Web Image Gathering with Region-based Bag-of-features and Multiple Instance Learning

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Keiji Yanai

The University of Electro-Communications, Tokyo, Japan

1. Objective & Background
2. Related Work
3. System & Methods
4. Experimental Results

1. Objective & Background

Background

Web is the largest image DB.
It is also a very noisy DB.

- To remove noise, image analysis is needed.
- Since 2001, we have been working on Web Image Gathering with **image analysis**
  - Keiji Yanai: Image Collector: An Image-Gathering System from the World-Wide Web Employing Keyword-based Search Engines, ICME 2001, Tokyo, Japan, pp.704-707 (2001/08). (ACMMM 2003,..)
  - Non-interactive. No feedback. Fully-automatic.
  - To gather **visual knowledge of many concepts for object recognition from the Web**

Objective of this paper

- **Import region-based bag-of-features to our Web image "gathering" system**
  [Yanai et al. ICME01, ACM MM03, ACM MIR 05, ICME08]

new combination !

【Image representation】
region-based bag-of-features
[Ravinovich et al. ICCV 07]

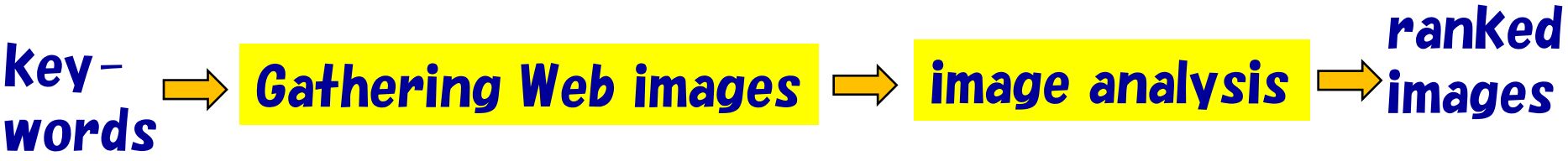
【Classifier】
mi-SVM (multiple instance learning)
[Andrew et al. NIPS 03]

2. Related Work

General Framework: Web image search + Object Recognition Technique

- Firstly, gather images from the Web using Web (image) search engines such as Google, Ask.com and MSN search by providing given keywords.

- Secondly, re-rank the results from the Web search engines with object/scene recognition methods

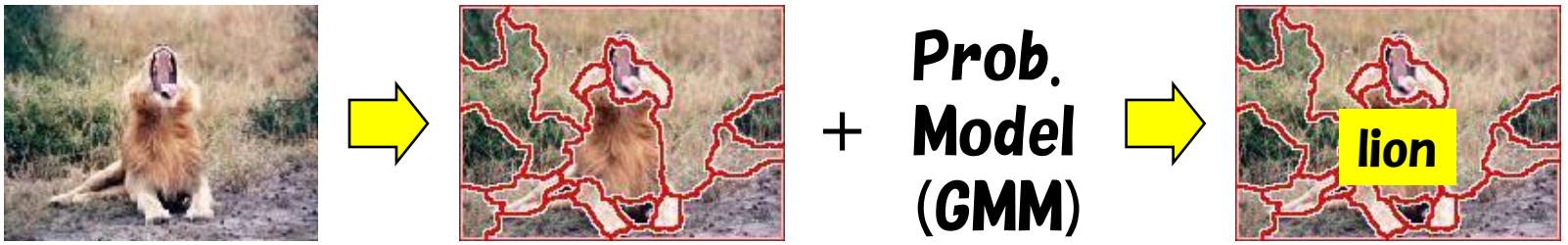


Literature: Web image search + Object Recognition Technique

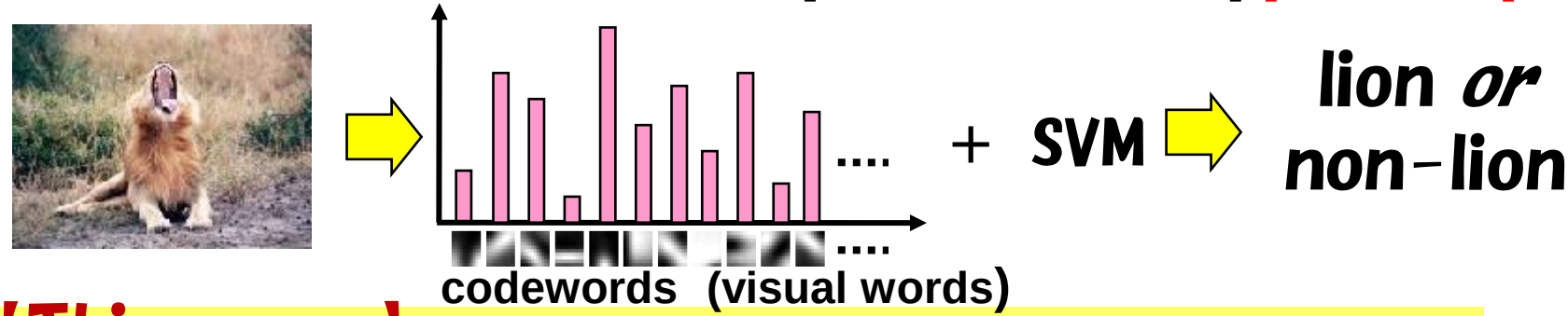
- Color histogram + k-means [Yanai ICME01]
- Color signature + EMD + k-NN [Yanai ACM MM03]
- Constellation model + RANSAC [Fergus ICCV04]
- JSEG + GMM (image-word translation model) [Yanai & Barnard ACM MIR 05]
- Bag-of-features (BoF)+ pLSA [Fergus ECCV05]
- Bag-of-features + HDP(Hierarchical Dirichlet Process) (OPTIMOL) [Li CVPR07]
- Bag-of-features + SVM [ICCV Schroff 07] [Yanai 07]
- 【This paper】
  JSEG + region-based bag-of-features + mi-SVM (multiple instance learning)

Literature: Web image search + Object Recognition Technique

- **JSEG + GMM (image-word translation model)**
  [Yanai & Barnard ACM MIR 05]



- **Bag-of-features + SVM** *[ICCV Schroff 07]* *[Yanai 07]*



- **[This paper]**
  JSEG + region-based bag-of-features + mi-SVM (multiple instance learning)

Contribution of this paper

- Import region-based bag-of-features to our Web image "gathering" system

【Image representation】
region-based bag-of-features
[Ravinovich et al. ICCV 07]

【Classifier】
mi-SVM (multiple instance learning)
[Andrew et al. NIPS 03]

3. Methods

Basic framework of our system [Yanai ICME01]～

Collection stage

Unchanged since [ICME01]

Gather image and HTML files using Web search engines.

Select ***pseudo-training images*** by **HTML analysis**

Selection Stage

Use supervised object rec. methods with pseudo-training images

Train a classifier and rank images based on estimated relevancy

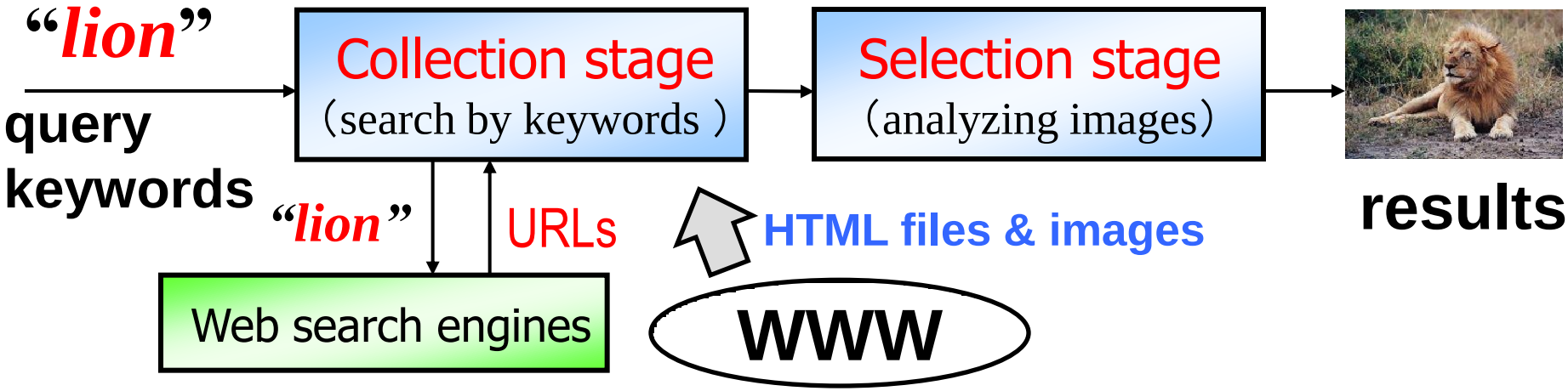
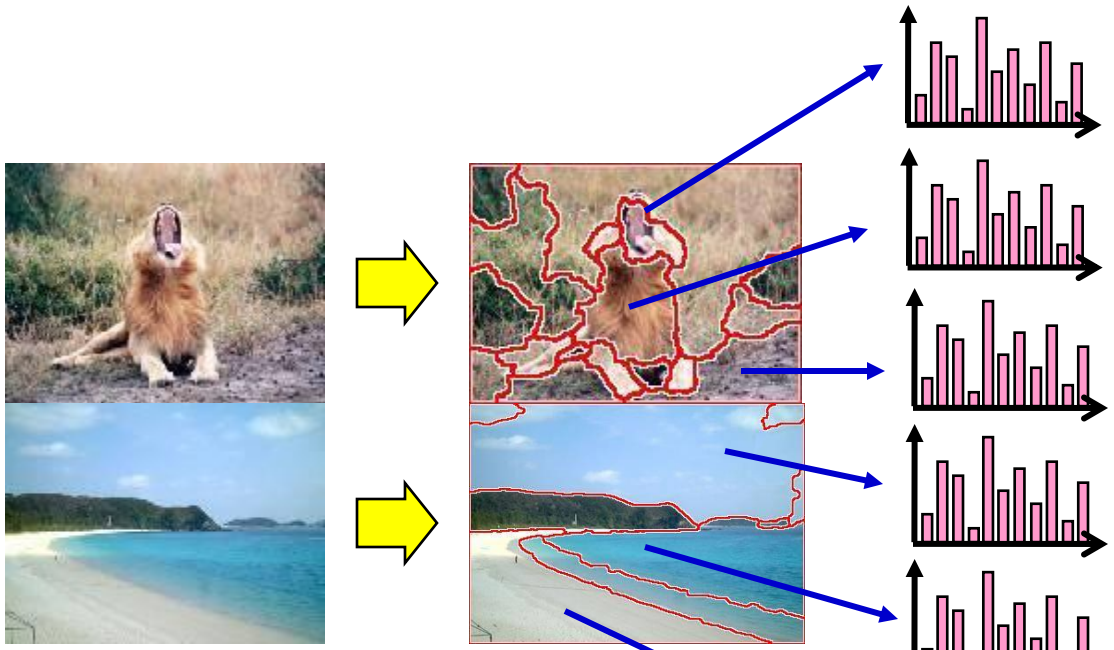


Image features

- **Divide each image into regions by JSEG**
  (8 regions on the average)

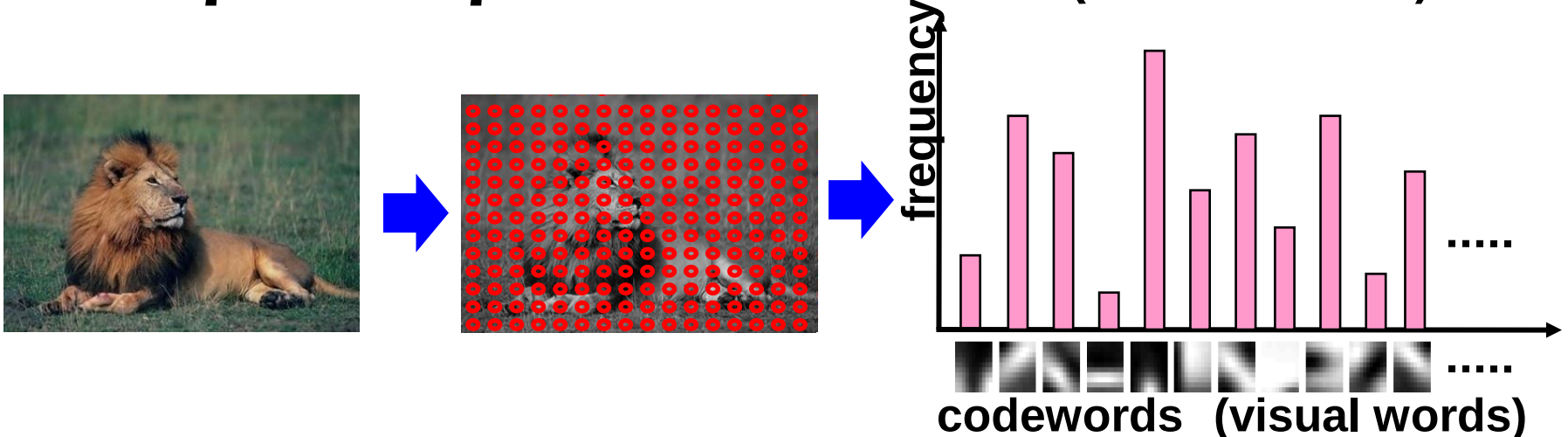


Segmentation (JSEG)

Bag-of-features (BoF) histograms (1000-dim)

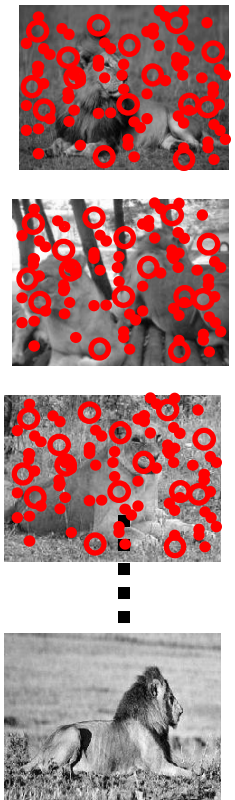
[image representation] Bag-of-features

- Represent an image as sets of features
  1. Densely-sample points along regular grids
  2. Represent local patterns around sampled points with SIFT descriptor
  3. Vector-quantize SIFT vectors based on pre-computed visual words (codewords)



How to obtain visual words

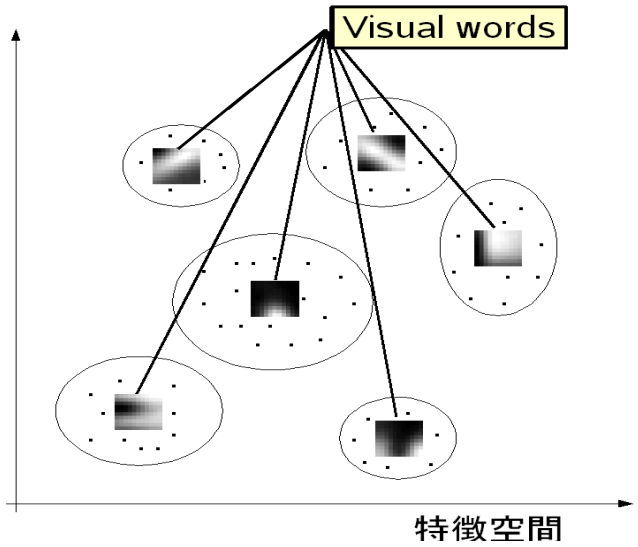
- Extract many SIFT vectors from positive and negative training samples
- Perform k-means clustering



center of clusters are “visual words”.

v_i v_i v_i v_i v_i v_i v_i v_i v_i

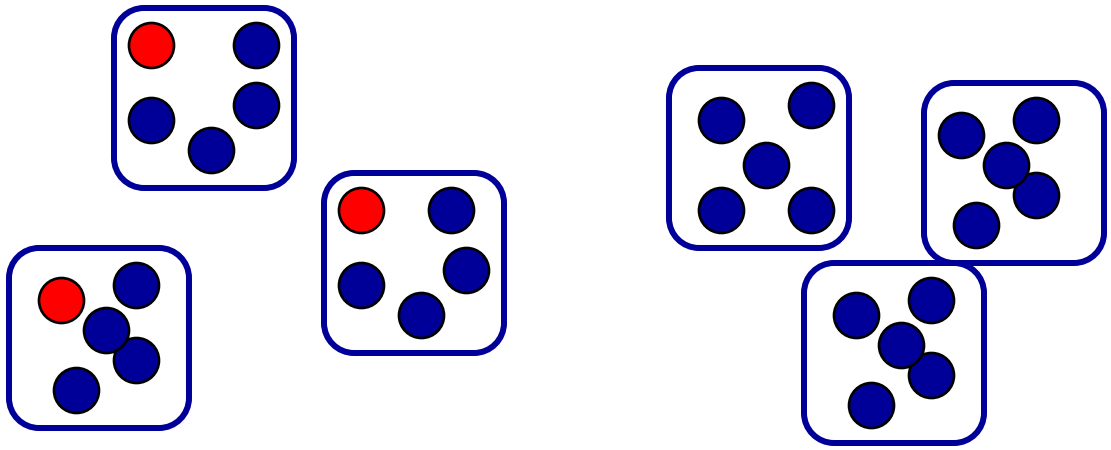
SIFT vectors



“Visual words” are representative local patterns.

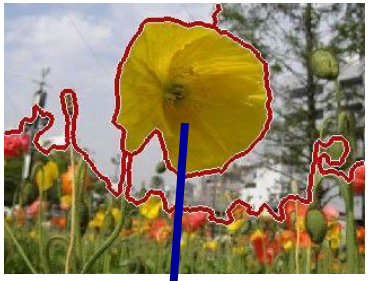
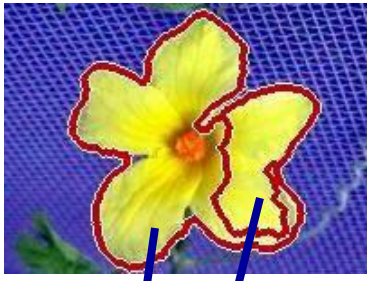
Multiple Instance Setting

- Positive bags / Negative bags

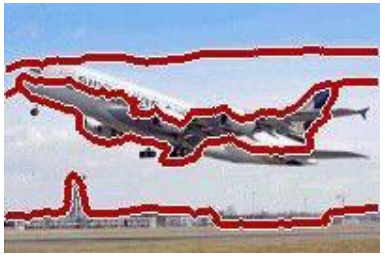


positive ins. (foreground)

negative ins. (background)

Positive instances of "flower"

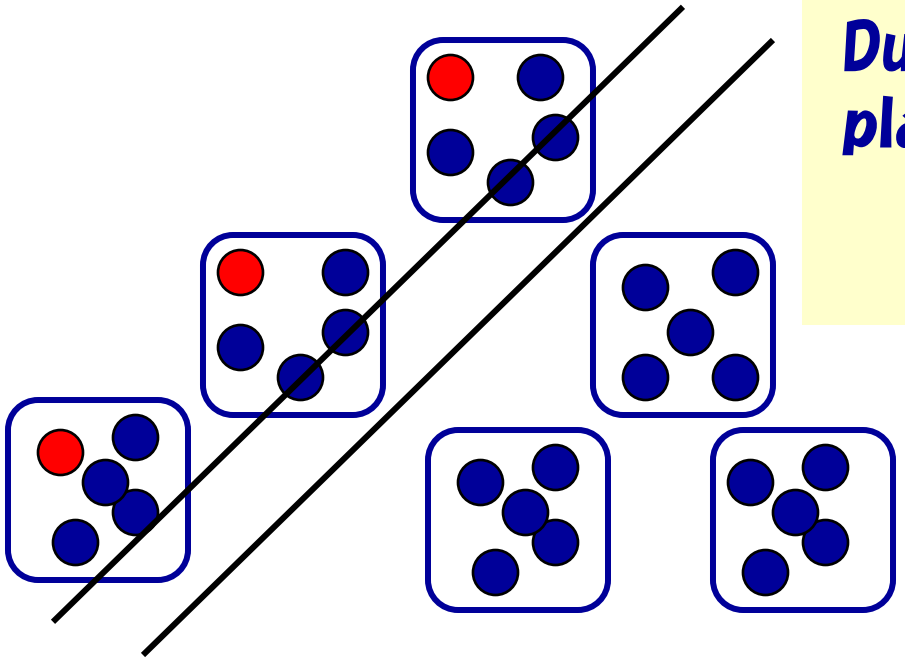
The rest of regions are negative regions.

pseudo-training images

random images

mi-SVM [Andrew et al. NIPS 03]

- Apply soft-margin SVM iteratively
  - Training → classifying → training → classifying → ……



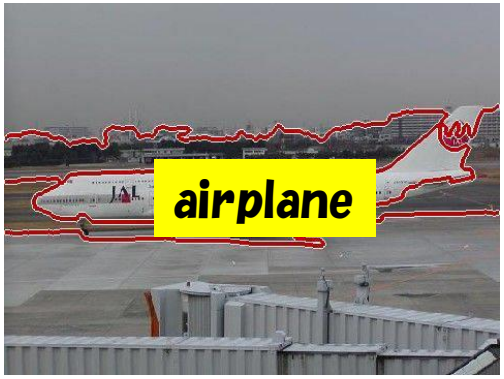
During the iteration, the hyper-plane is approaching the optimal plane to discriminate positive instances from negative ones.

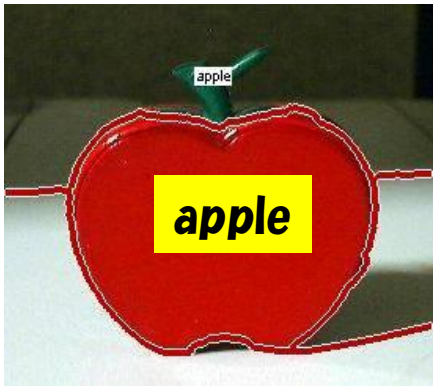
positive ins. (foreground)

negative ins. (background)

Final Image Re-ranking

- Regard the best SVM output score of the regions within an image as the score of the image
  - An image having one positive region at least is a positive image !
- Rank images based on the scores





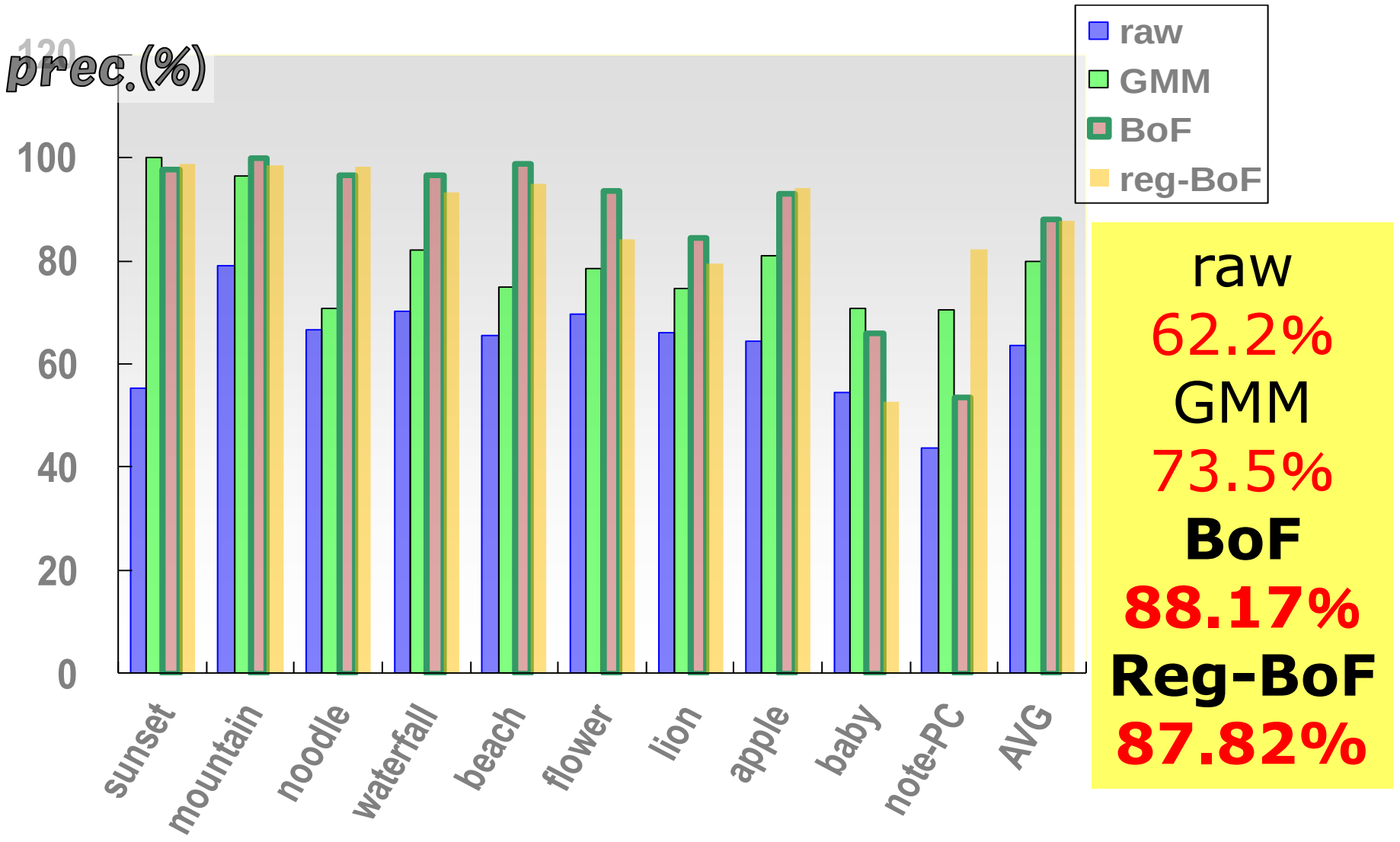
4. Experimental results

Experiments for 10+5 words

- sunset, mountain, waterfall, beach, (4scenes)
  noodle, flower, lion, apple, baby, laptop-PC, (6objects)
  airplane, guitar, leopard, motorbike, watch (5objects)

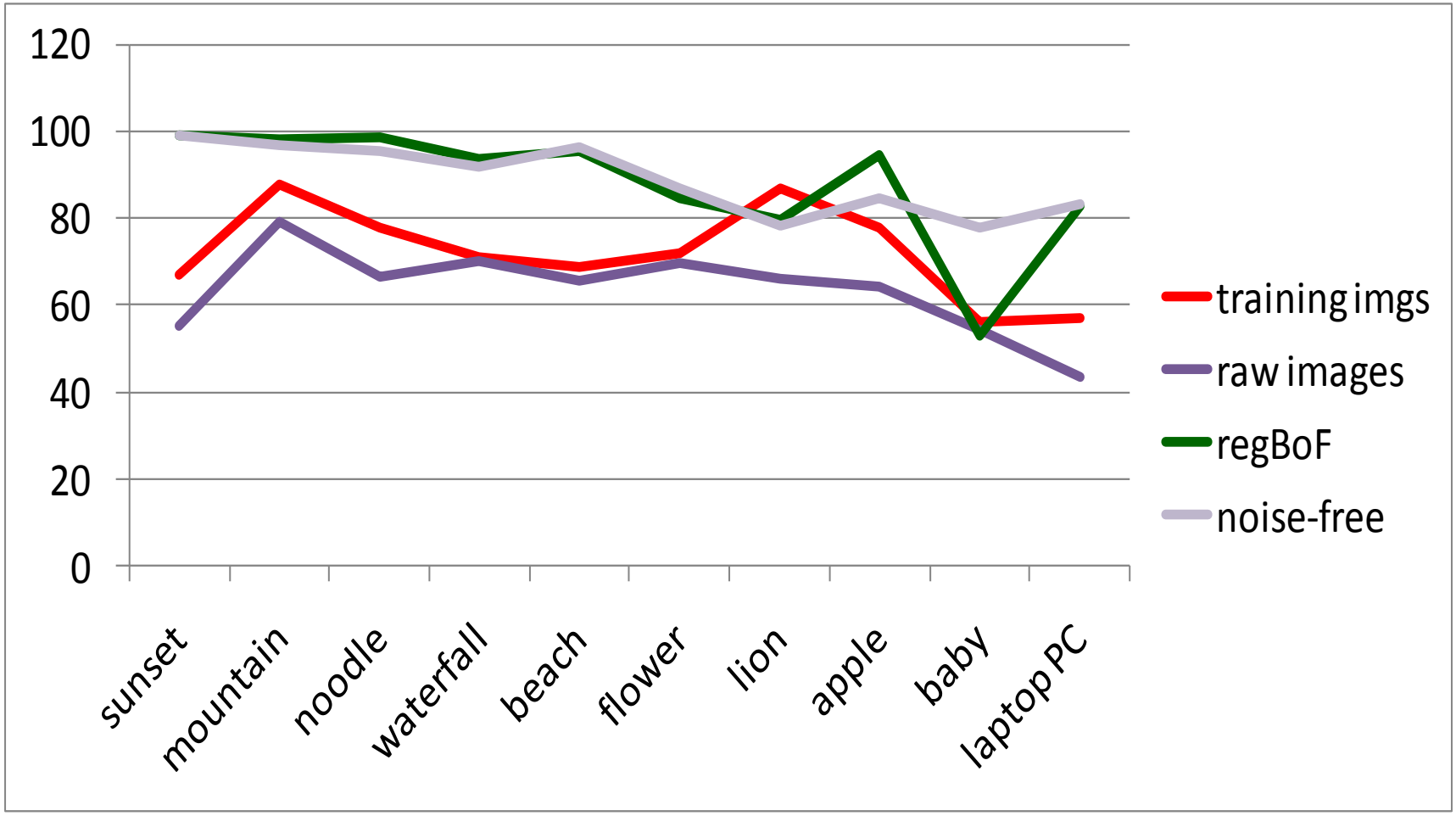
- Method:
  [raw data] raw (only HTML analysis) 39,143 images for 15 words
  [baseline1] GMM-based region probabilistic model [ACM MIR05]
  [baseline2] BoF + SVM
  [proposed] region-based BoF + SVM
- Evaluation: precision at 15% recall
  the same as [ICCV Schroff 07]

Comparison of 4 methods (raw, GMM, BoK, reg-BoF)



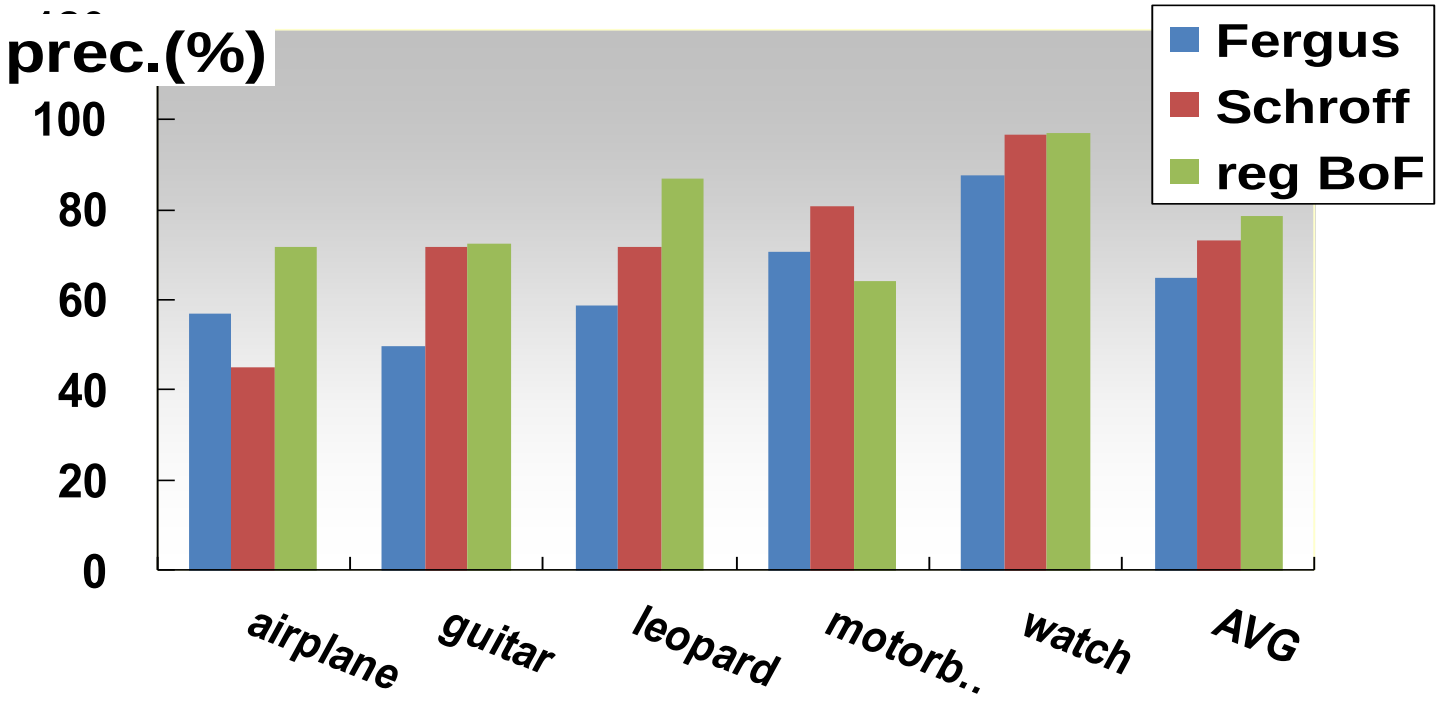
raw
62.2%
GMM
73.5%
BoF
88.17%
Reg-BoF
87.82%

Pseudo-training image sets and results by perfect training set (noise-free)



Comparison with related work

- [Fergus ICCV05] Bag-of-features + pLSA
- [Schroff ICCV07] Bag-of-features + SVM
- **[new] Region-based BoF + mi-SVM**



Many result images

- Laptop-PC (positive and negative)
- Mountain
- Waterfall
- Flower
- Airplane
- ???
- As by-products, we can estimate representative regions of images. (different from standard BoF)

Conclusion

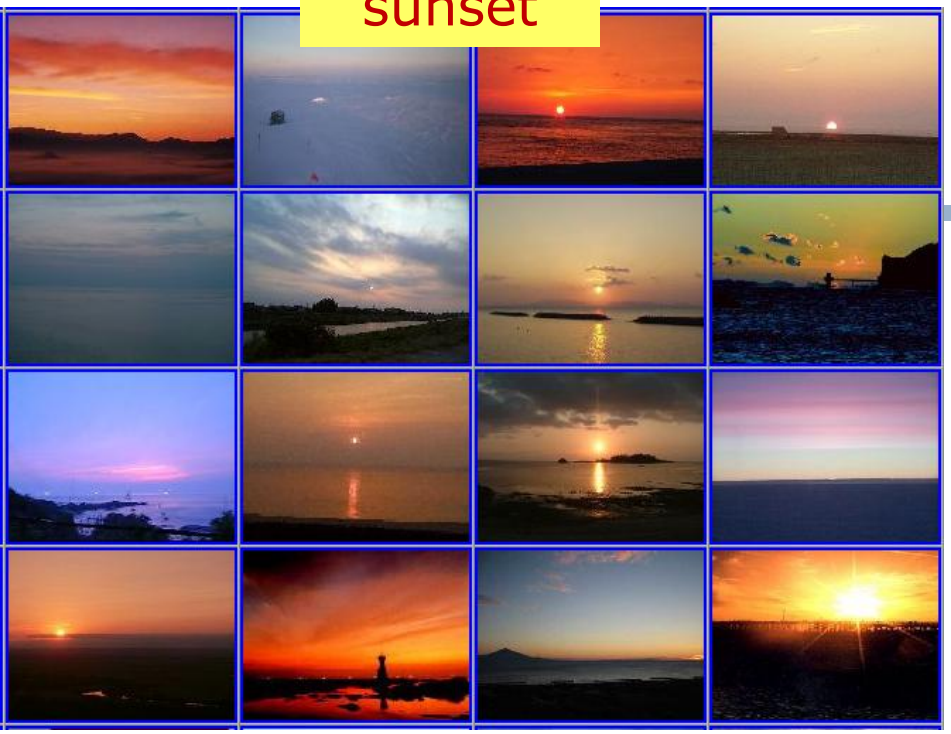
- Import region-based bag-of-features (BoF) and mi-SVM into the Web image gathering task.
  - In spite of noisy training data, the proposed method worked well.
  - It was especially effective for object concepts.

Future work

- Large-scale experiments
  - More than concept for 1000 concepts
- Improve the text analysis part to obtain more accurate psudo-training samples
  - Use co-occurrency of tags
  - Use taxonomy dic. (Wordnet, Wikipedia)


Thank you!

"sunset"



"waterfall"

Rejected "sunset"

Rejected "waterfall"


“Chinese noodle”


“notebook PC”


“lion”


“baby”