pix2pix

process

OUTPUT



save

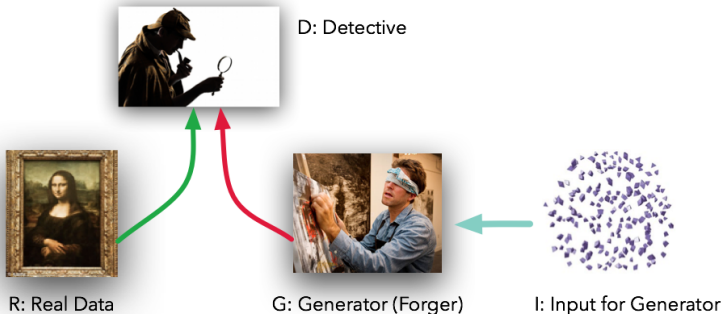


Figure 3. Example results by our proposed StackGAN, GAWWN [20], and GAN-INT-CLS [22] conditioned on text descriptions from CUB test set. GAWWN and GAN-INT-CLS generate 16 images for each text description, respectively. We select the best one for each of them to compare with our StackGAN.

Outline

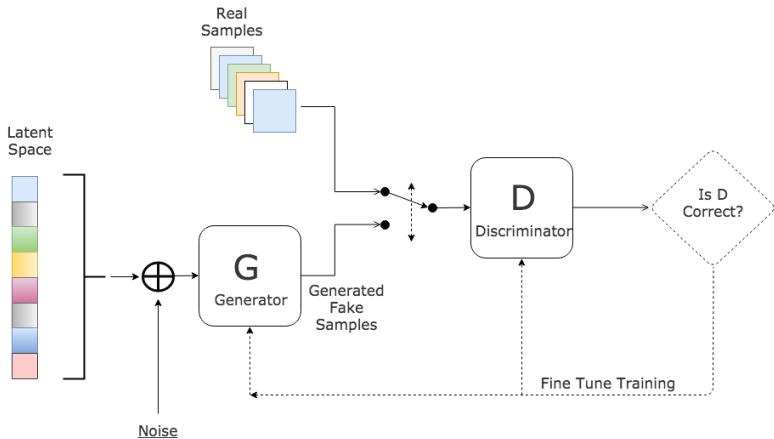
- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory**
- 4 “Quick, Draw!” Implementation

Goodfellow et al. (2014)



Goodfellow et al. (2014)

Generative Adversarial Network



1-D Gaussian

Approximating a Toy Distribution

[video]

Outline

- 1 Deep Learning Projects
- 2 Applications
- 3 Essential Theory
- 4 “Quick, Draw!” Implementation

[Quick, Draw!]

Unit 9 —
GANs

Your Projects

Applications

Theory

In Practice

[*generative adversarial network* notebook]