

# IMAGAN: Learning Images from Captions Samir Sen, Trevor Tsue, Karan Singhal | {samirsen, ttsue, ksinghal}@stanford.edu

# Motivation

- o Our task is to generate new images of flowers from text captions of flowers
- We were inspired by the idea that a machine could learn to
- We hope this work opens avenues toward a more complete understanding of textual information

# Data

# Oxford Flowers Dataset

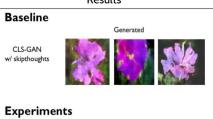
- 8189 images of flowers 5 human-labeled captions per flower

- To overall categories of flowers represented
   GloVe Word Embeddings
   300 dim representation of individual words
- trained on 6B words from Gigaword5 + Wikipedia2014

# Skipthoughts

- 4800-dim learned representation of sentences
   Trained on ~900 million words from BookCorpus

# Results



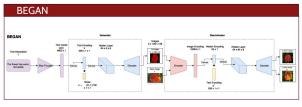


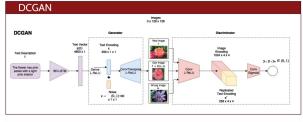
BEGAN w/ LSTM

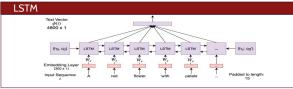




# Generative Adversarial Network (GAN) Architecture









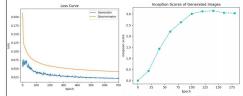
# **Future Work**

- Explore different model architectures like StackGAN, AttnGAN
- Spend more time tuning architecture and hyperparameters to produce better images
- Video reconstruction from text

# **Analysis**

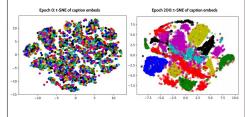
#### **Model Performances**

We perform a variety of architectural enhancements from our baseline, including Boundary Equilibrium (BEGAN), Deep Convolution (DCGAN)



# Learned Embedding Representations

We performed t-SNE dimensionality reduction on learned caption features over the categories of flowers generated



# Training Techniques

#### LSTM

- Initialize hidden states to small random values
   Batch size = 64

# GAN

- Dataset augmentation with random crops + flips
   Gaussian noise for generator
- Normal initialization

- Better resolution/eliminated checkerboard with upsampling One-sided label smoothing Leaky ReLU instead of ReLU
- Bidirectional, 1-layer Xavier weights initialization
- SGD instead of Adam for D Tanh for output
- Batchnorm
- Learning rate decay Average pool instead of max pool ELU instead of ReLU