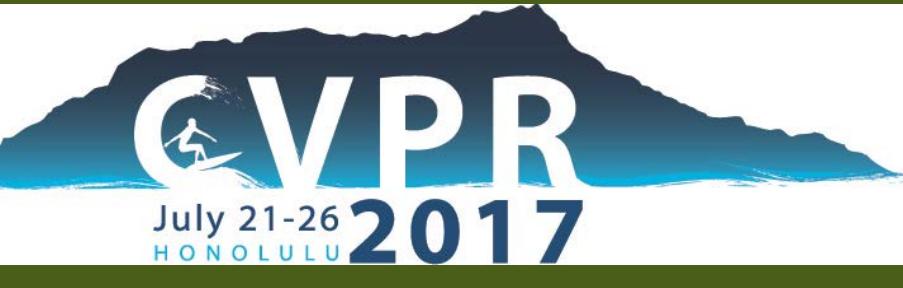




# Instance-Level Salient Object Segmentation

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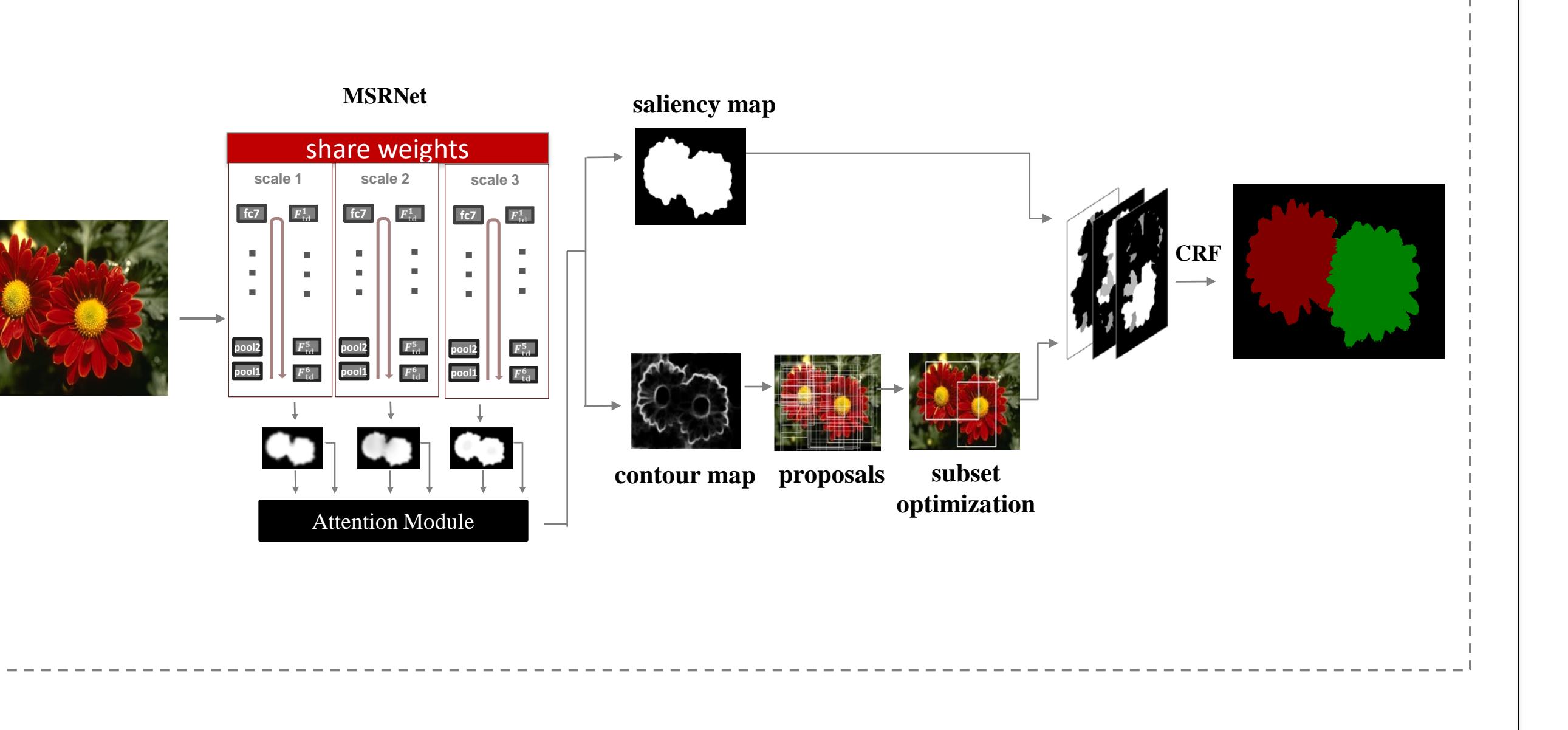
## Overview

### Instance level salient object segmentation pipeline:

1. Multiscale refinement network for saliency map estimation and salient object contour detection.
2. Salient object proposal generation and screening.
3. Refinement of salient instance segmentation based on CRF.

### Contributions:

- Develop a fully convolutional multiscale refinement network, called MSRNet, for salient region detection.
- MSRNet generalizes well to salient object contour detection, making it possible to separate object instances in detected salient regions
- A new challenging dataset for salient instance segmentation.

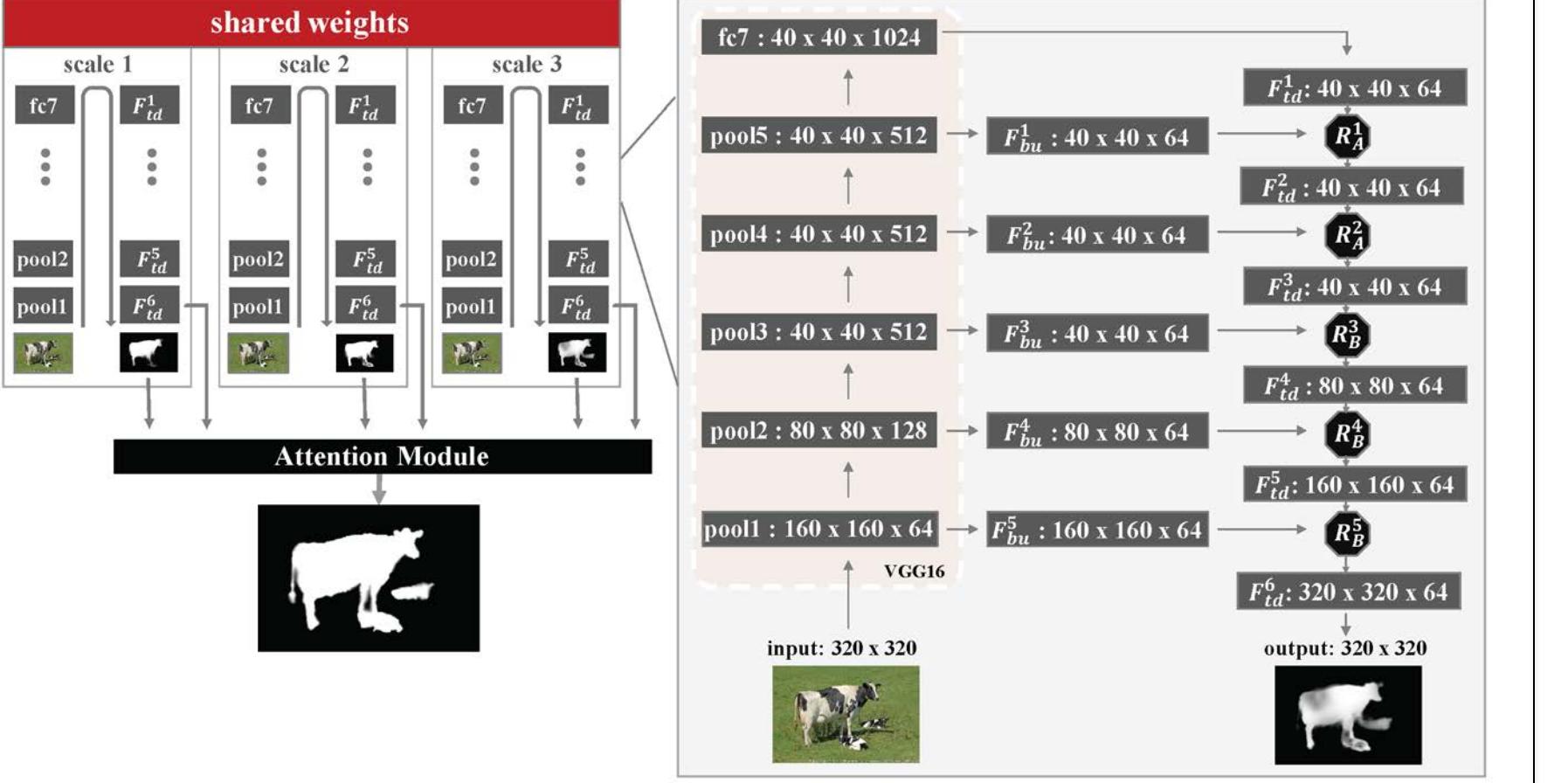


## Multiscale Refinement Network

Three refined VGG network streams with shared parameters and a learned attentional model for fusing results at different scales.

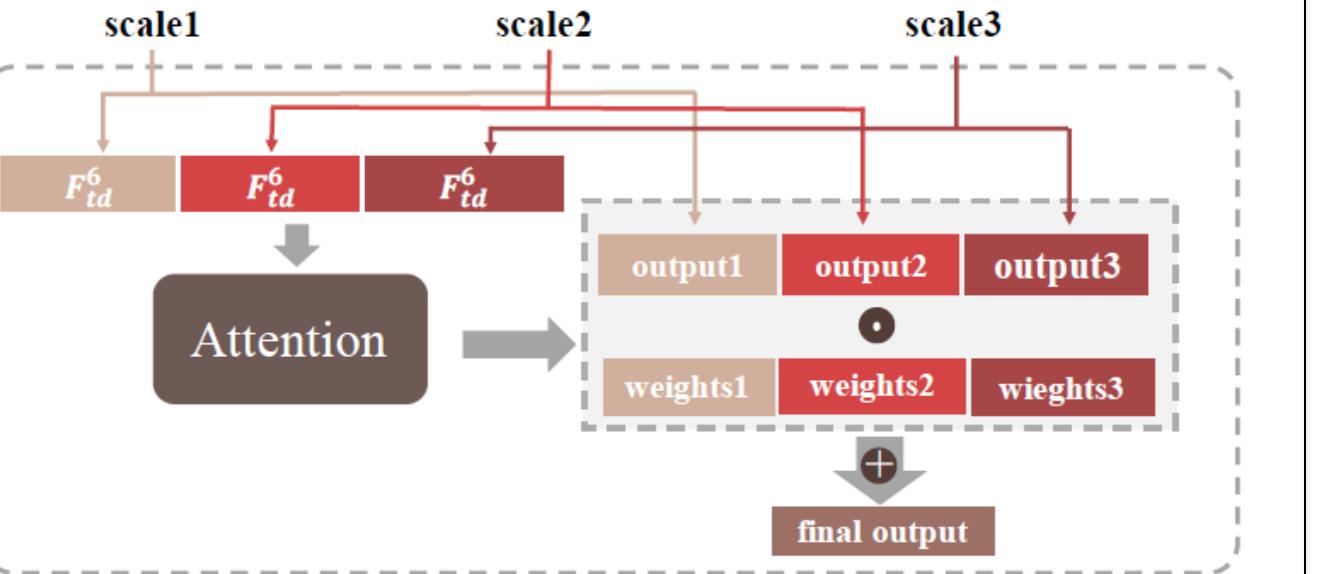
### Refined VGG Network

- Transform the original VGG16 into a fully convolutional network, which serves as our bottom-up **backbone network**.
- Augment the backbone network with a **top-down refinement stream**.
- The refinement stream consists of five stacked refinement modules.



### Refinement Module: R

- To invert the effect of each pooling layer and double the resolution if necessary.
- First concatenating  $F_{td}^i$  and  $F_{bu}^i$ , then feeding them to another  $3 \times 3$  convolutional layer with 64 channels.

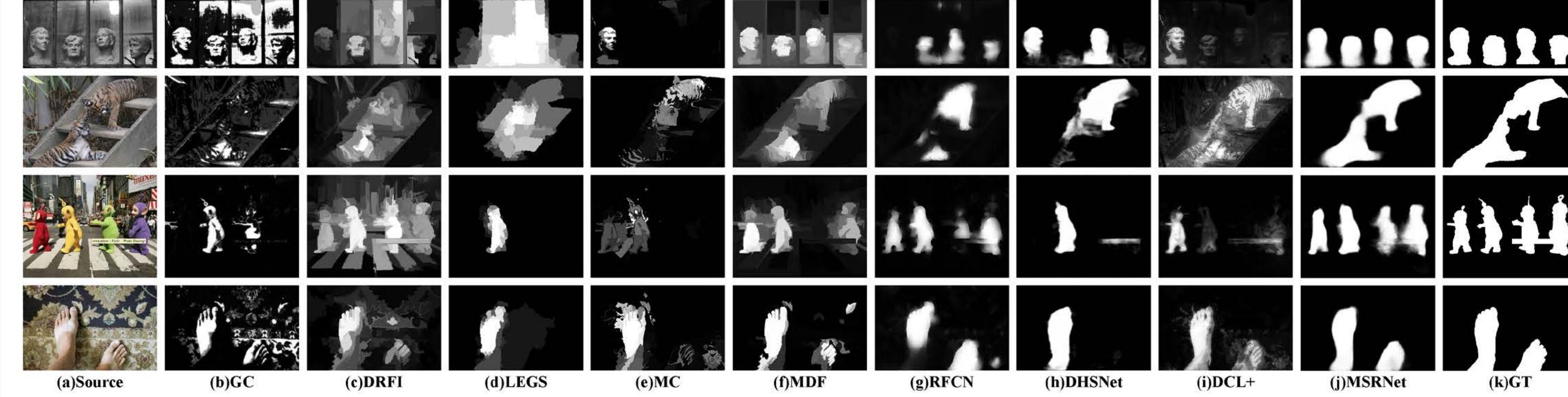


### Multiscale Fusion with Attentional Weights

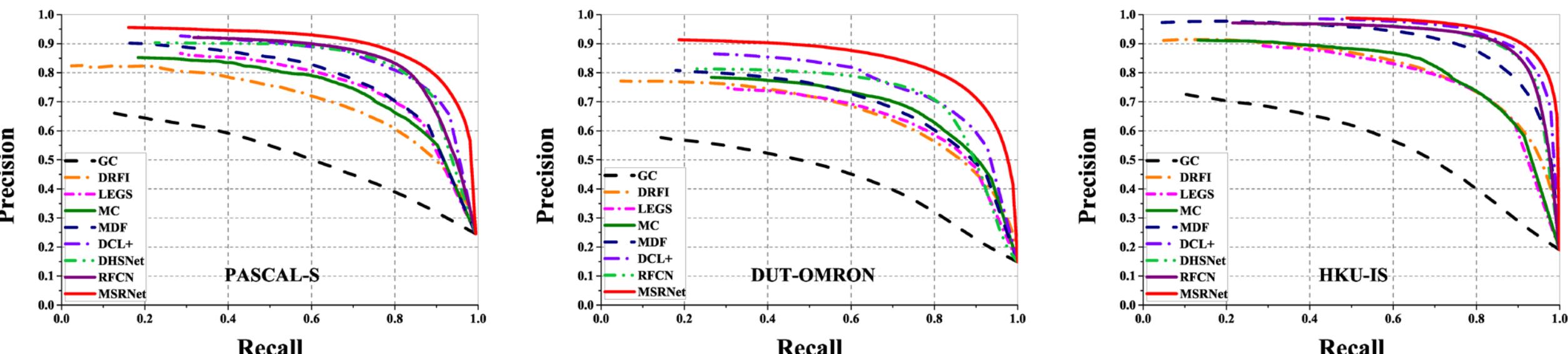
$$F_c = \sum_{s \in \{1, 0.75, 0.5\}} W^s \odot M_c^s$$

## Experiments

### Evaluation on Salient Region Detection:

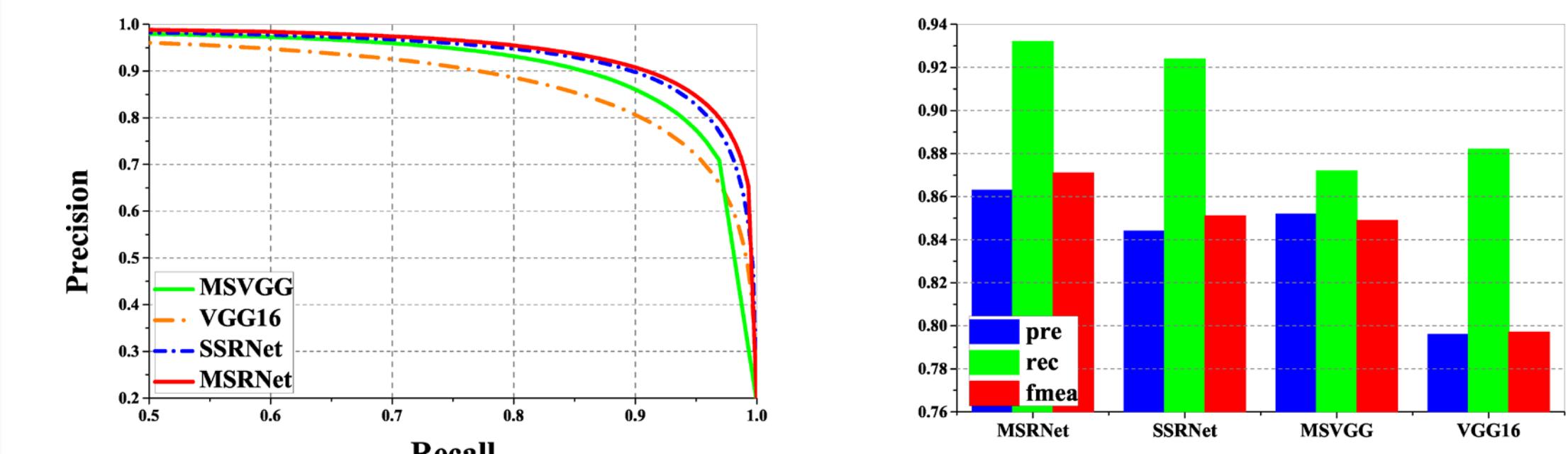


Visual comparison of saliency maps from state-of-the-art methods, including our MSRNet. MSRNet consistently produces saliency maps closest to the ground truth.



Dataset	Metric	GC	DRFI	LEGS	MC	MDF	RFCN	DHSNet	DCL+	MSRNet
MSRA-B	maxF	0.719	0.845	0.870	<b>0.894</b>	0.885	—	—	<b>0.916</b>	<b>0.930</b>
	MAE	0.159	0.112	0.081	<b>0.054</b>	0.066	—	—	<b>0.047</b>	<b>0.042</b>
PASCAL-S	maxF	0.539	0.690	0.752	0.740	0.764	<b>0.832</b>	<b>0.824</b>	0.822	<b>0.852</b>
	MAE	0.266	0.210	0.157	0.145	0.145	0.118	<b>0.094</b>	<b>0.108</b>	<b>0.081</b>
DUT-OMRON	maxF	0.495	0.664	0.669	0.703	0.694	<b>0.747</b>	—	<b>0.757</b>	<b>0.785</b>
	MAE	0.218	0.150	0.133	<b>0.088</b>	0.092	0.095	—	<b>0.080</b>	<b>0.069</b>
HKU-IS	maxF	0.588	0.776	0.770	0.798	0.861	<b>0.896</b>	0.892	<b>0.904</b>	<b>0.916</b>
	MAE	0.211	0.167	0.118	0.102	0.076	0.073	<b>0.052</b>	<b>0.049</b>	<b>0.039</b>
ECSSD	maxF	0.597	0.782	0.827	0.837	0.847	0.899	<b>0.907</b>	<b>0.901</b>	<b>0.913</b>
	MAE	0.233	0.170	0.118	0.100	0.106	0.091	<b>0.059</b>	<b>0.068</b>	<b>0.054</b>
SOD	maxF	0.526	0.699	0.732	0.727	0.785	0.805	<b>0.823</b>	<b>0.832</b>	<b>0.847</b>
	MAE	0.284	0.223	0.195	0.179	0.155	0.161	<b>0.127</b>	<b>0.126</b>	<b>0.112</b>

### Effectiveness of Multiscale Refinement Network



Scan here!

Examples of salient instance segmentation results by our MSRNet based framework.

Salient Contour Detection			Salient Instance Segmentation		
ODS	OIS	AP	mAP <sup>r</sup> @0.5(%)	mAP <sup>r</sup> @0.7(%)	
0.719	0.757	0.765	65.32	52.18	

Quantitative benchmark of our new dataset.