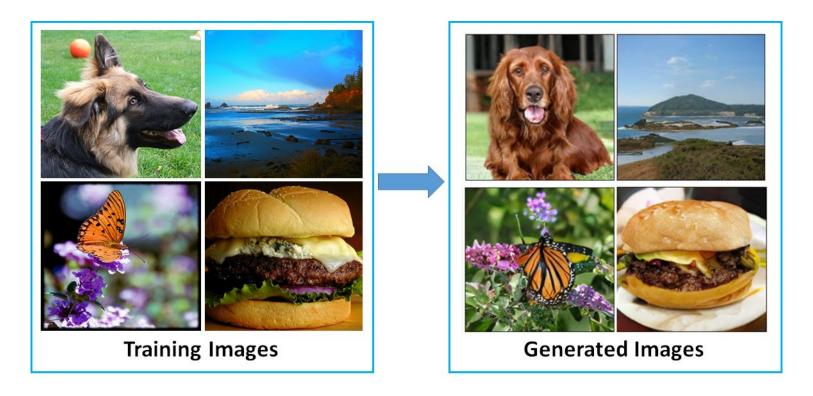


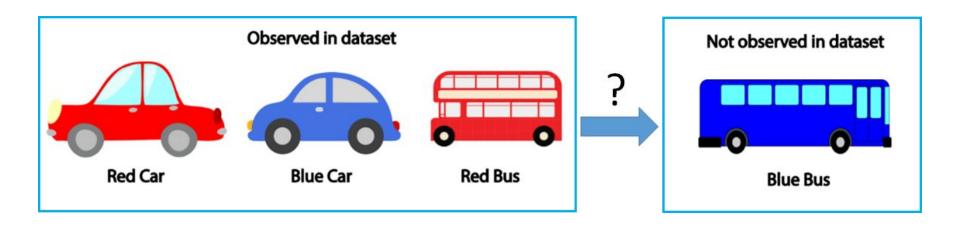
Bias and Generalization in Deep Generative Models

Shengjia Zhao*, Hongyu Ren*, Arianna Yuan, Jiaming Song, Noah Goodman and Stefano Ermon

Success in Generative Modeling of Images

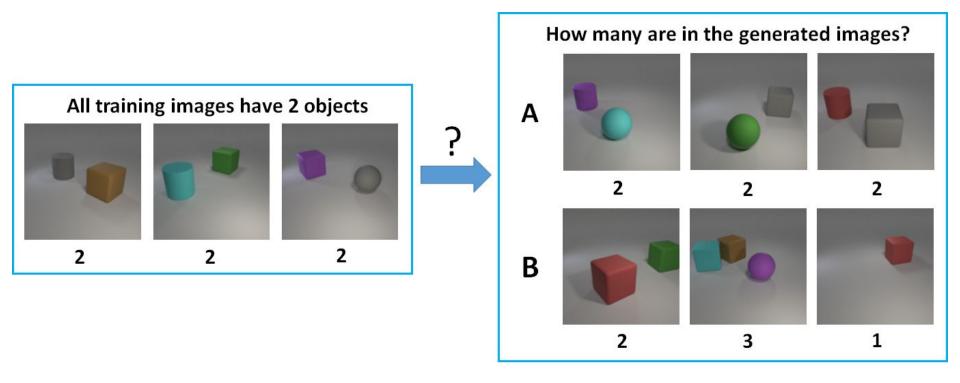


Goal: Understanding Generalization



How do generative models generalize?

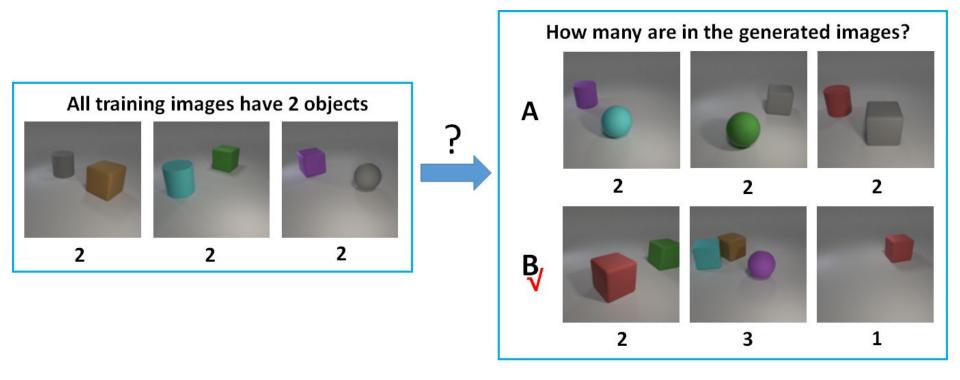
Generalization Example: Object Count



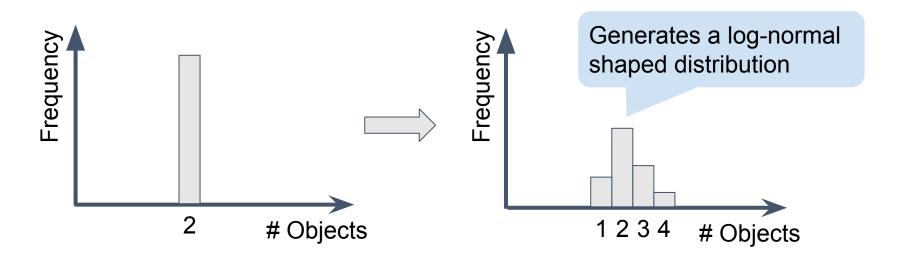
Empirical Study of Generalization: Method

- Design datasets
- Train generative models (VAE, GAN, PixelCNN)
- Observe generalization behavior
- Find common patterns

Generalization Example: Object Count

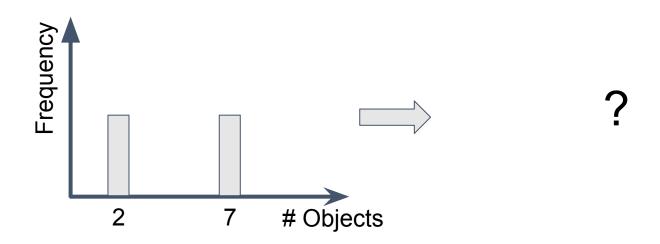


Generalization in Feature Space: Numerosity



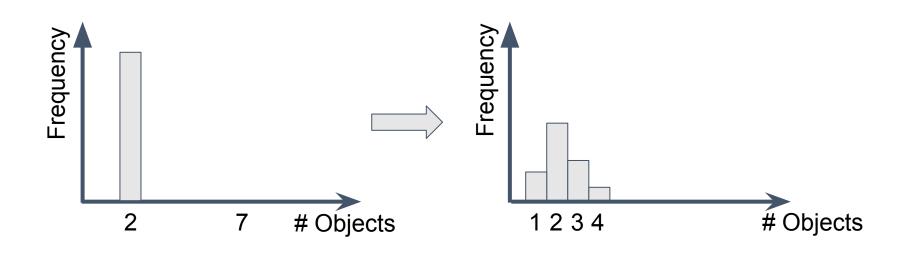
Training Distribution

Multiple Numerosities



Training Distribution

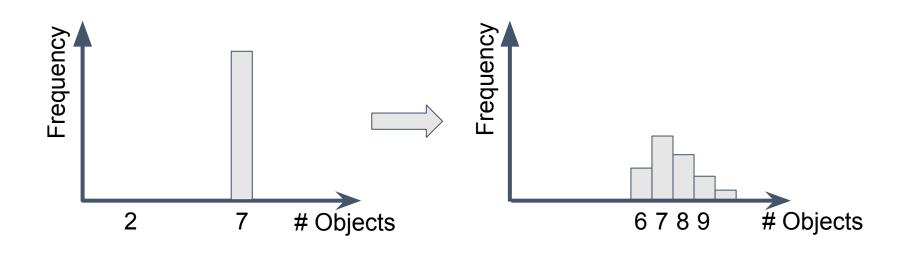
Multiple Numerosities: Only 2



Training Distribution

Generated Distribution

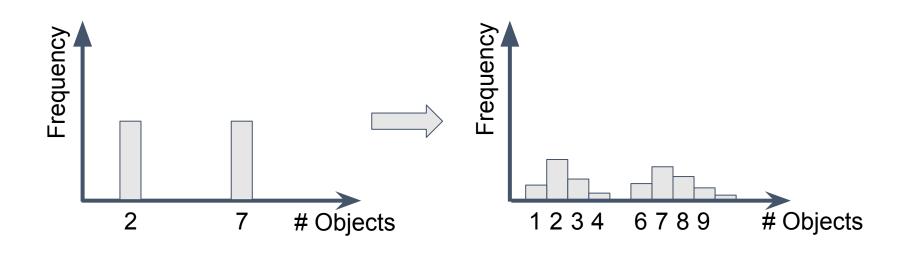
Multiple Numerosities: Only 7



Training Distribution

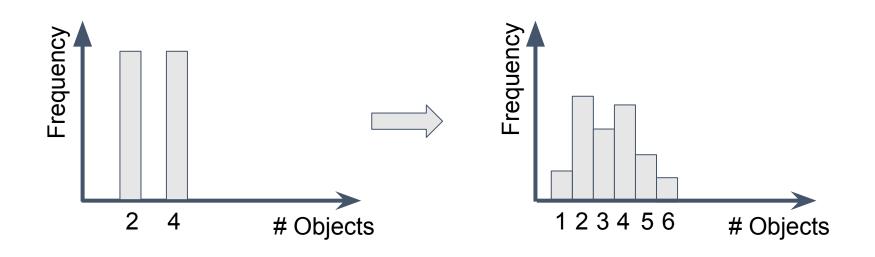
Generated Distribution

Multiple Numerosities: Additive Hypothesis



Training Distribution

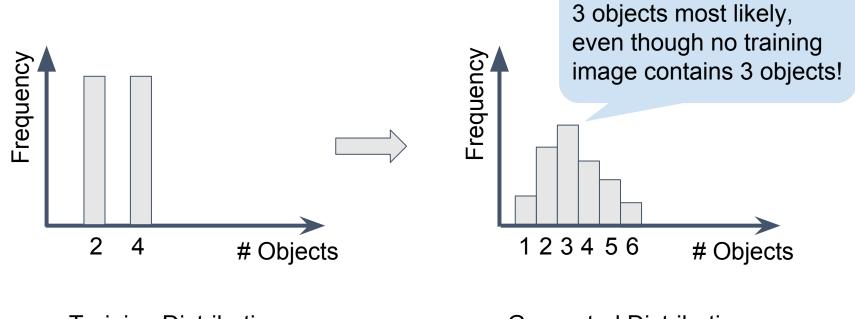
Additive Hypothesis with 2 and 4 Objects



Training Distribution

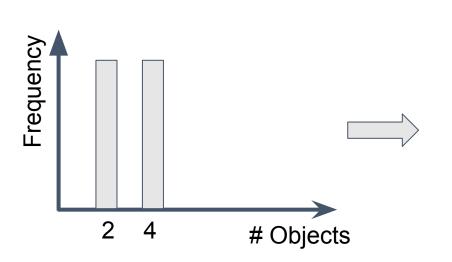
Generated Distribution (Hypothesized)

Actual Result: Prototype Enhancement

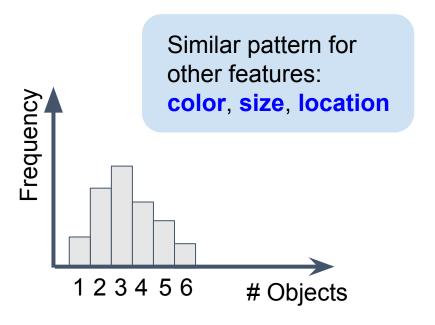


Training Distribution

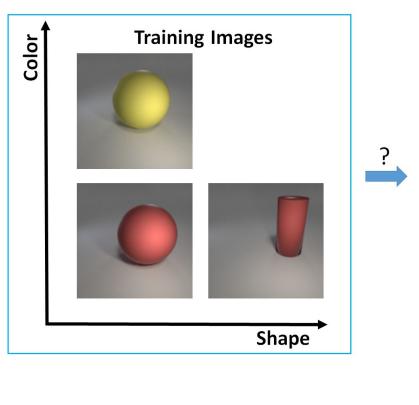
Prototype Enhancement

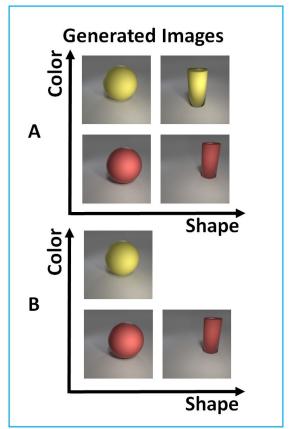


Training Distribution

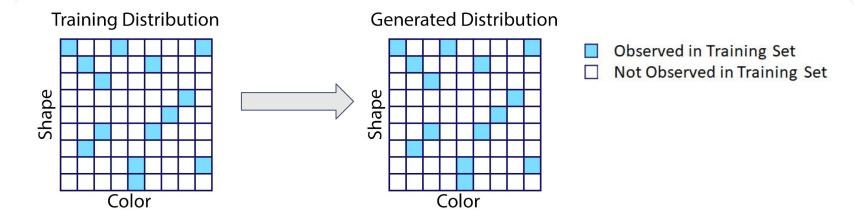


Multiple Features

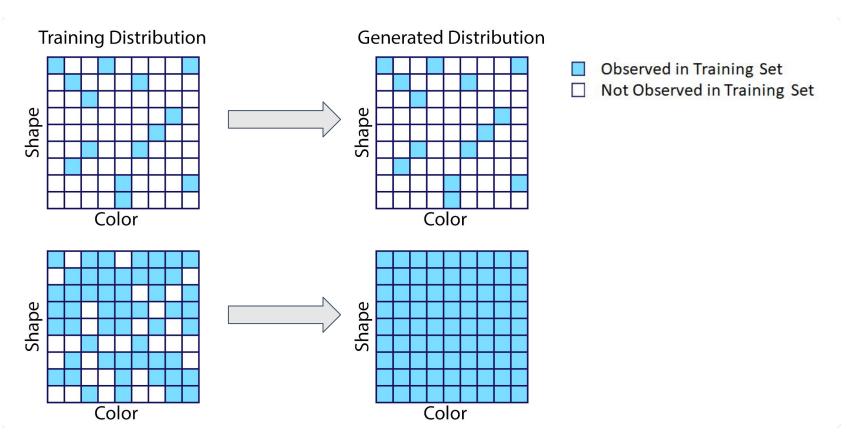




Memorization vs. Generalization



Memorization vs. Generalization



Different Setups, Similar Results

- Different features (shape, color, size, numerosity, etc.)
- Different models: (VAE, GAN, PixelCNN, etc.)
- Different architectures (fully connected, convolutional, etc.)
- Different hyper-parameters (network size, learning rate, etc.)

Conclusion

- New methodology: design datasets to probe generative models
- Observed common patterns across different setups

Welcome to our poster session for further discussions!

Tuesday 5-7pm @ Room 210 & 230 AB #6

Code available at github.com/ermongroup/BiasAndGeneralization