

# Constrained Parametric Min-Cuts for Automatic Object Segmentation

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# Overview

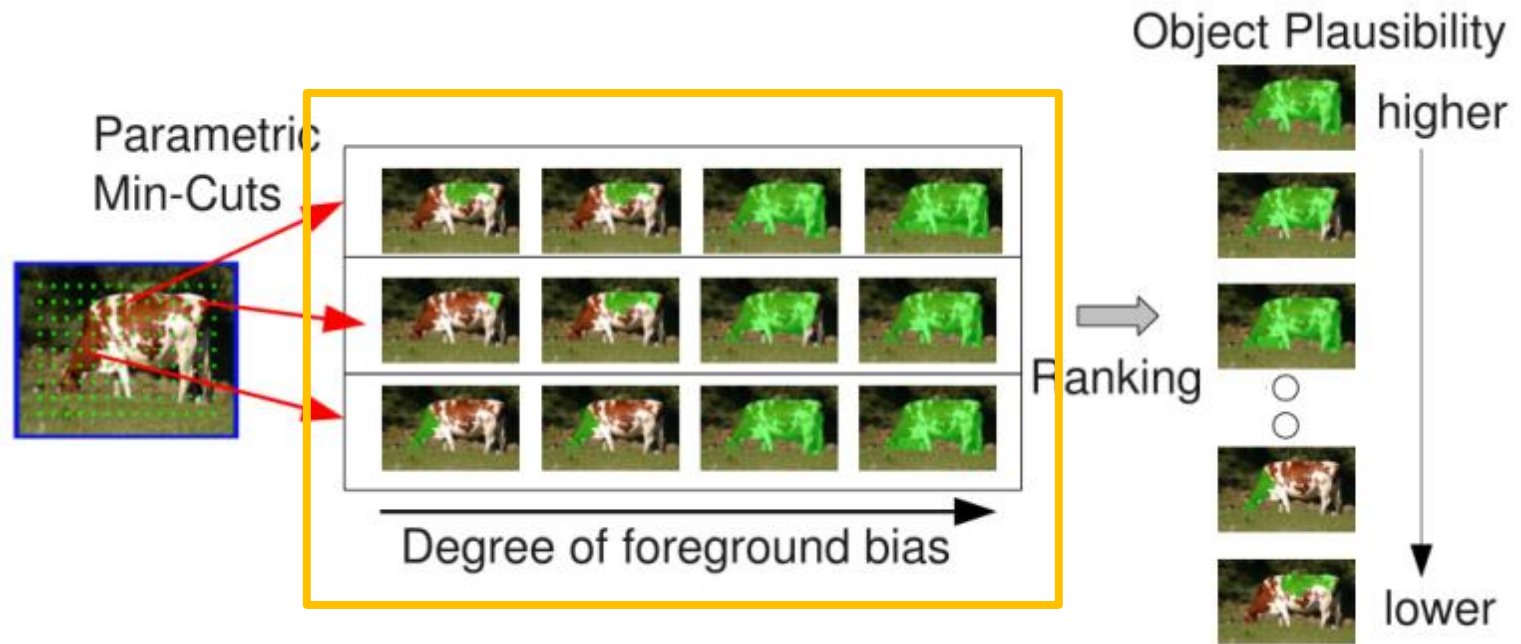


Figure credit: Joao Carreira *et al.*

# Constrained Parametric Min-Cuts (CPMC)

- Graph-based segmentation algorithm
  - Similarity between neighboring pixels is encoded as edges.

$$E^\lambda(X) = \sum_{u \in V} D_\lambda(x_u) + \sum_{(u,v) \in E} V_{uv}(x_u, x_v)$$

$$V_{uv}(x_u, x_v) = \begin{cases} 0 & , \text{ if } x_u = x_v \\ g(u, v) & , \text{ if } x_u \neq x_v \end{cases}$$

$$g(u, v) = \exp \left[ -\frac{\max(gPb(u), gPb(v))}{\sigma^2} \right]$$

where  $gPb$  is the output of the multi-cue contour detector.

# Constrained Parametric Min-Cuts (CPMC)

- Multi-Cue Contour Detector
  - Estimate the posterior probability of a boundary.

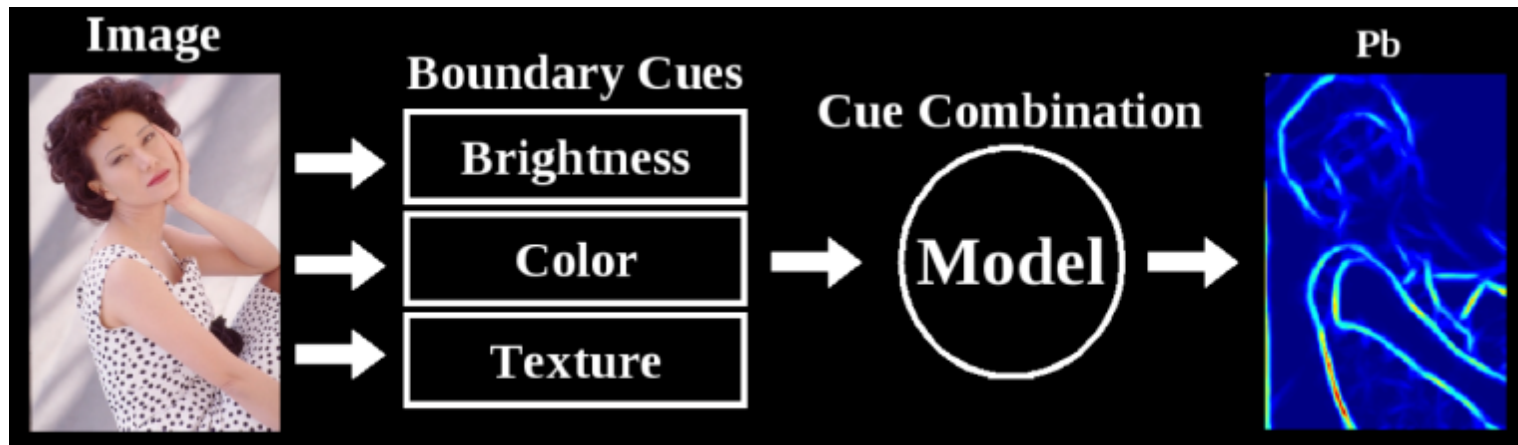


Figure credit: Michael Maire *et al.*

# Experiments

- Segmentation Covering

$$C(S, S') = \frac{1}{N} \sum_{R \in S} |R| * \max_{R' \in S'} O(R, R')$$

$$O(R, R') = \frac{|R \cap R'|}{|R \cup R'|}$$

$S$  : the ground-truth segmentation

$S'$  : the object hypotheses

$|R|$  : number of pixels in the ground-truth segment

# Experiments

- Example Images

0.923105



0.922365



0.915743



0.906883



# Experiments

- Example Images

0.978446



0.753905



0.369020



0.465728



0.681151





# Experiments – Distorted Images

- Will different distortions in images affect the segmentation performance?
- Will the distortion degrade the quality of the estimated posterior probability of boundary?
- LIVE Image Quality Database
  - Gaussian blur
  - JPEG compression
  - White noise

# Test Images

Reference



Blur



JPEG



White Noise



# Probability of Boundary Map

Reference



Blur



JPEG



White Noise



# Experiments

- Reference



# Experiments

- Blur



# Experiments

- JPEG



# Experiments

- White Noise



# Ranking Object Hypotheses

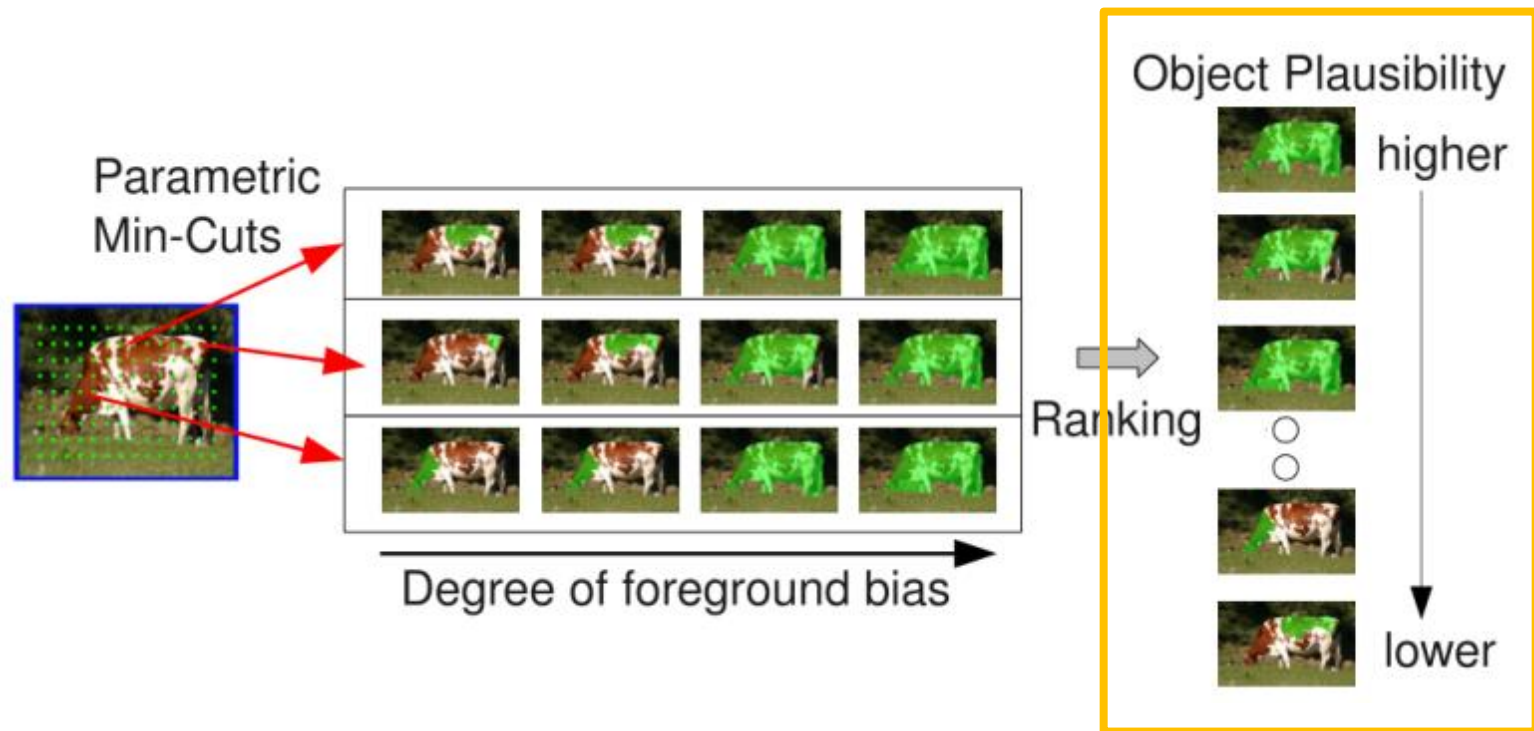


Figure credit: Joao Carreira *et al.*



# Experiments

- Can depth cues help rank the object hypotheses?
  - Depth are continuous; however, objects can be seen as residing in different depth planes.
- Middlebury Stereo Datasets
  - Ground-truth disparity maps
- LIVE Color+3D Database
  - Ground-truth range maps

# Experiments

- Append the feature with depth/disparity cues and retrain the ranking model with multiple linear regression.

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_p x_{ip} \text{ for } i = 1, 2, \dots, n$$

$$\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} x_{11} & \cdots & x_{1p} \\ \vdots & \ddots & \vdots \\ x_{n1} & \cdots & x_{np} \end{bmatrix} \begin{bmatrix} \beta_1 \\ \vdots \\ \beta_p \end{bmatrix} + \beta_0$$

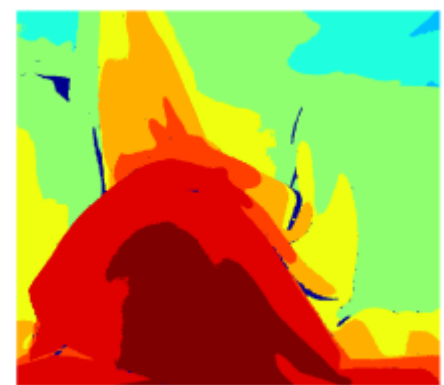
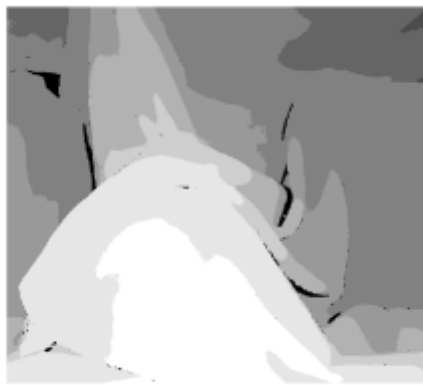
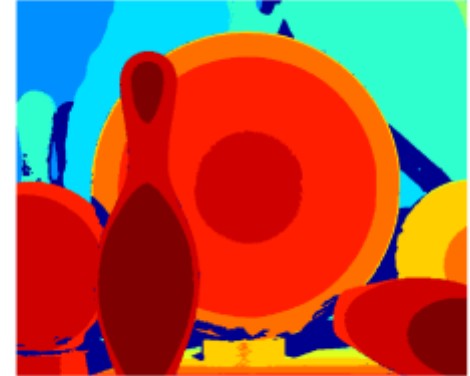
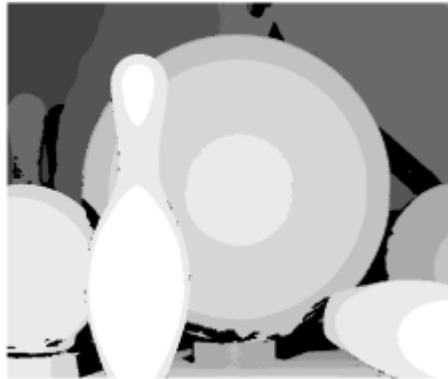
where  $[x_{i1}, \dots, x_{i(p-3)}]^T$  is the original feature vector containing graph partition, region, and gestalt properties,  
 $[x_{i(p-2)}, x_{i(p-1)}, x_{i(p)}]^T$  is the appended feature vector containing depth STD, depth gradient mean, and depth gradient STD.

# Experiments

- Middlebury Stereo Datasets
  - Indoor scenes with ground-truth disparity maps
  - Different types of objects
  - Ranking model is trained on LIVE Color+3D database.



# Experiments



# Original Features

0.264348



0.332096



0.219279



0.624507



0.220123



0.329886



# New Features and Regressor

0.426186



0.283115



0.219279



0.578269



0.206854



0.221965



# Original Features

0.629228



0.745103



0.463812



0.191724



# New Features and Regressor

0.467783



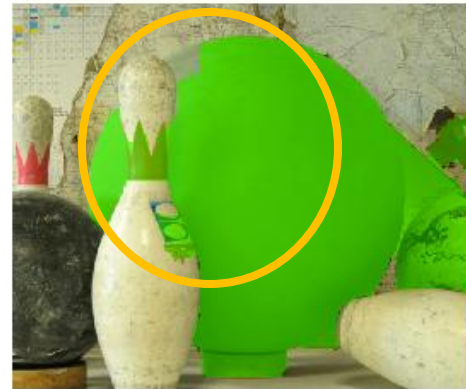
0.745103



0.403424



0.187363





# Original Features

0.196388



0.452087



0.505323



0.615173



# New Features and Regressor

0.196388



0.424314



0.490003



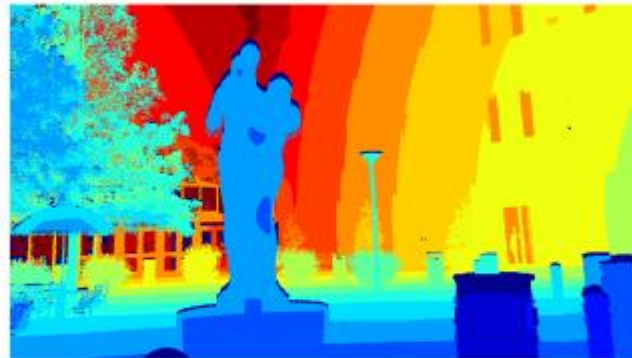
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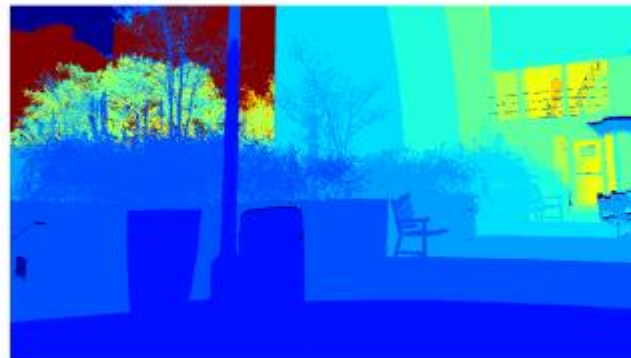
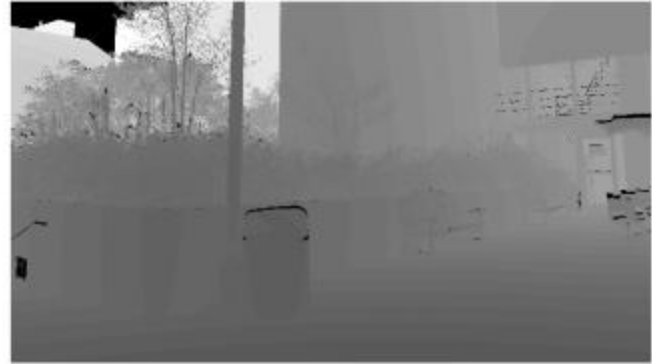
# Experiments

- LIVE Color+3D Database
  - Natural scenes with ground-truth range maps
  - Quantize actual range values to generate depth planes.
  - Ranking model is trained on Middlebury stereo datasets.

# Experiments



# Experiments



# Original Features

0.191496



0.338860



0.315339



0.251558



0.115919



0.165082



# New Features and Regressor

0.191496



0.193174



0.279806



0.180339



0.108559



0.165082

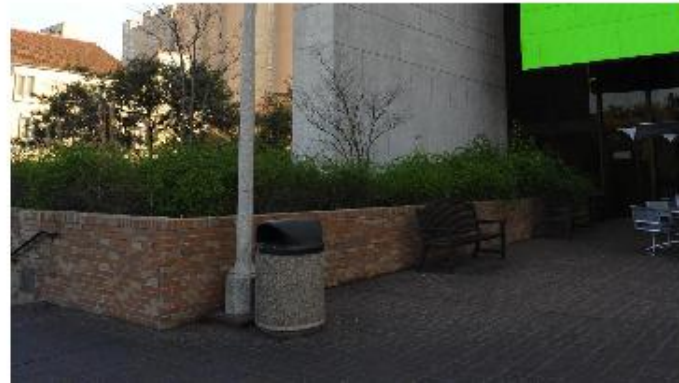


# Original Features

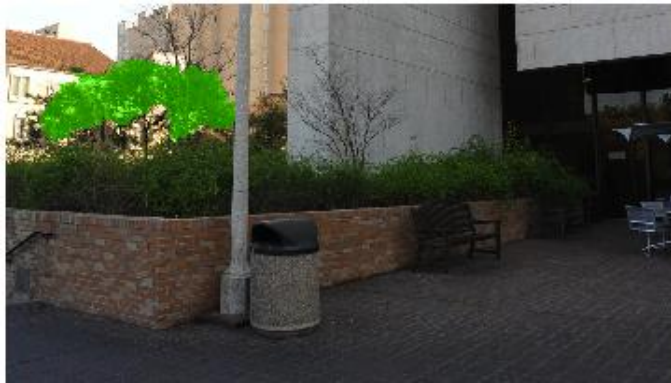
0.407832



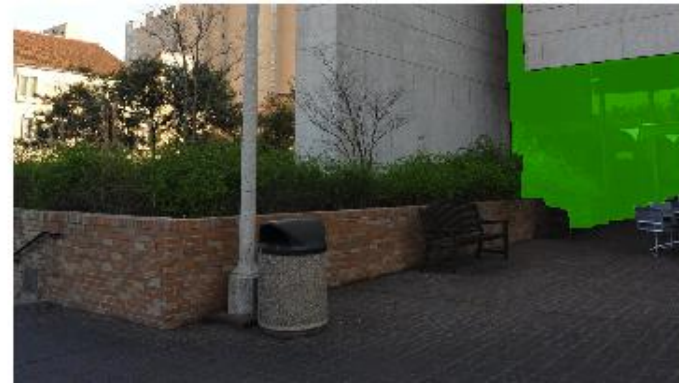
0.337091



0.133830



0.187111



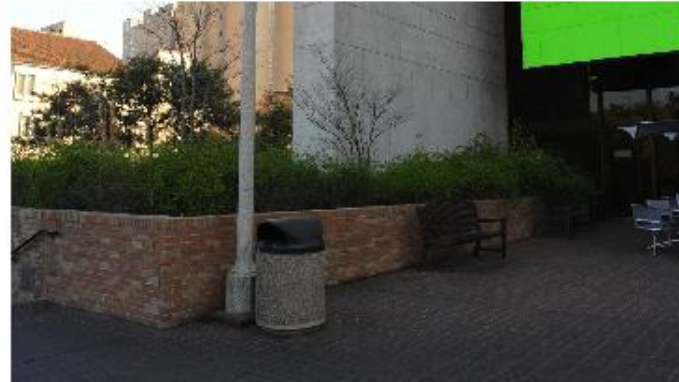


# New Features and Regressor

0.407832



0.333177



0.133830



0.179389



# Discussion

- Different types of distortions in images can affect the segmentation results.
  - Probability of boundary map is distorted.
  - CPMC generates incorrect figure-ground (object) hypotheses.
- Ranking model can be governed by different types of segment features and properties.
  - Depth cues could possibly help recognize objects, and vice versa.