

Image Segmentation and Synthesis

Saturday, March 23, 2019 8:21 PM

★ Image Segmentation

Instance segmentation : label all individual objects of a given instances

Semantic segmentation : per-pixel markup of image where label corresponds to a specific object

Object extraction : selecting a given object

Co-segmentation: Exclude the same object from different images

Unsupervised segmentation: Grouping pixels

1. Over segmentation

Segment the object into fragments/subcomponents.

- Heuristic methods (region growing, split and merge regions)
- Graph-based methods
- Energy-based methods (snakes, TurboPixels)
- Clustering-based approaches (Mean shift, QuickShift, SLIC)

Mean Shift: For each point, find the nearest mode of the distribution density, Split the entire sample into clusters according to the modes closest to them.

Simple Linear Iterative Clustering(SLIC): initialize the clusters over the grid at a distance s, use k-means procedure, stop when the change is less than the threshold.

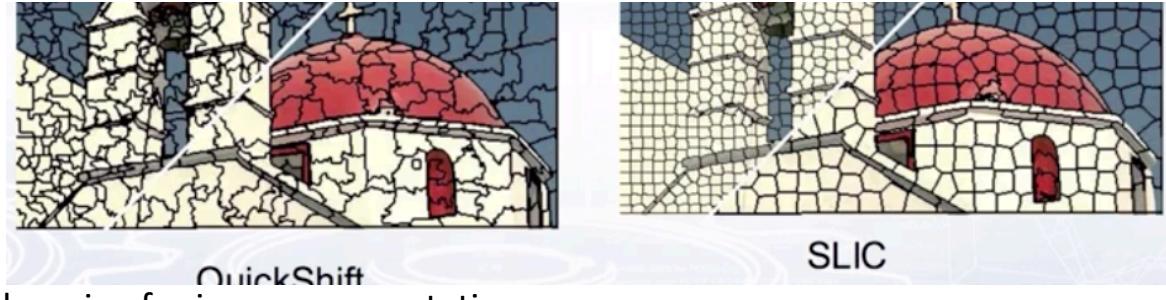


Efficient Graph-Based



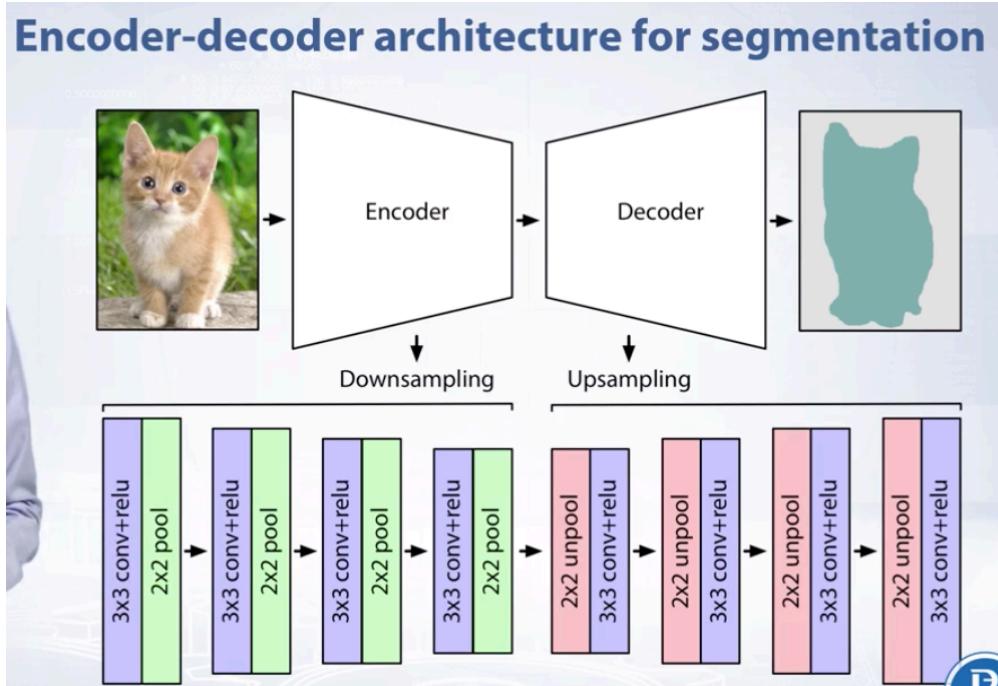
TurboPixel





2. Deep learning for image segmentation

- a. Naïve approach: fully convolutional network
Take a pre-trained CNN, convolutionalize fully-connected layers
Use multi-class cross entropy loss function
Fine-tune the fully-convolutional network
- b. Encoder-decoder architecture for segmentation

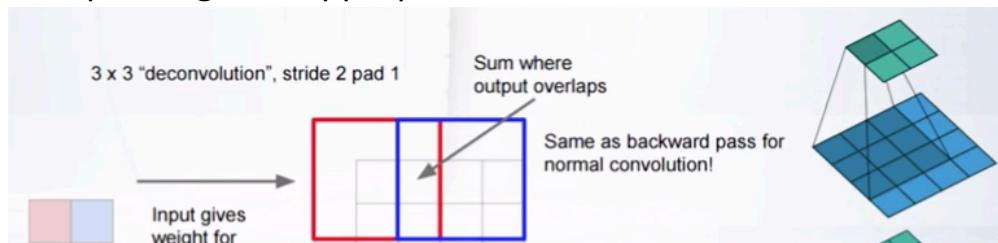


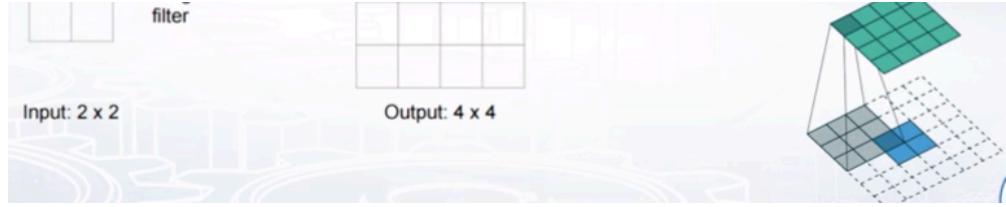
E.g. SegNet Model:

topologically identical to the 13 convolutional layers in the VGG16 network, transposed convolution in the decoder network, unpooling mechanism with Max Location switches.

- i. The transposed convolution layer

Kernel's and stride's sizes remain the same, but now we should use zero padding with appropriate size.





ii. The unpooling layer

- 1) Bilinear interpolation
- 2) Bed of nails
- 3) Max location switches

E.g. U-net

Typical downsampling-upsampling architecture, skip connections with concatenation.

3. Human pose estimation as image segmentation

Top-down approach: person detector -> single-person pose estimation for each detection

- a. Naïve regression task

Bottom-up approach:

- a. DeepCut: initial detections/part candidates and pairwise terms -> cluster parts belong to one person

★ Image Segmentation

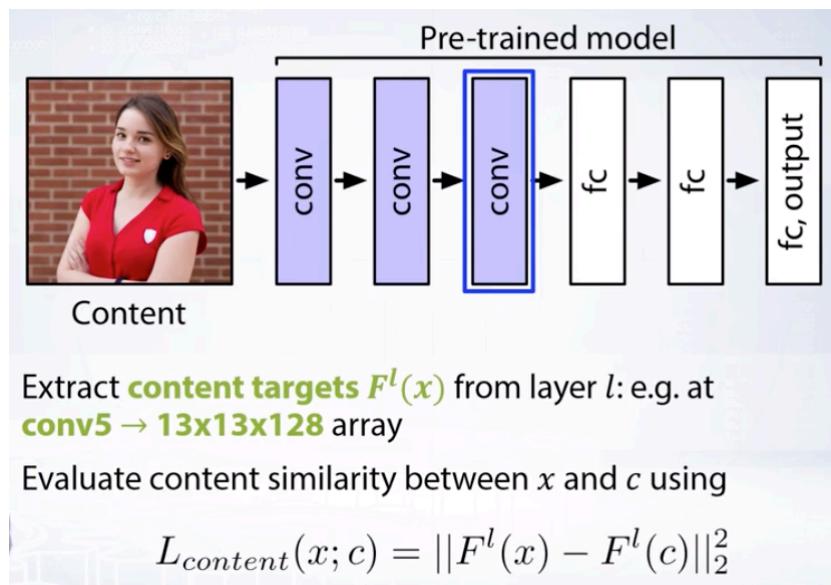
1. Style transfer

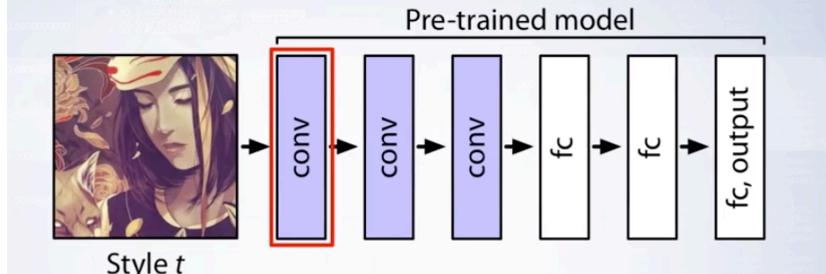
- a. Formulation:

Content c and Style t are presented ->

Goal: produce output x matching the content and the style ->

Using feature activations in a convolutional network as a similarities metric



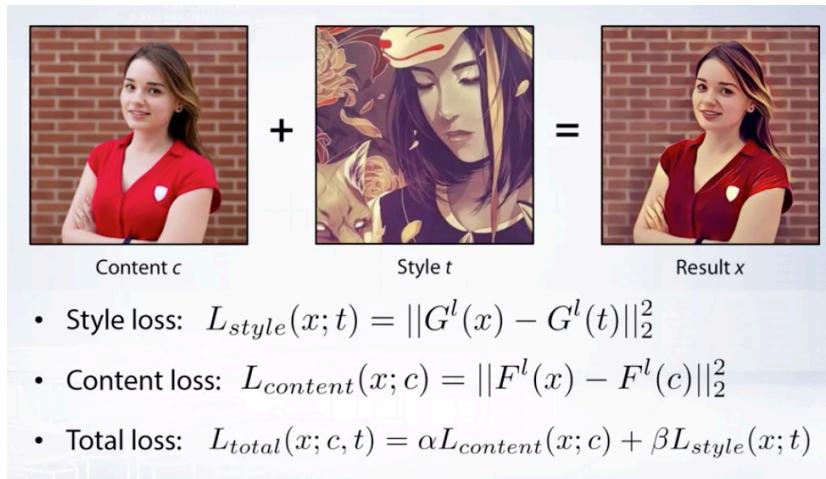


- Extract **style targets (Gram matrices)** $G^l(x)$ from layer l :
e.g. at **conv1** with **224x224x64** array of activations \rightarrow **64x64** Gram matrix

$$G_{i,k}^l = \sum_k F_{i,k}^l F_{j,k}^l$$

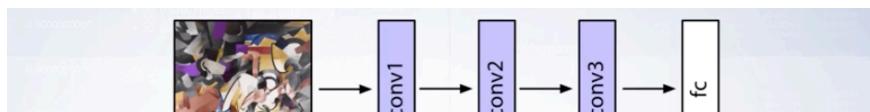
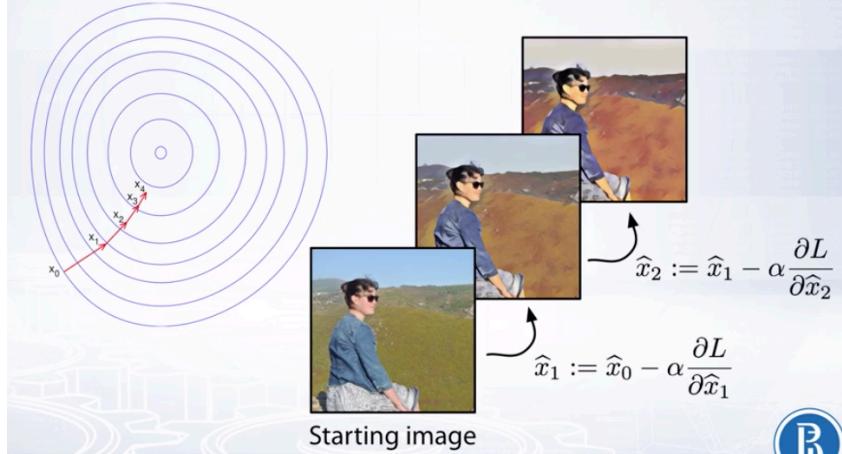
- Evaluate style similarity between x and t using

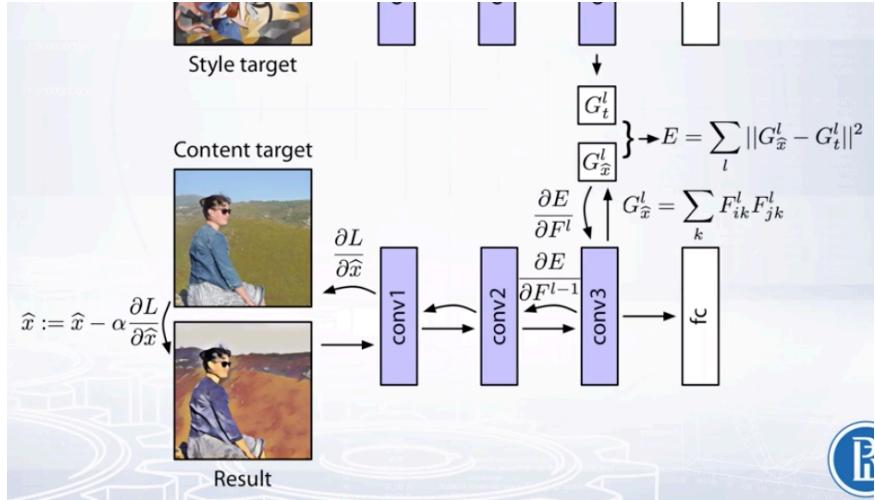
$$L_{style}(x; t) = \|G^l(x) - G^l(t)\|_2^2$$



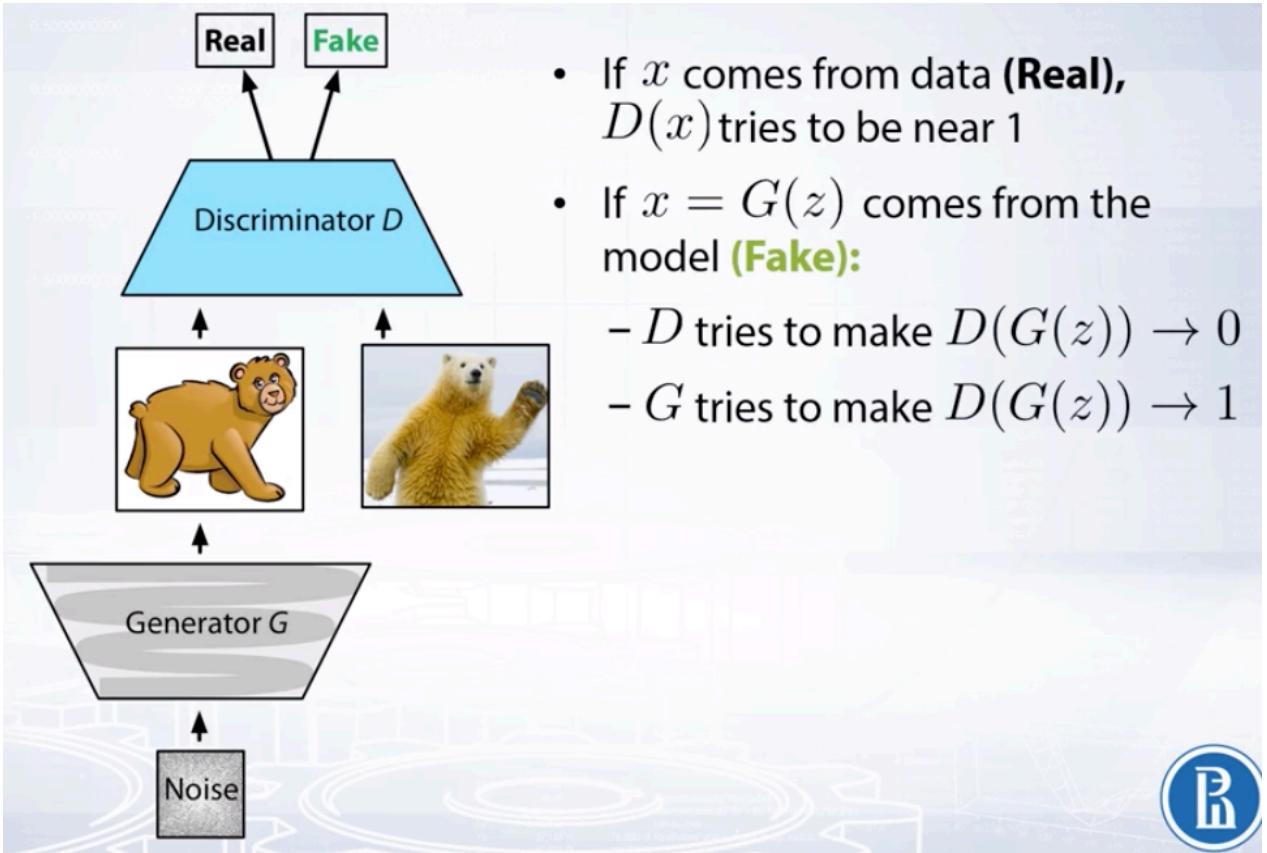
b. Image generation by optimization

Produce image x such that $x^* = \operatorname{argmin}_x \mathcal{L}(x)$





2. Generative adversarial network (GAN)



a. Training objectives for GAN:

- Discriminator objective (*standard cross-entropy*):

$$J(D) = \mathbb{E}_{x \sim p_{data}} \log D(x) + \mathbb{E}_{z \sim p(z)} \log(1 - D(G(z)))$$

- Generator objective:

$$J(G) = \mathbb{E}_{z \sim p(z)} \log(1 - D(G(z)))$$