DEEP LEARNING FOR COMPUTER GRAPHICS

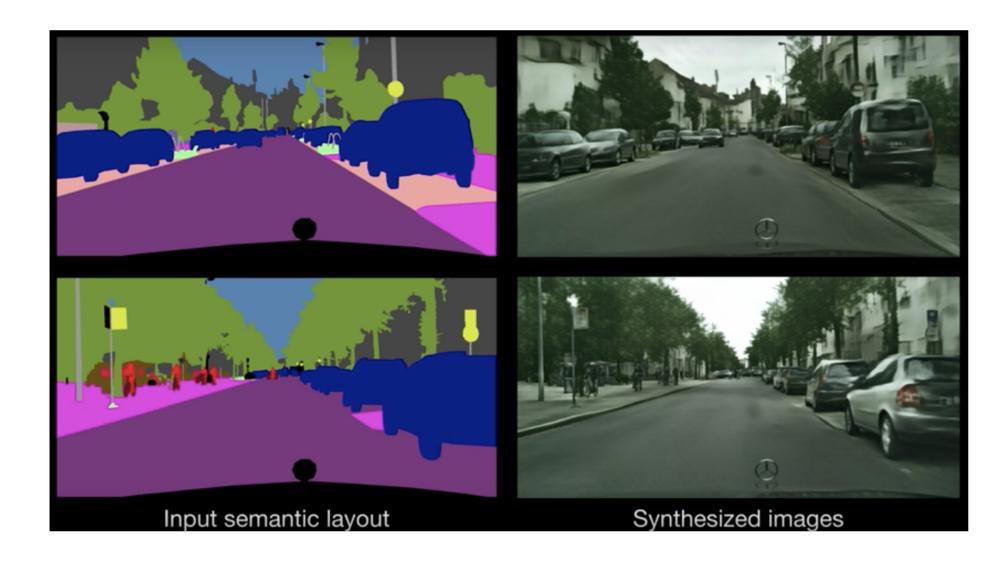
Final Project

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Objective

- To produce an image with photographic appearance that conforms to the input layout
- Input:
 - The input layout is given in the form of a semantic label map
 - Semantic classes are coded by color

Expected Input – Output Example



Proposal

- Paper:
 - Photographic Image Synthesis with Cascaded Refinement Networks
- Authors:
 - Qifeng Chen, Vladlen Koltun

Approach Presented

- Synthesizing photographic images by a direct supervised learning of single feedforward convolutional network trying to minimize regression loss
- Works seamlessly for high image resolutions (2 megapixels)

Training Dataset

- Inverse semantic segmentation
- Cityscapes dataset
 - https://www.cityscapes-dataset.com/
 - 2048X1024



gtFine_trainvaltest.zip (241MB) [md5]

fine annotations for train and val sets (3475 annotated images) and dummy annotations (ignore regions) for the test set (1525 images)

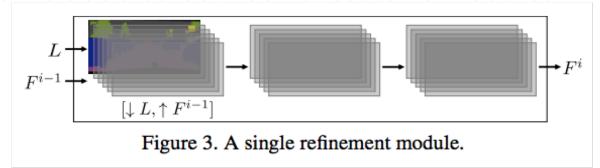


leftImg8bit_trainvaltest.zip (11GB) [md5]

left 8-bit images - train, val, and test sets (5000 images)

Architecture

- Consists of Cascaded Refinement Network (CRN)
- Each module M_i operates at a given resolution
- Resolution is doubled between successive modules



- Each module has 3 layers:
 - Input layer: $w_i x h_i x (d_{i-1} + c)$
 - Intermediate layer: w_i x h_i x d_i
 - Output layer: w_i x h_i x d_i

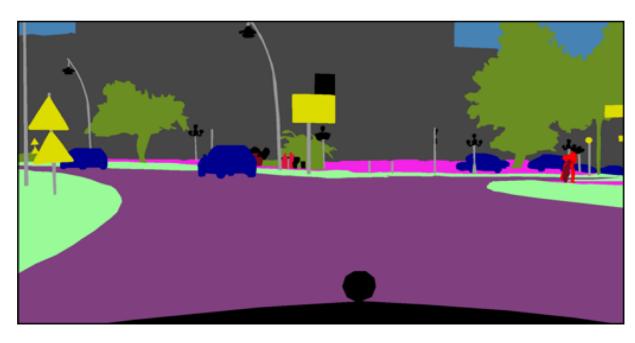
Architecture

- Each layer is followed by 3x3 convolutions, layer normalization & LReLU non-linearity
- The output layer of final module is not followed by above layers. A linear projection is applied to map Fⁱ to generate output color image (w_i x h_i x 3)
- Total number of refinement modules depends on the output resolution
- Number of refinement modules for 8x4 to 512x256 is 7

Challenges

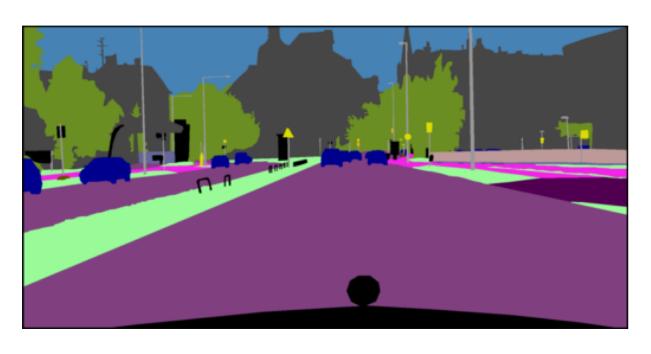
- Coding Challenges
- Execution Challenges
 - AWS
 - Google Colab
 - TAMU HPRC
 - Google Cloud Platform

Results





Results





Thank You