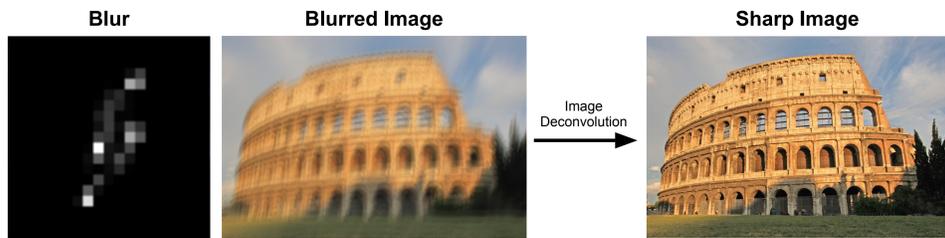


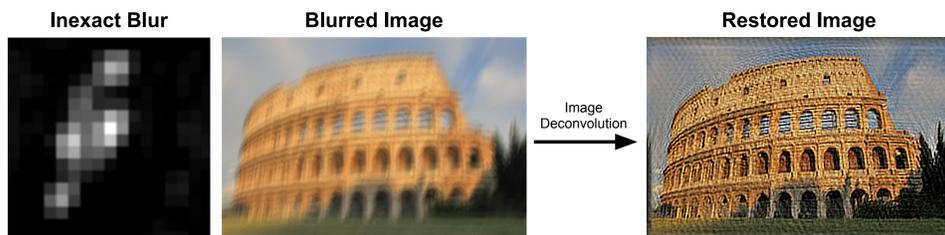


PROBLEM

The problem we consider is non-blind deconvolution



when the PSF (blur) is not fully known or exactly estimated by a blind deconvolution method, artifacts appear in the sharp reconstruction.

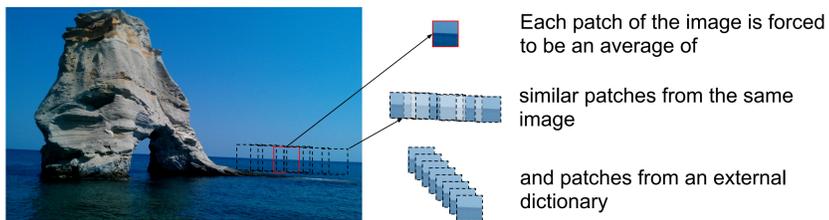


- We address deblurring when blur is uncertain by introducing a prior on the sharp image.
- The prior is built by exploiting the partial knowledge of the blur.

CONTRIBUTIONS

- A novel non-local image prior built directly from the blurry input.
- A unified framework that combines dictionary-based and non-local approaches.
- A consensus strategy that exploits partial knowledge about the blur to discard correspondences due to the blur.

PATCH-BASED FRAMEWORK



Averaging of non-local corresponding patches is effective to remove artifacts such as noise.

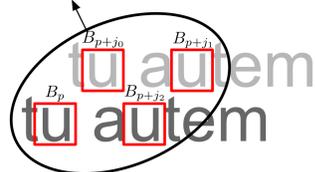
SELF-SIMILARITIES IN BLURRED IMAGES

When we apply the above procedure to a blurred image incorrect correspondences may be generated.



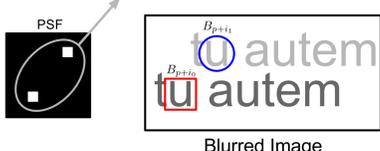
Sharp Image

Set of non-local corresponding (self-similar) patches
 $C_p = \{j \in \mathbb{Z} : \|B_p - B_{p+j}\|^2 \leq \epsilon^2\}$

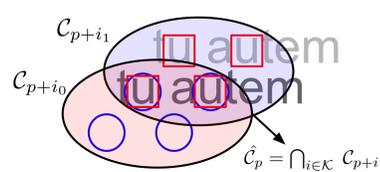


c) Set of similar patches from the blurred image. Which of these are correspondences also in the sharp image?

set of dominant blur peaks
 $\mathcal{K} = \{i \in \mathbb{Z} : |\max(k) - k_i| \leq \tau\}$



Blurred Image



$$\hat{C}_p = \bigcap_{i \in \mathcal{K}} C_{p+i}$$

d) The correlated patches shown in (a) are obtained by overlapping the correspondence sets of the two patches.

b) Blurred version of the previous image.

REFERENCES

- [1] S. Cho, J. Wang and S. Lee. Handling Outliers in Non-blind Image Deconvolution In ICCV '11

ALGORITHM

The image prior is connected to the patch-based framework as

$$\begin{array}{l} \text{Dictionary formed of} \\ \text{only patches from the same} \\ \text{image} \end{array} \longrightarrow \begin{array}{l} \text{Image Prior} \\ A = I - C^T \\ b = 0 \end{array}$$

where C is a matrix that defines correspondences between patches. For any pair of patches i, j we define $C_{i,j} = c$ if $j \in \hat{C}_i$ and 0 otherwise. c is defined such that $\sum_j C_{i,j} = 1$.

$$\begin{array}{l} \text{patches from the same image} \\ \text{and patches from an external} \\ \text{dictionary} \end{array} \longrightarrow \begin{array}{l} A = I - C_{self}^T \\ b = (D_{ext} C_{ext})(k) \end{array}$$

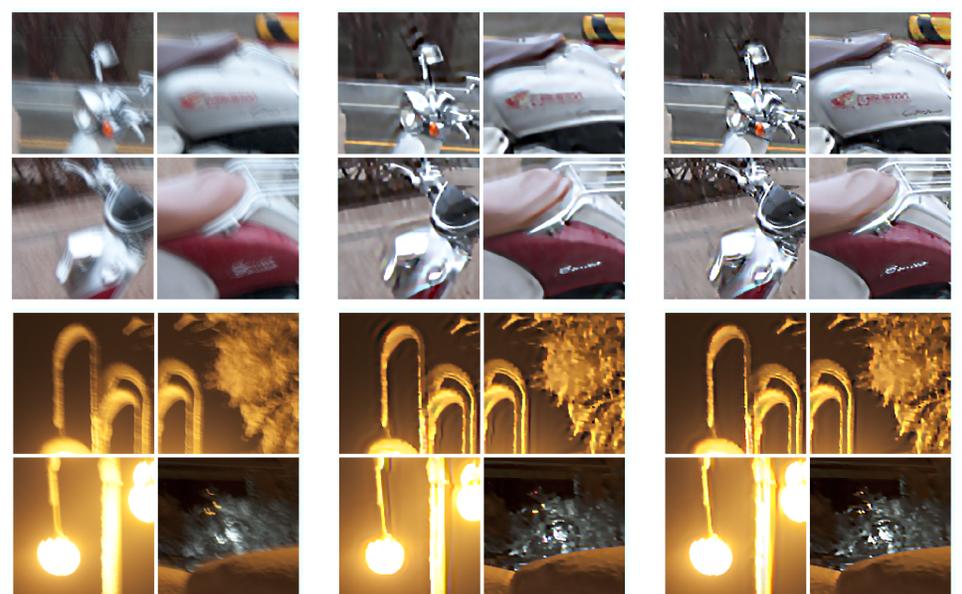
where k identifies the central row of the matrix

where C_{self} defines similar patches within the blurred image, and C_{ext} defines for each patch of the blurred image similar patches in the dictionary D_{ext} .

The sharp image is then estimated by minimising the following energy

$$\left\| \begin{array}{c} \text{Blur} * \text{Sharp Image} - \text{Blurred Image} + \text{Sparse Noise} \\ \text{Unknown} \quad \text{Unknown} \end{array} \right\|_2 + \left\| \begin{array}{c} A \cdot \text{Sharp Image written as vector} - b \\ \text{Sparse Noise} \end{array} \right\|_1$$

RESULTS



Selected areas from real blurred images.

Deblurring results with the algorithm of Cho et al. [1].

Deblurring results with our algorithm.

WEB

Project webpage
<http://www.iam.unibe.ch/~cvd/dperrone/uncertainblur/>



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