



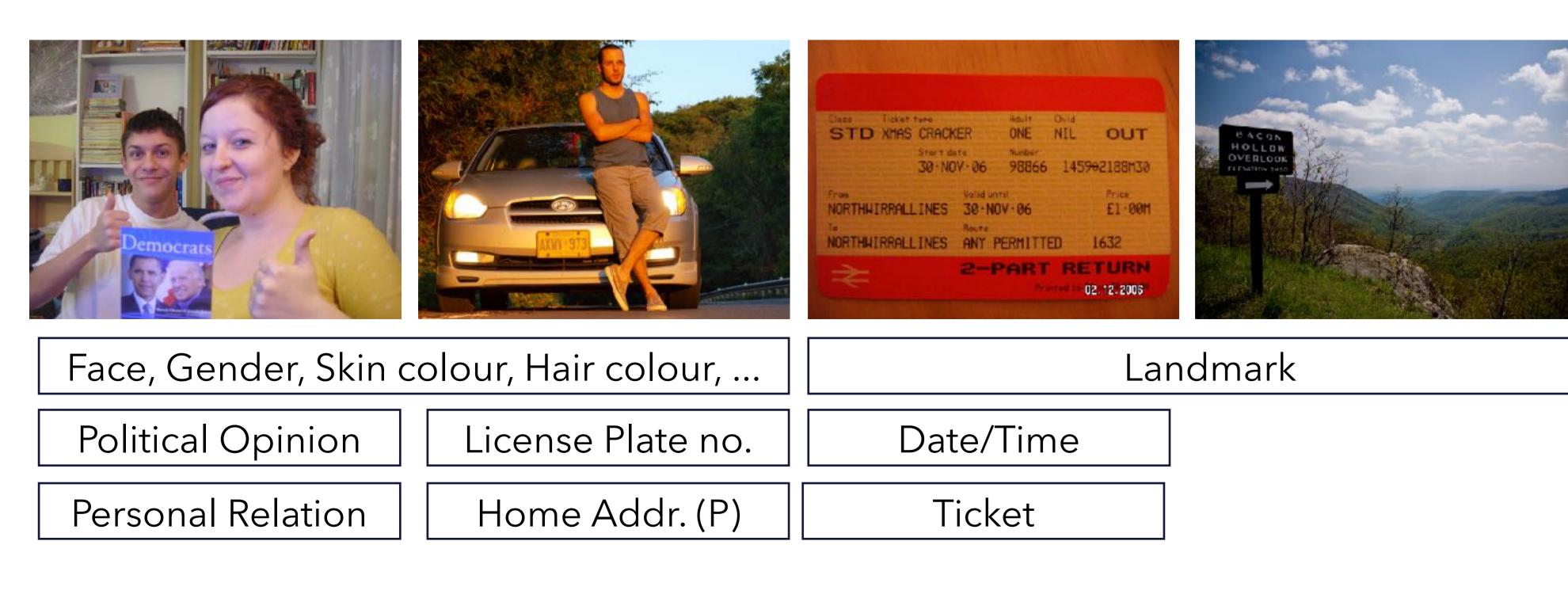
Motivation



- Images contain a broad range of private information
- Users unintentionally expose such information when sharing images online (e.g, Twitter, Flickr, Facebook)
- Can we extend the concept of "privacy settings" to visual content?

The Visual Privacy (VISPR) Dataset

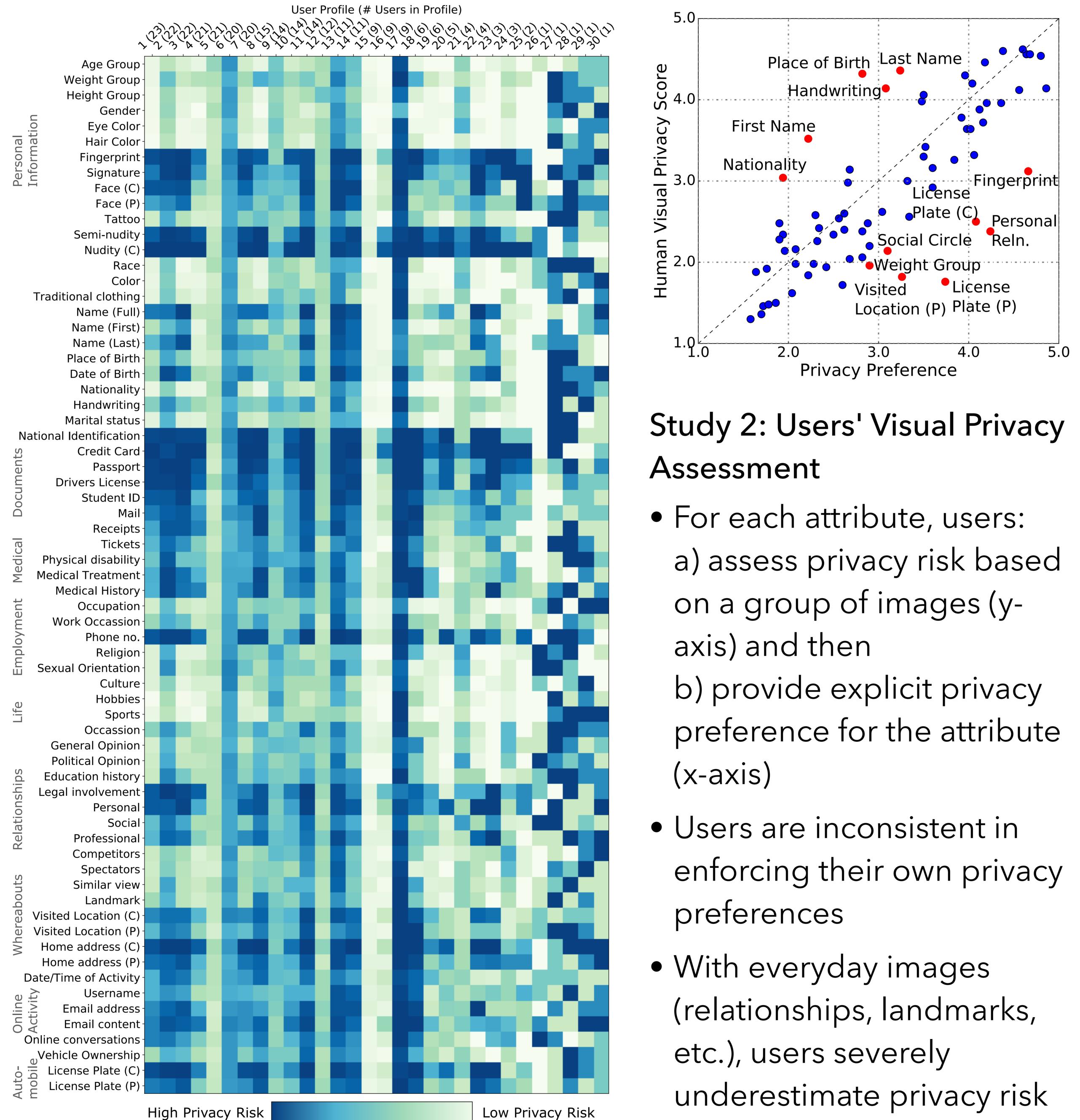
- ~22k publicly available Flickr images
- Challenges from natural everyday scenes: foreground and background clutter, text content
- 68 Privacy Attributes across 9 groups
- Attributes gathered from: EU Data Protection Directive, US Privacy Act of 1974, Social network website rules
- ~116k assigned attributes, 5.22 per image



Towards a Visual Privacy Advisor: **Understanding and Predicting Privacy Risks in Images**

Max Planck Institute for Informatics, Saarbrücken, Germany







- Diverse Preferences \Rightarrow Same image, different privacy risks
- Some users especially sensitive to some attributes

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Approach

Privacy Attribute Prediction

- User-independent multilabel classification task: <sup>^w/₄
 </sup> given an image, predict multiple attributes
- We compare multiple baseline methods for this task
- ResNet-based model achieves an MAP of 47.45 %

Personalizing Privacy Risk

Let Privacy Risk = max_a (privacy rating of attribute a in image)

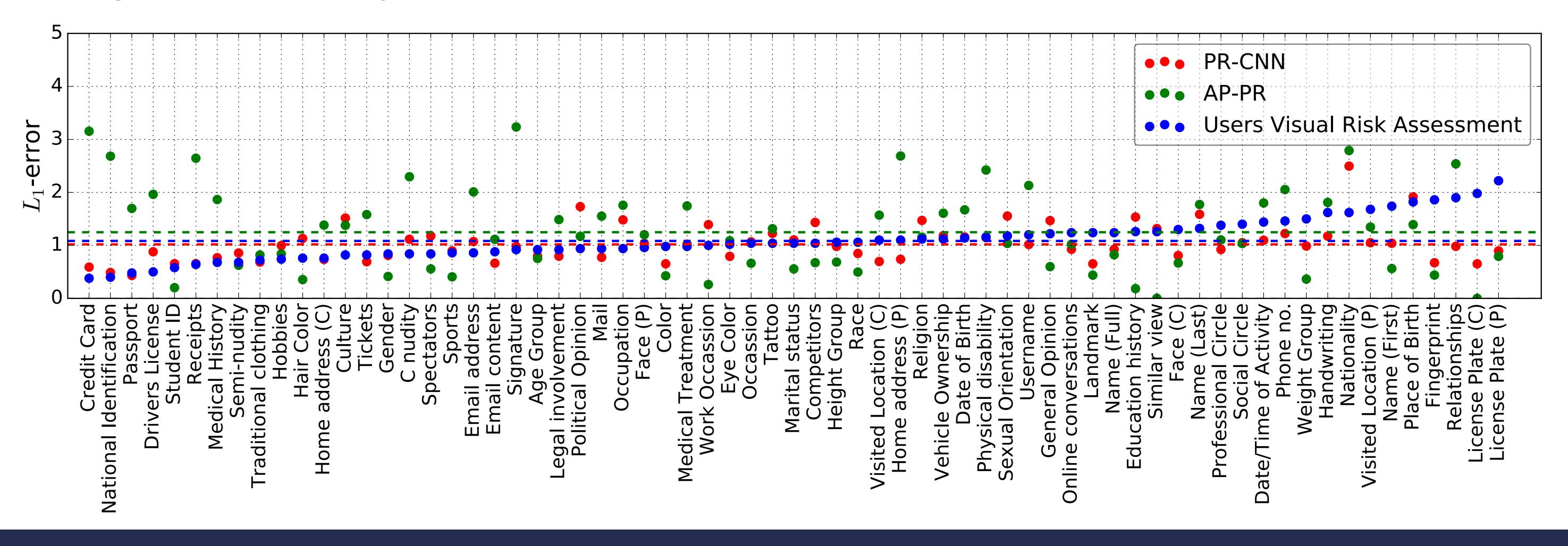
Two proposed approaches:

1. AP-PR: Directly uses attribute predictions (AP) to estimate privacy risk (PR) 2. PR-CNN: End-to-end learning to predict

user-specific privacy risk from images. This is better at handling noisy attribute predictions

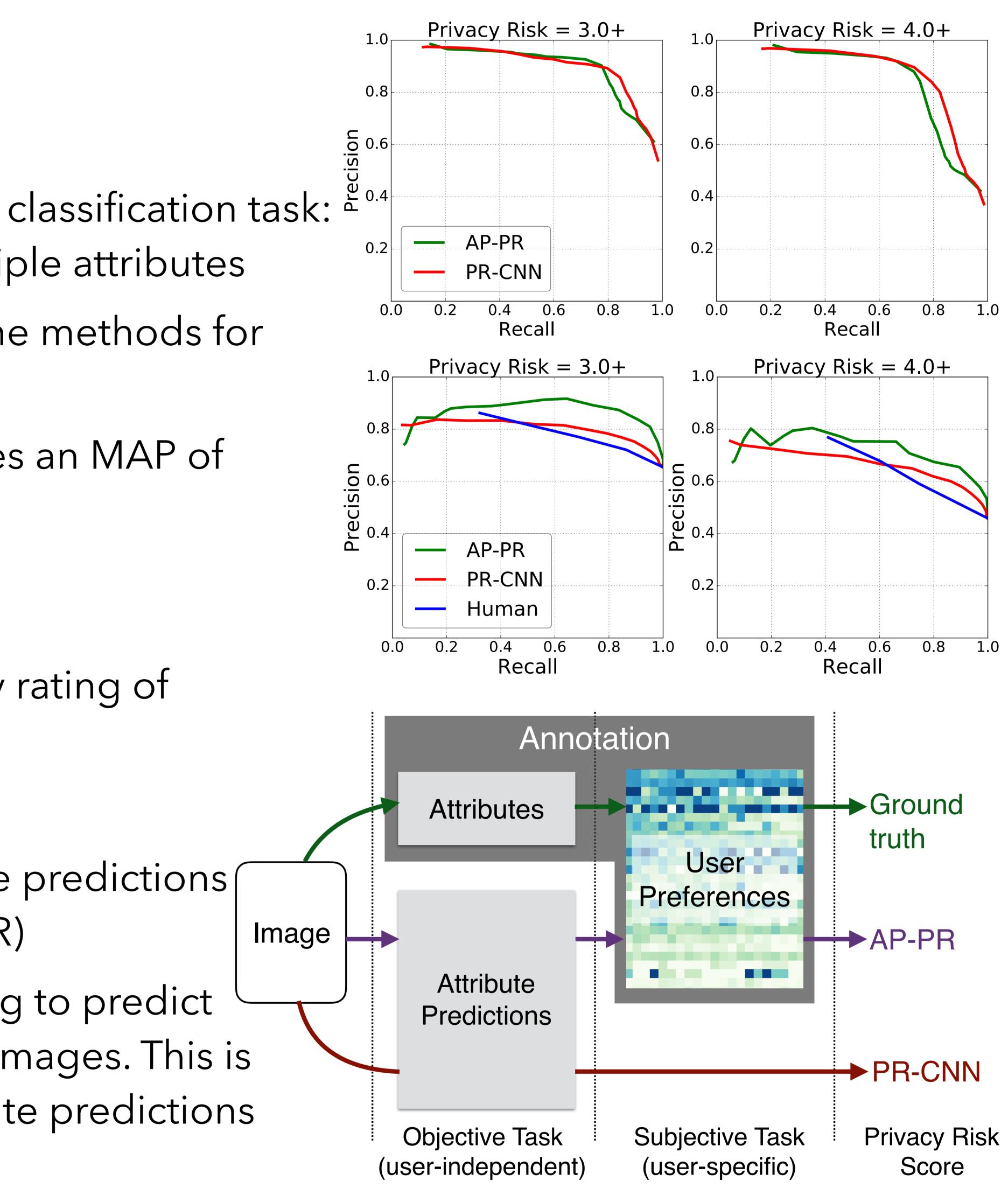
Humans vs. Machine

- images, when compared to users themselves









• We compare these approaches to users' visual privacy assessment (Study 2) • Our approach achieves better Precision-Recall and L1 scores for the same