# Probabilistic Joint Image Segmentation and Labeling

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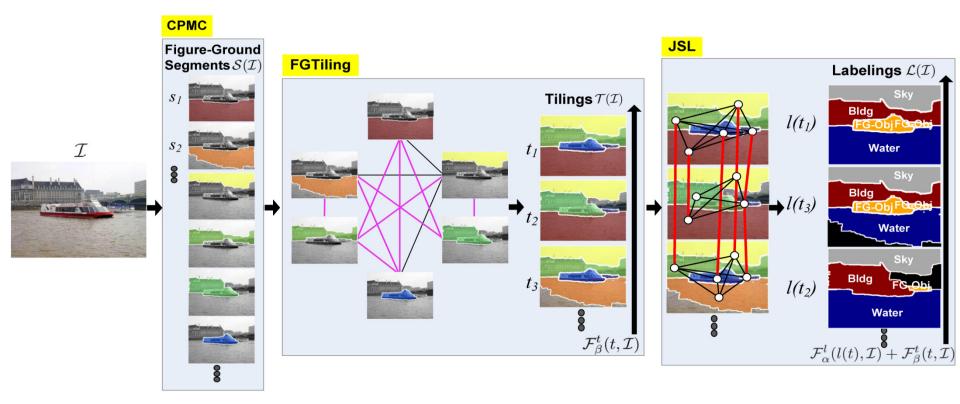
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#### How to segment and how to label images?

- Image segmentation and labeling are inter-dependent
  - At pixel level recognition is poorly defined for many semantic categories, e.g. people, chairs...
  - Given regions with sufficiently large spatial support, reliable recognition is possible
- Low-level segmentation can produce useful spatial support hypotheses, but these are rarely accurate in any single segmentation
- A recognition model should allow segment recombination and produce semantic label distributions rather than point estimates

## Model and Computational Principle



We explore figure/ground methods to generate large segment pools, then recombine and recognize subsets, within a sound statistical framework

## Joint Segmentation and Labeling Model

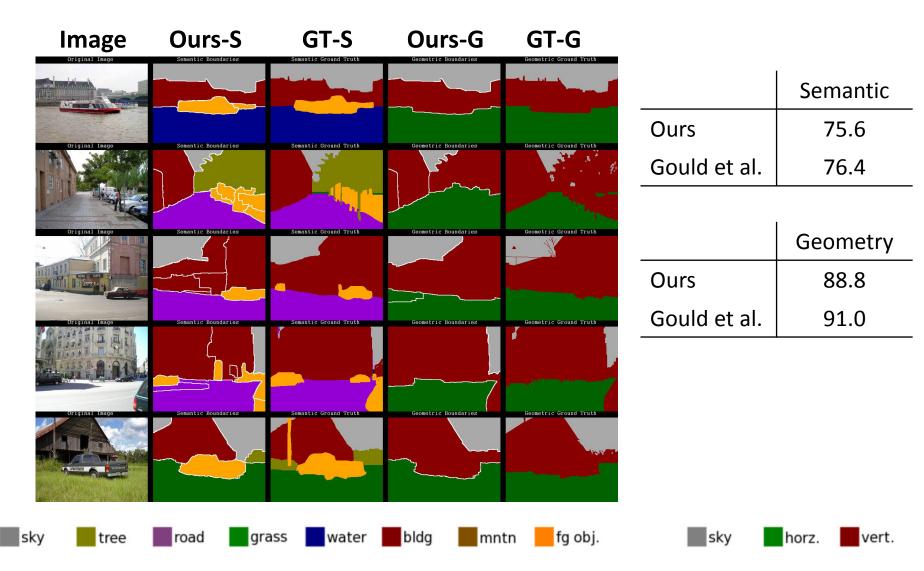
$$p_{\theta}(l(t), t, I) = \frac{1}{Z_{\theta}(I)} \exp F_{\theta}(l(t), t, I)$$

$$Z_{\theta}(I) = \sum_{t} \sum_{l(t)} F_{\theta}(l(t), t, I)$$

$$F_{\theta}(l(t), t, I) = F_{\alpha}^{l}(l(t), I) + F_{\beta}^{t}(t, I)$$
category dependent category independent

- Learn parameters using Maximum Likelihood
- Novel incremental partition function estimation
  - Sum over subset of configurations (multiple cliques, labeled)
  - Include incorrect configurations the model rates probable

#### Qualitative Scene Geometry (Stanford class + geometry)



#### Pascal VOC 2010

#### Segmentation and object class recognition

Average





	score
Ours	41.7
CVC-HARMONY-DET	40.1
UOCTTI_LSVM_MDPM	31.8
BROOKES_AHCRF	30.3
STANFORD_REGLABEL	29.1





