

# Learning Continuous Implicit Representation for

# Near-Periodic Patterns

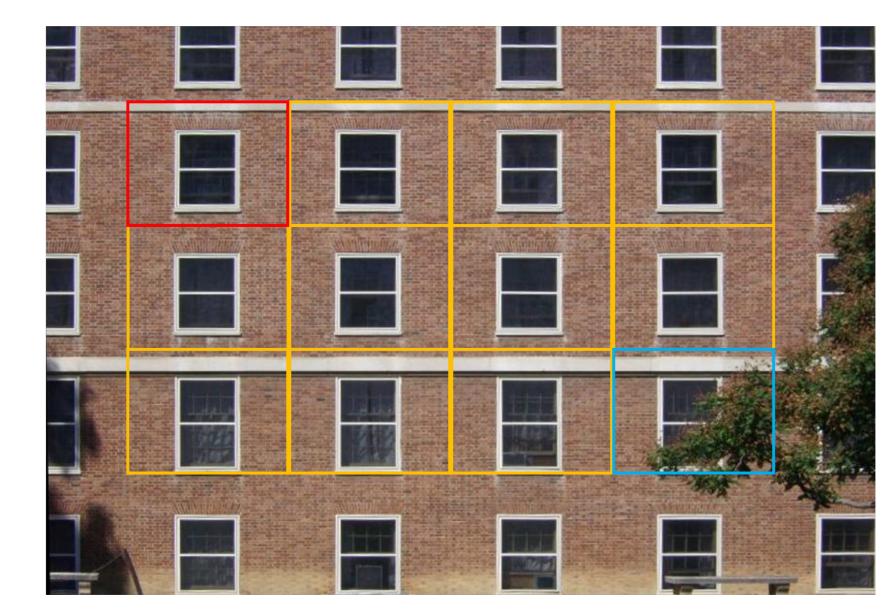
Bowei Chen, Tiancheng Zhi, Martial Hebert, Srinivasa G. Narasimhan

paper, code, and video are available:

https://armastuschen.github.io/projects/NPP\_Net/

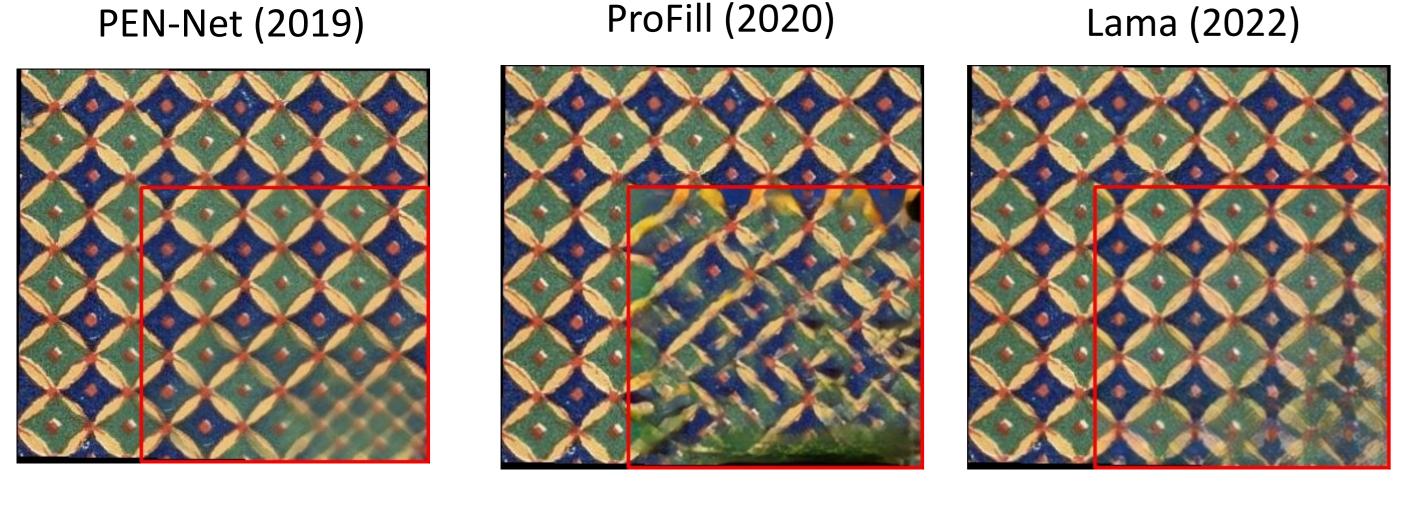
Carnegie Mellon University

#### What is Near-Periodic Patterns (NPP)?



- (1) Global Consistency: a motif (red box) tiles spatially across image (orange boxes).
- (2) Local Variations: appearance variations in different motifs due to design choice and occlusions (red and blue boxes).

# Completing NPP using Existing Methods



Existing methods designed for general scenes fail in this simple NPP. Pixels inside red boxes are completed (unknown).



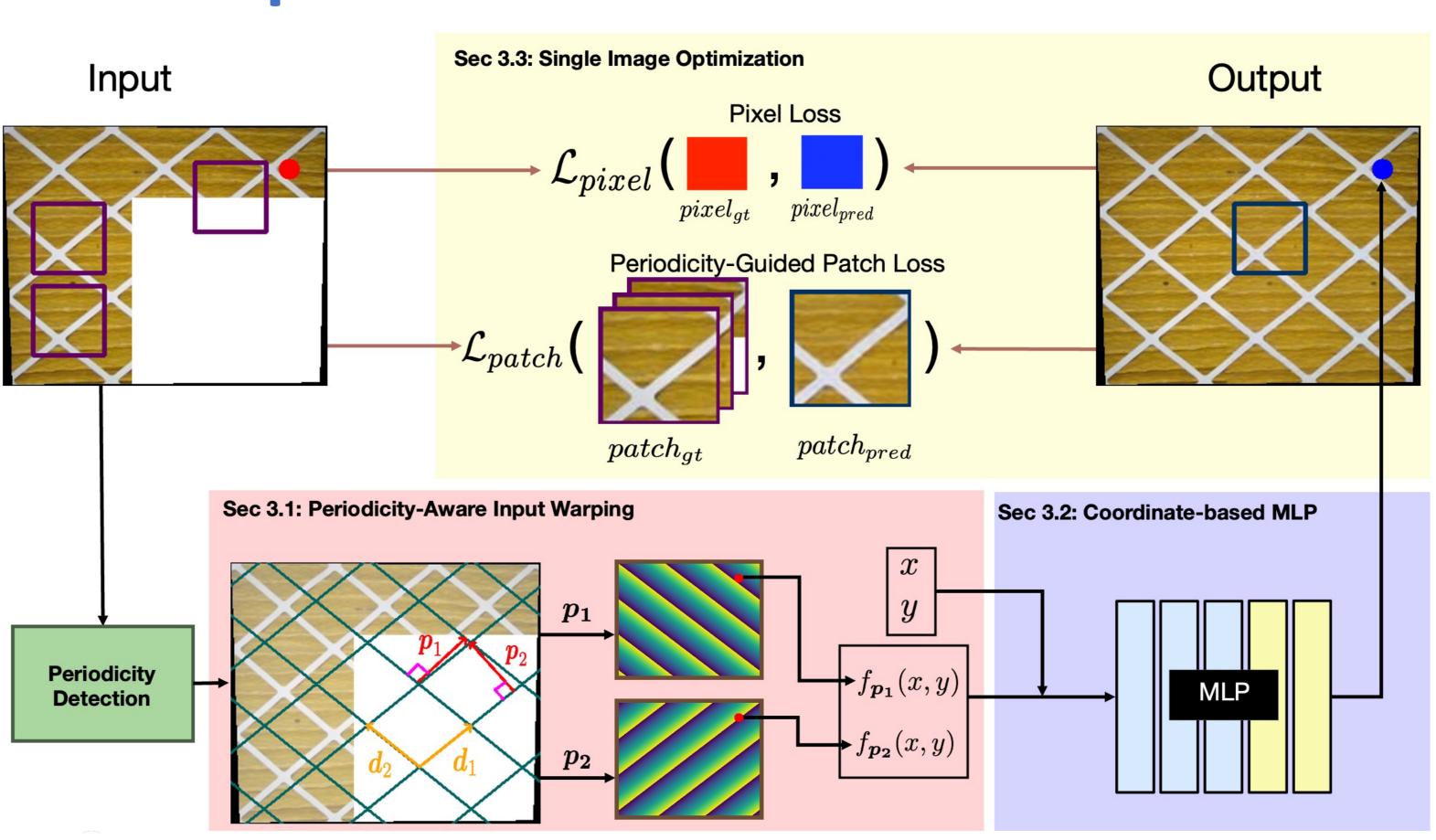
Although existing methods designed for NPP work in the previous simple NPP, they fail in this complicated façade NPP. Pixels inside red boxes are completed (unknown).

### Core Idea:



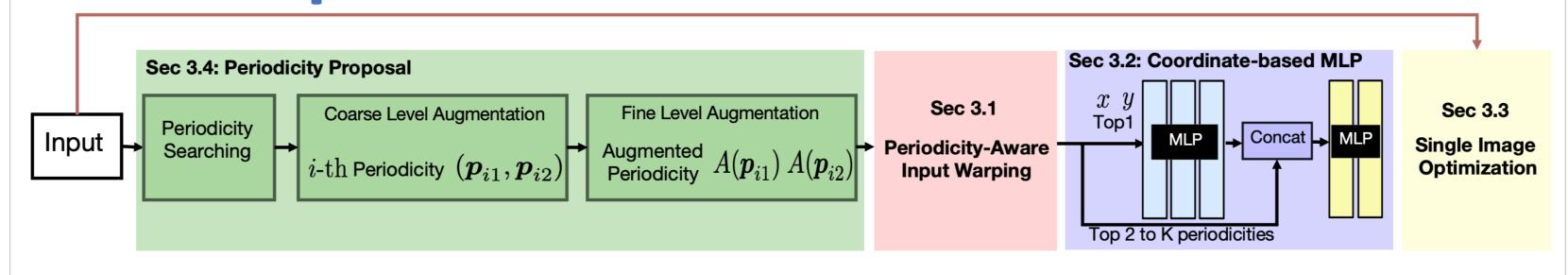
Our primary application is NPP completion, but we also study segmentation and remapping as extension.

# Initial Pipeline:



- Periodicity-Aware Input Warping (pink) warps input coordinates using detected periodicity.
- Coordinate-based MLP (blue) maps warped and input coordinate features to an RGB value.
- Single Image Optimization (yellow) uses pixel loss and periodicityguided patch loss on a single NPP image.

## Final Pipeline:



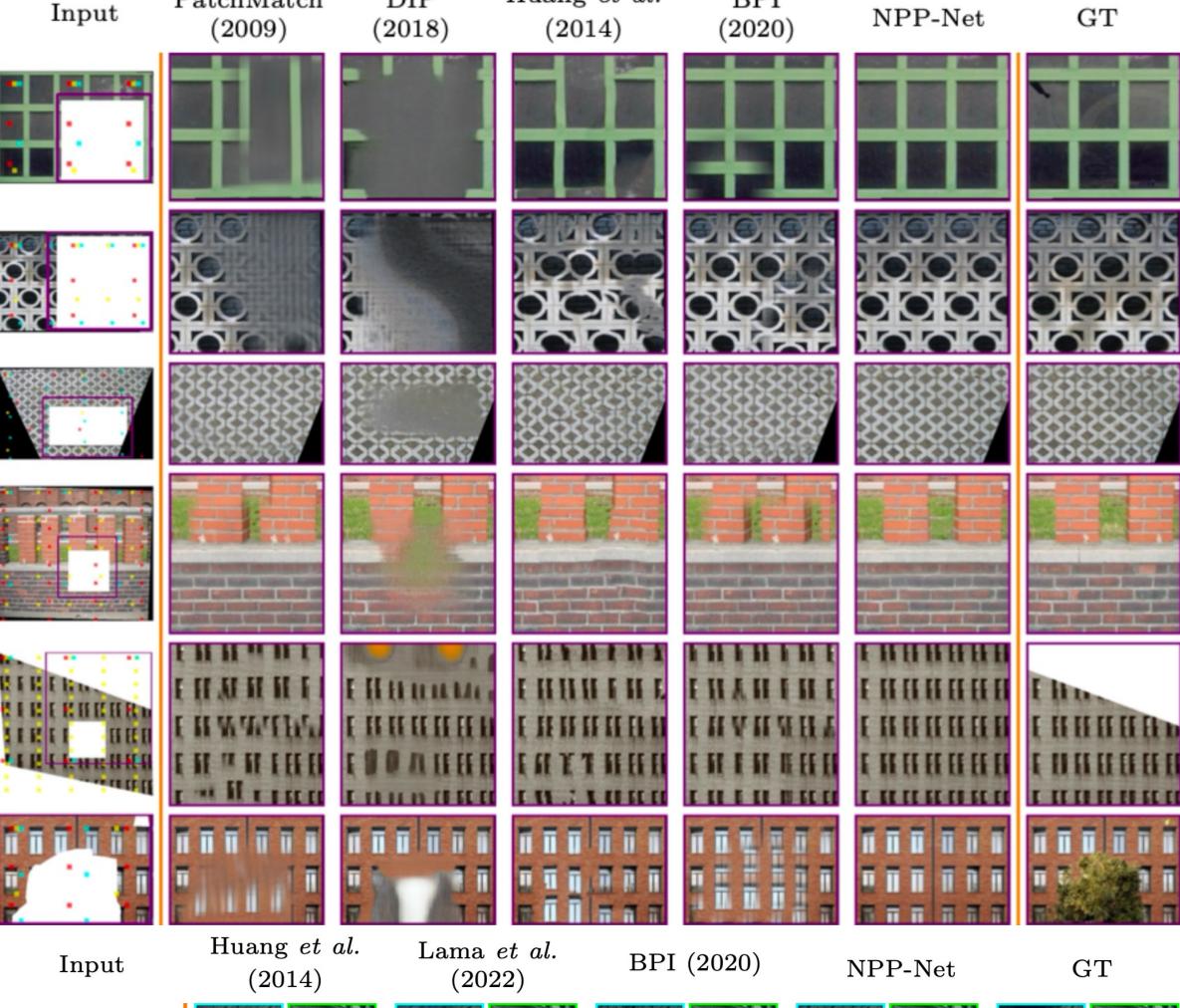
Periodicity Proposal (green) automatically searches and augments the input periodicity to handle inaccurate periodicity detection and encourage the global consistency.

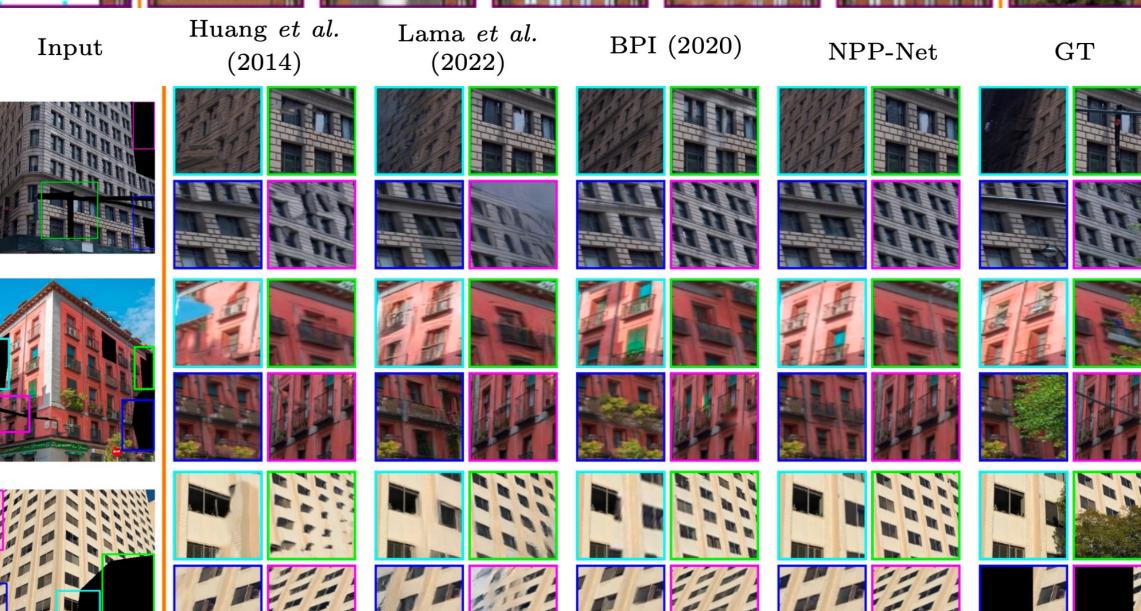
#### **Datasets**

- Subset from NRTDB, DTD, Façade datasets.
- Contain 532 images, larger than those used by existing methods (157 images at most).

## **Completion Result**

Category	Method	NRTDB [1]			DTD [ <u>10</u> ]			Facade [48]		
		LPIPS ↓	SSIM ↑	PSNR ↑	LPIPS ,	↓ SSIM ↑	PSNR ↑	LPIPS ↓	SSIM ↑	PSNR ↑
Large Datasets	PEN-Net [57] ProFill [58] Lama [46]	0.497 $0.401$ $0.196$	$0.452 \\ 0.300 \\ \underline{0.551}$	17.97 16.35 <u>18.64</u>	0.473 $0.443$ $0.274$	$0.365 \\ 0.249 \\ \underline{0.479}$	15.81 14.30 <u>16.39</u>	0.426 0.374 <b>0.207</b>	0.444 $0.391$ $0.468$	$   \begin{array}{r}     \underline{15.78} \\     14.73 \\     15.24   \end{array} $
Single Image	Image Quilting [11] PatchMatch [2] DIP [50] Siren [44] Huang et al. [16] BPI [23]	0.428 0.263 0.554 0.636 0.287 0.254	0.074 0.542 0.292 0.084 0.410 0.442	13.25 18.14 16.46 14.38 16.99 16.86	0.415 0.361 0.659 0.762 0.302 0.303	0.077 0.383 0.181 0.080 0.320 0.305	12.18 15.47 13.15 13.11 14.88 14.82	0.550 0.369 0.582 0.780 0.387 0.458	0.002 $0.341$ $0.258$ $0.052$ $0.279$ $0.173$	10.28 14.22 15.22 12.00 13.75 12.20
NPP-Net	Top3 + Offsets	0.188	0.679	21.01	0.249	0.504	18.32	0.263	0.485	15.93
Input	put PatchMatch DIP F (2009) (2018)		Iuang $et \ al.$ (2014)		BPI (2020)	NPP-Net		$_{ m GT}$		





#### More extensions and comparisons in the paper.

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