

Jiwei Li, NLP Researcher

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By Pragya Arora & Piyush Ghai

Introduction

- Graduated from Stanford University in 2017
- Advised by Prof. Dan Jurafsky
- Closely worked with Prof. Eduard Hovy from CMU and Prof. Alan Ritter from OSU
- Affiliated with The Natural Language Processing Group at Stanford University

Research Interests

- Jiwei's research interests focus on computational semantics, language generation and deep learning. His recent work explores the feasibility of developing a framework and methodology for computing the informational and processing complexity of NLP applications and tasks.
- His PhD thesis was on “*Teaching Machines to Converse*”
- Has over 1200¹ citations on Google Scholar.
- Has over 38¹ scholarly publications.

1 : [Google Scholar Site](#)

Teaching Machines to Converse

- Jiwei's primary research focus and his thesis work was on conversational models for machines.
- Some of his publications in this domain are :
 - [Deep Reinforcement learning for dialogue generation \[2016\], J Li, W Monroe, A Ritter, M Galley, J Dao, D Jurafsky](#)
 - [A persona based neural conversation model \[2016\], J Li, M Galley, C Brockett, GP Spithourakis, J Gao, B Dolan](#)
 - [Adversarial Learning for Neural Dialogue Generation \[2017\], J Li, W Monroe, T Shi, A Ritter, D Jurafsky](#)

Adversarial Learning for Neural Dialogue Generation

Co-Authors

- Will Monroe, PhD Student @Stanford
- Tianlin Shi, PhD Student @Stanford
- Sebastien Jean, PhD Student @NYU Courant
- Alan Ritter, Assistant Professor, Dept of CSE, Ohio State University
- Dan Jurafsky, Professor, Dept of CSE, Stanford University

Goal

“To train and produce sequences that are indistinguishable from human-generated dialogue utterances”.

This paper trended on social media as well...



A screenshot of a tweet from Ian Goodfellow (@goodfellow_ian) posted on January 31, 2017, at 3:37 AM. The tweet text is "Seq2seq RL GANs for dialogue generation: [arxiv.org/pdf/1701.06547...](https://arxiv.org/pdf/1701.06547v1.pdf)". The tweet has 181 retweets and 437 likes. The interface includes a profile picture, name, handle, a "Follow" button, and icons for replies, retweets, likes, and direct messages.

 **Ian Goodfellow**
@goodfellow_ian Follow ▼

Seq2seq RL GANs for dialogue generation:
[arxiv.org/pdf/1701.06547...](https://arxiv.org/pdf/1701.06547v1.pdf)

3:37 AM - 31 Jan 2017

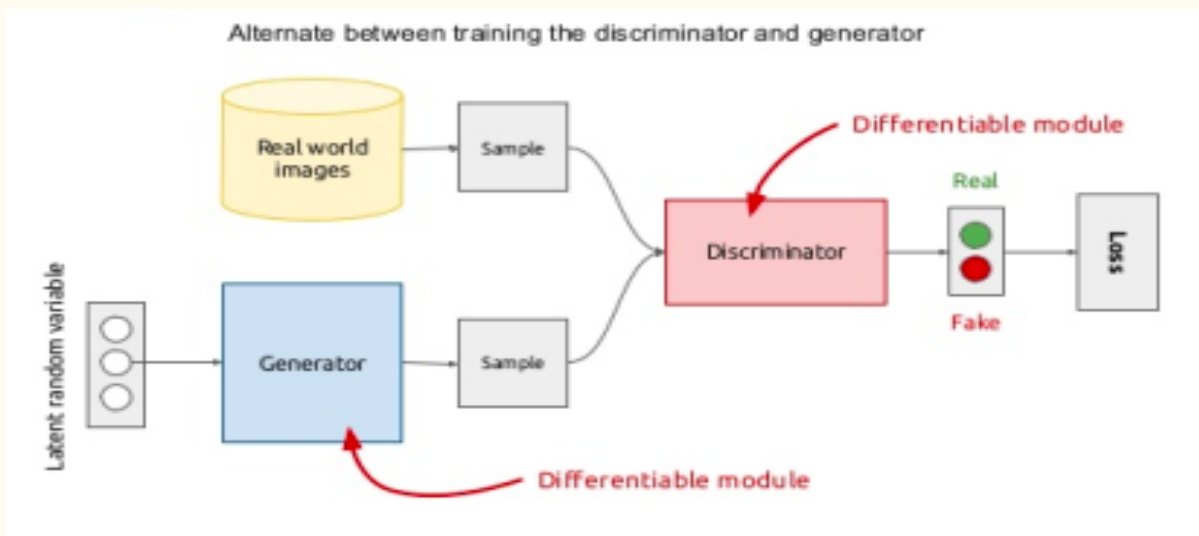
181 Retweets 437 Likes



 4  181  437 

Adversarial Models

It's a Min-Max game between a Generator & Discriminator



Model Used

- Earlier REINFORCE Algorithm was used, which had it's own drawbacks.
 - The expectation of reward is approximated by only one sample and reward associate with it is used for all the samples.

Input : *What's your name*
human : *I am John*
machine : *I don't know*

- Vanilla REINFORCE will assign the same negative weight for all the tokens - [I, don't, know], even though [I] matched with the human utterance.

REGS - Reward Generation for Every Step

- They reward the sequence generated at intermediate steps as well.
- They essentially train their discriminator for rewarding partially decoded sequences.
- They also use Teacher Forcing as well, where the human responses are also fed to the generator, with a positive reward. This helps it to overcome the problems where it can get stuck in Minimas and it would not know which update steps to take.

```
For number of training iterations do
.   For  $i=1, D$ -steps do
.       Sample  $(X, Y)$  from real data
.       Sample  $\hat{Y} \sim G(\cdot|X)$ 
.       Update  $D$  using  $(X, Y)$  as positive examples and
.        $(X, \hat{Y})$  as negative examples.
.   End
.
.   For  $i=1, G$ -steps do
.       Sample  $(X, Y)$  from real data
.       Sample  $\hat{Y} \sim G(\cdot|X)$ 
.       Compute Reward  $r$  for  $(X, \hat{Y})$  using  $D$ .
.       Update  $G$  on  $(X, \hat{Y})$  using reward  $r$ 
.       Teacher-Forcing: Update  $G$  on  $(X, Y)$ 
.   End
End
```

Results

Input	tell me ... how long have you had this falling sickness ?
Vanilla-MLE	i 'm not a doctor .
Vanilla-Sample	well everything you did was totally untrue .
REINFORCE	i don 't know how long it 's been .
REGS Monte Carlo	A few months, I guess .
Input	so i had the doctors test sammy 's response to conditioning .
Vanilla-MLE	sammy wrote the test sammy wrote the test .
Vanilla-Sample	objects pick objects up objects objects objects objects
REINFORCE	i 'm not sure that 's relevant .
REGS Monte Carlo	so he took the pills .
Input	what are you going to do ?
Vanilla-MLE	i 'm going to the movies .
Vanilla-Sample	get him outta here first !
REINFORCE	i 'm going to get you .
REGS Monte Carlo	i 'm going to get you a new car .
Input	they fear your power your intellect .
Vanilla-MLE	you 're the only one who knows what 's going on .
Vanilla-Sample	when they are conquered and you surrender they will control all of us .
REINFORCE	i 'm afraid i 'm not ready yet .
REGS Monte Carlo	i 'm not afraid of your power .

