

## 1. Overview

### Problem

Face Attributes Prediction in the Wild



Arched Eyebrows? Receding Hairline?  
Big Eyes? Mustache?

### Performance

	CelebA	LFWA
FaceTracer [ECCV08]	81%	74%
PANDA-W [CVPR14]	79%	71%
PANDA-I [CVPR14]	85%	81%
SC+ANet	83%	76%
LNets+ANet(w/o)	83%	79%
<b>LNets+ANet</b>	<b>87%</b>	<b>84%</b>

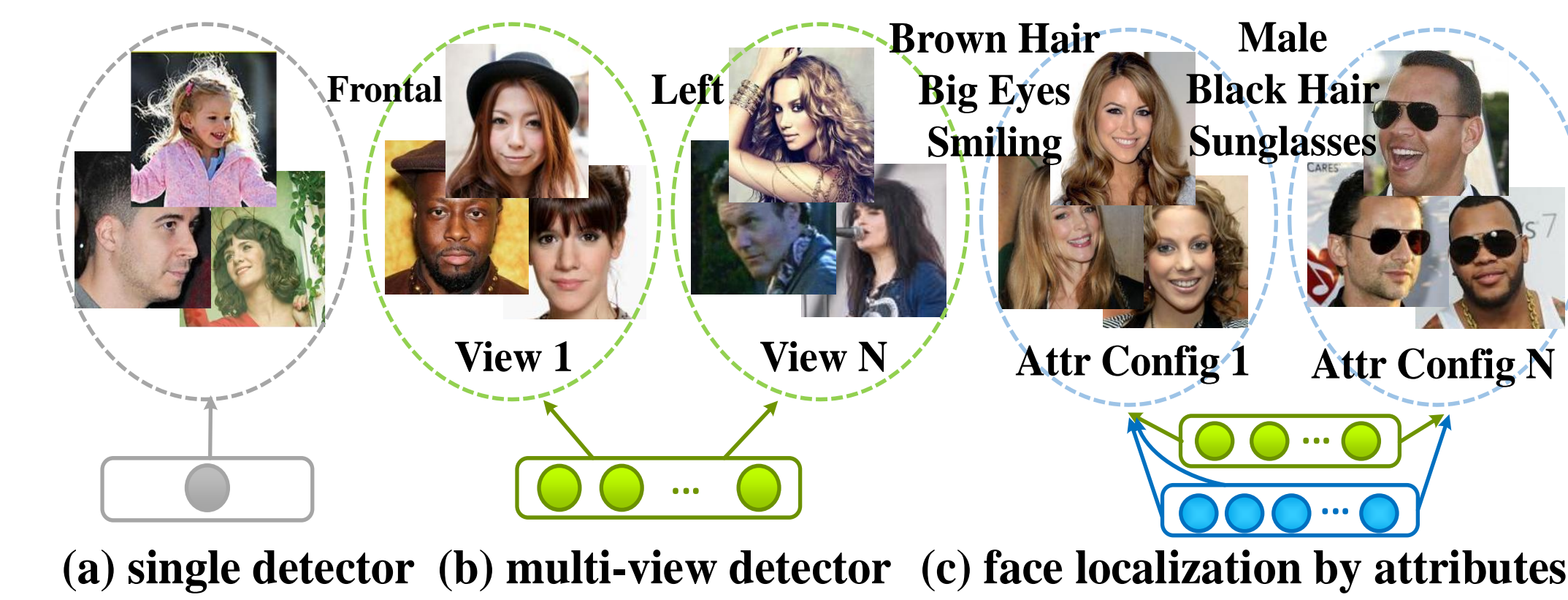
### Running Time

LNets: 35ms, ANet: 14ms

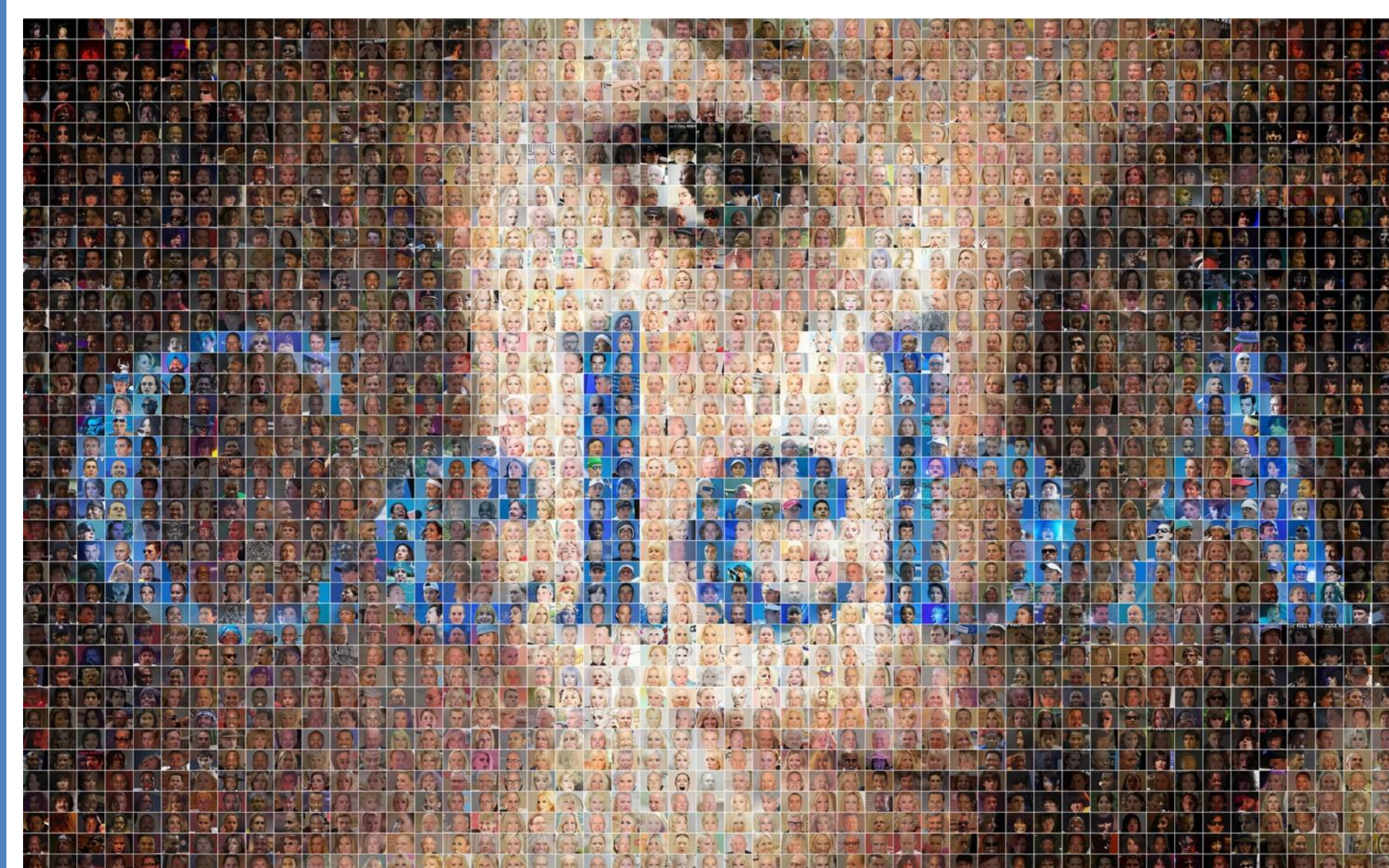
- Project Page: <http://personal.ie.cuhk.edu.hk/~lz013/projects/FaceAttributes.html>

## 2. Motivation

- Existing methods: global and local methods
- Global methods: not robust to deformations of objects
- Local methods: rely on face localization and alignment, which would fail under unconstrained face images with complex variations
- Our idea: joint face localization and attribute prediction using only image-level attribute tags



## 3. Large-scale CelebFaces Attributes Dataset

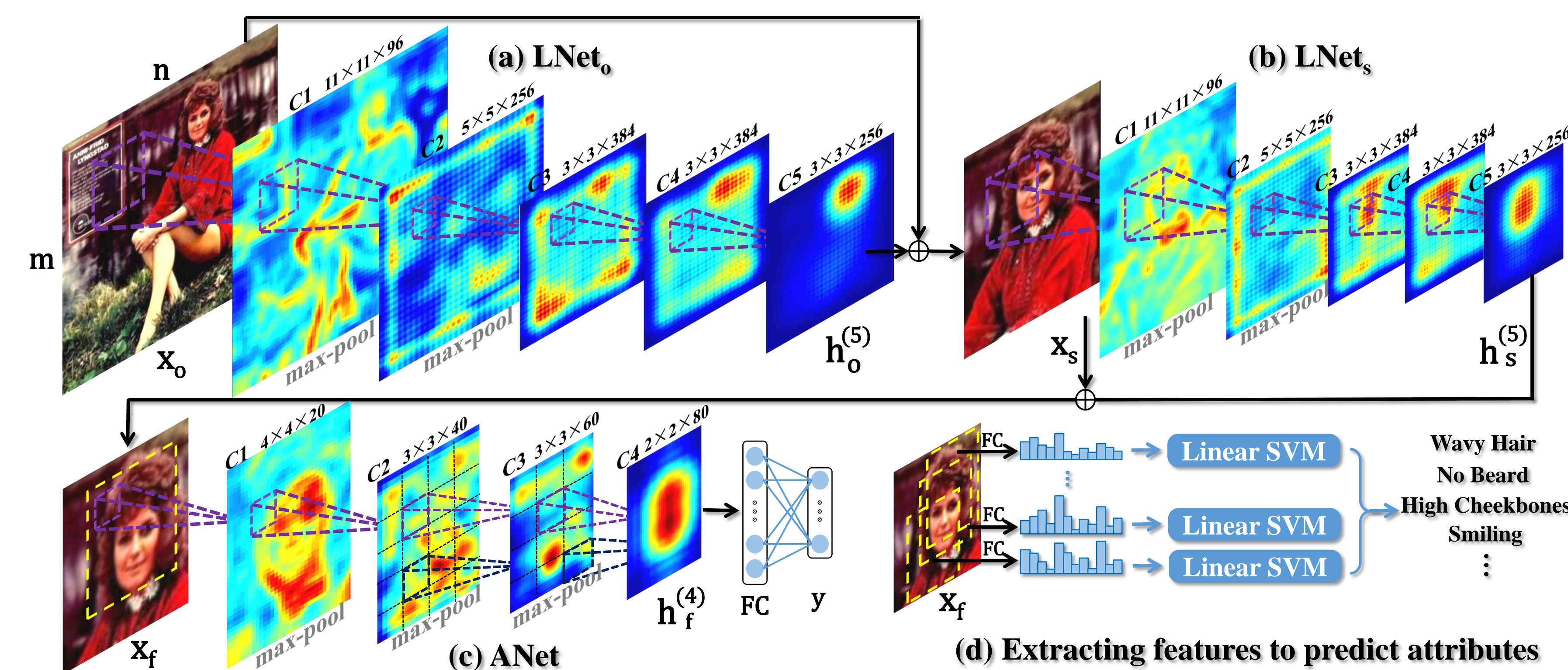


- 202,599 face images
- 10,177 human identities
- 5 landmarks per image
- 40 attributes per image

20x larger than previous

- Available at: <http://mmlab.ie.cuhk.edu.hk/projects/CelebA.html>

## 4. Overall Pipeline

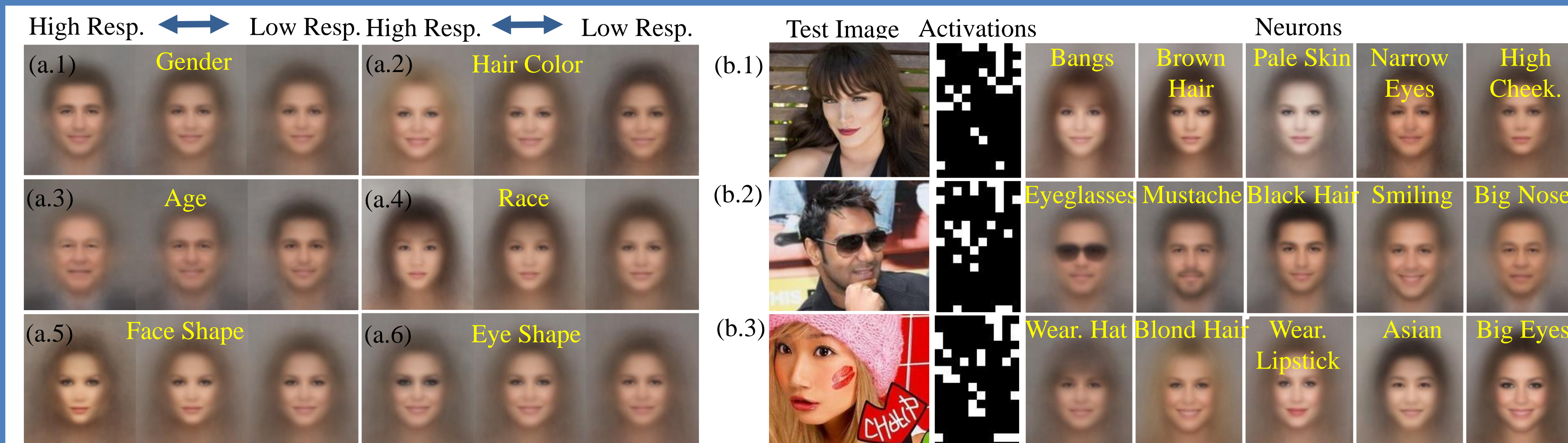


- LNets:
  - pre-trained with massive general objects
  - face localization with weak supervision
- ANet:
  - pre-trained with massive face identities
  - attribute prediction by leveraging local features
- LNets and ANet are jointly learned

## 5.1. Experimental Results (Face Localization)



## 5.2. Experimental Results (Attribute Prediction)



## 6. Conclusions

- With carefully designed pre-training strategies, our approach is robust to background clutters and face variations.
- We devise a new fast feed-forward algorithm for locally shared filters to save redundant computation.
- We have also revealed multiple important facts about learning face representation, which shed a light on new directions of face localization and representation learning.