

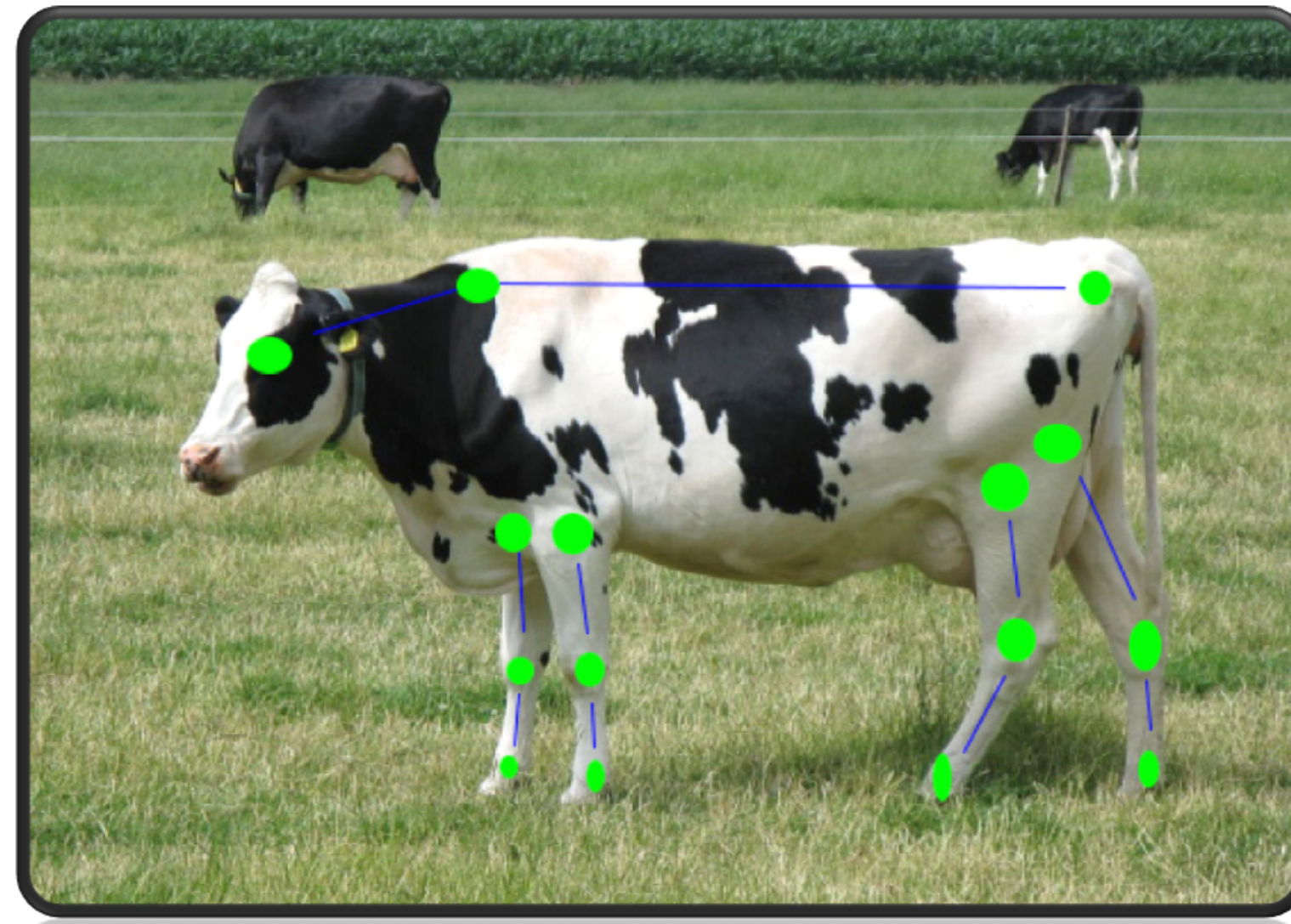


Unsupervised Landmark Discovery via Self-Training

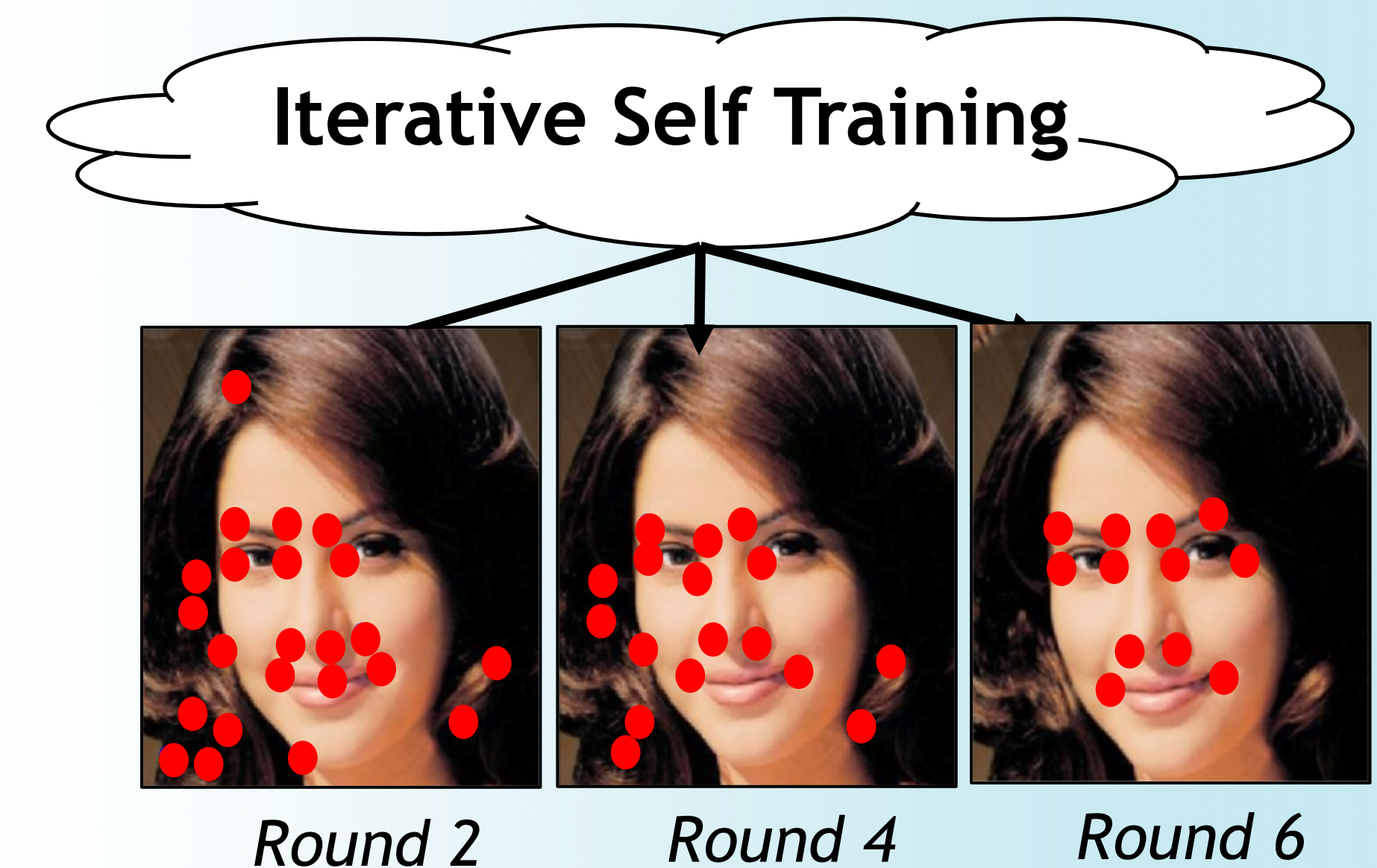
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Motivation

- Landmark localization can be used for accurate animal monitoring
- Landmarks capture the shape (limb configuration) of the animal
- Essential building block for most vision applications
- Strong landmark detectors are only available on the human-facing domain due to lack of annotated data

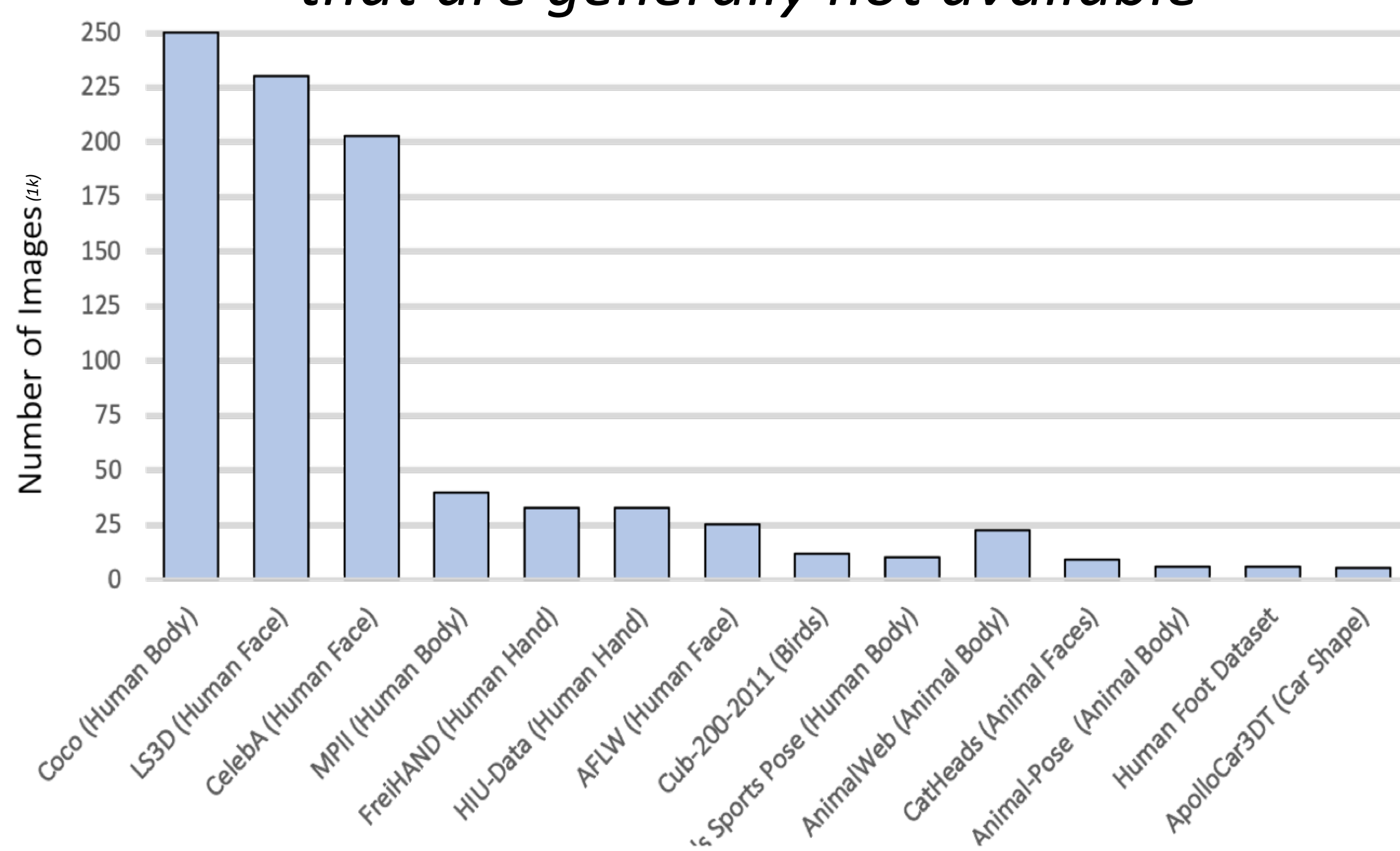


Core Idea



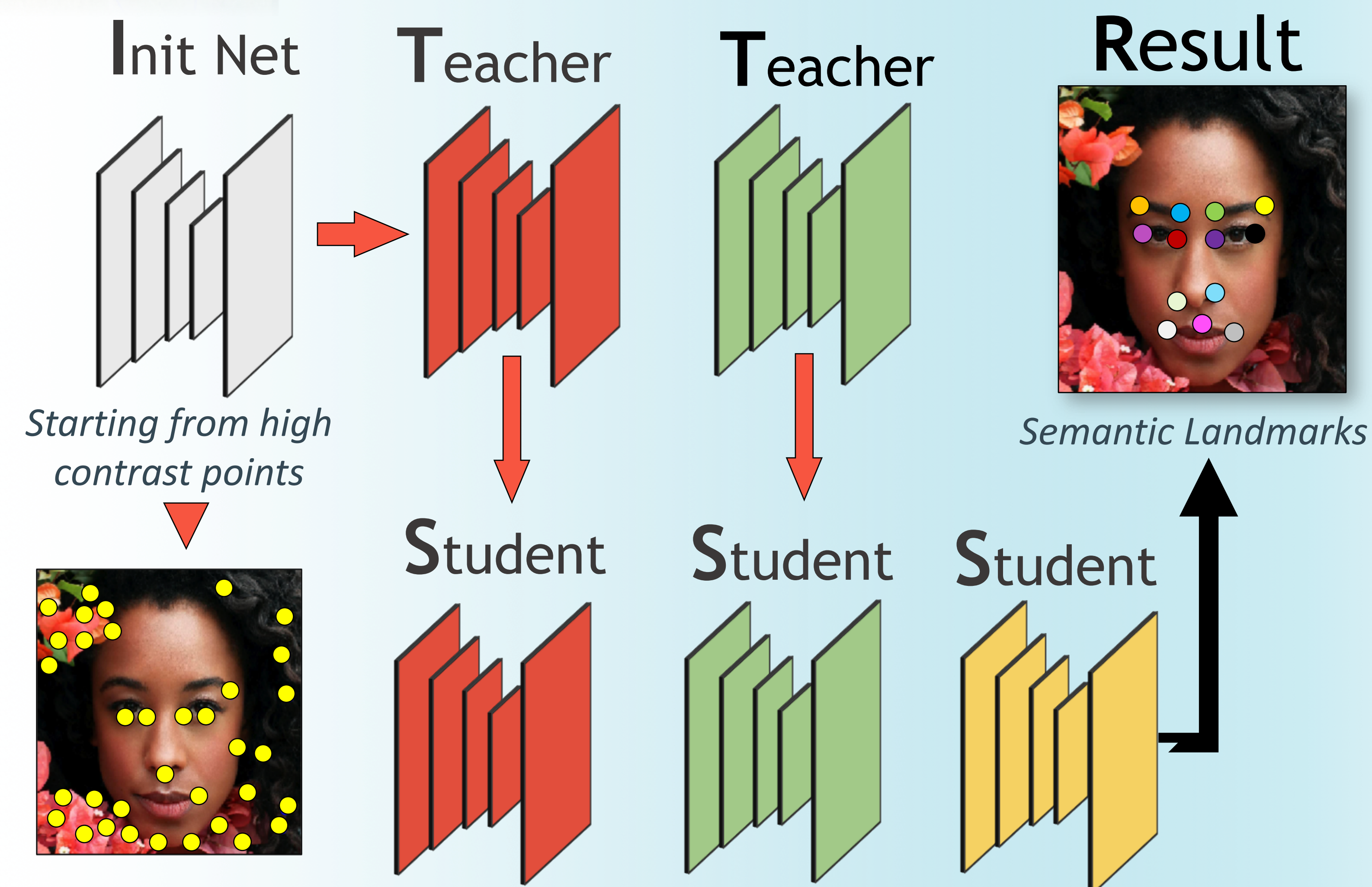
Research Challenge

Existing methods require vast amounts of labeled data that are generally not available



We are exploring the task of Unsupervised Landmark Detection from 2D unlabeled images.

Proposed Architecture



Quantitative Results

PennAction Raw Landmark Accuracy (%)

Method	Head	Shldrs	Elbws	Hands	Waist	Knees	Legs
Jakab	6.36	9.23	7.85	0.59	22.27	17.85	6.48
Ours	74.27	57.91	33.00	8.36	64.81	69.54	75.84

CatHeads Forward-NME (%)

Thewlis	Zhang	Lorenz	Ours
26.94	14.84	9.30	9.31

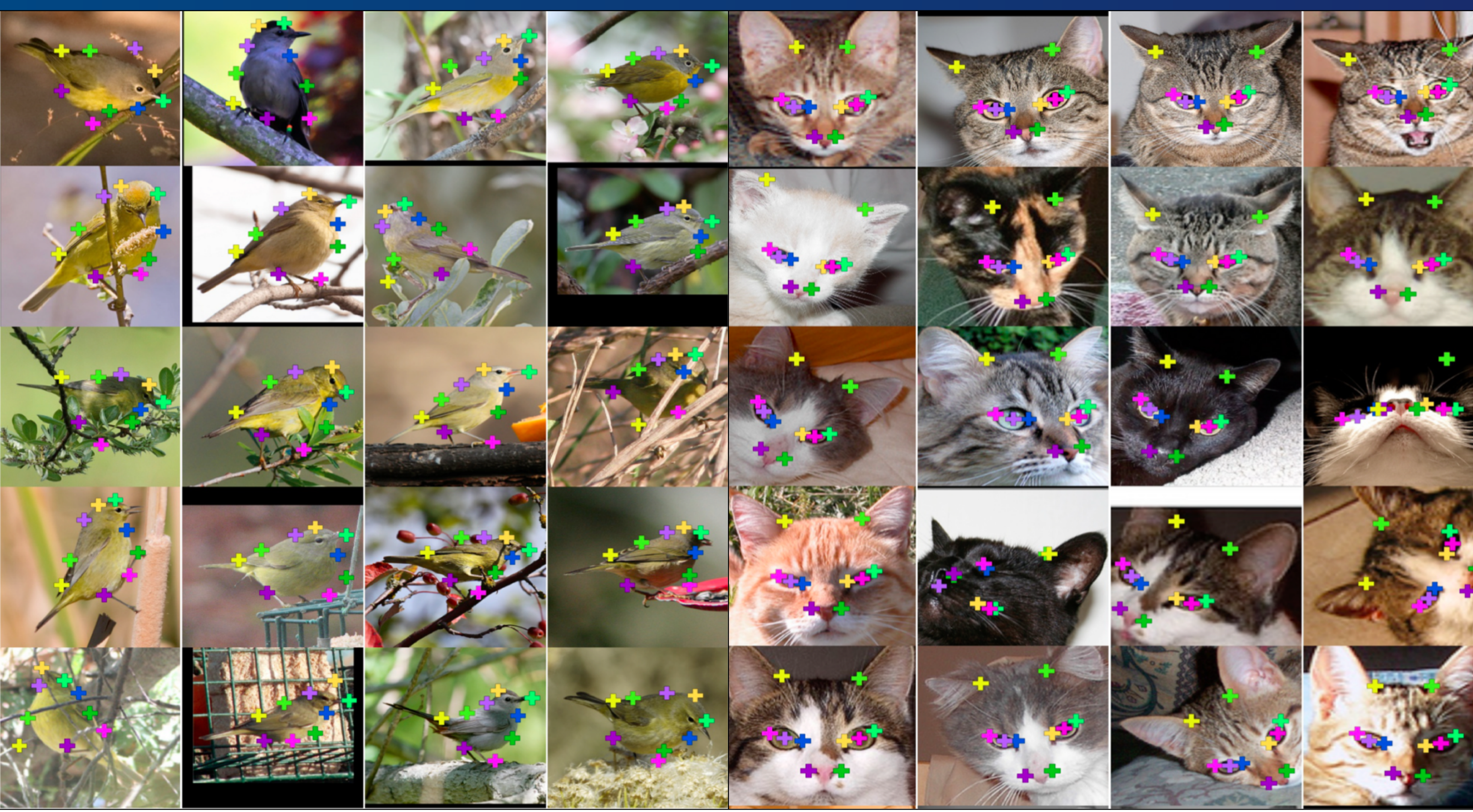
Conclusions

- Our method can detect meaningful landmarks without need for manual annotations.
- Results show that our method can perform better than other recently proposed methods.
- Can be applied without modification on arbitrary object categories

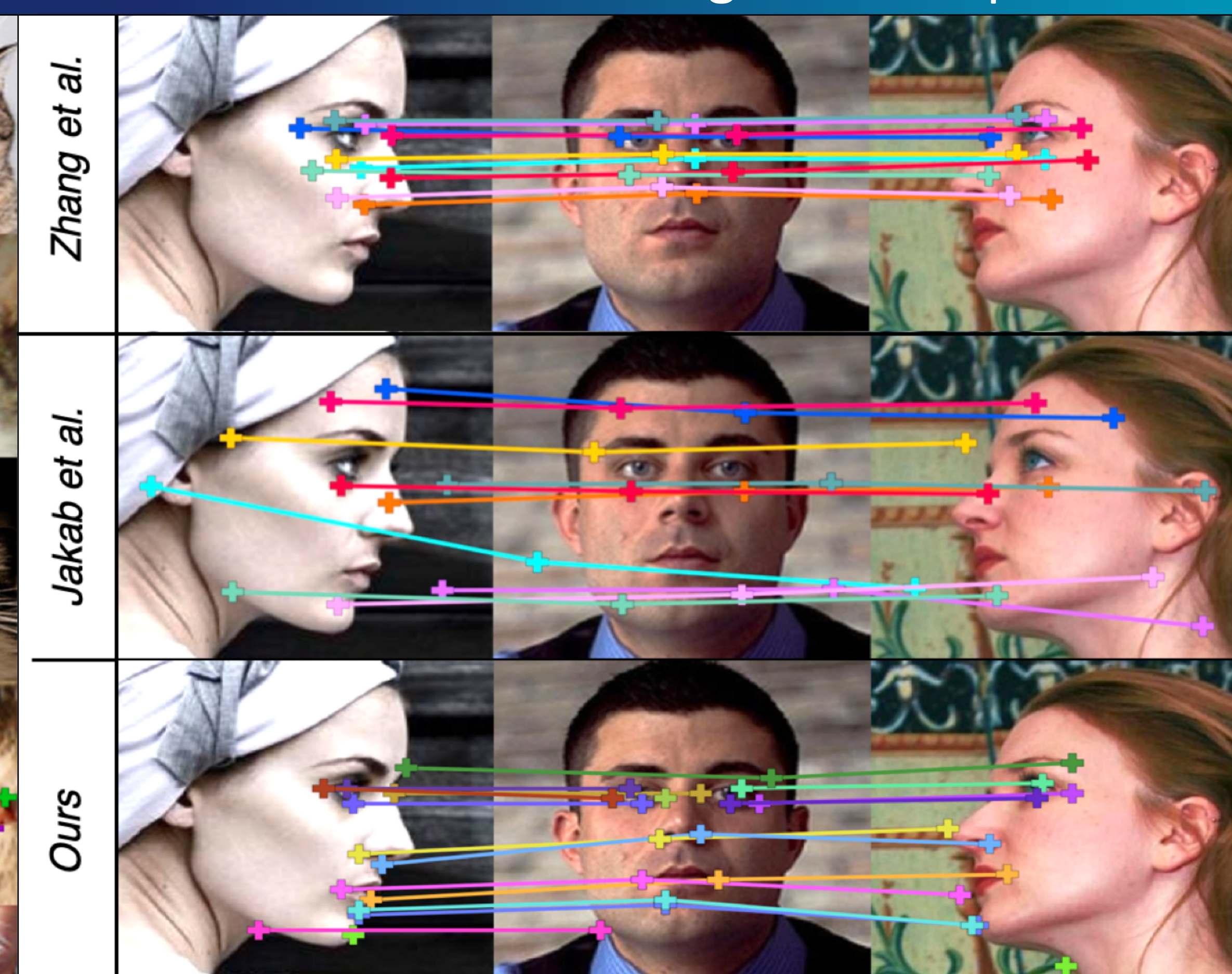
References

- Jakab et al. Unsupervised learning of object landmarks through conditional image generation. *NeurIPS* 2018
- Zhang et al. Unsupervised discovery of object landmarks as structural representations. *CVPR* 2018
- Thewlis et al. Unsupervised Learning of object landmarks by Factorized Spatial embeddings, *ICCV* 2017
- Lorenz et al. Unsupervised Part-Based Disentangling of Object Shape and Appearance. *CVPR* 2019

Qualitative Results on Animal datasets



Robustness to Changes in Viewpoint



Our approach can learn landmarks that are more flexible in terms of capturing large changes in viewpoint.