

FastJAM: a Fast Joint Alignment Model for Images





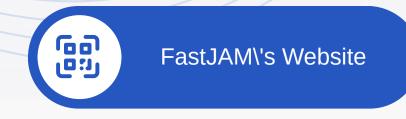


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Shira Ifergane

Oren Freifeld



Presented By:

Omri Hirsch

https://bgu-cs-vil.github.io/FastJAM/



















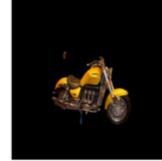
























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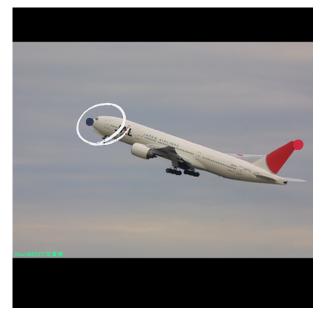
Motivation

Joint Alignment is crucial for many vision tasks

O1 Pairwise = Error Accumulation

Inefficiency of Existing Joint Alignment Methods

JA based on Dense Feature Losses









NEURAL INFORMATION PROCESSING SYSTEMS

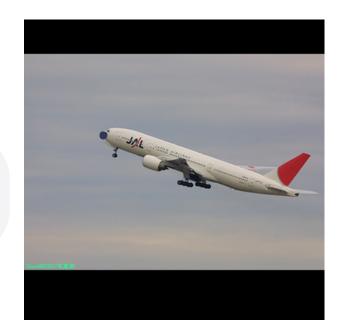
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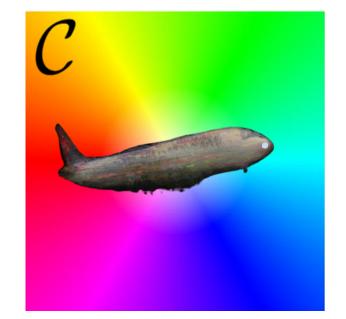
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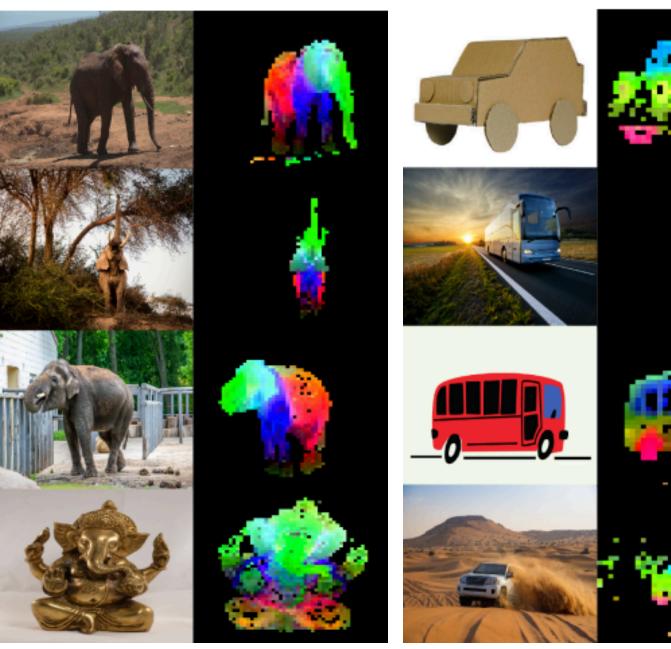
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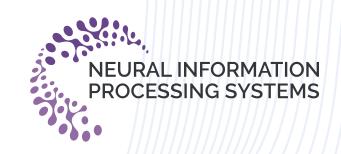
Inefficiency of Existing Joint Alignment Methods

JA Based on Dense Feature Losses



[1] Oquab et al., TMLR 2024





Challenges



Unsupervised: 01

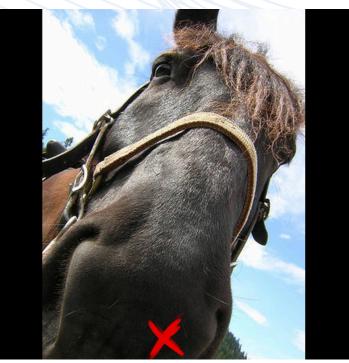
No "Known" Reference



Regularization: 02

Due to #1, Usually Previous Methods Require Regularization









Cross Instance: 03

Cross Instance JA is Even More Difficult

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Challenges





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No "Known" Reference



Regularization: 02

Due to #1, Usually Previous Methods Require Regularization



Cross Instance: 03

Cross Instance JA is Even More Difficult



[1] Barel et al., ECCV 2024

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Challenges



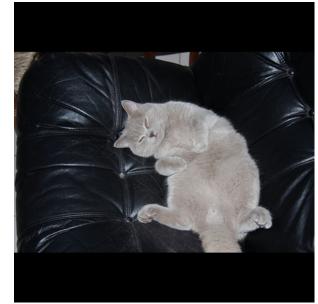
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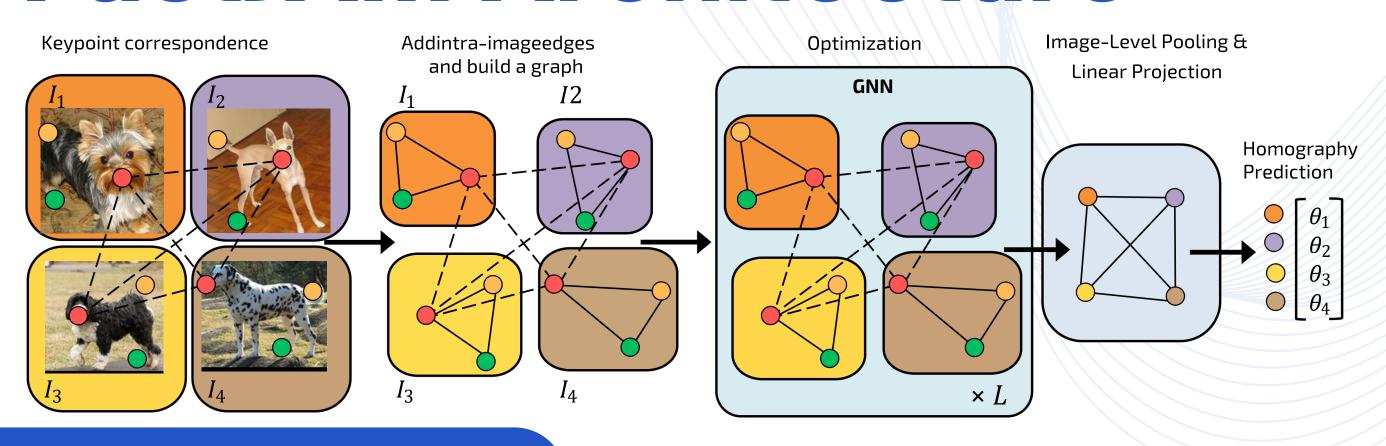
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FastJAM Architecture



INPUT:

Keypoint Coresponidence Graph

OUTPUT:

Homographies For Each Image Legend:

Intra-image edge

Inter-image edge

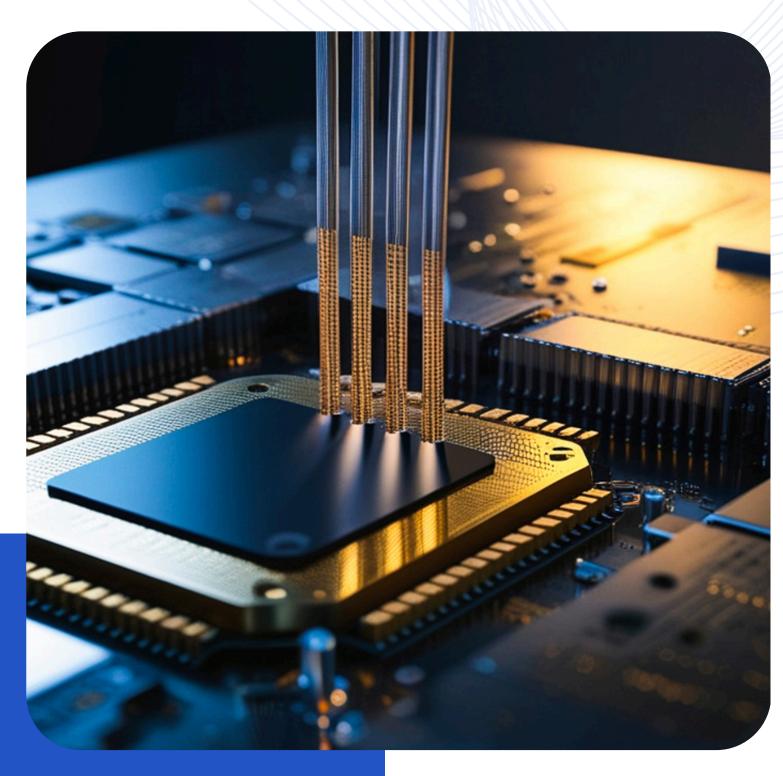
Corresponding Keypoints Between Images

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Runtime Comparison

FastJAM runtime is in Seconds Compared to Minutes or Hours (hh:mm:ss)

Neural Congealing 01:18:30 ± 00:06:18

→ ASIC 01:06:38 ± 00:00:38

→ SpaceJAM 00:06:00 ± 00:00:12

> FastJAM (Ours) 00:00:49 ± 00:00:04

*For the Full Table Please Check Our Paper

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Quantitative Results

FastJAM gets either better or comparable results.

Based on PCK		SPair-71K Avg. All Classes	CUB-200 Subsets
→	Neural Congealing	_	63.6
→	ASIC	37.0	_
→	SpaceJAM	<u>45.7</u>	<u>69.9</u>
\rightarrow	FastJAM (Ours)	53.4	73.6

^{*}For the Full Table Please Check Our Paper

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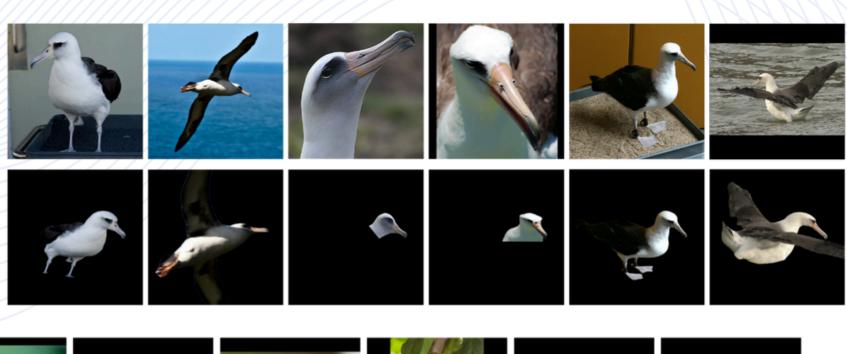
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Qualitative Results



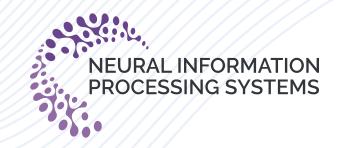




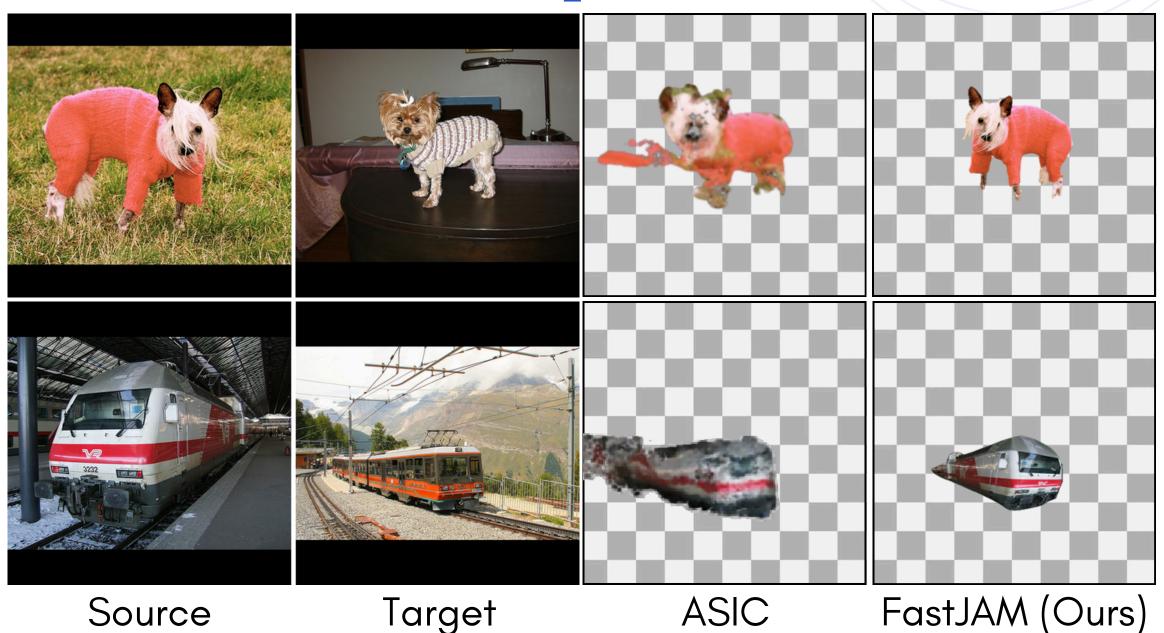
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Qualitative Comparison

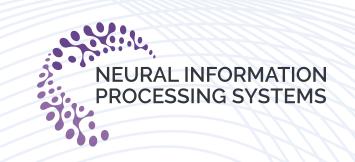


*For More Results
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Qualitative Comparison

Sheep

SpaceJAM

FastJAM (Ours)

















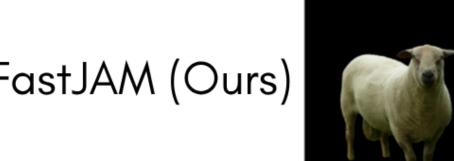














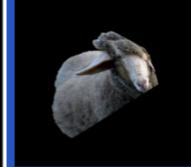
















Thank You.

Thank you for your attention.

I welcome any questions or feedback.

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