# Learning Neurosymbolic Generative Models via Program Synthesis

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### **Generative Models and Global Structure**



Full Image



⅓ of Image



**Baseline Completion** 

#### **Our Approach: Global Structure as Programs**



# **Application to Image Completion**



Full Image







**Our Completion** 

Third Image

Baseline Completion (No representation of global structure)













#### **Comparison to Baselines**

Model	Synthetic		Facades	
	SGM	BL	SGM	BL
GLCIC	106.8	163.66	141.8	195.9
CycleGAN	91.8	218.7	124.4	251.4
VED	44570.4	52442.9	8755.4	8636.3

Our approach (Synthesis-Guided Generative Model, SGM) significantly **outperforms the baseline in 5 out of 6 experiments** in image completion (as well as in all image generation experiments).

(Scores for GLCIC/CycleGAN represent Fréchet Inception Distance, scores for VED represent negative log likelihood).

# **Experimental Results**



Original Image (Synthetic)



Original Image (Facades)

# **Experimental Results**



Original Image (Synthetic)



Original Image (Facades)



SGM (GLCIC, Synthetic)





SGM (GLCIC, Facades)

# **Experimental Results**



Original Image (Synthetic)





Original Image (Facades)







Baseline (GLCIC, Synthetic)





SGM (GLCIC, Facades)



Baseline (GLCIC, Facades)

# **Future Work**

•More expressive programs

•Better ways of incorporating program structure

Domains beyond images