GAN and RL: A Perspective

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Motivation



Figure: Self-Driving Cars [Erie]

- What's the deal about GAN?
- Was it RL that Alpha-Go used to defeat Lee Sedol?
- Can we possibly devise an idea linking these two?

Generative Adversarial Networks (GAN)



Figure: Fake Faces [NVidia]

- The high-level idea
- Can something still go wrong with GANs?
- (Yes, instability, mode collapses, ...)
- Enter, Wasserstein GAN.
- Earth-Mover Distance

Objective: $min_G max_D E_{x \sim P_r}[log(D(x))] + E_{\tilde{x} \sim P_g}[log(1 - D(\tilde{x}))]$

Architecture

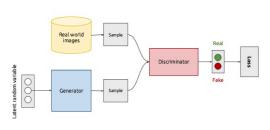


Figure: Skeleton of a GAN model [UPC 2016]

- Even WGAN by itself isn't good enough!
- Lipschitz constraint ⇔
 Weight clipping
- Think model capacity, unbounded gradient
- Thus, Gradient Penalty (WGAN-GP)

Objectives of the current study

- Classification Accuracy ↑ with GANs
- Augmenting available datasets for RL models

(Don't ask what RL is. The professor just taught it last week. You should know)

Reinforcement Learning (RL)



Figure: Go v/s Lee [Shelly Palmer]

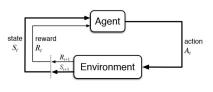


Figure: RL Architecture [Barto/Sutton]

A Pre-Trained RL Model

- Model Courtesy Stanford's CS238 course (Surag et al.)
- MCTS (Monte-Carlo Tree Search) for Policy Improvement
- Policy Iteration through Self-Play

High Level Challenges

- Connecting the dot between GAN images and RL inputs
- Classification Accuracy?
- Training an RL model from scratch.
- Application to Self-Driving Cars

Experiments

- Applied WGAN to Flower Dataset
- Co-GAN did not produce good results with small batch size
- Applied WGAN separately to each class

Generated Daisies



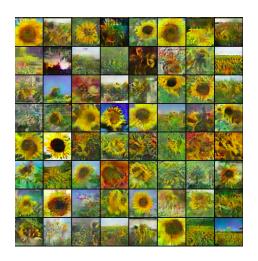
Generated Dandelions



Generated Roses



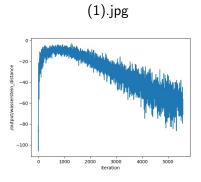
Generated Sunflower



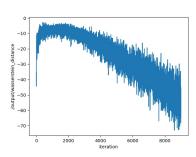
Generated Tulips



Wasserstein Distance Convergence



(2).jpg



Results

- Achieved accuracy of 85.2 using GAN
- Used a DNN classifier to calculate the accuracy of the vavlidation set till 100 epochs.

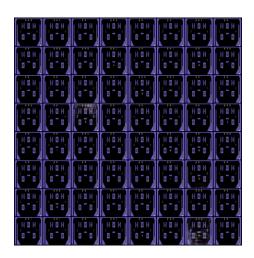
Experimental Challanges

- Initial Learning rate was poor due to weak Critic at the beginning
 - Used more critic iterations initilaly to correct this
- Wasserstein distance moves too slowly
 - Changing learning rates help
- Long training time(spent 500 worth of credits)

Next Steps

- Use conditional GANs on the cloud
 - It might improve tulips, roses, etc. by improving basic features detection
- Select only images which the critic approves (have to see how it would work)
- Figure out the gaps in our RL approach

Generated Pinball images



Questions?

The ball's in your court. Thanks for listening.