



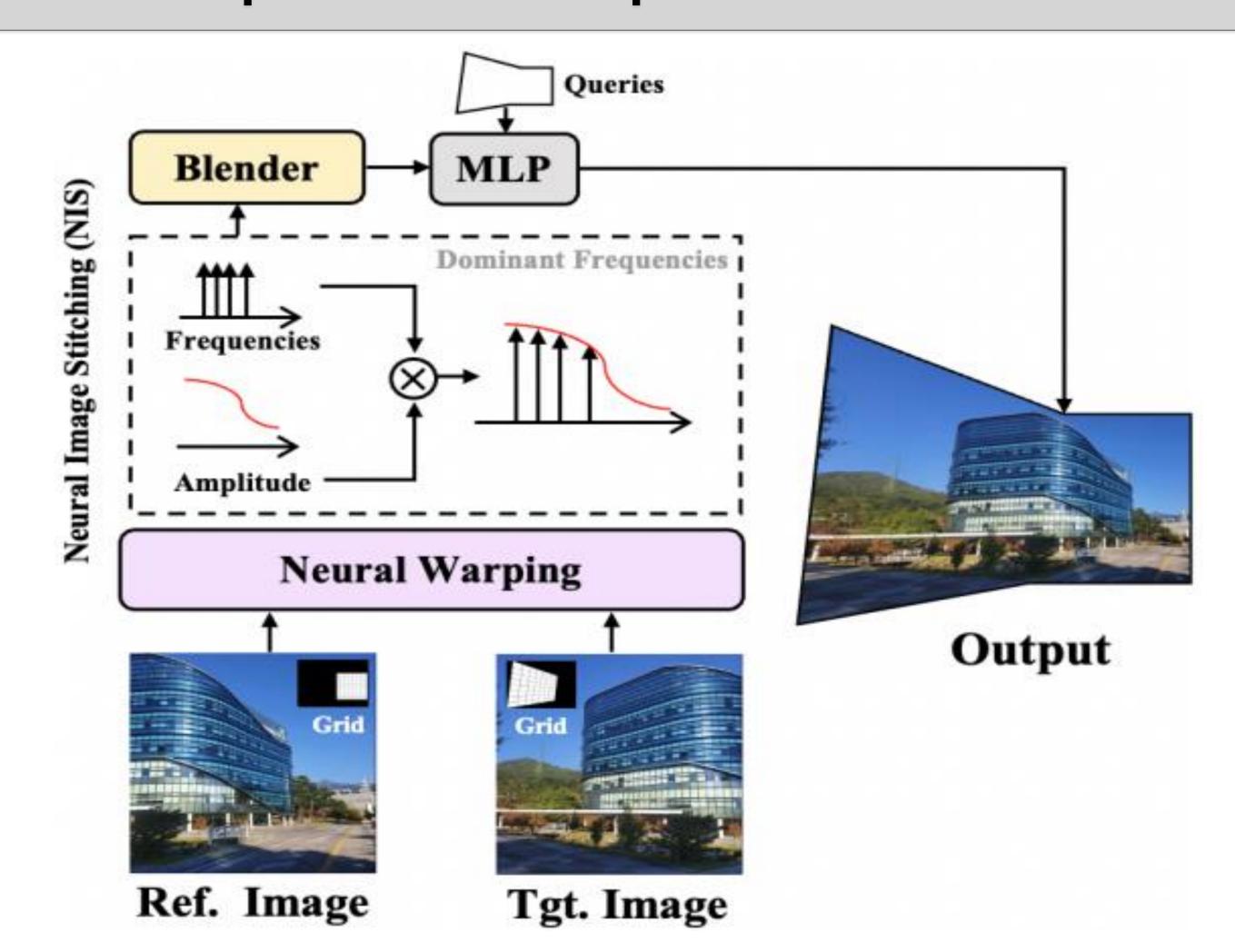
Implicit Neural Image Stitching With Enhanced and Blended Feature Reconstruction



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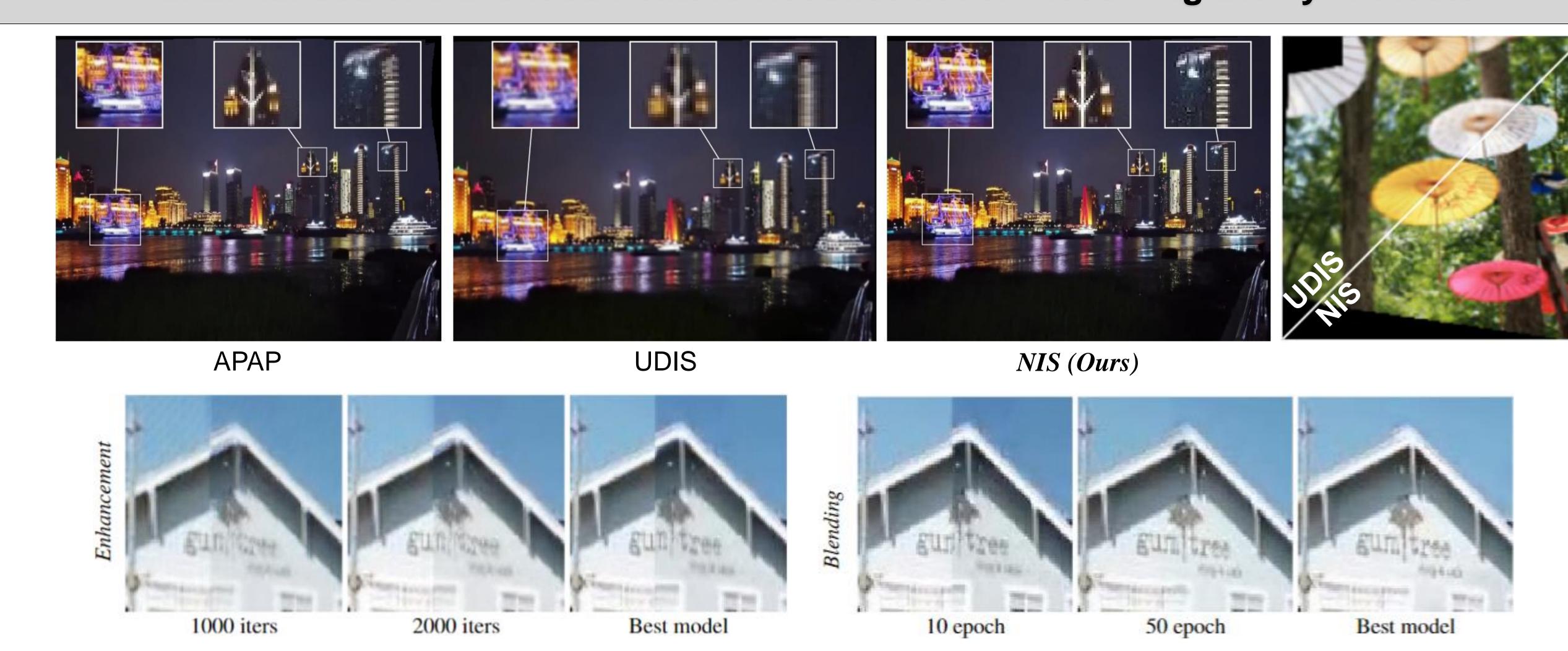
Overview

1. Novel Implicit Neural Representation for Stitching



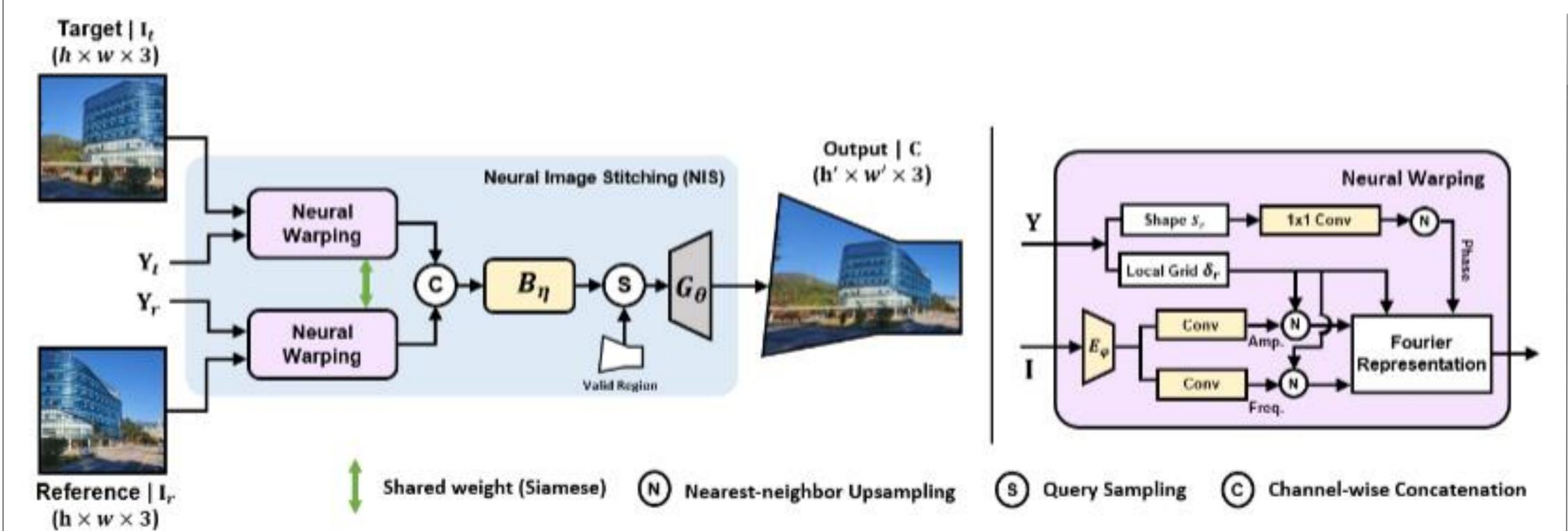
Takes in Aligning Transformation, Corresponding Images, and valid region Queries. Outputs Representation of a stitched image.

2. Enhanced and Blended Feature Reconstruction Resolving Blurry Artifacts



- Top: Resolving severe blurry artifacts of previous learning-based work's stitched feature reconstruction.
- Bottom: Our 2-stage training consisting of <u>learning Enhancement</u> (left) and <u>automatically Blended stitching</u> (right).

Method



1. Estimation of Fourier Coefficients

Amp. = $g_a(\mathbf{z}')$, Freq. = $g_f(\mathbf{z}')$, Phase = $g_p(\mathbf{s}; \mathbf{y})$, $\mathbf{z}' = \mathbf{E}_{\varphi}(\mathbf{I}^{IN})$, $g_{(\cdot)} \coloneqq \text{CNN Coeff. Estimator,}$

 $\mathbf{s} \coloneqq [\text{Jacobian and Hessian of given mesh}] \in \mathbb{R}^{10}$.

2. Capturing Dominant Frequency via warped Fourier Rep.

Fourier Representation: $z[y] = A[y] \cdot (\cos \mathcal{F} + \sin \mathcal{F})$, where $\mathcal{F} = \pi(\langle \mathbf{F}[\mathbf{y}], \mathbf{c}_m \rangle + \mathbf{P}[\mathbf{y}])$, $\mathbf{c}_m = \mathbf{y} - \mathbf{X}[\mathbf{y}]$, $\mathbf{X} \coloneqq \{\mathbf{x} | \mathbf{x} \in \mathbb{R}^2\}$, X[y] := nearest neighbor sampling X with coordinate y.

3. Stitched representation

 $\widehat{\mathbf{C}}[\mathbf{y}_{u}] = G_{\theta}(\mathbf{B}_{\eta}(\mathbf{z}_{t}, \mathbf{z}_{r}), \mathbf{y}_{u});$

 $\mathbf{y_u} \in \mathbf{Y_u} \subset \mathbf{U}$ denotes a coordinate of valid region, $\mathbf{U} \coloneqq [0, h] \times [0, w)$ is a uniform mesh.

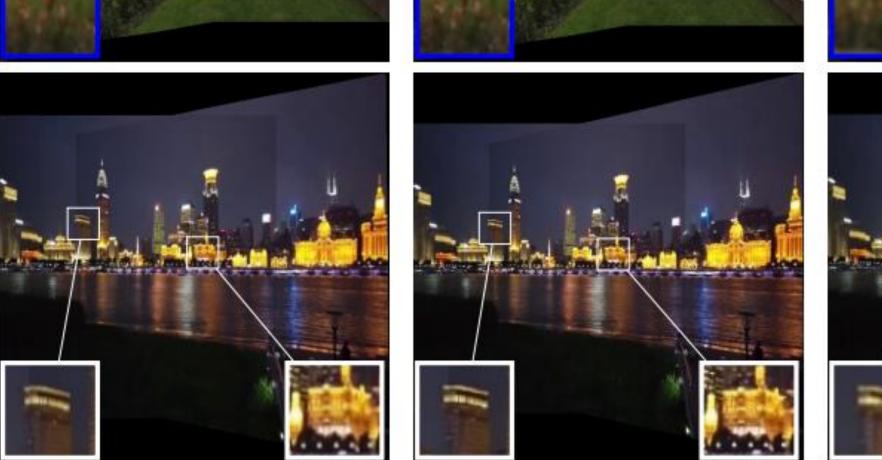
4. Strategy for enhancing and blending

Enhance: Minimizing L_1 loss between GT and \widehat{C} ,

Blending: Seam Loss of UDIS.

Results

Qualitative Comparison





Robust ELA

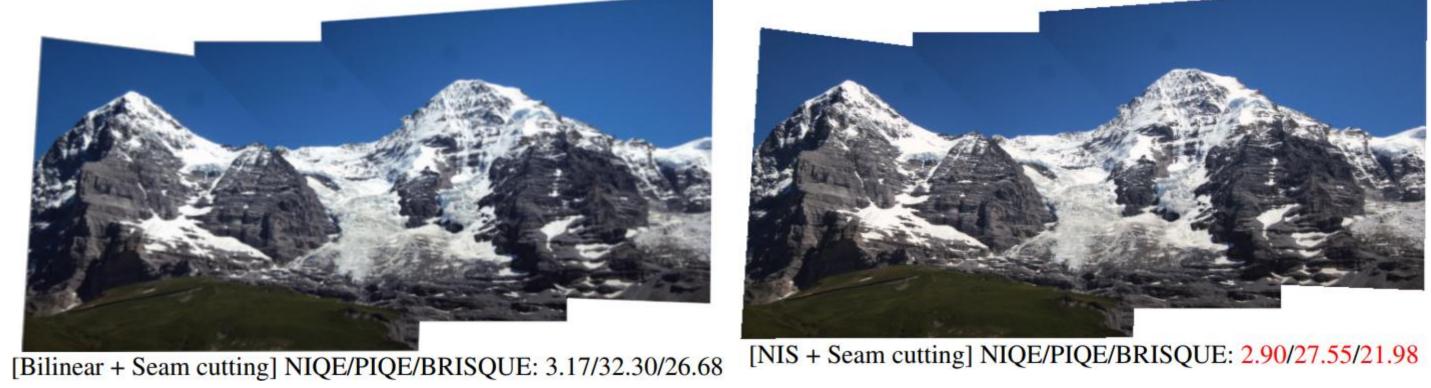
APAP

UDIS

NIS (Ours)

Applying Seam cutting on NIS

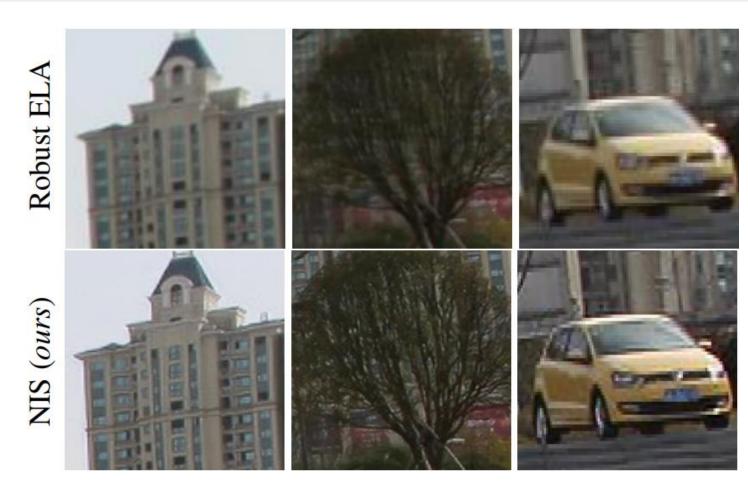




Association of a unified feature: $\hat{\mathbf{C}}[\mathbf{y}_u] = \mathbf{G}_{\theta}(\mathbf{M}_t \cdot \mathbf{z}_t, +\mathbf{M}_r \cdot \mathbf{z}_r, \mathbf{y}_u), \quad \mathbf{M}_{(\cdot)} \coloneqq \text{Seam mask.}$

Discussion

Generalization on elastic warps

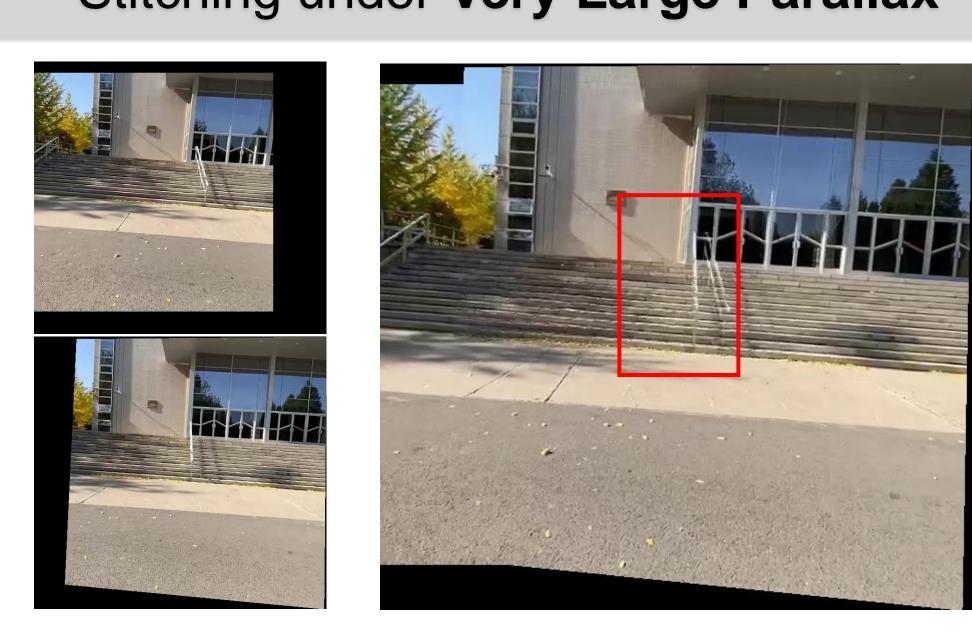


Top: Robust ELA's Stitching w/o NIS, Bottom: Robust ELA w/ NIS.

Quantitative Comparison

Benchmark	UDIS-D			MS-COCO (Synthetic)		Param.
Metric	NIQE	PIQUE	BRISQUE	mPSNR	mSSIM	-
APAP	3.30	46.95	34.72	-	-	-
Robust ELA	3.59	53.67	37.78	-	-	-
LPC	3.37	50.81	37.15	-	-	-
LPC + Graph Cut	3.50	50.63	37.14	-	-	_
Bilinear	-	-	-	34.78	0.96	_
Bicubic	-	-	-	36.25	0.97	-
UDIS	3.43	50.01	36.71	33.45	0.97	8.0 M
NIS (Enhance)	3.28	46.21	33.17	-	-	3.2 M
NIS (Enhance + Blend)	3.15	43.05	31.14	38.69	0.98	3.2 M

Stitching under Very Large Parallax



No trivial method for escaping ghosting artifacts.