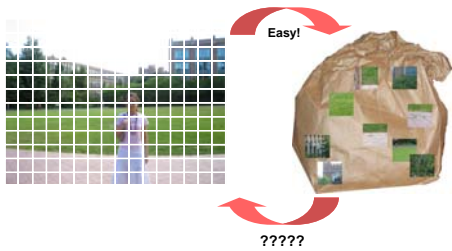


The patch transform and its applications to image editing

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Motivation

From a set of non-overlapping patches from an image, can we reconstruct the image?



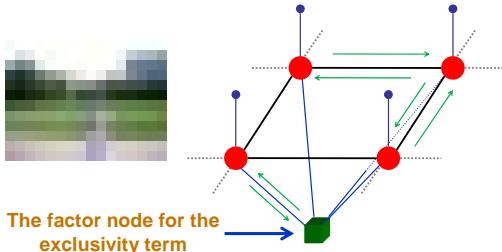
Solving jigsaw puzzle is NP-complete!

The image model

$$P(\mathbf{x}) = \frac{1}{Z} \prod_{i=1} \prod_{j \in \mathcal{N}(i)} \underbrace{p(y_i | x_i)}_{\text{Local evidence}} \underbrace{p_{i,j}(x_j | x_i)}_{\text{Patch prior}} \underbrace{p(x_i)}_{\text{Exclusivity term}} E(\mathbf{x})$$

- Adjacent patches should plausibly fit next to each other.
- Each patch should not be used more than once (we name it the **exclusivity term**.)
- The image structure and user's constraints should be maintained.

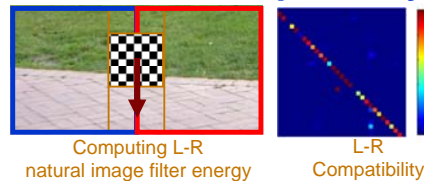
The inverse patch transform



The factor node for the exclusivity term

Belief propagation is used to solve for patch assignments

Pair-wise compatibility



$$\psi_{i,j}^A(k,l) = \frac{1}{Z} \prod_{l,m} \sum_{q=1}^J \left\{ \frac{\pi_q}{\sigma_q} \exp(-w_l^T x_m(k,l)) \right\}$$

Exclusivity term

$$m_{f_i}(x_i = l) \approx \prod_{t \in S \setminus \{i\}} \sum_{x_t=1}^M \psi_{F_t}(x_t | x_i = l) m_{t,f}(x_t) \\ = \prod_{t \in S \setminus \{i\}} (1 - m_{t,f}(x_t = l))$$

The factor node tells others not to use the patch already claimed by another node.

Image editing applications

Subject reorganization



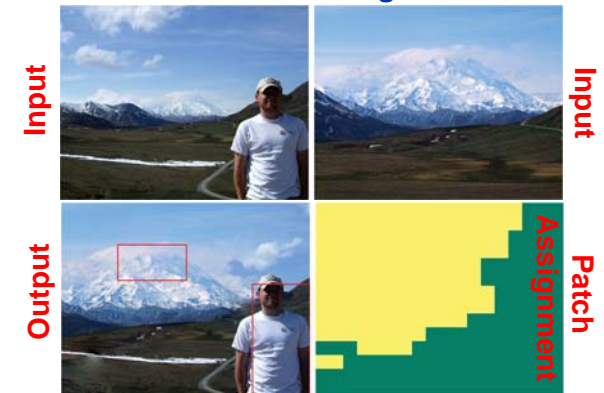
Texture control



Image retargetting

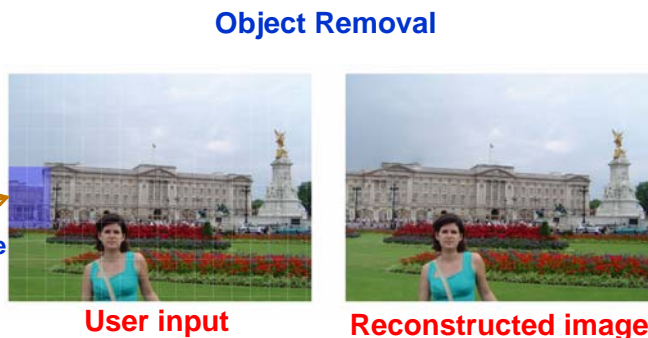
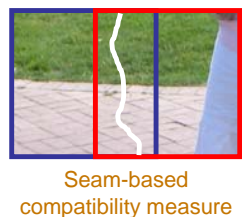


Photomontage

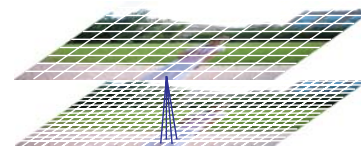


Extensions

The patch transform using overlapping patches can reduce artifacts.



Multiscale Patch Transform



The reconstructed image using coarse patches is refined with smaller patches.