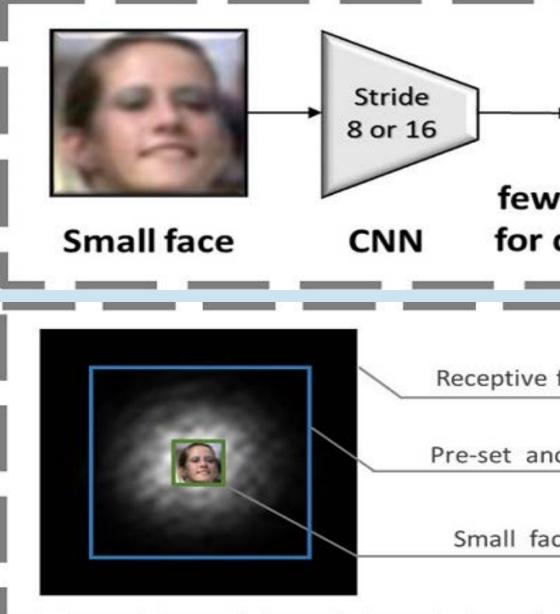
# Problem

Though tremendous strides have been made in object detection, one of the remaining open challenges is detecting small objects, since anchor-based detectors deteriorate dramatically as the objects become smaller. In this work, we analyze this problem in three aspects and propose three corresponding solutions to solve it.

# Reasons

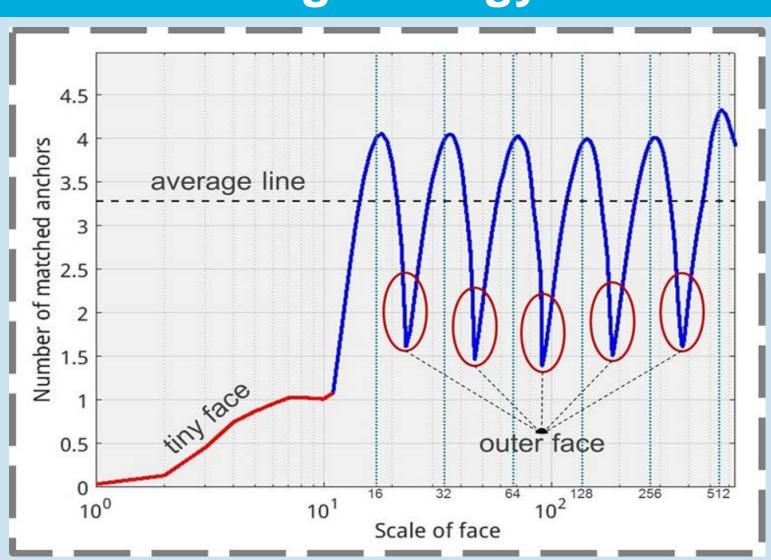
### **Reason1: Biased framework**

- Small faces have few features at detection layer
- > Small face, anchor scale and receptive field are mutual mismatch



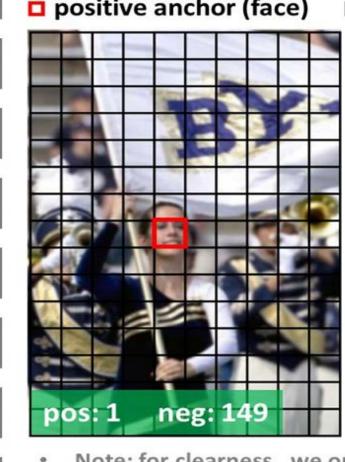
### **Reason2: Anchor matching strategy**

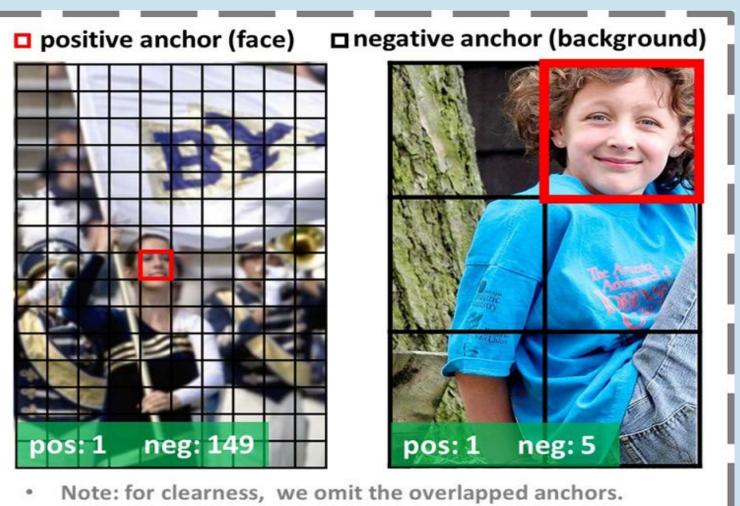
Some faces can not match enough anchors, such as tiny and outer face



### **Reason3: Background from small anchors**

Detecting small faces needs to tile plenty of small anchors, which may bring about many false positive faces













# S<sup>3</sup>FD: Single Shot Scale-invariant Face Detector Shifeng Zhang, Xiangyu Zhu, Zhen Lei, Hailin Shi, Xiaobo Wang and Stan Z. Li CBSR & NLPR, Institute of Automation, Chinese Academy of Sciences, University of Chinese Academy of Sciences

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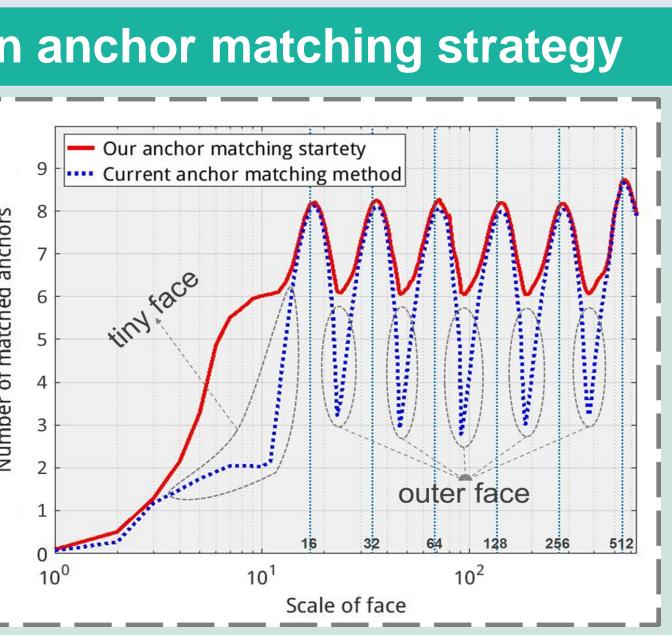


#### Solutions Solution1: Scale-equitable framework **Base Convolutional Layers Extra Convolutional Layers** VGG16 through Pool5 laver) Normalization Lavers Detection Layers Predicted Conv3\_3 Convolutional Conv4\_3 Layers Conv5\_3 Conv\_fc7 Max-out BG label Conv6\_2 Conv7 2 Multi-task Loss Layer: SoftmaxLoss and SmoothL1Lc

- > Anchor locations: To ensure various scales of faces have adequate features for detection, we associate anchors with layers whose stride size gradually double from 4 to 128
- > Anchor scales: To make anchors match the effective receptive field and keep them have the same density, we design scales from 16 to 512 pixels based on effective receptive field theory and equal-proportion interval principle

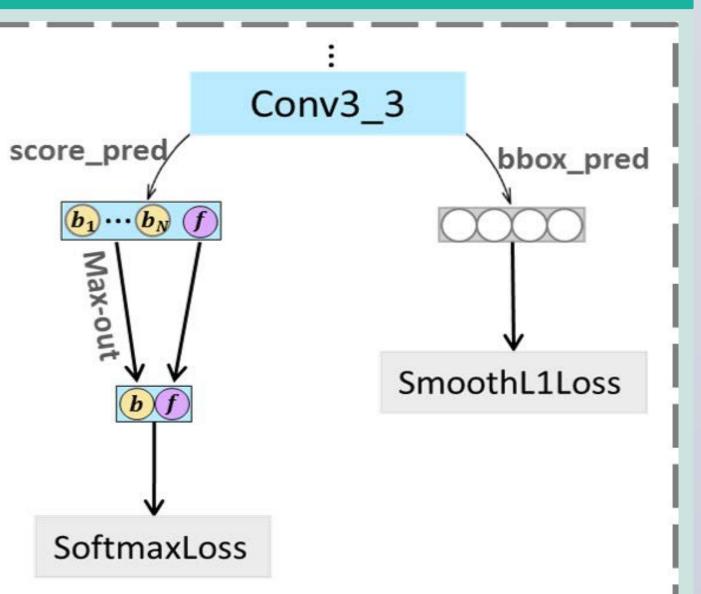
### Solution2: Scale compensation anchor matching strategy

For these faces that can not match enough anchors, we increase their matched anchors via an extra step that relaxes the matching threshold then selects the top N anchors

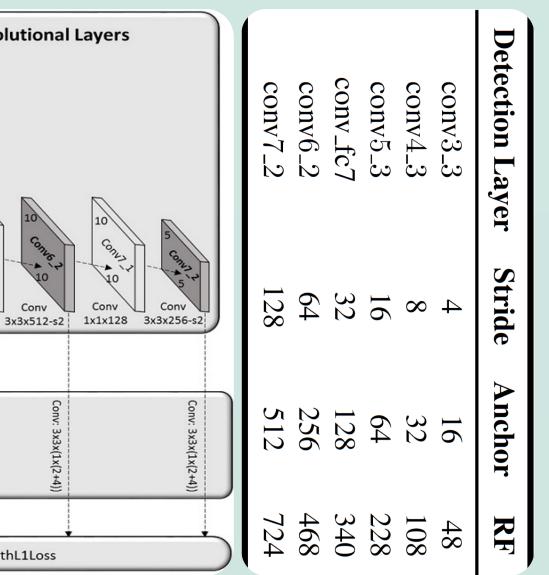


## Solution3: Max-out background label

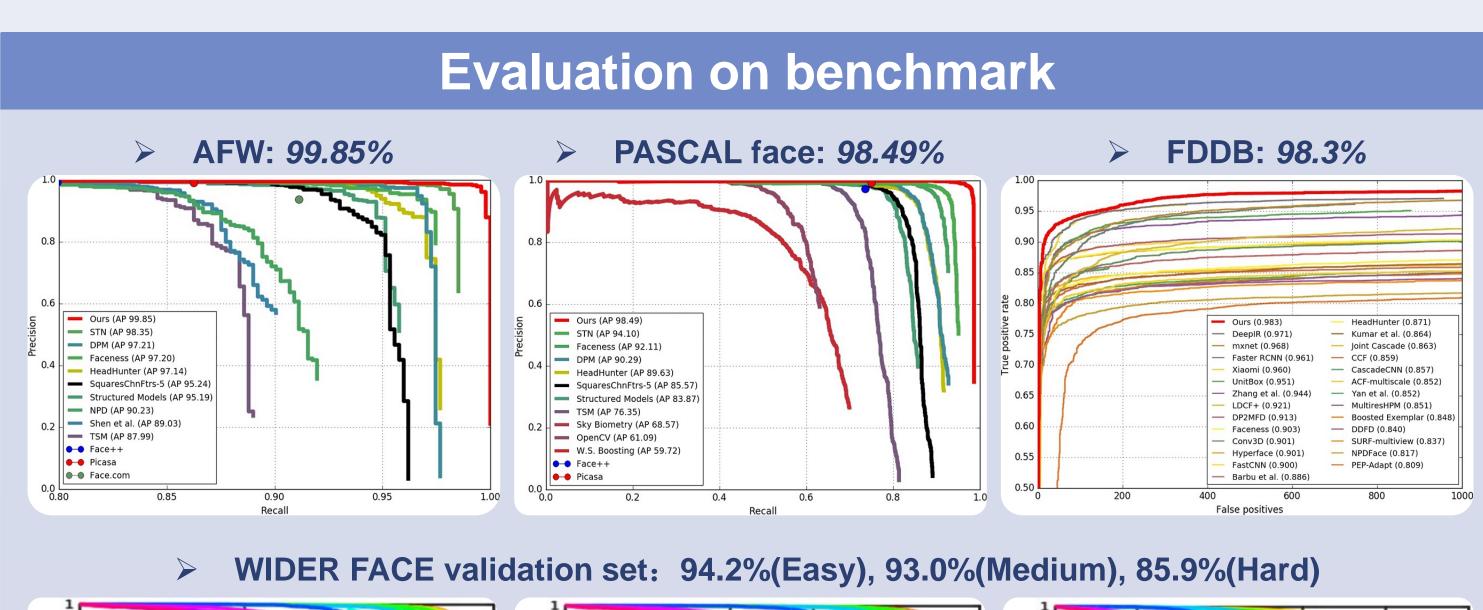
For each of the small anchors on the lowest layer, we predict N<sub>m</sub> scores for background label and then choose the highest as its final score, so as to reduce the false positive rate of small faces

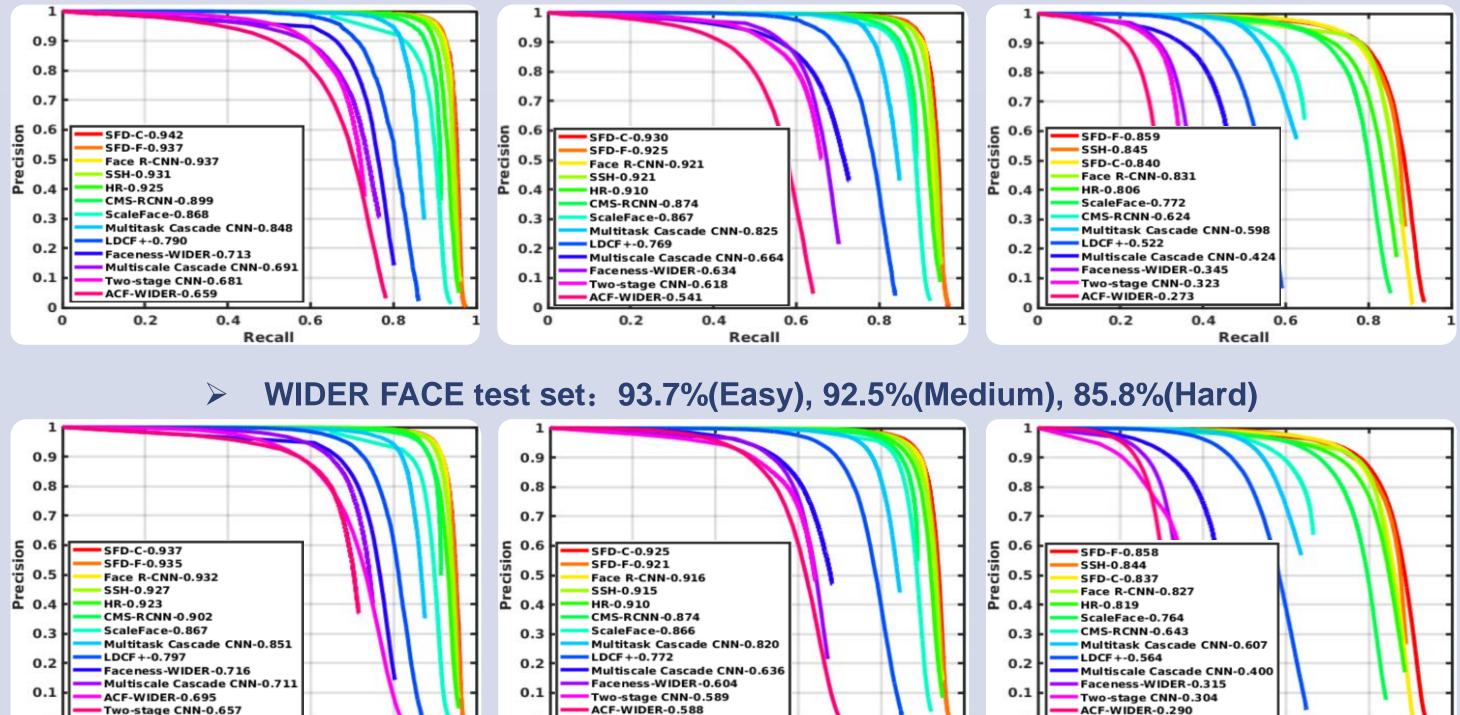


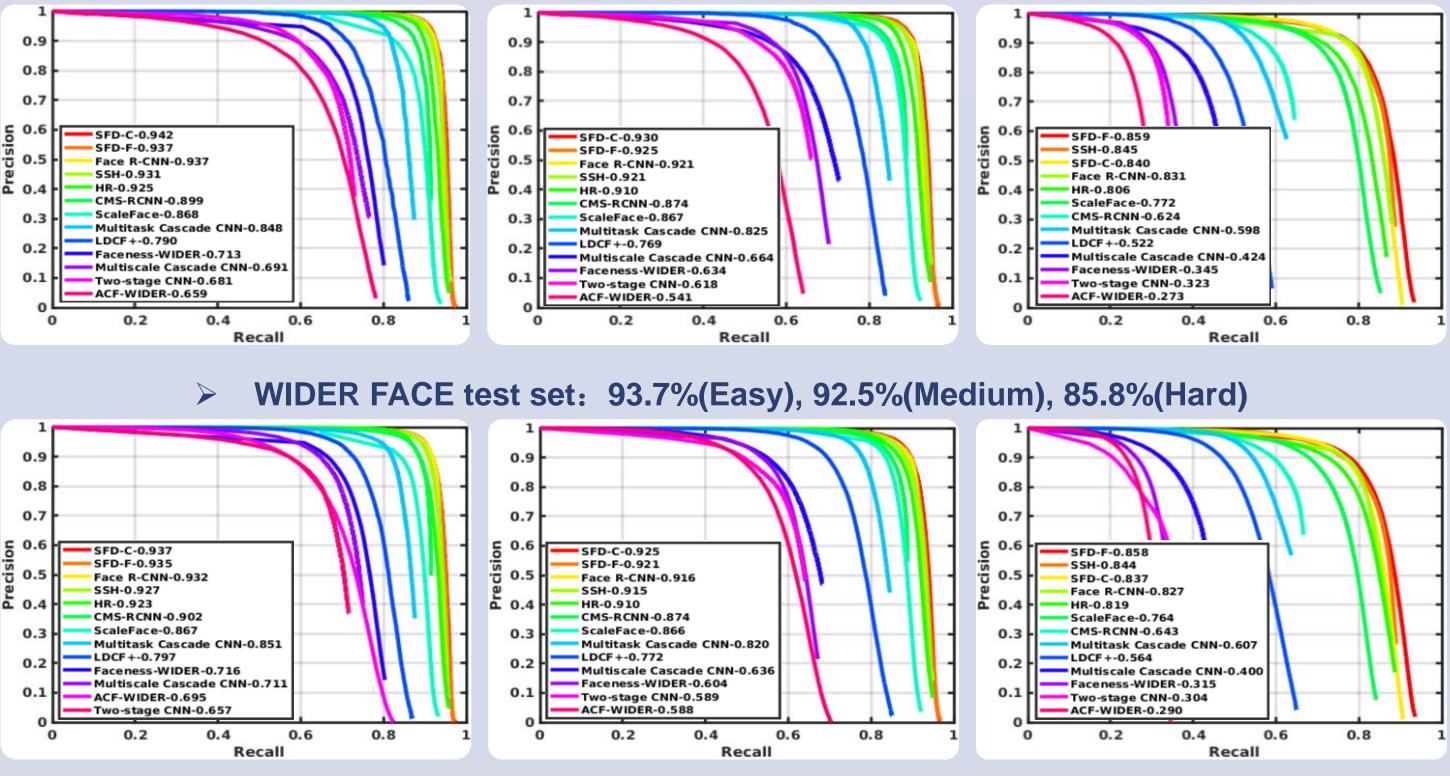




- Scale-equitable frame crucial
- Scale compensation matching strategy is
- Max-out background promising







We propose a scale-equitable framework, a scale compensation matching strategy and a max-out background label to solve the problem of finding small faces. Consequently, our detector achieves state-of-the-art detection performance on all the common face detection benchmarks, including the AFW, PASCAL face, FDDB and WIDER FACE datasets. Besides it can run at 36 FPS on a NVIDIA Titan X (Pascal) for the VGAresolution images.

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# **Results**

Model analysis					
ework is	mAP(%) Subsets	Easy Medium Hard			
	Methods				
anchor	RPN-face	91.0	88.2	73.7	
better	SSD-face	92.1	89.5	71.6	
label is	$S^{3}FD(F)$	92.6	91.6	82.3	
	$S^{3}FD(F+S)$	93.5	92.0	84.5	
	$S^{3}FD(F+S+M)$	93.7	92.4	85.2	

#### Conclusion

