## Style transfer

© 2019 Philipp Krähenbühl and Chao-Yuan Wu

#### Style transfer

- Content of source
- Style of target
- How do we measure style and content?

Image generative by <a href="https://github.com/">https://github.com/</a>
<a href="pytorch/examples/tree/master/">pytorch/examples/tree/master/</a>
<a href="fast\_neural\_style">fast\_neural\_style</a>



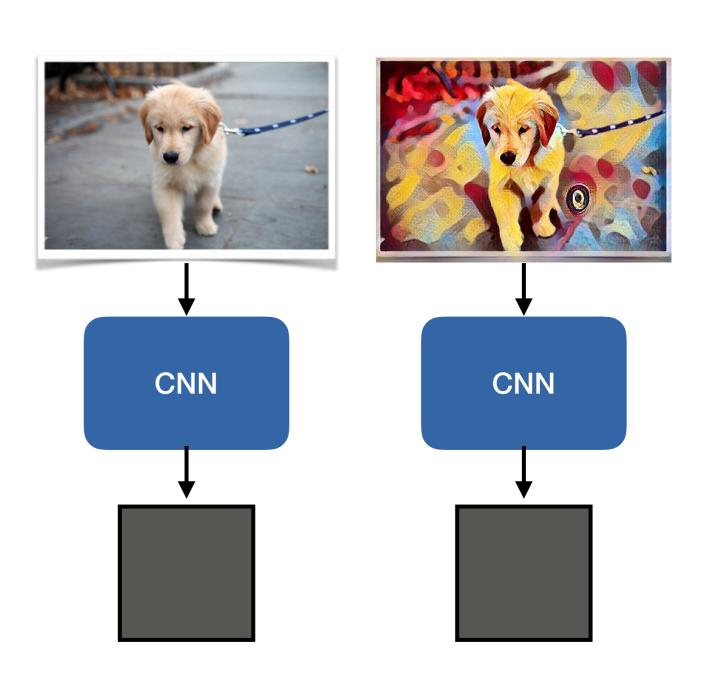




#### Content

 High level activations of a CNN

• e.g. VGG



#### Style

- Low level activations?
  - Contain both style and content
    - Spatial arrangement

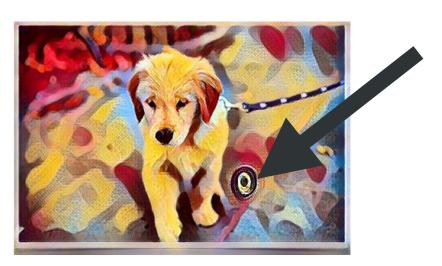




## Style

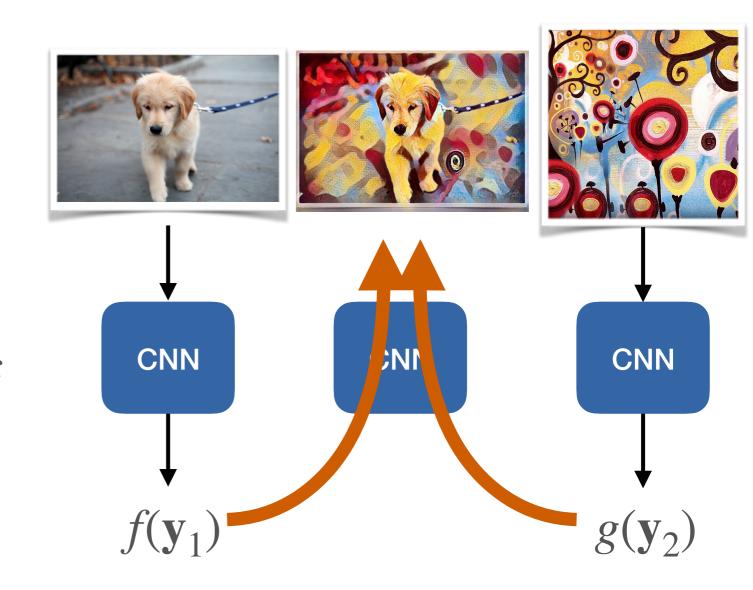
- Solution: Statistics of low level activations
  - No (global) spatial information
  - Local patterns only





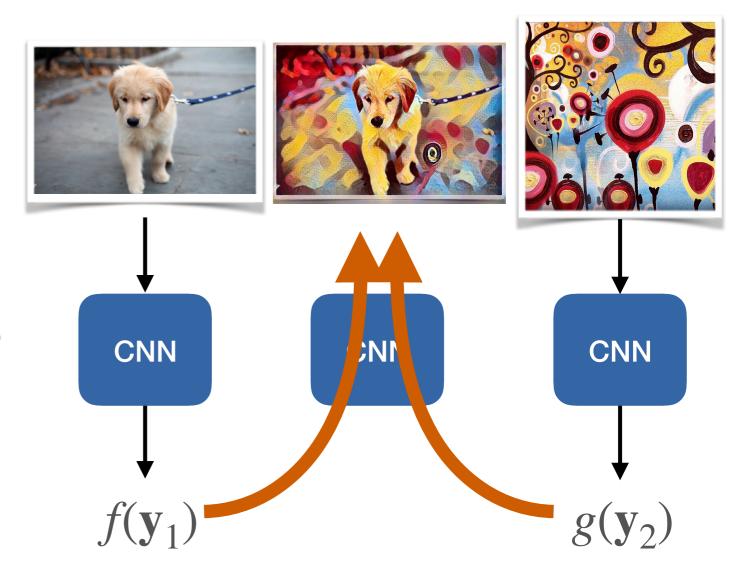
## Neural style transfer

- Optimize for image X
  - With high level activations  $f(\mathbf{y}_1)$
  - and Gram statistic of low level activation  $g(\mathbf{y}_2)$



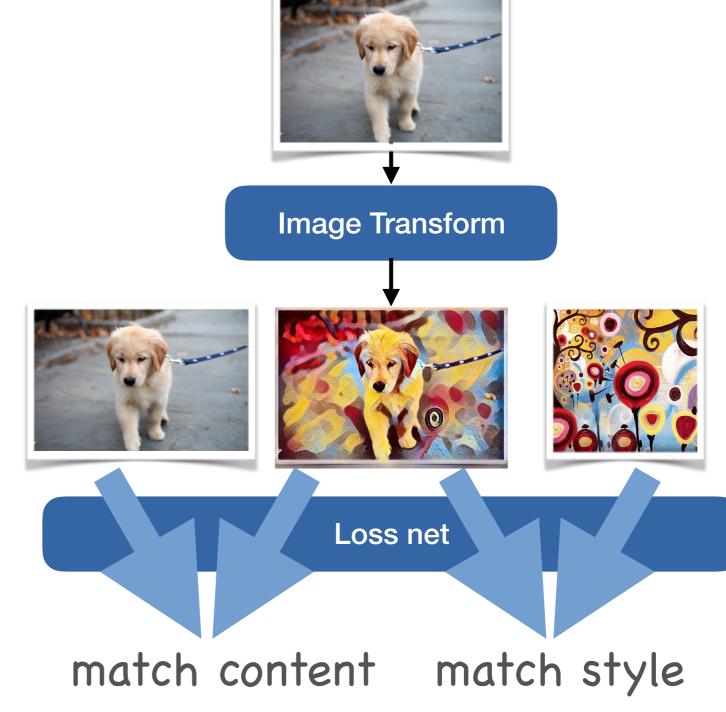
# Neural style transfer -Issues

- Optimization is slow
  - 100-1000
     backpropagation steps
     per image



## Fast neural style transfer

- Train a network to do style transfer
  - Objective
    - Match activations
    - Match Gram statistics
- Fast inference
- Slower training



Perceptual Losses for Real-Time Style Transfer and Super-Resolution, Johnson et al., ECCV 2016