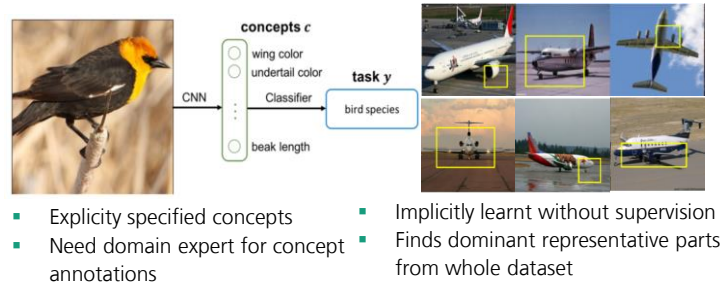


## Concept-based vs Prototype-based Learning



Can AI models be reliably explained using human-interpretable concepts?  
 Can we still learn human-interpretable concepts without requiring concept annotations?  
 We need **prototype-based learning methods** which do not need concept annotations, are interpretable-by-design and global.

**Motivation:**

- How interpretable are these prototypes towards assessment of Image Classification models?
- What are the conditions these prototypes should fulfill towards a truly human-interpretable model?

**Key Contributions:**

- Proposed a **Desiderata** for truly human-interpretable prototypes
- Designed a common setup to evaluate the existing methods (ProtoPNet, ProtoTree, PRP) in the light of these properties with real (CUB, IM30) and synthetic (3D Shapes) datasets
- Validated our findings quantitatively by conducting a user-study
- Demonstrated the application of prototype-based learning on real-world use-case of *OOD Detection*

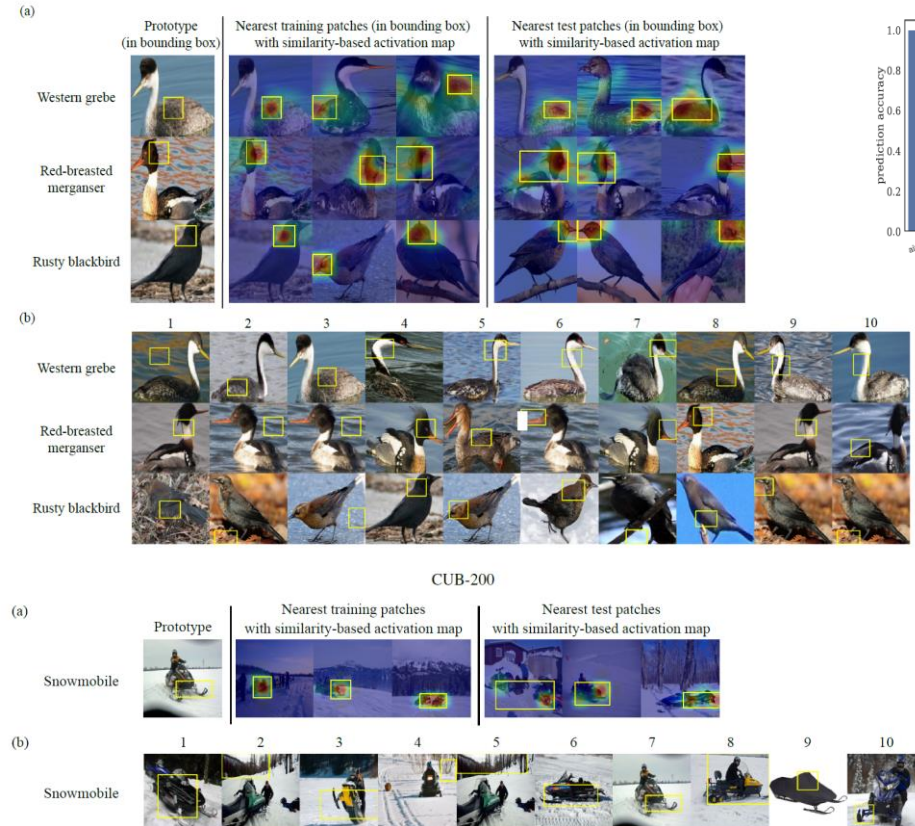
**Desiderata for the prototypes to be truly human-interpretable:**

- Human-understandable/ interpretable
- Semantically disentangled
- Semantically transformation invariant
- Relevant to the learnt task



Full Paper

## Qualitative Results (Real-world Dataset)

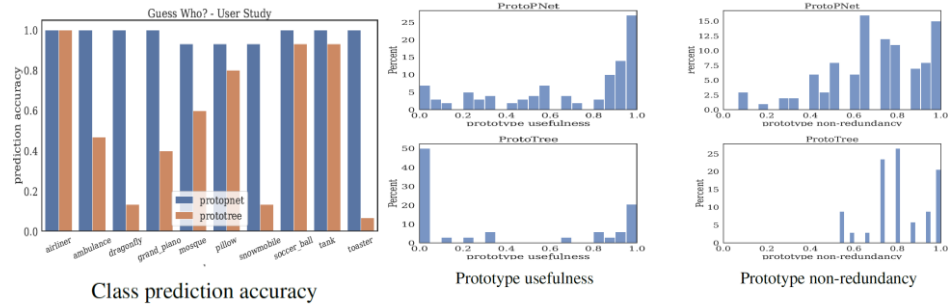


Results for ProtoPNet using CUB-200 (top) and ImageNet-30 (bottom): (a) shows for a prototype from a given class - the nearest training and test patches (yellow box) with similarity score based activation maps, (b) shows all the 10 prototypes (yellow box) learnt for the respective classes in (a).



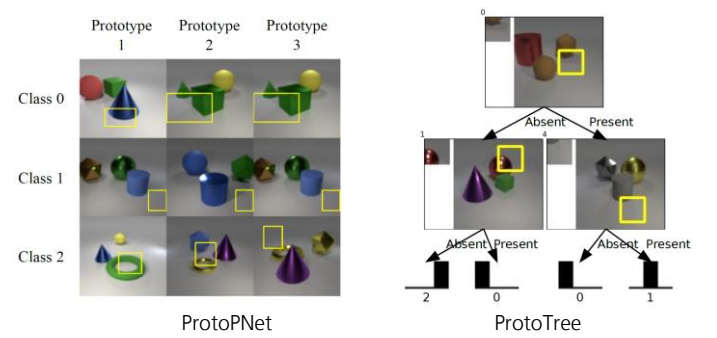
Results from ProtoTree on ImageNet-30: For each test image, corresponding path taken in the decision tree towards final prediction is shown. The node and absent prototypes are shown in yellow and red.

## Quantitative Results (User Study)



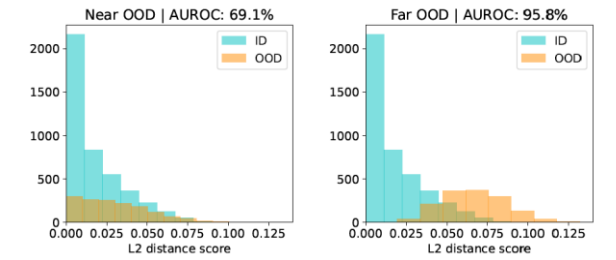
Quantitative results from user study conducted on prototypes from ImageNet-30 classes

## Qualitative Results (Synthetic Dataset)



Results using synthetic 3D-Shapes dataset (V1) showing in (a) prototypes from each class using ProtoPNet and (b) the node prototypes along with decision tree (depth=2) using ProtoTree. Yellow boxes show the prototypes.

## Real-world application (OOD Detection)



Histogram showing distribution of L2 distances to closest prototypes for Near vs Far OOD samples for a model trained on first 150 classes of CUB-200 dataset.