

# Automated Place Detection Based on Coherent Segments

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## Introduction

- Problem definition
- Related work
- Related work
- General approach

## Method

- Region Adjacency Graphs
- Temporal RAG Tracking
- Coherency score
- Place Detection
- Segments Summary Graphs

## Experiments

## Conclusion

# Introduction

- ▶ Goal: Automated appearance-based place detection
- ▶ Place is a specific spatial unit or area
- ▶ Place detection is a prior step to
  - ▶ Place recognition
  - ▶ Topological mapping
  - ▶ Semantic scene understanding

# Introduction

- ▶ Appearance-based approach
  - ▶ Suitable for scene content analysis
  - ▶ Geometric or odometric data may not be available
- ▶ Challenges
  - ▶ Appearance variability
  - ▶ Indiscriminate boundaries

# Introduction

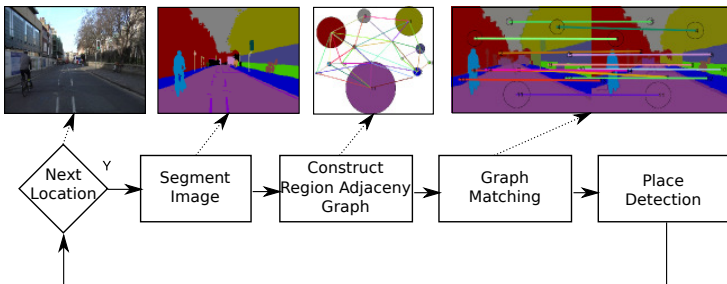
## ► Related work

- Partitioning of incoming sensory data based on similarity
- Clustering
- Feature types:
  - Global: Histograms, Census Transform, GIST ✗ Sensitive
  - Local: SIFT, SURF ✗ Low level, Matching
  - Hybrid: BoW, Bubble Space ✗ Low level
- Detecting transition regions (i.e. doors, passages, corridors) ✗  
Fails if transitions are not obvious

# Introduction

- ▶ Proposed approach
  - ▶ Visual segments
  - ▶ Smooth body or head motion assumption
  - ▶ Spatio-temporal coherence of visual segments
- ▶ Advantages:
  - ▶ More stable features
  - ▶ Segments Summary Graphs representation

# General Approach



# Region Adjacency Graphs



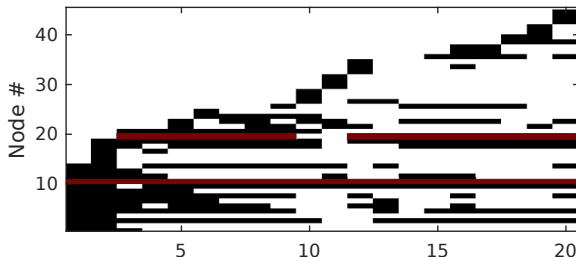
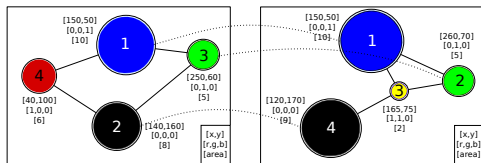
- ▶ Segments  $\Rightarrow$  Nodes
- ▶ Spatial relations  $\Rightarrow$  Edges
- ▶ Graph based segmentation method [Felzenszwalb, Huttenlocher, 2004]

- ▶ Nodes
  - ▶ Color
  - ▶ Position
  - ▶ Size
- ▶ Edges
  - ▶ Mean color difference



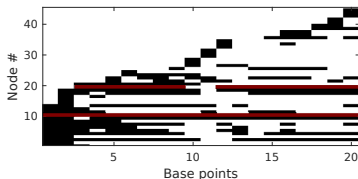
# Temporal RAG Tracking

- ▶ Matching consecutive RAGs:
  - ▶ Cost matrix  $C^{kl}$  with  $c_{ij} = \delta(s(\mathcal{N}_i^k), s(\mathcal{N}_j^l))$
  - ▶ Optimal match by Hungarian method
  - ▶ Remove nodes with matching cost  $> \tau_m$
- ▶ Nonmatched nodes - Matching via backtrack



# Coherency score calculation

- ▶ Coherency over temporal window
- ▶ Factors:
  - ▶  $\tau_w$  - window size
  - ▶ # appearing nodes
  - ▶ # disappearing nodes
  - ▶ node weights -  $\rho_i^l$

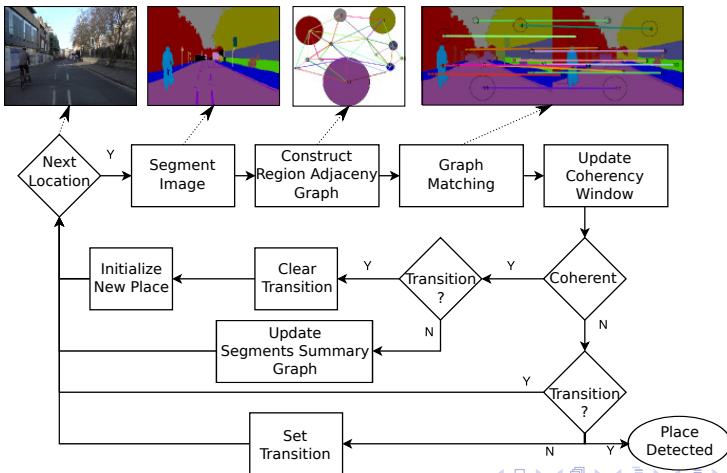


where

$$\varphi^k = 1 - \sum_{l=k-\tau_w}^k \sum_{i=1}^{|n^l|} \rho_i^l (a_i^l + b_i^l) \quad (1)$$

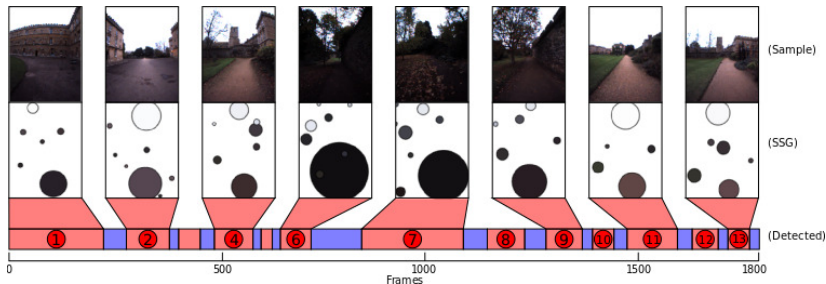
$$\rho_i^l \propto s_3(N_i^l) \times \sum_{k=l-\tau_w}^l M_{ki} > 0 \quad (2)$$

# Place Detection



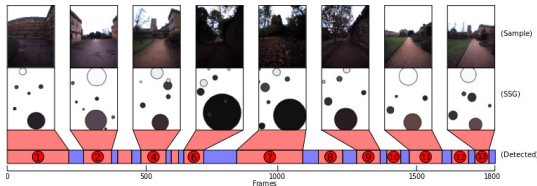
# Segments Summary Graphs

- ▶ Contains coherent segments only
- ▶ Encodes spatial relations

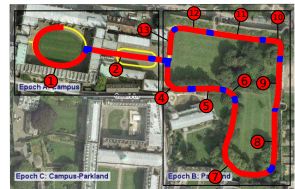


# Outdoor experiments

- ▶ New College dataset
- ▶ 1800 basepoints 550 m
- ▶ Contains gradual changes

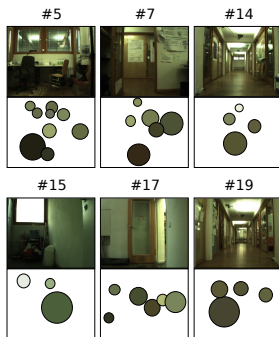


## New College Map

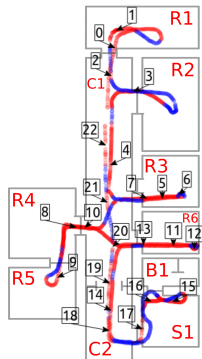


# Indoor experiments

- Freiburg (Fr), Saarbrücken (Sa) and Ljubljana sites of COLD Dataset



Freiburg site



# Conclusion

## Segments based Place Detection

- ▶ Stable under wide range of view-points and dynamical changes compared to low-level descriptors
- ▶ Reliable place detection
- ▶ SSG enables semantic content analysis

## Future work

- ▶ Use semantic segmentation
- ▶ Use SSG for place recognition and hierarchical place representation