Multi-criteria Energy Minimization with Boundedness, Edgedensity and Rarity, for Object Saliency in Natural Images



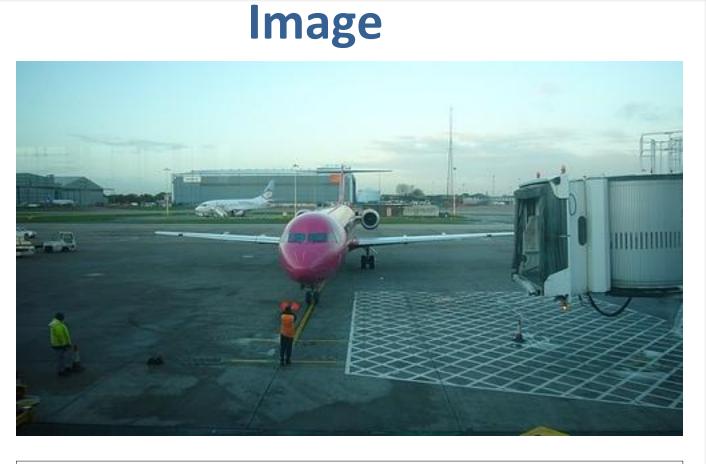
Sudeshna Roy and Sukhendu Das Visualization and Perception Lab.

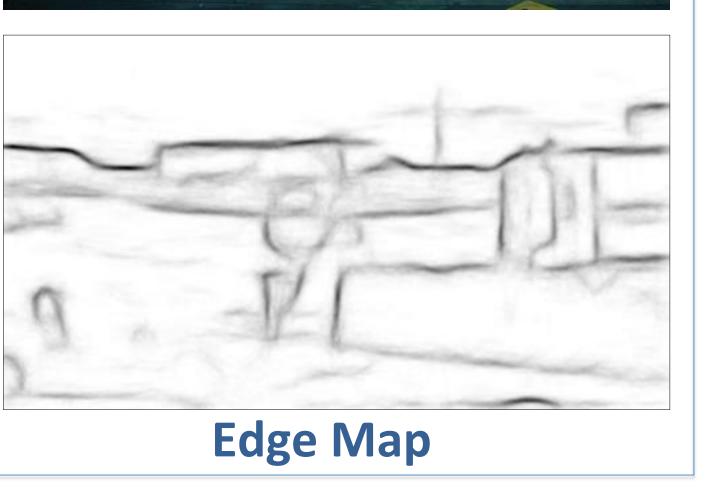
Dept. of Computer Science and Engineering, IIT Madras, India

Abstract

- Motivated by saliency and category independent object segmentation methods, we predict a single object proposal which captures all the salient objects in an image, within O(number of superpixels)³.
- The CRF parameters are learned using a margin-rescaled struct-SVM approach with (1- IoU) loss function.
- Fast but effective objectness criteria, viz. boundedness and edge-density. Submodular CRF is minimized using graph-cut. Results are shown on the challenging PASCAL VOC 2012 dataset.

Algorithm Overview

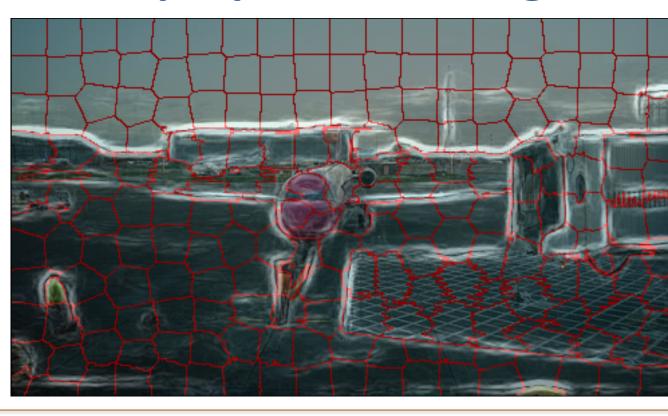


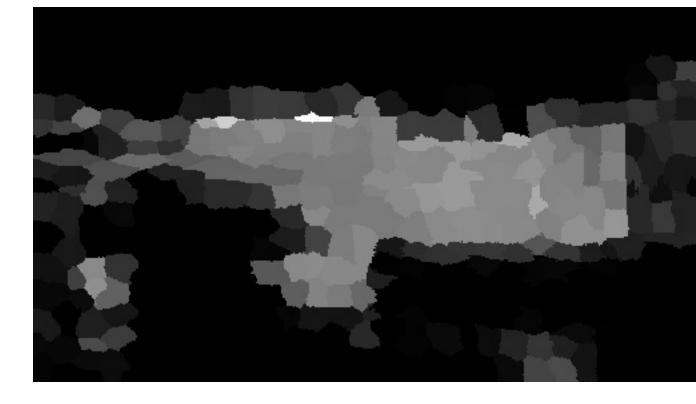


Input

Energy Function over superpixels:

Superpixels and Edges

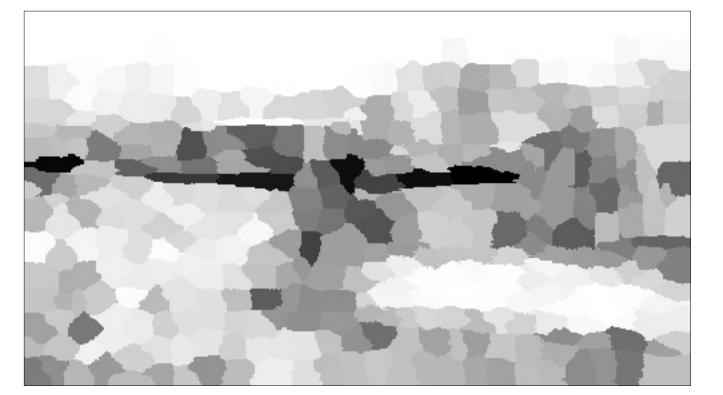




Boundedness (b_i)

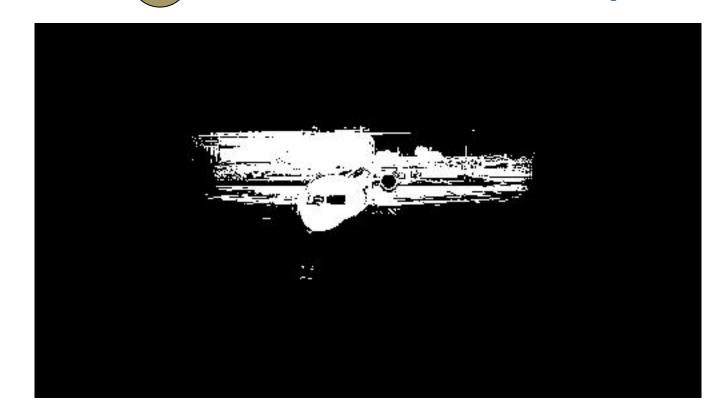
Objectness





Edge Density (ed_i)

Unary Potential
(Saliency, Objectness) $\phi_{i}^{(1)}(y_{i}, x_{i} | w_{1})$ Pairwise Potential
(based on color) $\phi_{i,j}^{(2)}(y_{i}, y_{j}, x_{i}, x_{j} | w_{2})$ Submodular CRF
Inference: Graph-cut



Proposed Output

 $E(\mathbf{y}, \mathbf{x}, \mathbf{w}) = \sum_{i \in V} \phi_i^{(1)} + \lambda \sum_{(i,j) \in \mathcal{E}} \phi_{i,j}^{(2)}$

Salient Object Likelihood:

 $\phi_i^{(1)}(y_i, x_i^{(1)} | w_1) = w_s ((1 - y_i) (1 - s_i) + y_i s_i) + w_b ((1 - y_i) (1 - b_i) + y_i b_i) + w_{ed} ((1 - y_i) ed_i + y_i (1 - ed_i))$ Edge Cost:

 $\phi_{i,i}^{(2)}$ (y_i, y_i, x_i⁽²⁾, x_j⁽²⁾) = |y_i-y_j| exp(-k_c||c_i-c_j||)

 c_i denotes color of ith superpixel in lab space and k_c indicates the sensitivity of color similarity.

CRF parameter Learning:

- Convex in \mathbf{w} , given \mathcal{L}
- Iteratively optimized, by finding $\mathbf{y_n}$ using graph-cut, and thus \mathcal{L}
- M is the number of training instances.
- $\bullet w = [w_1 \ w_2] = [w_s \ w_b \ w_{ed} \ \lambda]^T$

$$\min_{\mathbf{W}} \frac{1}{2} \|\mathbf{w}\|^2 + \frac{1}{M} \sum_{n=1}^{M} \xi_n$$

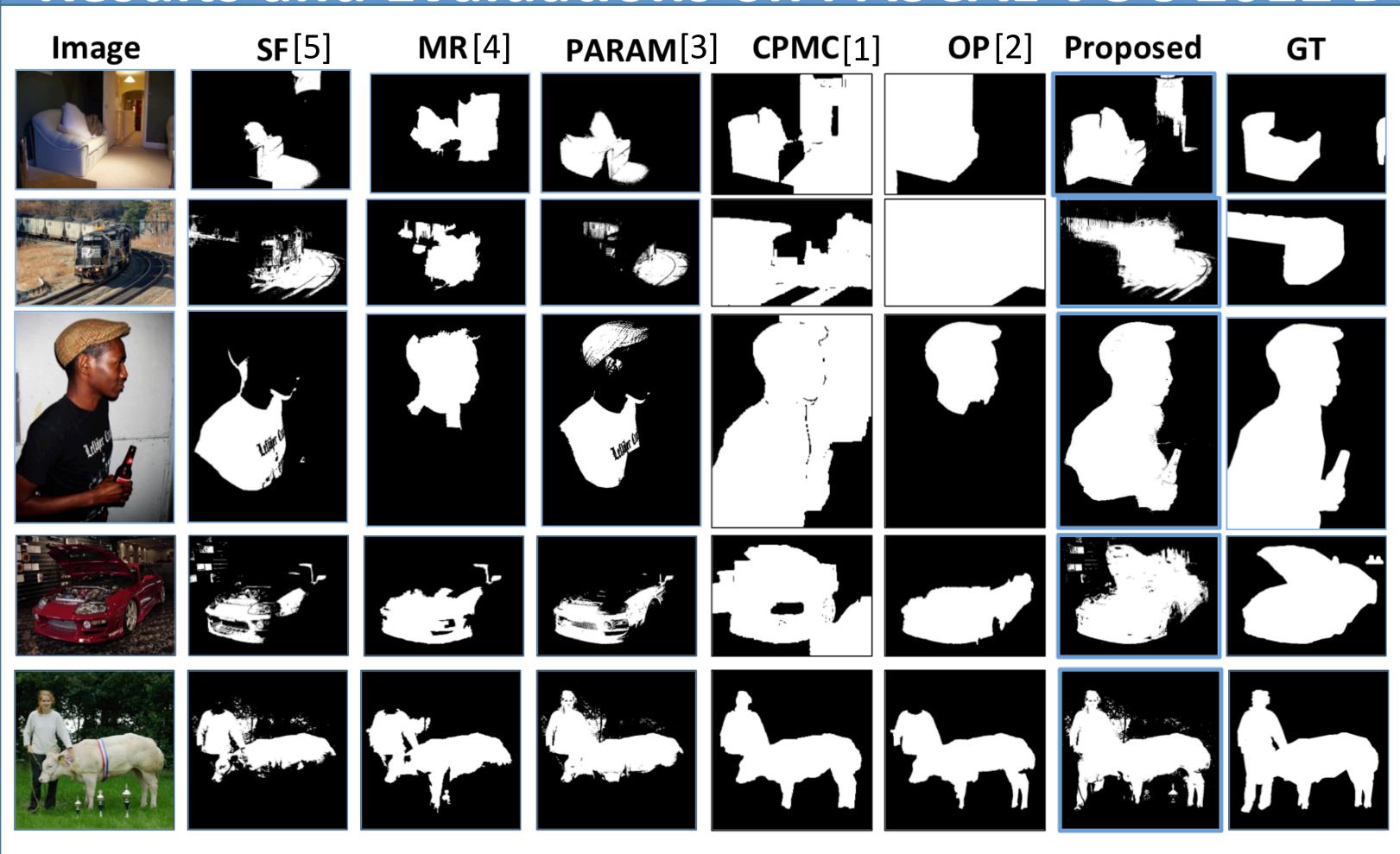
subject to,
$$\mathbf{w}^{\mathbf{T}}\mathbf{v} = \mathbf{1}$$

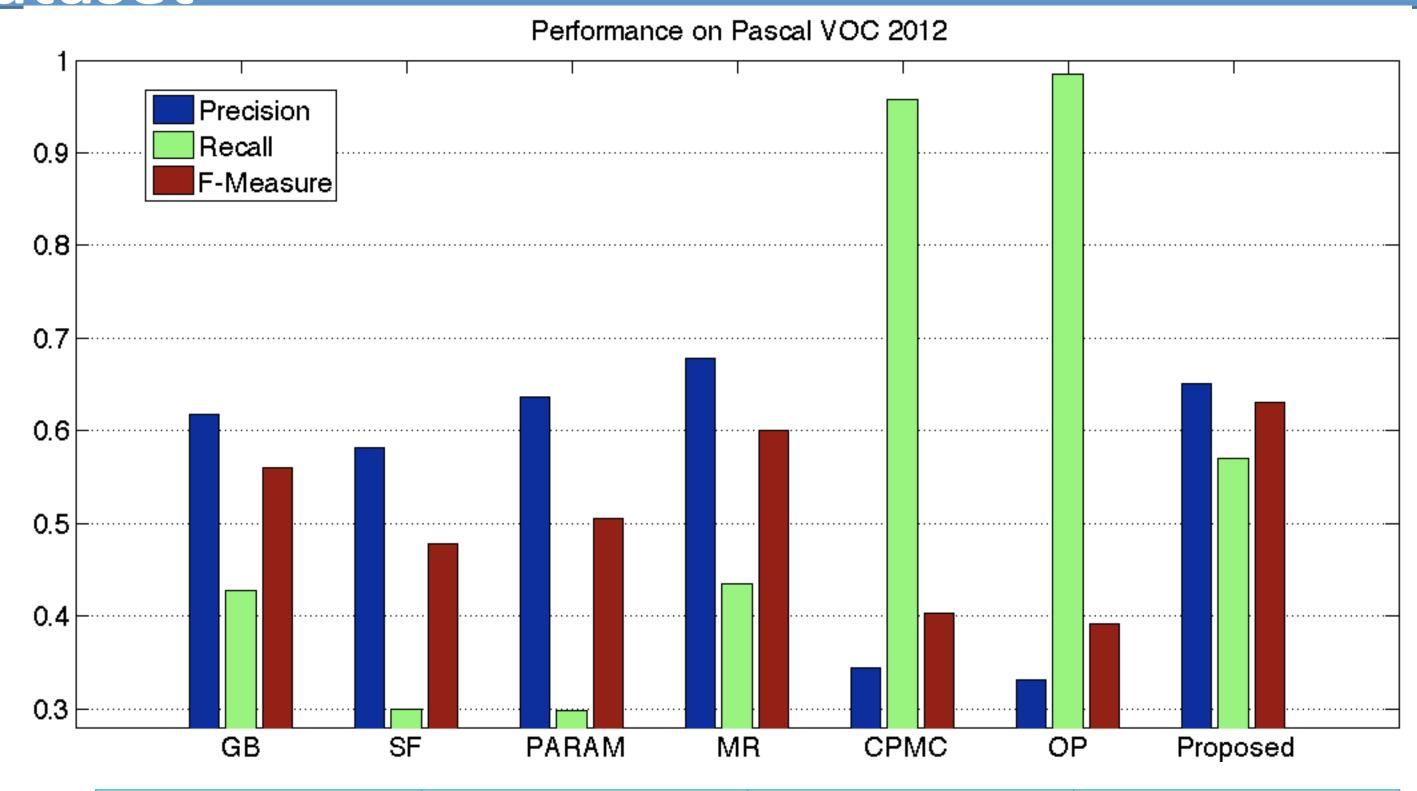
$$\mathbf{w}^T \phi_n - \mathbf{w}^T \hat{\phi}_n \le \mathcal{L}(\mathbf{y}_n, \hat{\mathbf{y}}_n) + \xi_n$$

$$\xi_n \ge 0, w_2 > 0$$

$$\mathcal{L} = (1-\text{loU})$$

Results and Evaluations on PASCAL VOC 2012 Dataset





G.D.			
Method	CPMC [1] *	OP [2] *	Proposed
IoU Score	0.3319	0.3266	0.4097

^{*} Considering top 10 proposal masks

Conclusion

- An efficient method based on saliency in conjunction with objectness cues to predict a category-independent object segmentation, using a CRF based formulation.
- Can be a preprocessing step for many high-level vision task.

References

- 1. Carreira and Sminchisescu, "Constrained parametric min-cuts for automatic object segmentation," CVPR, 2010.
- 2. Endres and Hoiem, "Category independent object proposals," ECCV, 2010.
- 3. Roy and Das, "Saliency detection in images using graph-based rarity, spatial compactness and background prior," VISAPP, 2014.
- 4. Yang, et. al., "Saliency detection via graph-based manifold ranking," CVPR, 2013.
- 5. Perazzi et. al., Saliency filters: Contrast based filtering for salient region detection. CVPR. 2012.