

# NeRD++: Improved 3D-mirror symmetry learning from a single image

Task: detecting mirror symmetry planes from single-view images  
- Input: a single RGBA image  
- Output: perspective mirror symmetry plane ( $w_x + 1 = 0$ )

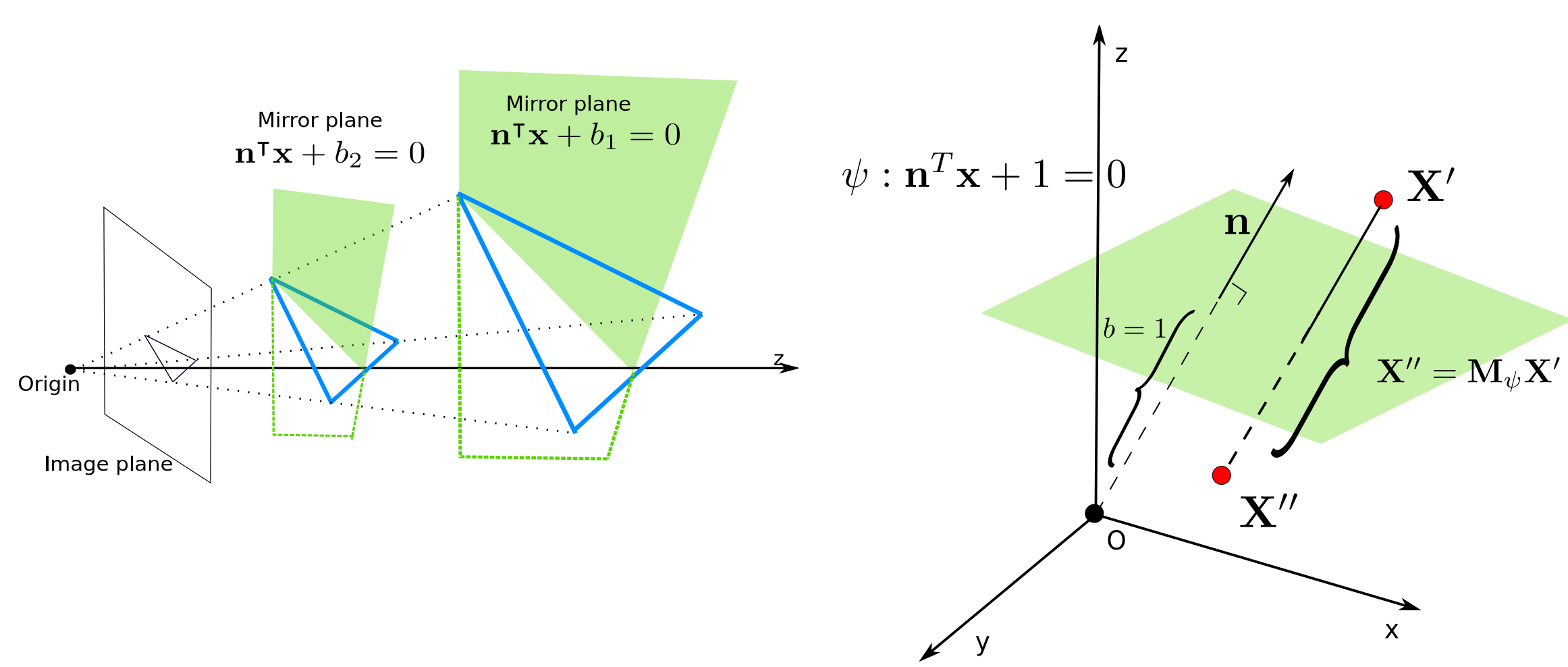
Mirror symmetries from single-view images:



## (I) Contributions:

- Improved data-efficiency over sota NeRD [1]  
Solution space is a hemisphere - spherical convs for localization
- Improved compute efficiency (20x faster):  
Feature correlations indicate similarity between pixels and mirrors.

## (II) Background knowledge



- The bias (scale) can not be determined from an image.
- Mirrors can be mathematically calculated [2].

## (III) NeRD Recap

Given a candidate plane

For each point (pixel):

Calculate its mirrors over depth;

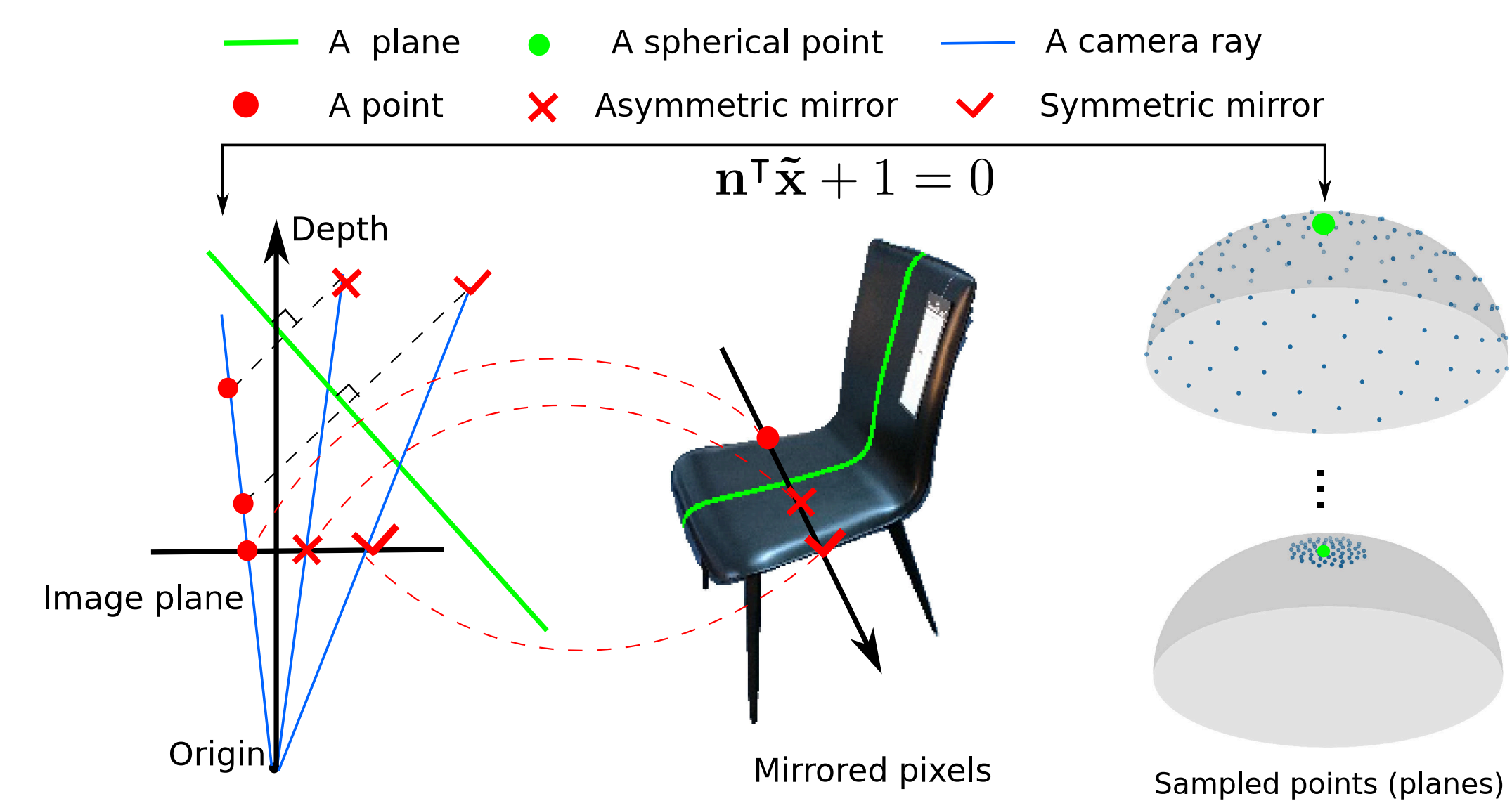
Concatenate the feats;

Construct an expensive 4D feature volumes  $[2C, D, H, W]$ ;

Apply learning and Classification (how likely the given plane is a symmetry plane);

Re-sample nearby points from the hemisphere

Iterate over 3 times (coarse-to-fine) to reach a desired resolution.



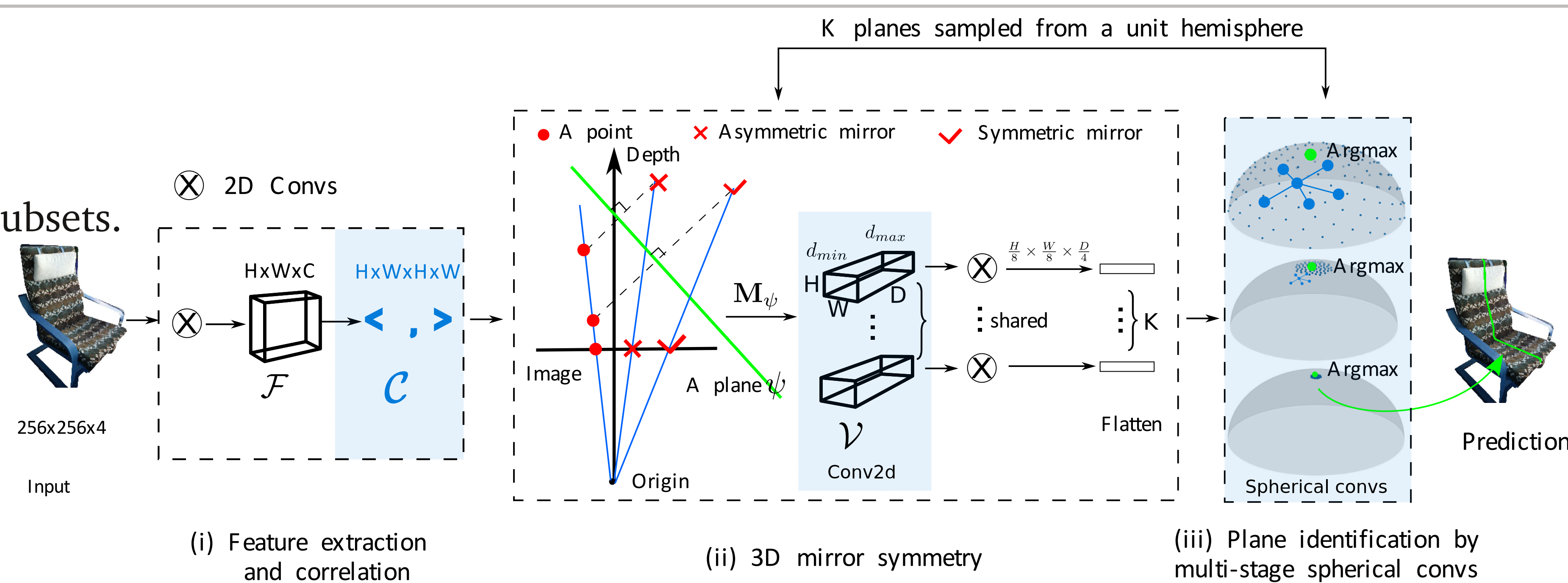
## (III) NeRD++

However,

- computationally expensive: 4D feature volumes;
- data-hungry: massive training data/inferior results on subsets.

We propose:

- calculate feature correlations explicitly,  
Correlation indicates similarity ( $[D, H, W]$ );
- use another inductive prior:  
Solution space is a hemisphere, spherical convs.



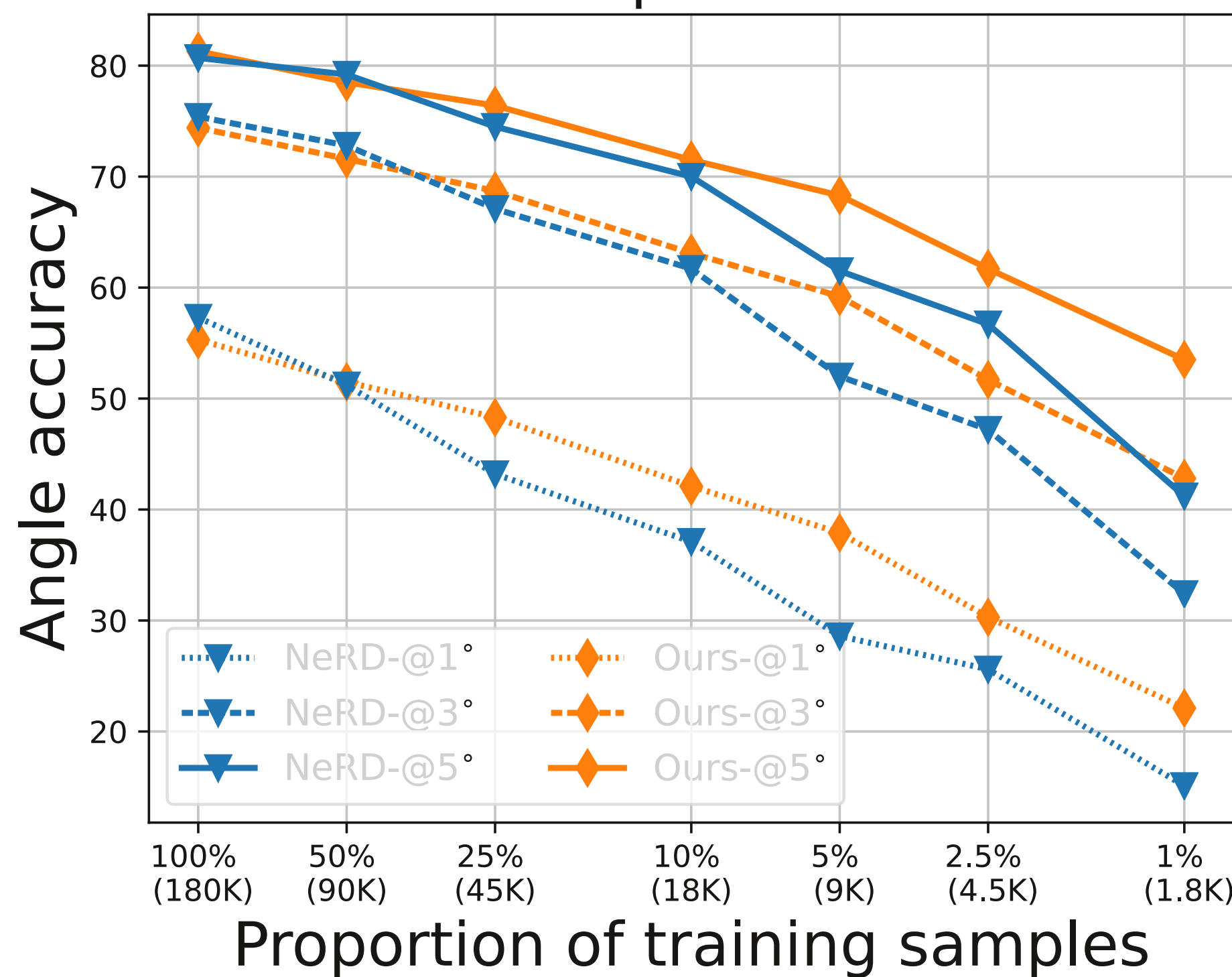
## (IV) Experiments

Two datasets: ShapeNet[3], Pix3D[4]

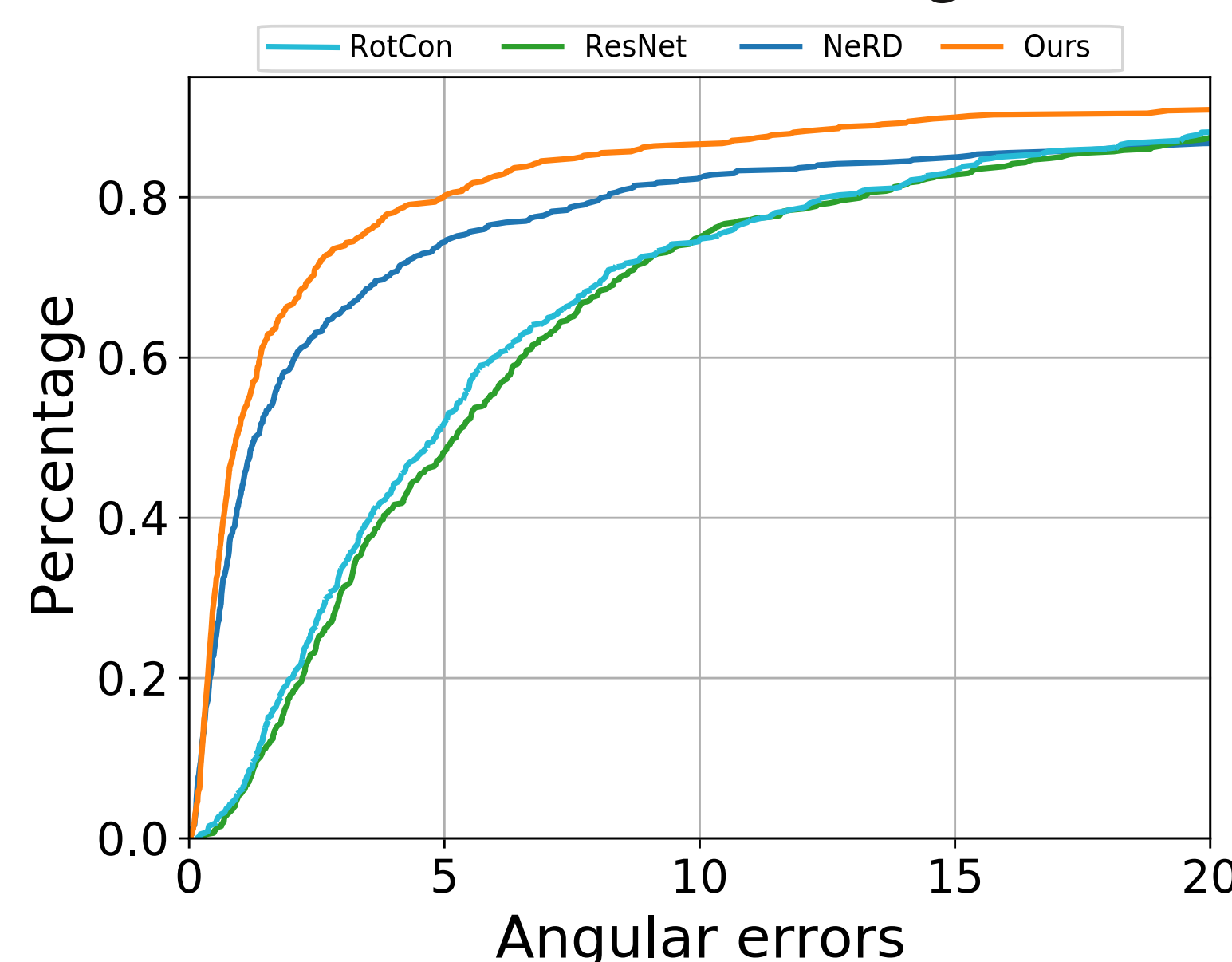
Evaluation: angular errors in the camera space (Angular Accuracy).

### 1. Data-efficiency on ShapeNet

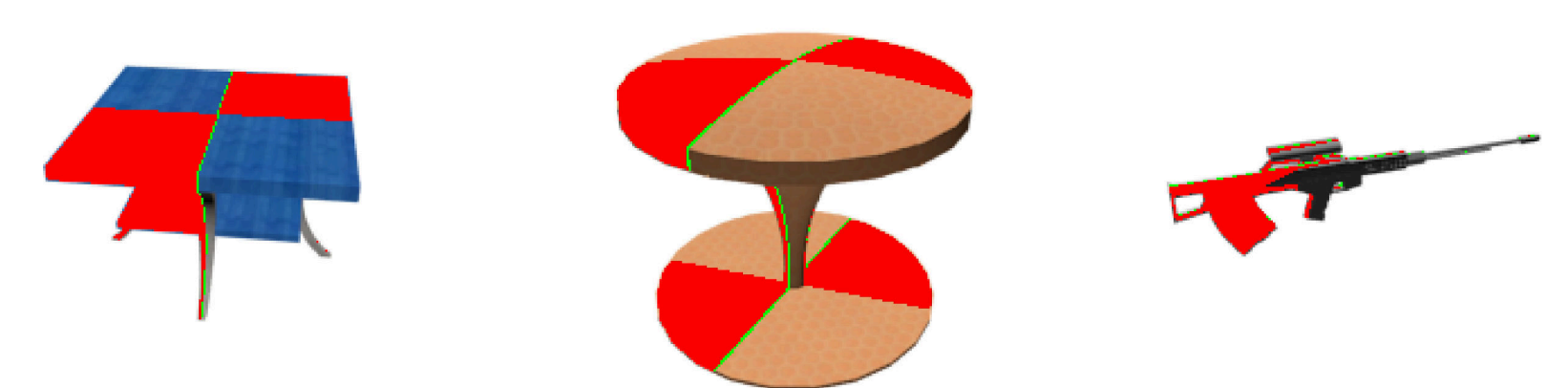
AA on ShapeNet subsets



### 2. Pix3D (~5K images)



### 3. Failure cases



## (V) Limitations

- Rely on entirely on appearance features (textureless surface/repetitive patterns);
- Iterate over all possible depth;
- Detect only the dominant symmetry plane.

## (V) Future work

- How to handle occlusion explicitly?
- Downstream tasks using symmetries (depth, shape completion, 3D reconstruction).
- Detect intrinsic symmetries (of non-rigid objects).

## Reference

- Y Zhou et al, NeRD: Neural 3D Reflection Symmetry Detector, CVPR 2021
- D Cailliere et al, 3d mirror symmetry detection using hough transform, ICIP 2008.
- A Chang et al, Shapenet: An information-rich 3d model repository, CVPR 2015.
- X Sun et al, Pix3d: Dataset and methods for single-image 3d shape modeling, CVPR 2018.
- Y Zhou et al, On the continuity of rotation representations in neural network, CVPR 2019