

Vessel Classification using Graph Kernel

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Outline

- Introduction
 - Vessel Satellite Images
 - Graph
 - Graph Kernel
- Convert Image to Graph
- Graph Kernels
 - Shortest Path Graph Kernel
 - Unordered Neighboring Graph Kernel
- Results



Outline

• Introduction

- Vessel Satellite Images



Vessel Satellite Images

• Four Classes, each class has 200 images





Barge

Cargo



Container





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Graph

- A graph G is a set of vertices V and edges E,
 where E ⊂ V²
- A graph G may have labels on vertices and/or edges
- The adjacency matrix **A** of G is defined as $[\mathbf{A}_{ij}] = \begin{cases} 1 & if(v_i, vj) \in \mathbf{E} \\ 0 & otherwise \end{cases}$



Labelled Undirected Graphs

vertices





Labelled Undirected Graphs

vertices

edges







Labelled Undirected Graphs





Labelled Undirected Graphs





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Graph Kernel

- To Calculate the similarities between two graphs in polynomial time
 - Random Walk Kernel
 - Compare all walks in two graphs **G** and **G'**
 - Shortest Path Kernel
 - Compare all pairs shortest paths for **G** and **G'** via Floyd-Warshall
 - Subtree Kernel
 - Compare subtree-like patterns in two graphs **G** and **G'**
 - Cyclic Pattern Kernel
 - Compare simple cycles in two graphs **G** and **G'**
 - Graphlet Kernel
 - Count subgraphs of limited size K in G and G'



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Convert Image to Graph

- Cut image into tiles
- Feature Extraction on each tile
 - Each tile is represented by a histogram
 - Local Binary Pattern (LBP)
 - Scale Invariant Feature Transform (SIFT)
- Eliminate uninteresting tiles
 - Set a threshold for the mean value of histogram
 - 0.175 in our case
- Connect the interesting tiles to form a graph
 - Each tile is a node
 - Each node connects with its vertical and horizontal neighbors



Cut Image into Tiles





Cut Image into Tiles







Local Binary Pattern

- Each Pixel compares with its eight neighbors
 - Write 1 if the pixel's value greater then neighbor
 - Write 0 otherwise
 - Form a 8-digit binary number(256 patterns)
- Compute histogram of the frequency of each pattern
 - The image is represented by a 256-dimension vector



Feature Extraction using LBP







Feature Extraction using LBP



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Feature Extraction using LBP

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Feature Extraction using LBP

Feature Extraction using SIFT

- SIFT Frame
 - (X,Y) coordinates
 - Scale
 - Orientation
- SIFT Descriptor
 - Spatial histogram of image gradients
 - 128-dimension vector
- If a image tile contains multiple SIFT key point
 - Average them to get a mean value 128-dimension vector ²¹

Feature Extraction using SIFT

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- Convert graph to all pair shortest path graph
 - Floyd-Warshall Algorithm

Floyd-Warshall

Original Graph

Shortest Path Graph

Floyd-Warshall

Original Graph

Shortest Path Graph

•
$$K_{sp(G,G')} = \sum_{e \in E} \sum_{e' \in E'} K_{walk}(e,e')$$

- $K_{sp(G,G')} = \sum_{e \in E} \sum_{e' \in E'} K_{walk}(e,e')$
- $K_{walk}(e, e') = Knode(u, u') \cdot K_{edge}(e, e') \cdot Knode(v, v')$

- $K_{sp(G,G')} = \sum_{e \in E} \sum_{e' \in E'} K_{walk}(e,e')$
- $K_{walk}(e, e') = Knode(u, u') \cdot K_{edge}(e, e') \cdot Knode(v, v')$
- K_{node} is a valid kernel function for comparing two vertices
 - Gaussian kernel
 - Intersection kernel
- K_{edge} is a valid kernel function for comparing two edges

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Unordered Neighboring Graph Kernel

 Given node v, let N(v) be the set contains all neighbors of node v

$$k_{UNORD}(G,G') = \sum_{v \in V} \sum_{v' \in V'} k_{node}(v,v') * (1 + \sum_{n \in N(v)} \sum_{n' \in N(v')} K_{node}(n,n'))$$

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Data sets

- BCCT200-original
 Unchanged images
- BCCT200-rotated
 - images are rotated
- BCCT200-cropped
 - images are rotated and cropped
- BCCT200-resized
 - images are rotated, cropped, aligned and resized

SVM results using LBP feature

| Data | SP-Gaussian | SP-Intersect | Unord-Gaussian | Unord-Intersect |
|------------------|-------------|--------------|----------------|-----------------|
| BCCT200-original | 79.5 | 75.8 | 76.7 | 75.1 |
| BCCT200-rotated | 81.3 | 78.2 | 80.3 | 75.8 |
| BCCT200-cropped | 83.8 | 80.3 | 80.9 | 77.7 |
| BCCT200-resized | 88.8 | 86.9 | 89.2 | 86.8 |

SVM results using SIFT feature

| Data | SP-Gaussian | SP-Intersect | Unord-Gaussian | Unord-Intersect |
|------------------|-------------|--------------|----------------|-----------------|
| BCCT200-original | 87.8 | 84.5 | 88.9 | 82.7 |
| BCCT200-rotated | 87.3 | 84.2 | 87.3 | 83.1 |
| BCCT200-cropped | 88.7 | 85.1 | 87.7 | 83.9 |
| BCCT200-resized | 91.8 | 90.3 | 92.4 | 91.1 |

Thanks! Questions?