Localizing 3D Cuboids in Single-view Images

Jianxiong Xiao

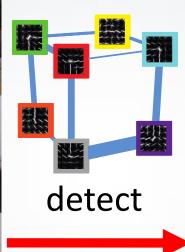
Bryan C. Russell

Antonio Torralba

3D Cuboid Detector

Input image





Output detections

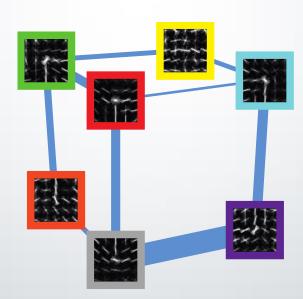


Synthesized New Views

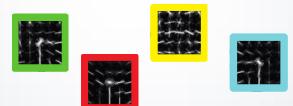


$$S(I, p) = \sum_{i \in \mathcal{V}} w_i^H \cdot \text{HOG}(I, p_i) + \sum_{ij \in \mathcal{E}} w_{ij}^D \cdot \text{Displacement}^{2D}(p_i, p_j)$$

$$+ \sum_{i \in \mathcal{V}} w_{ij}^E \cdot \text{Edge}(I, p_i, p_j) + w^S \cdot \text{Shape}^{3D}(p)$$



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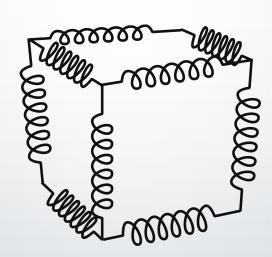








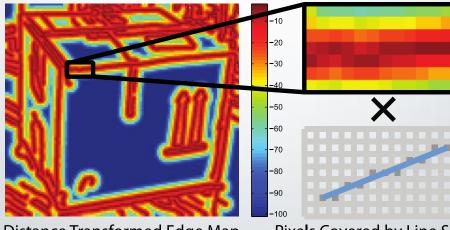
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Image



Distance Transformed Edge Map

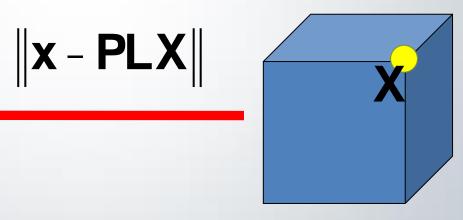
Pixels Covered by Line Segment

Dot-product is the Edge Term

$$S(I, p) = \sum_{i \in \mathcal{V}} w_i^H \cdot \text{HOG}(I, p_i) + \sum_{ij \in \mathcal{E}} w_{ij}^D \cdot \text{Displacement}^{2D}(p_i, p_j) + \sum_{ij \in \mathcal{E}} w_{ij}^E \cdot \text{Edge}(I, p_i, p_j) + w^S \cdot \text{Shape}^{3D}(p)$$

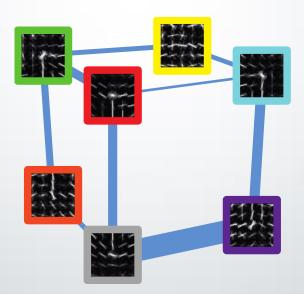


Image (2D)



Unit Cuboid (3D)

$$S(I, p) = \sum_{i \in \mathcal{V}} w_i^H \cdot \text{HOG}(I, p_i) + \sum_{ij \in \mathcal{E}} w_{ij}^D \cdot \text{Displacement}^{\text{2D}}(p_i, p_j)$$
$$+ \sum_{ij \in \mathcal{E}} w_{ij}^E \cdot \text{Edge}(I, p_i, p_j) + w^S \cdot \text{Shape}^{\text{3D}}(p)$$



Inference

Select initial cuboid configuration from 2D view



Step 1: Approximation by a tree (dynamic programming + distance transform).

$$S(I,p) = \sum_{i \in \mathcal{V}} w_i^H \cdot \mathrm{HOG}(I,p_i) + \sum_{ij \in \mathcal{T}} w_{ij}^D \cdot \mathrm{Displacement^{2D}}(p_i,p_j)$$
 Step 2: Local search (hill climbing or ICM).

Learning

Learn term weights of score function by supervised learning.

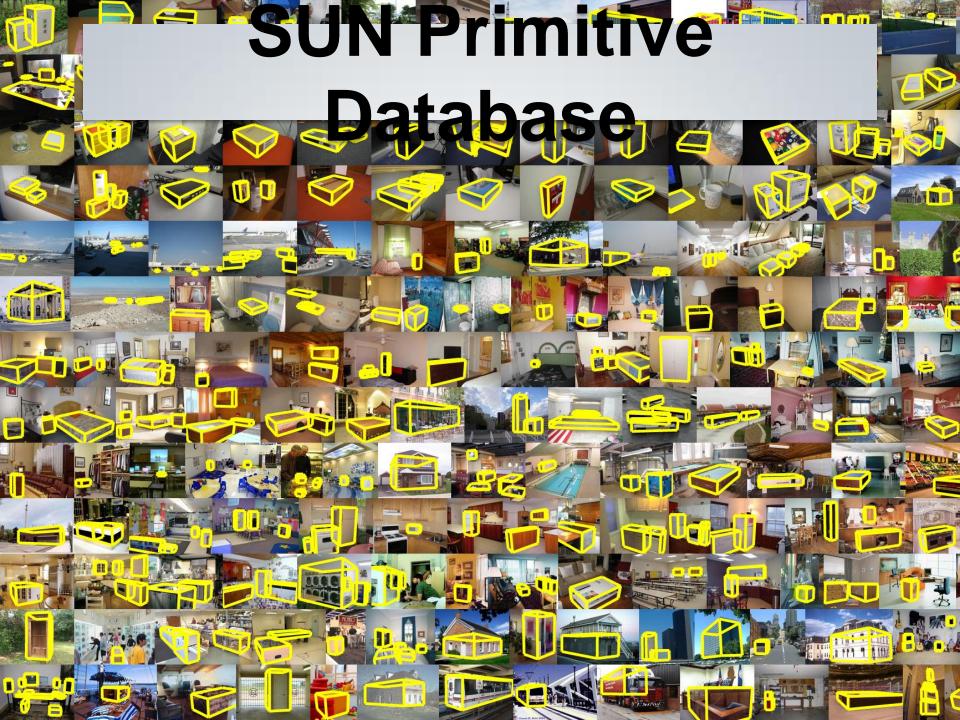
Annotate positive and negative corners and train a structural SVM

Supervised corner-location training with Structural SVM.

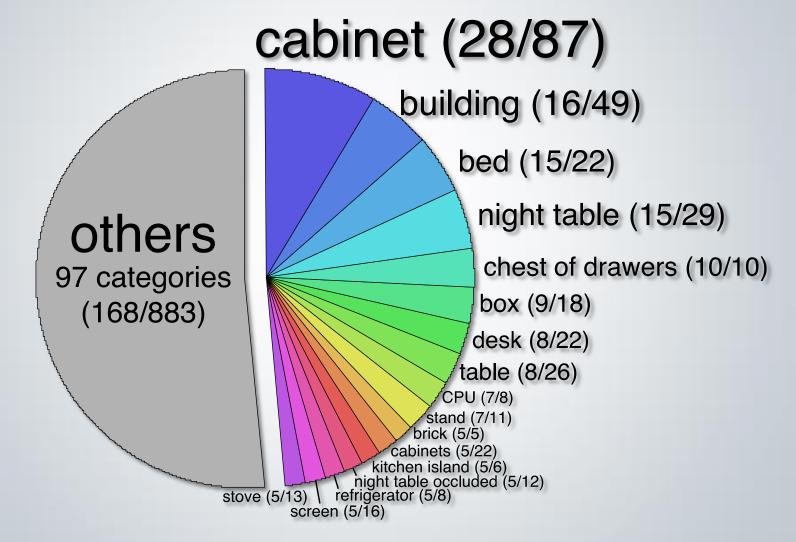
$$\min_{\beta,\xi \geq 0} \quad \frac{1}{2}\beta \cdot \beta + C \sum_{n} \xi_{n}$$

$$\forall n \in \text{pos} \quad \beta \cdot \Phi (I_{n}, p_{n}) \geq 1 - \xi_{n}$$

$$\forall n \in \text{neg}, \forall p \in P \quad \beta \cdot \Phi (I_{n}, p) \leq -1 + \xi_{n}$$

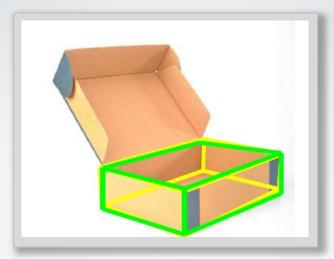


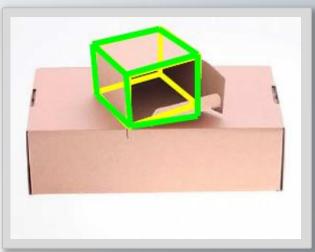
What objects are cuboids?

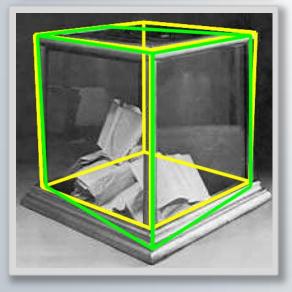


Easy Cases

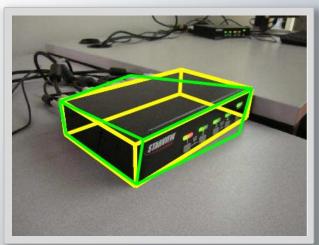












Detection Result







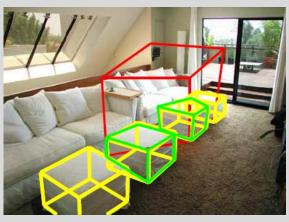




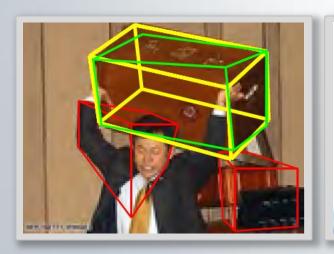


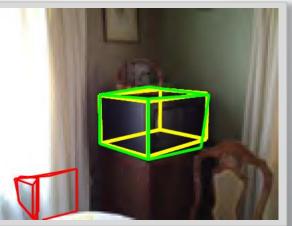
Multiple Instances

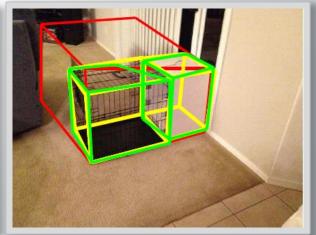












2D vs 3D









Ground truth



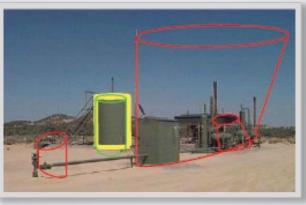
2D Detector

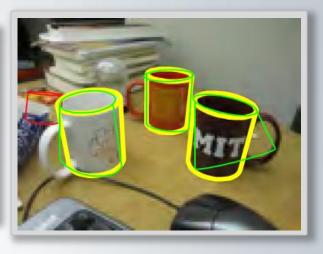


3D Detector

Cylinder Detection





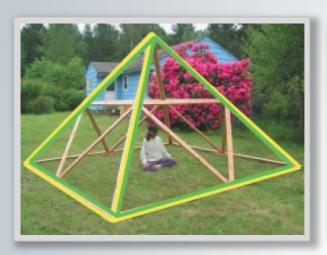


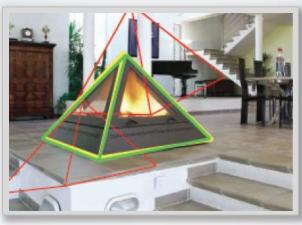


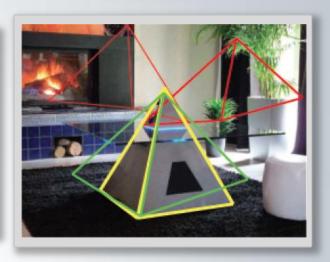




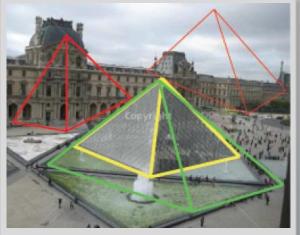
Pyramid Detection

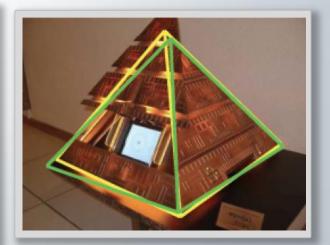




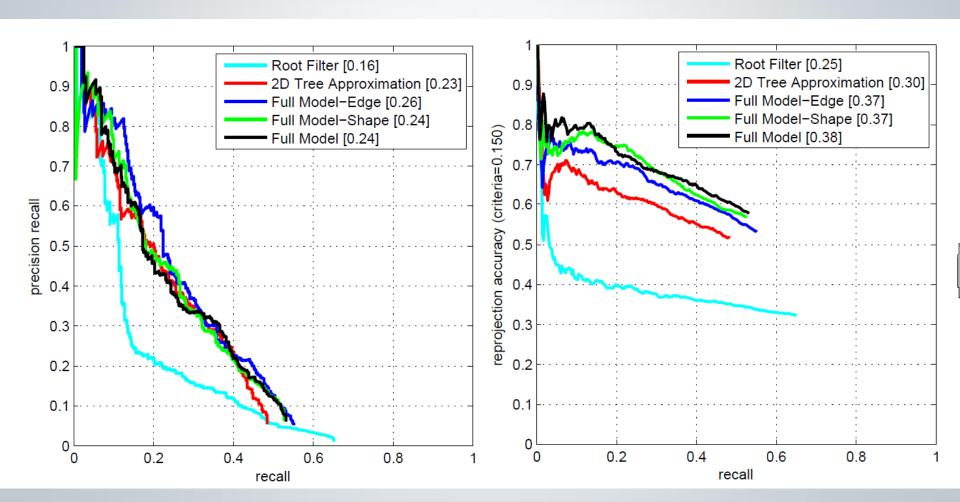








Quantitative Results



Thanks