Image Tag Assignment, Refinement and Retrieval

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ABSTRACT

This tutorial focuses on challenges and solutions for content-based image annotation and retrieval in the context of online image sharing and tagging. We present a unified review on three closely linked problems, i.e., tag assignment, tag refinement, and tag-based image retrieval. We introduce a taxonomy to structure the growing literature, understand the ingredients of the main works, clarify their connections and differences, and recognize their merits and limitations. Moreover, we present an open-source testbed, with training sets of varying sizes and three test datasets, to evaluate methods of varied learning complexity. A selected set of eleven representative works have been implemented and evaluated. During the tutorial we provide a practice session for hands-on experience with the methods, software, and datasets. For repeatable experiments all data and code are online at http://www.micc.unifi.it/tagsurvey.

Categories and Subject Descriptors

H.3.3 [INFORMATION STORAGE AND RETRIEVAL]: Information Search and Retrieval; H.3.1 [INFORMATION STORAGE AND RETRIEVAL]: Content Analysis and Indexing—Indexing Methods

General Terms

Algorithms, Experimentation, Performance

Keywords

Content-based image retrieval, social tagging, tag relevance, tag assignment, tag refinement, tag retrieval

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Table 1: Methods evaluated in this tutorial. The media column characterizes what essential information a specific method exploits, while the learning SemanticField depicts how such information is exploited.

<table>
<thead>
<tr>
<th>Method</th>
<th>Media</th>
<th>Learning</th>
<th>Code</th>
</tr>
</thead>
<tbody>
<tr>
<td>TagCooccur [9]</td>
<td>tag</td>
<td>Instance based</td>
<td>Python</td>
</tr>
<tr>
<td>TagRanking [6]</td>
<td>tag + image</td>
<td>Instance based</td>
<td>C + Python</td>
</tr>
<tr>
<td>KNN [7]</td>
<td>tag + image</td>
<td>Instance based</td>
<td>C + Python</td>
</tr>
<tr>
<td>TagVote [4]</td>
<td>tag + image</td>
<td>Instance based</td>
<td>Python</td>
</tr>
<tr>
<td>TagCooccur+ [4]</td>
<td>tag + image</td>
<td>Instance based</td>
<td>Python</td>
</tr>
<tr>
<td>TagProp [2]</td>
<td>tag + image</td>
<td>Model based</td>
<td>C + Matlab + Python</td>
</tr>
<tr>
<td>TagFeature [1]</td>
<td>tag + image</td>
<td>Model based</td>
<td>C + Python</td>
</tr>
<tr>
<td>RelExample [3]</td>
<td>tag + image</td>
<td>Model based</td>
<td>C + Python</td>
</tr>
<tr>
<td>RobustPCA [10]</td>
<td>tag + image</td>
<td>Transduction based</td>
<td>C + Matlab + Python</td>
</tr>
<tr>
<td>TensorAnalysis [8]</td>
<td>tag + image + user</td>
<td>Transduction based</td>
<td>–1</td>
</tr>
</tbody>
</table>

by various research groups. A selected set of eleven representative works, i.e., SemanticField [11], TagRanking [6], KNN [7], TagVote [4], TagProp [2], TagCooccur [9], TagCooccur+ [4], TagFeature [1], RelExample [3], RobustPCA [10], TensorAnalysis [8], have been implemented and evaluated on the test bed for tag assignment, refinement, and/or retrieval. An overview of the methods is given in Table 1. The interested reader is referred to [5] for a comprehensive comparison between these methods.

During the tutorial, we also provide a practice session for hands on experience with the methods, software, and datasets. For each method a front-end pipeline is implemented, allowing users to conduct tag relevance learning from scratch, obtain tag ranks and image ranks accordingly, and report multiple performance metrics including image-centric Mean image Average Precision (MiAP), tag-centric Mean Average Precision (MAP), and Normalized Discounted Cumulative Gain (NDCG). In addition, Python wrappers for C and Matlab code are given for the ease of cross-platform use.

We conclude the course with our perspective on the many challenges and opportunities ahead for the multimedia community.


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2. REFERENCES


\footnote{1}Results provided by the method developers.