

SemanticPaint

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Outline

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- 2 State of the Art
- 3 Pipeline
- 4 Results
- **5** Discussion and Outlook



Introduction



Introduction



State of the Art



Acquisition and Reconstruction



Scene Understanding



Pipeline



Voxel Oriented Patch features

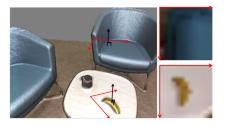


Figure: Colours shown in RGB for illustration purposes.

$$(\mathbf{p} - \mathbf{p}_i) \cdot (n)_i = 0$$

 $r \times r$, $r = 13px$ with $10\frac{mm}{pixel}$
CIELab
Rotated to dominant gradient direction



Streaming Random Forest



SRF - Reservoir Splitting

$$K = 10 \qquad \mathcal{R}_n = \left(\mathcal{T}_n = \left\{\begin{array}{c} \bullet & \bullet \\ \bullet & \bullet \\ \end{array}\right\}, m_n = 20 \right) \qquad P(l \mid \mathcal{T}_n)$$

$$\bar{\mathcal{T}}_n^{\mathrm{L}}(\theta_n) = \left\{\begin{array}{c} \bullet & \bullet \\ \bullet & \bullet \\ \end{array}\right\} \qquad P(l \mid \bar{\mathcal{T}}_n^{\mathrm{L}})$$

$$\bar{\mathcal{T}}_n^{\mathrm{R}}(\theta_n) = \left\{\begin{array}{c} \bullet & \bullet \\ \bullet & \bullet \\ \end{array}\right\} \qquad P(l \mid \bar{\mathcal{T}}_n^{\mathrm{R}})$$

$$m_n^{\mathrm{L}} = \left|\bar{\mathcal{T}}_n^{\mathrm{L}}\right| \frac{m_n}{K} = 6 \qquad m_n^{\mathrm{R}} = \left|\bar{\mathcal{T}}_n^{\mathrm{R}}\right| \frac{m_n}{K} = 14$$

$$\mathcal{R}_n^{\mathrm{L}} = \left(\mathcal{T}_n^{\mathrm{L}} = \left\{\begin{array}{c} \bullet & \bullet \\ \bullet & \bullet \\ \end{array}\right\}, m_n^{\mathrm{R}} = 14 \right)$$



Dynamic Conditional Random Field

$$P(\mathbf{x}|\mathbf{D}) = \prod_{i \in \mathcal{V}} \left(\psi_i(x_i) \prod_{j \in \mathcal{E}_i} \psi_{ij}(x_i, x_j) \right)$$
 (1)

$$E_t(\mathbf{x}) = \sum_{i \in \mathcal{V}} \left(\phi_i(x_i) + \sum_{j \in \mathcal{E}_i} \phi_{ij}(x_i, x_j) \right) + K$$
 (2)



CRF - User Interactions

Touching:

$$\phi_i(l) = \begin{cases} 0 & \text{if } l = l_S \\ \infty & \text{otherwise} \end{cases}$$
 (3)

Encircling:

$$\phi_i(l) = \begin{cases} \log P_E(fg|\mathbf{a}_i) & \text{if } l = \text{fg} \\ \log(1 - P_E(fg|\mathbf{a}_i)) & \text{if } l = \text{bg} \end{cases}$$
(4)



CRF - Predictions and Smoothnes

Predictions:

$$\phi_i(l) = -\log P_F(x_i = l|\mathbf{D}) \tag{5}$$

Smoothnes:

$$\phi_{ij}(x_i, x_j) = \theta_p e^{-||\mathbf{p}_i - \mathbf{p}_j||} + \theta_a e^{-||\mathbf{a}_i - \mathbf{a}_j||} + \theta_n e^{-||\mathbf{n}_i - \mathbf{n}_j||}$$
 (6)



Mean-Field Inference

 $P(\mathbf{x})$ approximated by $Q(\mathbf{x})$ under KL(Q||P):

$$Q_i^t(l) = \frac{1}{Z_i} e^{M_i(l)}, \ t = 1, \dots, T$$
 (7)

$$M_i(l) = \phi_i(l) + \sum_{l' \in \mathcal{L}} \sum_{j \in \mathcal{N}(i)} Q_j^{t-1}(l')\phi_{ij}(l, l')$$
(8)

Frame at time t initialized with:

$$\widetilde{Q}_{i}^{t}(x_{i}) = \gamma Q_{i}^{t-1}(x_{i}) + (1 - \gamma)P_{F}^{t-1}(x_{i} = l|\mathbf{D}), \ \gamma \in [0, 1]$$
 (9)



Results



Segmentation

Table: Segmentation Results

Component	LivingRoom	Bedroom	Kitchen	Desk	Average
User Interaction	99.35%	97.61%	96.09%	97.73%	97.7%
Forest Prediction	94.57%	88.31%	82.58%	90.29%	88.94%
Final Inference	96.26%	95.19%	90.69%	95.55%	94.42%



Features

Table: Feature Comparison

Feature	LivingRoom	Bedroom	Kitchen	Desk	Average
VOP	94.57%	88.31%	82.58%	90.29%	88.94%
\triangle RGB mean	80%	71.84%	76.29%	73.42%	75.39%
Depth Probe	77.54%	61.79%	84.9%	68.9%	73.06%
Color Probe	56.39%	65.68%	60.77%	60.74%	60.9%
SURF	43.74%	67.12%	57%	58.13%	56.5%
SPIN	58.77%	43.22%	48.41%	36.1%	46.63%



Streaming Random Forest

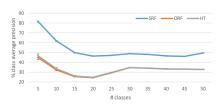


Figure: Average Precision

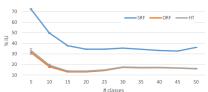


Figure: Intersection/Union

Data: 300 objects 51 classes full revolution 3 points of view

SRF - Streaming Random Forest ORF - Online Random Forest HT - Hoeffding Tree





Discussion and Outlook



Summary

- · customized models of 3D enviornments
- fully interactive
- online and real time
- no pretraining



Failures

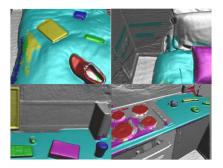


Figure: Failure cases.

- bleeding
- illumination change
- viewpoint change



Future Work

- class priors for different enviornments
- priors for class properties (vertical walls)
- discriminative geometrical features
- outdoor enviornments
- better scalability



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