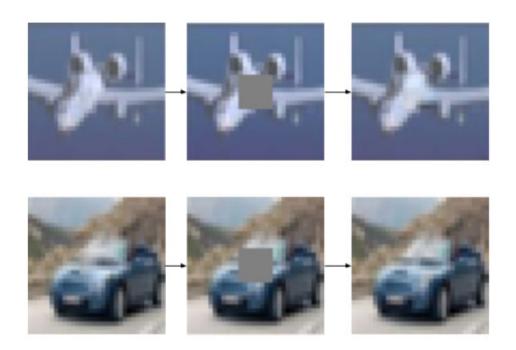
# Beyond CNNs: Encoded Context for Image Inpainting with LSTMs and Pixel CNNs

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## Image Inpainting?

Reconstructing the missing or damaged parts within the image



#### Some Common and Low Level Methods

- Convolutional Neural Networks (CNN in Raw form)
- Pixel-CNN
- Generative Adversarial Network (GAN and or forms)(High Level)
- Other methods based on Autoencoders and Attention Mechanism (High Level)

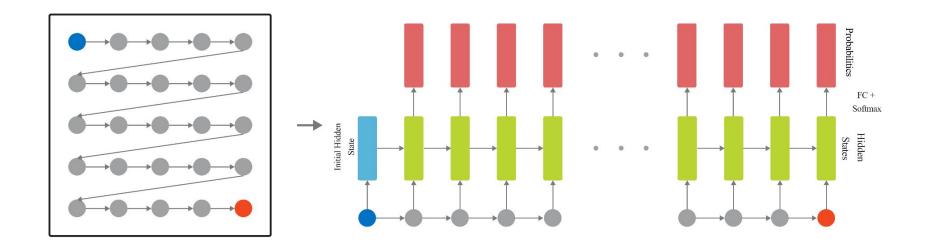
#### Pixel-CNN

- In CNN, the images are not considered in a sequence, instead we take the whole image as an input
- Then we apply filters using Conv Layers, Max/Avg Pooling, and Activation Function (if any in FC layers)
- In Pixel-CNN, we're doing the same but using the Masked Version
- That means we're masking certain values and apply filters to others
- Then those values are fed into fully connected layers to predict probabilities (results)

#### Our Model Extends Pixel-CNN :: Row Flattened LSTM

- Flattened the rows and append to each other to make a 1D array and then fed into an LSTM Network Architecture
- In this extended version of Pixel-CNN, the probability is computed based on each pixel from preceding rows
- That's why we called it, Row Flattened LSTM

#### **Row Flattened LSTM**



### Result

On the left are Ground Truths, while on the right are the reconstructed image









## Result Comparison :: L2 Loss

Model	L2 Loss
Row Flattened LSTM	7.63
Pixel CNN	6.98
Wasserstein GAN	4.26
CNN	7.49

## **Thanks**