# Generative Adversarial Networks with Memory for Text Generation

Emily Sheetz

### Overview

- Project Description
- Generative Adversarial Networks (GANs)
- Recurrent Network Architectures
- Learning Algorithms
- Tools and Project Timeline
- Socio-Technical Context

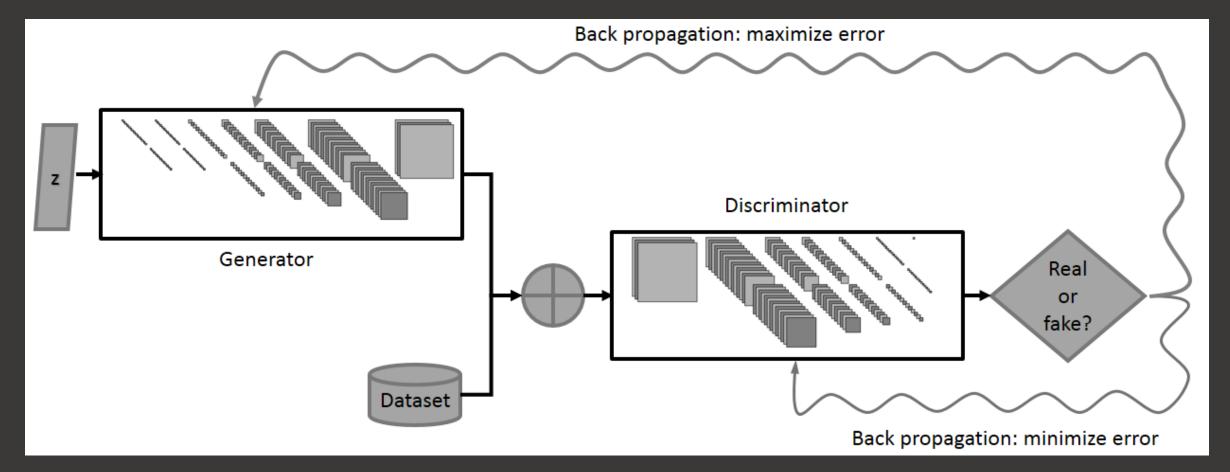
#### **Project Description**

- Train a neural network to generate sentences of horror text
- Generate samples at the character level
- Evaluate the quality of generated samples through survey
- Qualitative measurement

## Generative Adversarial Networks (GANs)

- Framework containing two neural networks
- Generator (G) attempts to generate samples that could be from the dataset
- Discriminator (D) attempts to determine which samples are from the generator or from the data set

#### GAN Training Process



Heinrich, Greg. "Photo Editing with Generative Adversarial Networks (Part 1)." NVIDIA, Parellel Forall (2017). Accessed Aug. 2017.

#### Recurrent Neural Networks

- Generator and discriminator need memory for long-term structure
- Recurrent neural networks (RNNs) feed outputs from previous time steps back into network
- Output probability distribution of next element in sequence conditioned on previous elements

#### **Recurrent Network Architectures**

- RNNs do not train well within GAN framework
- Long Short-Term Memory Networks (LSTMs):
  - Selectively remember new information and forget information from previous time steps
  - Learn more complex long-term relationships than RNN
- Gated Recurrent Units (GRUs):
  - Comparable performance to LSTMs
  - Fewer parameters than LSTMs, meaning more computationally efficient

#### Work with Network Architectures

- GAN with GRUs for memory (Press et al.)
  - Python and TensorFlow
  - Problems running, but sample code for GAN
- TensorFlow tutorial with LSTMs
  - LSTMs for language modeling, predicting next word in sequence
  - Training time estimate, sample code for LSTM

## Learning Algorithms

- Curriculum Learning: increase length of generated sequences
- Variable Length Outputs: allow for output sequences less than or equal to maximum length
- Teacher Helping: generator conditioned on samples in training data

## Tools

- Project Gutenberg
- Python
- TensorFlow

## Project Timeline

- Checkpoint 1: Script for text preprocessing
- Checkpoint 2: Run code from previous work, extend to horror text, architecture of network
- Checkpoint 3: Code GAN framework, LSTMs for memory
- Checkpoint 4: Code LSTMs, implement learning algorithms
- Checkpoint 5: Train GAN
- Checkpoint 6: Survey opinions on quality of text

#### Socio-Technical Context

- Humans tell stories to communicate
- Artificial intelligence given human-like intelligence through storytelling
- Horror elicits strong emotions
- Emotions of horror fiction depend on person
- Al participating in human tasks

## **Concluding Remarks**

- GAN with LSTMs for memory
- Generate sentences of horror text
- Qualitative evaluation of quality of generated samples
- Human aspect of stories
- Artificial intelligence participating in human task of storytelling

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