



THE UNIVERSITY OF TOKYO

MIL at ImageCLEF 2013 Personal Photo Retrieval

Masaru Mizuochi, Takayuki Higuchi,
Chie Kamada, and Tatsuya Harada

Machine Intelligence Laboratory, The University of Tokyo

Machine Intelligence Laboratory

Subtask2: Personal Photo Retrieval

The system which can help users to retrieve images from a lot of personal photo collections using browsing data.



User's browsing data = actual browsing work



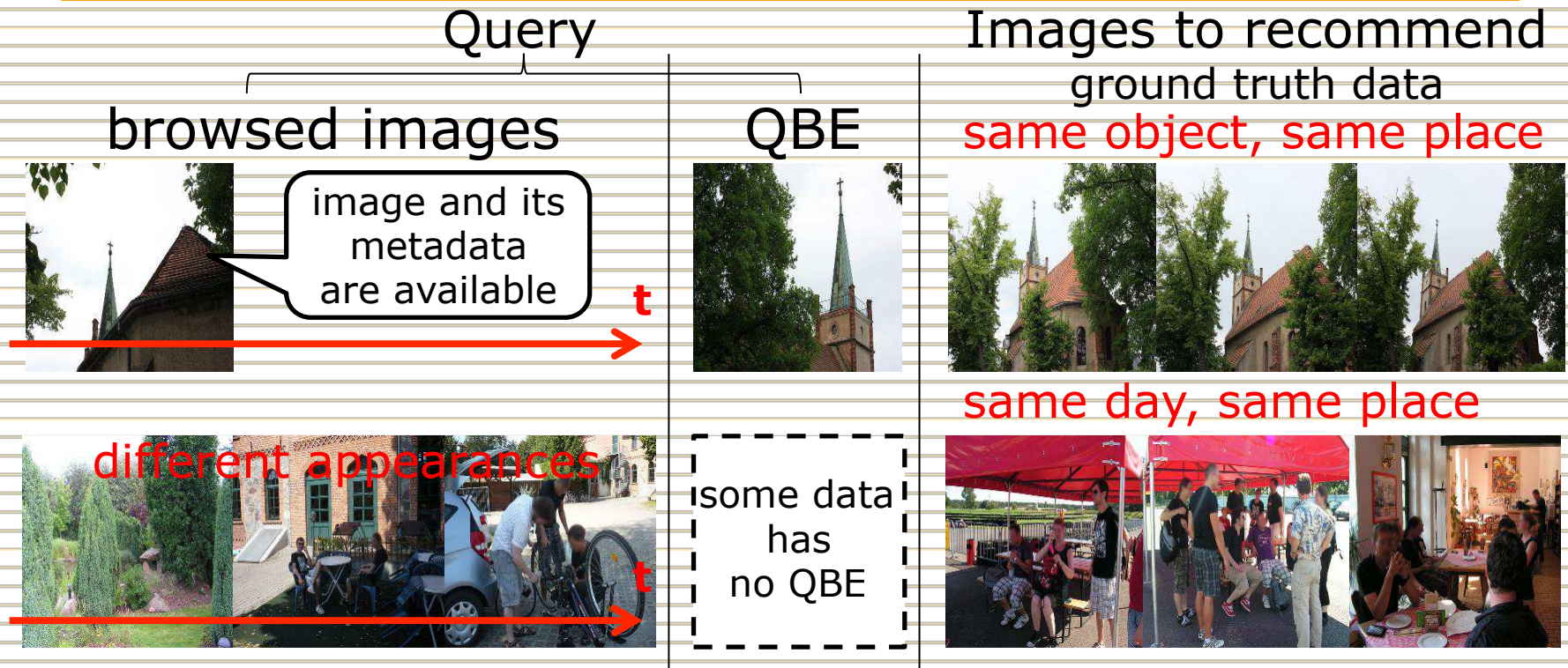
Query By Example
= User thinks best

Recommendation

"Are you looking for these images?"

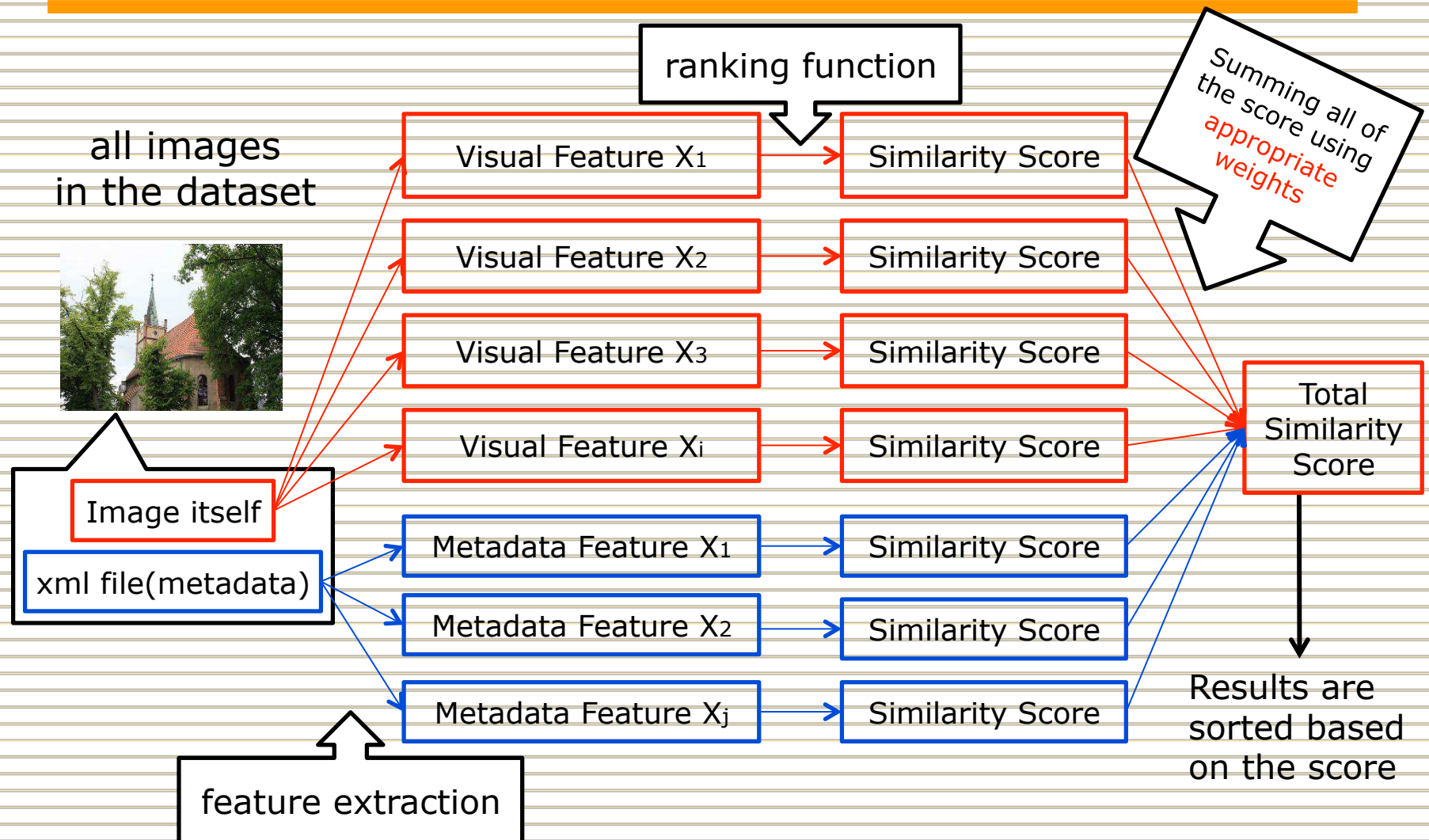


Motivation



- Task: estimating a topic from few query data and retrieve images which have the topic

General Photo Retrieval



Summing scores using appropriate weights

□ Relevance Feedback: A Power Tool for Interactive Content-Based Image Retrieval [Y. Rui et al., 1998]

- Learning with SVM classifier
- Several visual descriptor
- Similarity score is obtained by combining the scores of each feature with relevance feedback

- subjectivity of human's perception
- dynamically update weights

reviewed as correct



Calculate Similarity Score

- Learning to Rank for Content-Based Image Retrieval [F. Faria et al., 2010]
 - Learning a ranking by using the multi-stage evaluation by user

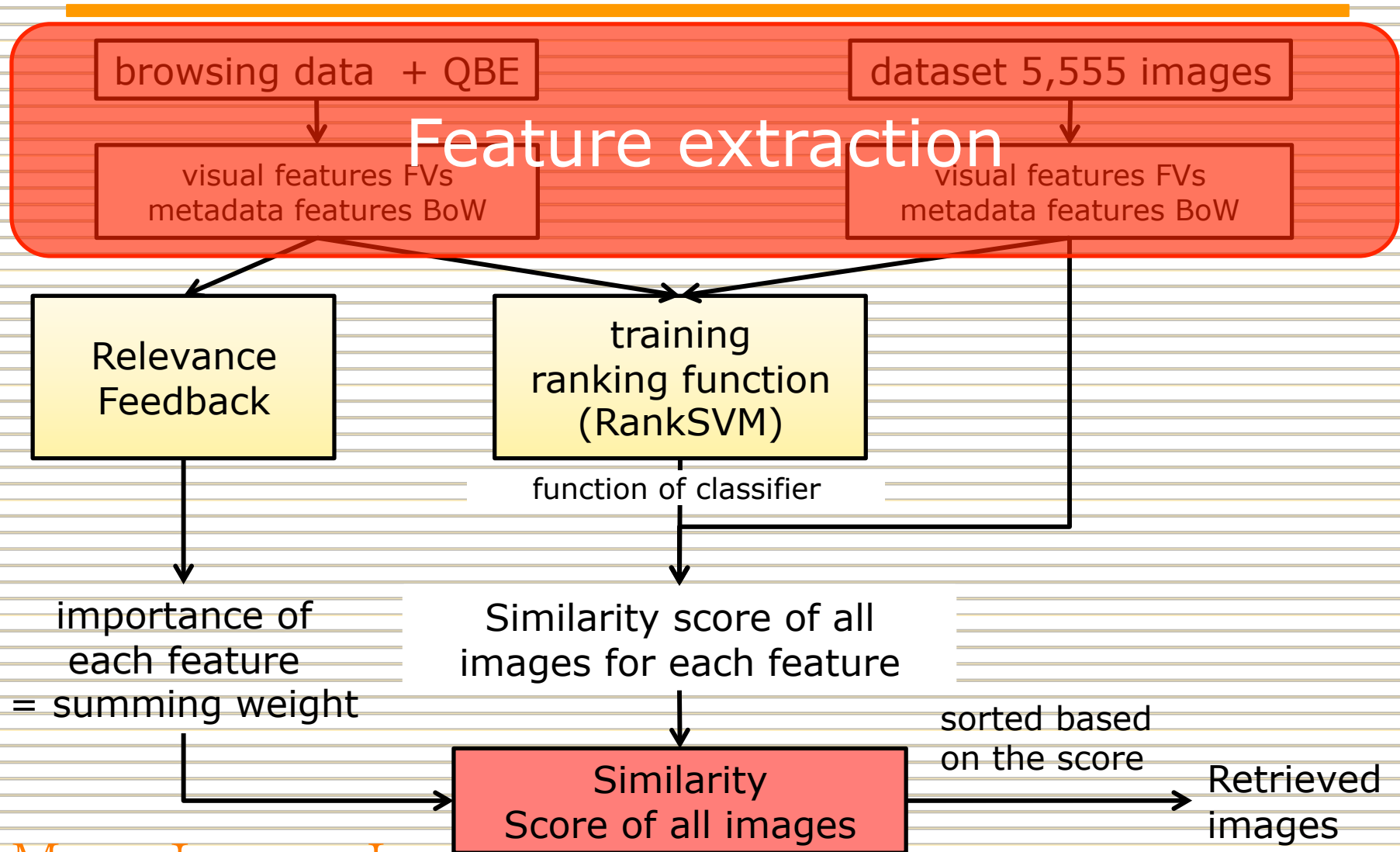
- K-Nearest Neighbors directed synthetic images injection[L. Piras et al., 2010]
 - No learning and simple

Methods depend on the query which is available.

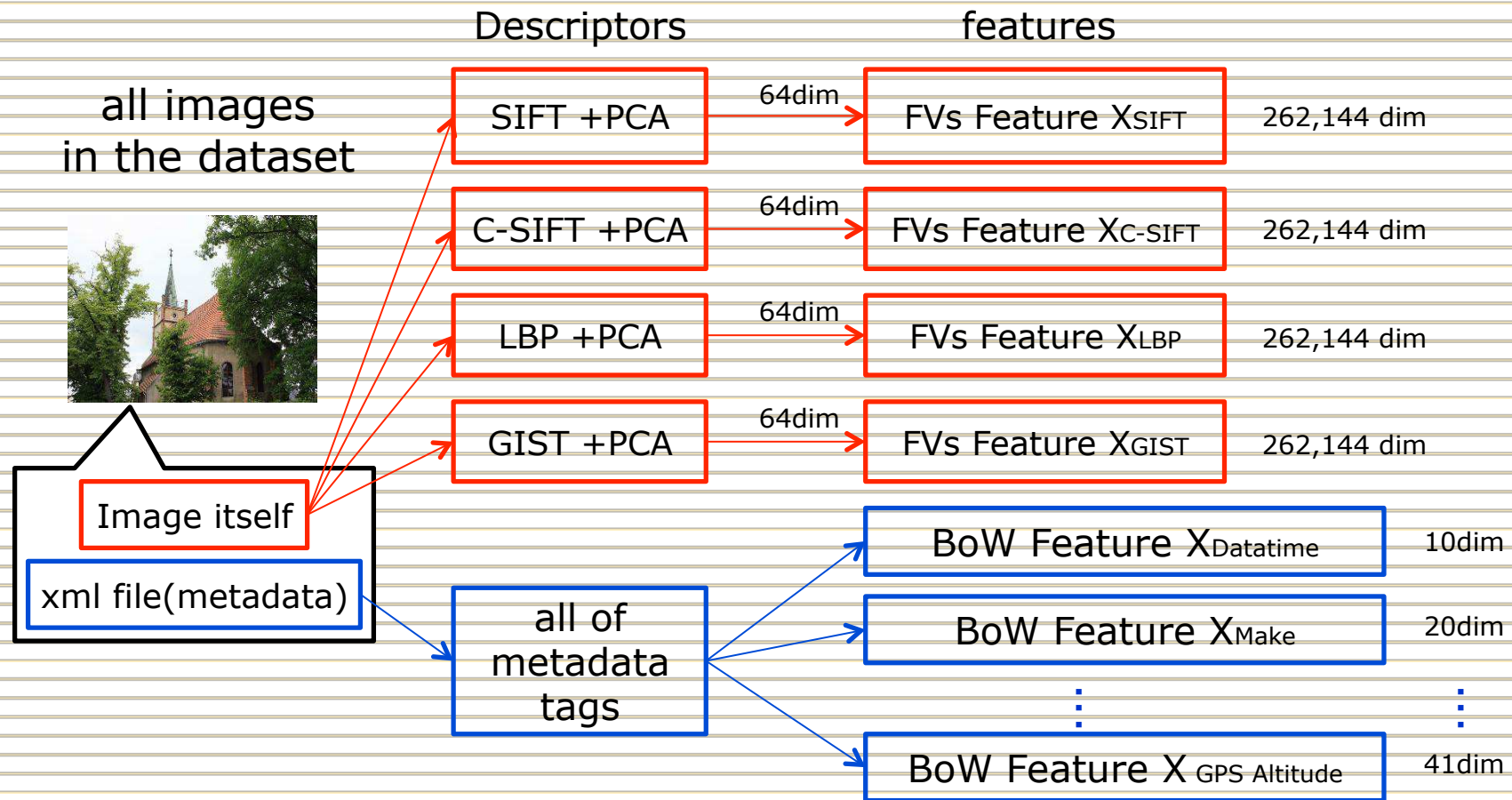
Subtask2: Personal Photo Retrieval

- Time-series data are available.
 - browsing data is obtained sequentially
 - ⇒ Images that user browses later represent the topic better
 - ⇒ Ranking SVM [T. Joachims, 2003]
- The task requires a higher level object recognition to topic detection
 - The latest feature coding for object recognition
 - ⇒ Fisher Vectors [F. Perronnin et al., ECCV 2010]

Methodology Overview



Feature Extraction Overview



Visual Feature Extraction

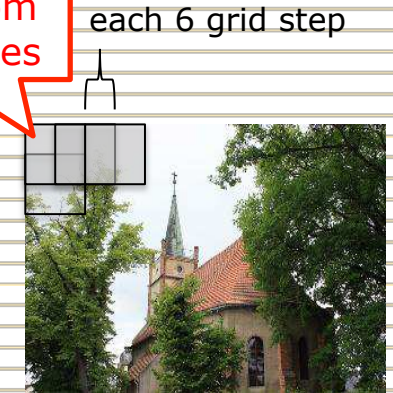
- We used the Improved Fisher Vectors (IFV)[F. Perronnin et al., ECCV 2010]

- Dimension of IFV = 262,144

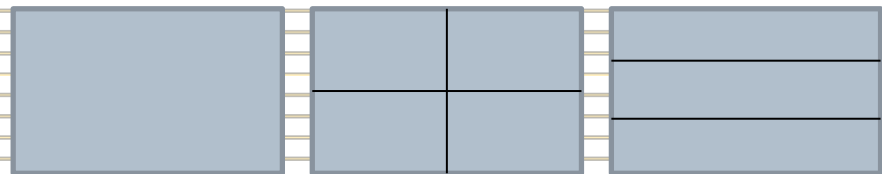
- Local descriptors

- 4 descriptors: SIFT, C-SIFT, GIST, LBP
- use Global descriptors as Local one
 - 5 scales of local patches
 - Sampling: each 6 grid step

extract global
feature from
local patches



- Dimension reduction of local feature with PCA :64
- components in GMM :256
- spatial pyramid divided into 1x1, 2x2, and 3x1 cells



Metadata Feature Extraction

- Bag of Words representation ($\Rightarrow [0,0,0,1,0,\dots]$)
- Extract 10 Exif data from xml file given

EXIF data name	dimension
Make (Canon, NIKON, SONY, ...)	20
Model (Canon PowerShot, CYBERSHOT, ...)	38
Flash (auto, fired, ...)	13
SceneCaptureType (Portrait, Night scene, ...)	4
DateTime (2011, 2009, ...)	10
GPS Altitude (0 metres , 102 metres, ...)	41
GPS Latitude Ref (S, N)	2
GPS Latitude (8°32'42", 8°17'16", ...)	143
GPS Longitude Ref (E, W, ...)	2
GPS Longitude (150°19'53.4", 6°15'33.6", ...)	151

didn't use about 30 metadata
"orientation" , "shutter speed", ...

Retrieval Methods

□ Similarity Score

- train the classifier so that QBE gets higher score than browsed images and Later browsed images are regarded as higher ranked than earlier ones.

score of QBE



score of browsed last

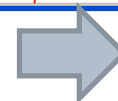


score of browsed 2nd last



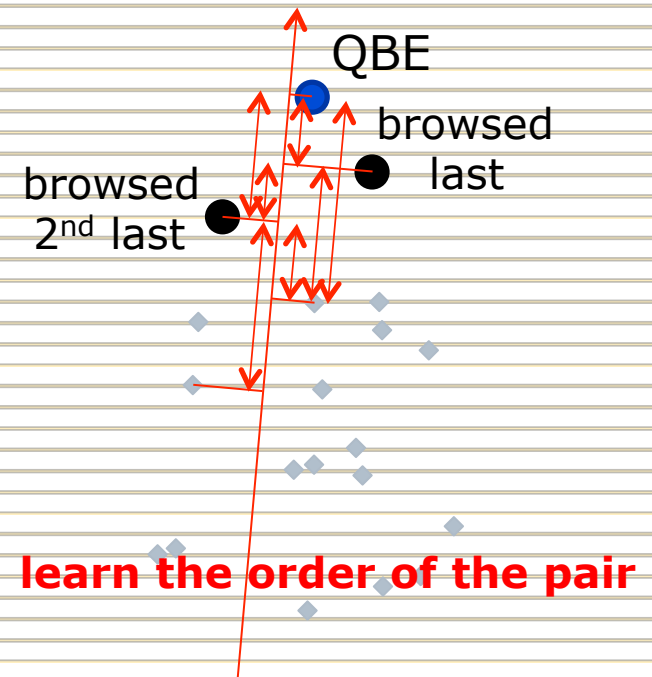
score of the others

Output from RankSVM

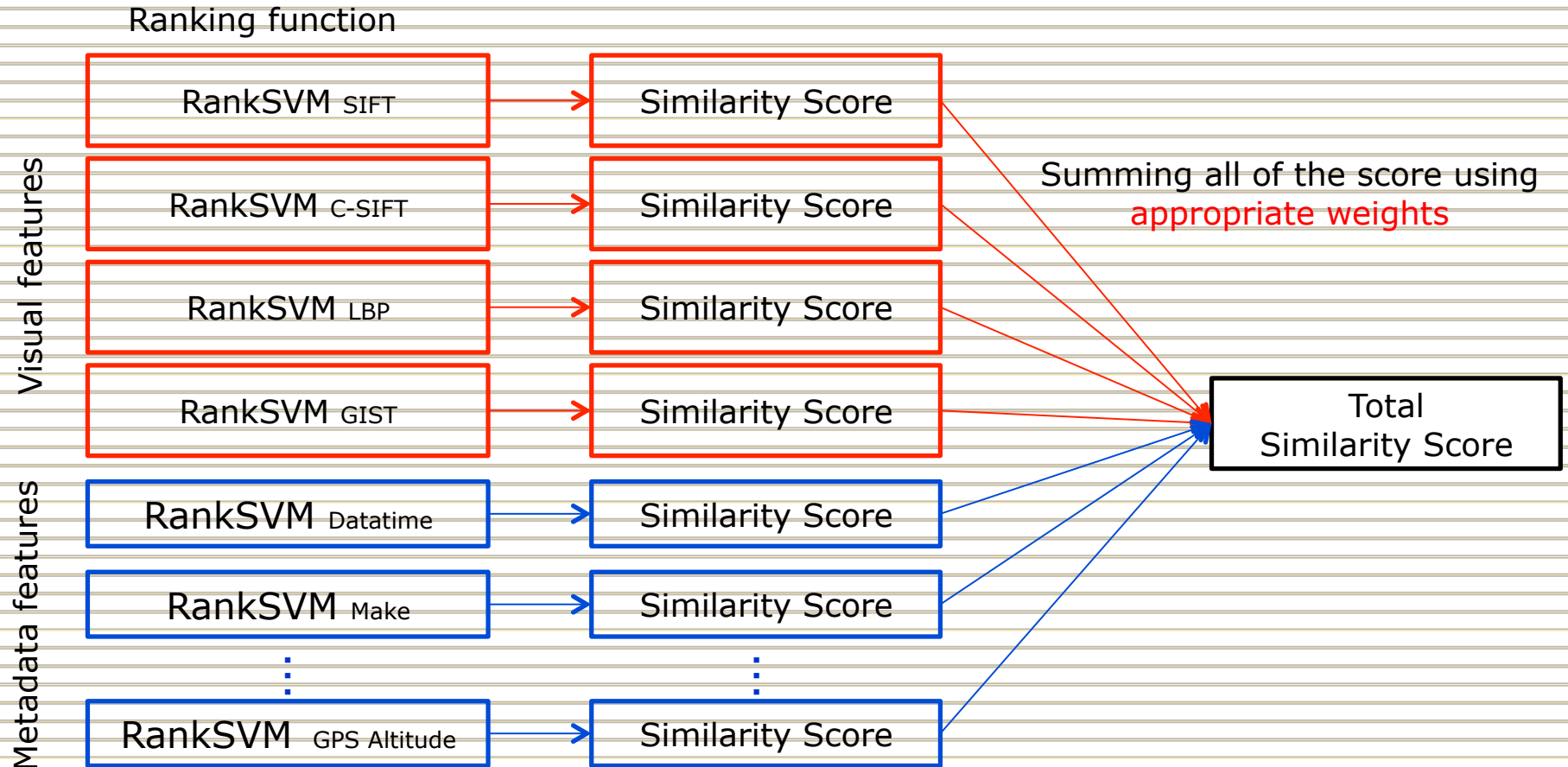


Score

RankSVM



Relevance Feedback



Relevance Feedback

- The weights are calculated by utilizing the browsing process.



firstly browsed image

[Visual feature]

[DateTime feature]

...

[GPS feature]

[Visual 2nd]

[DateTime 2nd]

...

[GPS 2nd]

Calculate variance

[Visual Large]

[DateTime 3rd]

...

[GPS small]

recal weights

Visual: 0.8 decrease

DateTime: 1.0

...

GPS: 1.2 increase

$$\omega_{l,t}^{new} = \frac{\sigma_l^I}{\sigma_l^{B_t}} \quad \leftarrow \text{variance in all images}$$

$$\quad \quad \quad \leftarrow \text{variance in query images}$$

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$$\omega_{l,t} = \alpha \times \omega_{l,t}^{new} + (1 - \alpha) \times \omega_{l,t}^{old}$$

Experiment

□ 1. Ranking function and Feature representations comparison (Visual features)

- RankSVM vs NN vs SVM

- FVs coding vs LLCs coding

 - LLCs (Locality-constrained Linear Coding) [Lin et al., CVPR 2011]

dimension = $1024 * 7 = 7168$

Local descriptors SIFT, C-SIFT, LBP and GIST

□ 2. Ranking function comparison (Metadata features)

- RankSVM vs NN vs SVM

□ 3. Combinations of visual and metadata features

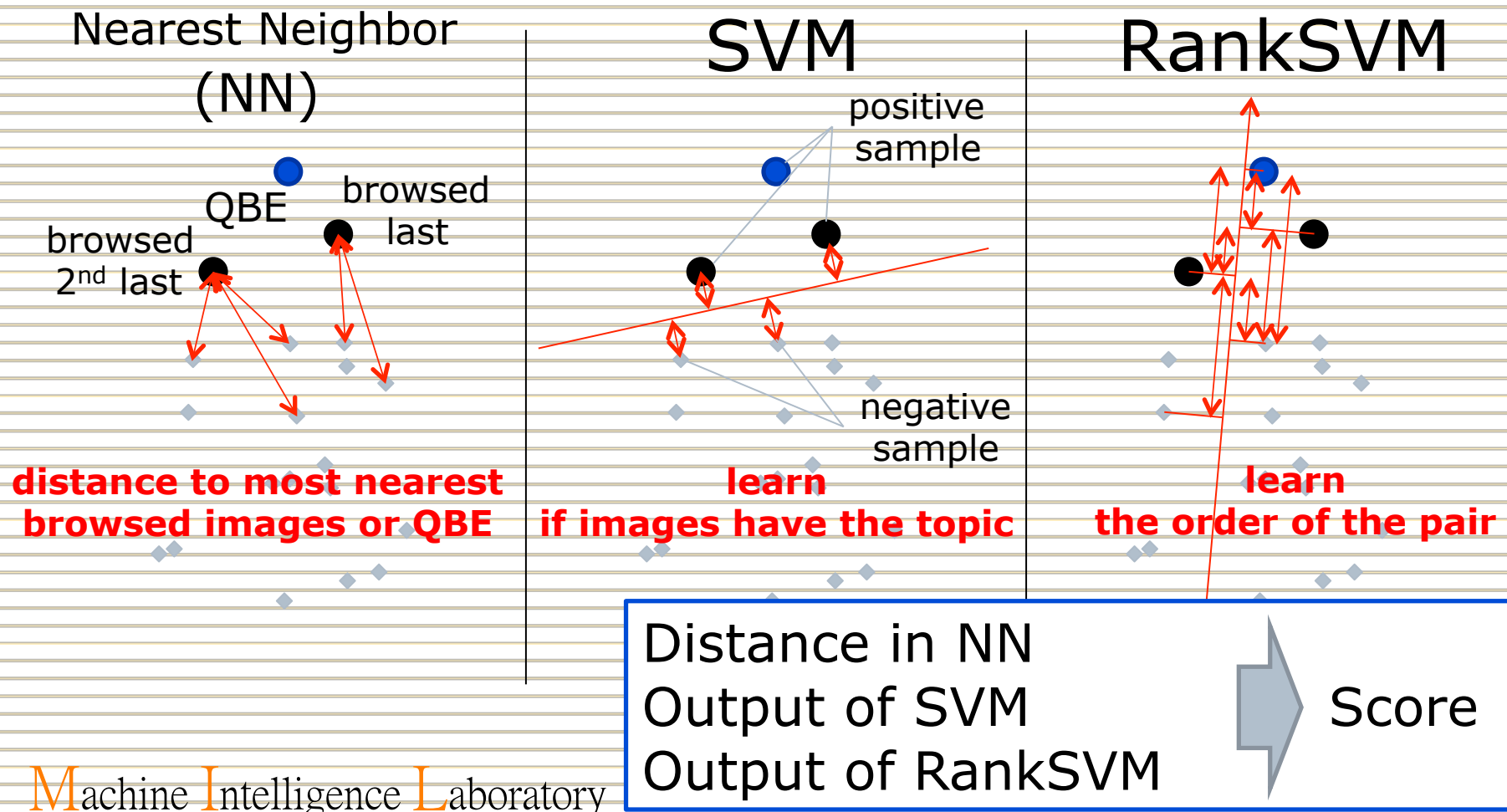
- number of topics(browsing data) : 74 • Dataset : 5,555 images

- browsing data and QBE : 1~4 images

- Evaluation : NDCG(ndcg_cut_100) on ground truth data

1. Ranking function comparison(Visual features)

- calculation methods of similarity score (Visual features)



1.Ranking function and Feature representations comparison (Visual feature only)

ndcg_cut_100	NN	SVM	rankSVM
LLCs+ SIFT	0.2946	0.3066	0.3308
LLCs+ C-SIFT	0.2856	0.2967	0.3257
LLCs+ LBP	0.3043	0.3199	0.3385
LLCs+ GIST	0.2796	0.2943	0.3175
FVs+ SIFT	0.3135	0.3278	0.3357
FVs+ C-SIFT	0.3492	0.3486	0.3696
FVs+ LBP	0.3636	0.3363	0.3861
FVs+ GIST	0.3376	0.3145	0.3572

▪ SVM < NN < RankSVM

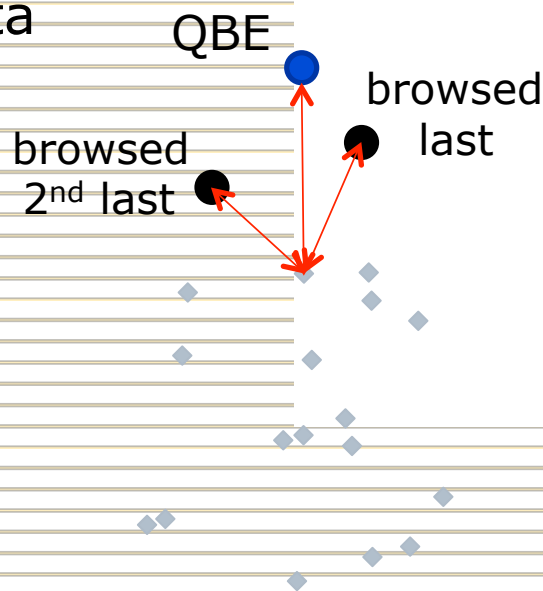
▪ LLCs < FVs

2. Ranking function comparison (Metadata features)

calculation methods of similarity score (Metadata features)

□ Nearest Neighbor

- Distance metric like RBF kernel between images
Euclidean distance is not appropriate for BoW
- Summing similarity scores of image and all browsed data



$$d(\mathbf{x}_i^m, \mathbf{x}_j^m) = 1 - e^{-\tau \|\mathbf{x}_i^m - \mathbf{x}_j^m\|^2}$$

$$c_{i,j} = \frac{1}{1 + d(\mathbf{x}_i^m, \mathbf{x}_j^m)}$$

$$s_{t,i}^m = \sum_{k \in B_t \cup q_t} c_{k,i}$$

calculation methods of score comparison

ndcg_cut_100

□ RankSVM \Rightarrow 0.6508

□ SVM \Rightarrow 0.6367

□ Nearest Neighbor \Rightarrow 0.6228
with RBF kernel

□ Nearest Neighbor \Rightarrow 0.6203
with RBF kernel
without Relevance Feedback

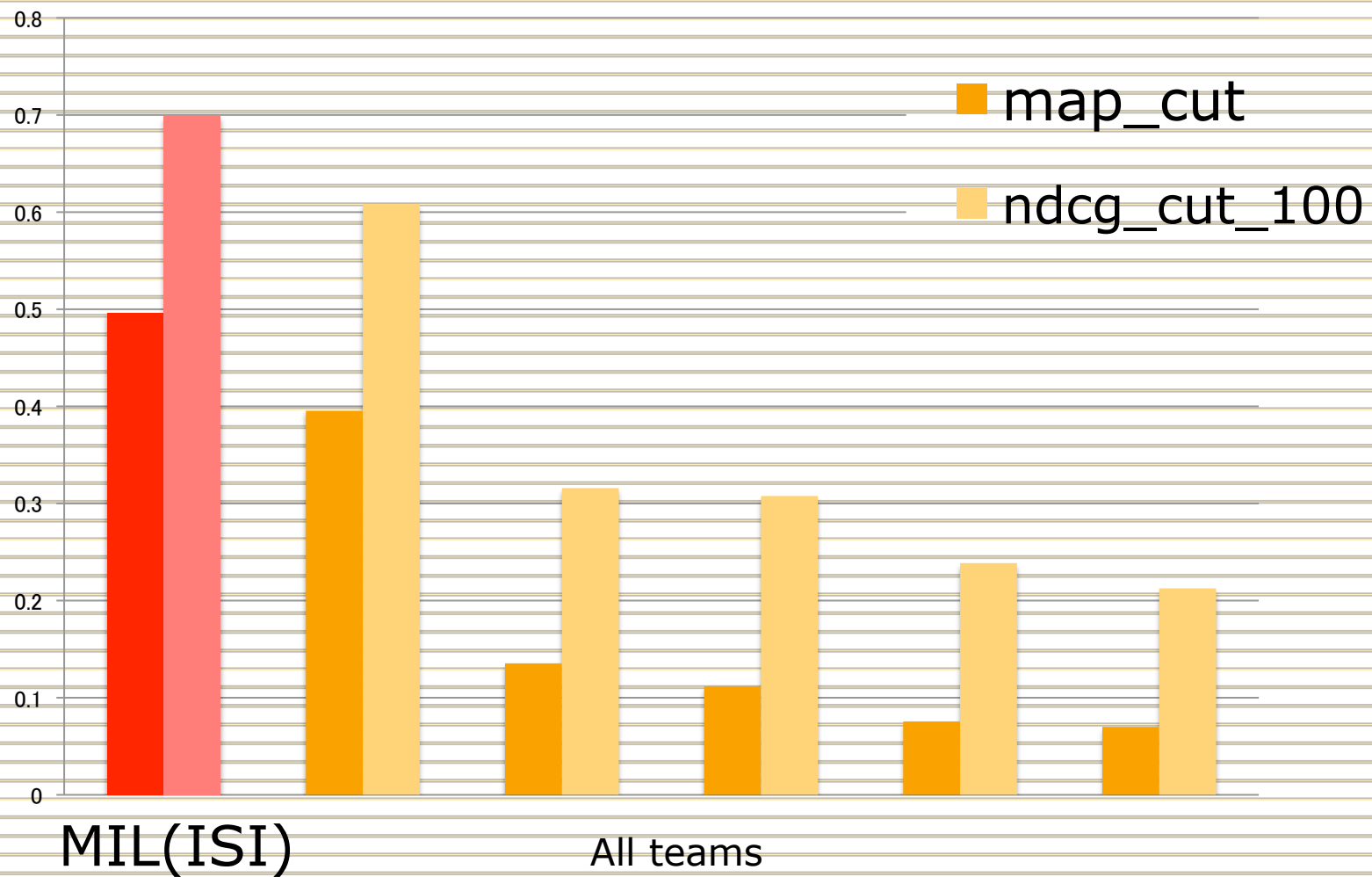
NN < SVM < RankSVM

3. Combinations of visual and metadata features

- We used **RankSVM** as ranking function and **FVs** for visual features.

SIFT	✓	—	—	—	✓	—
C-SIFT	✓	✓	—	✓	✓	✓
LBP	✓	✓	—	✓	✓	✓
GIST	✓	✓	—	✓	✓	—
10 Metadata	✓	✓	✓	—	—	—
ndcg_cut_100	0.7039	0.7040	0.6508	0.4236	0.4186	0.4166
ndcg_cut_20	0.7477	0.7463	0.6689	0.5193	0.5134	0.5171

Result



Conclusions

□ Motivation

Estimating a topic from few query data and retrieve images which have the topic

□ Methodology

Train **RankSVM** for visual features(**FVs** of SIFT, C-SIFT, LBP, GIST) and metadata features(**BoW** of 10 Exif data).—

Combine similarity score with relevance feedback

□ Result

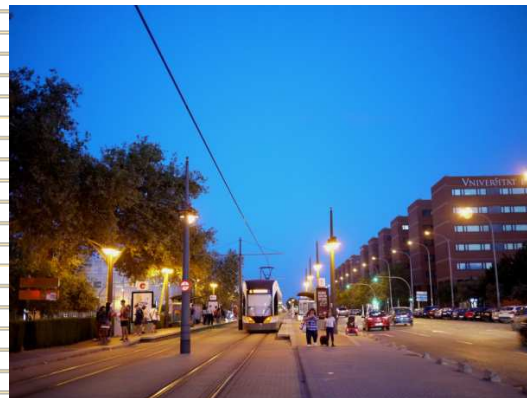
LLCs < FVs (Visual)

SVM < NN < RankSVM (Visual)

NN < SVM < RankSVM (Metadata)

Make
Model
Flash
SceneCaptureType
DateTime
GPS Altitude
GPS Latitude Ref
GPS Latitude
GPS Longitude Ref
GPS Longitude

□ Thank you for listening.



Topic:
CLEF2013@Valencia







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- Outline of subtask

- Methodology

- Outline
- Feature Extraction
- Retrieving Methods
- Relevance Feedback

- Results

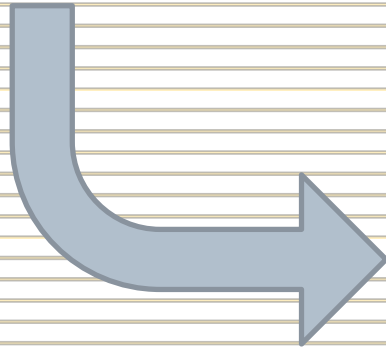
- Conclusions

What is needed?

□ another example

QBE

browsed images



same day, place, camera

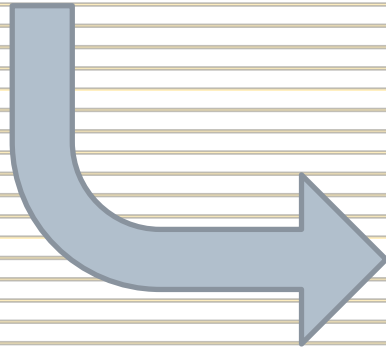


What is needed?

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QBE

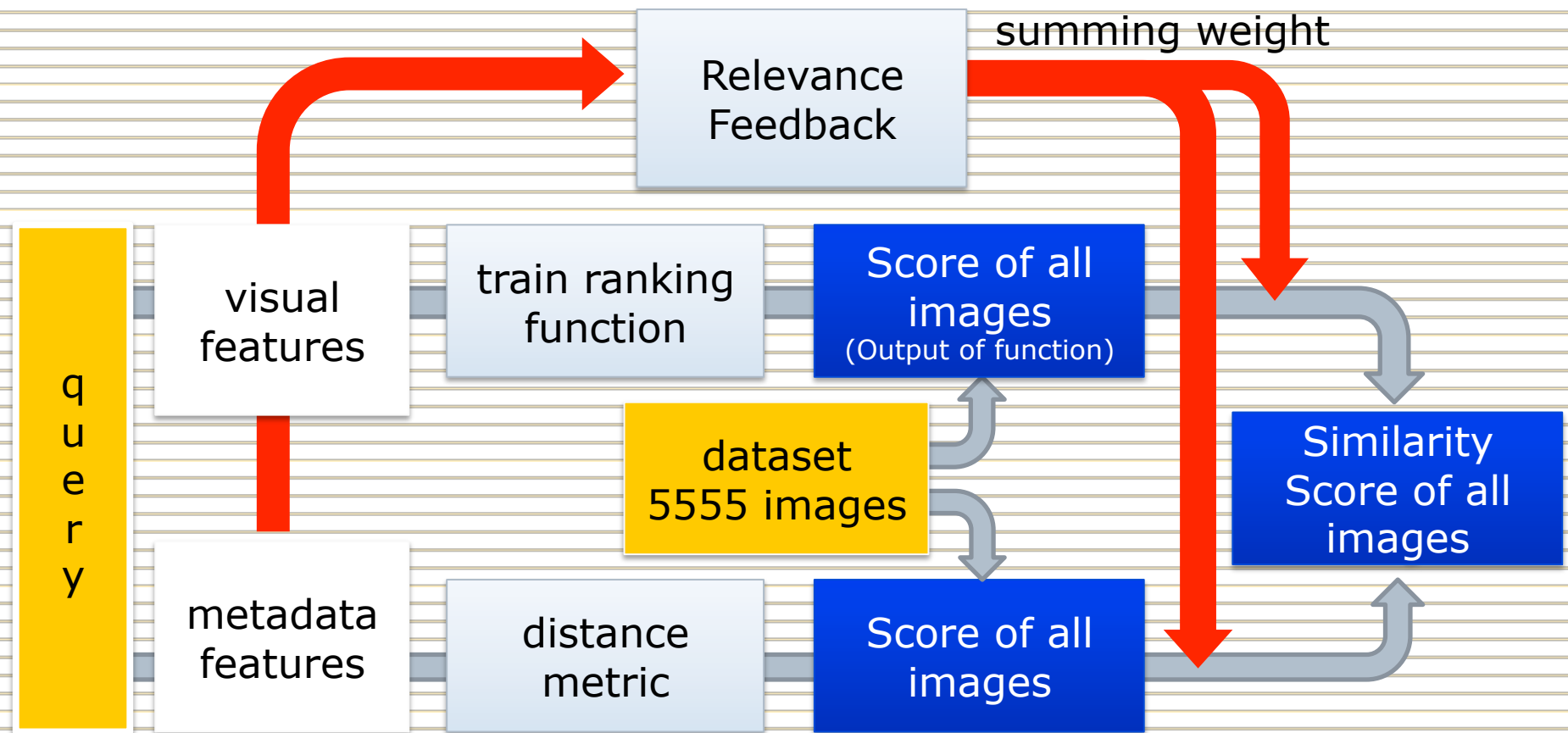
browsed images



same object



Methodology Outline



Retrieving Methods

□ Metadata Similarity Score

- Distance between image i & j

$$d(\mathbf{x}_i^m, \mathbf{x}_j^m) = 1 - e^{-\tau \|\mathbf{x}_i^m - \mathbf{x}_j^m\|^2}$$
$$c_{i,j} = \frac{1}{1 + d(\mathbf{x}_i^m, \mathbf{x}_j^m)}$$

- Similarity of image i for topic t

$$s_{t,i}^m = \sum_{k \in B_t \cup q_t} c_{k,i}$$

summing the scores with all of the query
browsed images

Result 2 Feature combination

Combinations of FVs visual features only

SIFT	–	✓	–	✓	–	✓
C-SIFT	✓	✓	✓	✓	–	✓
LBP	✓	✓	✓	✓	✓	–
GIST	✓	✓	–	–	✓	✓
10 Metadata	–	–	–	–	–	–
ndcg_cut_100	0.4236	0.4186	0.4186	0.4118	0.4058	0.4008

Result 3

□ metadata features only

SIFT	–	–
C-SIFT	✓	–
LBP	✓	–
GIST	✓	–
10 Metadata	–	✓
ndcg_cut_100	0.4236	0.6228

Result 4 Feature combination

- Top combinations of visual and metadata features

SIFT	✓	–	–	✓	✓	✓
C-SIFT	✓	✓	✓	✓	–	✓
LBP	✓	✓	✓	✓	✓	–
GIST	✓	✓	–	–	✓	✓
10 Metadata	✓	✓	✓	✓	✓	✓
ndcg_cut_100	0.6998	0.6986	0.6985	0.6983	0.6982	0.6967

We submitted this score

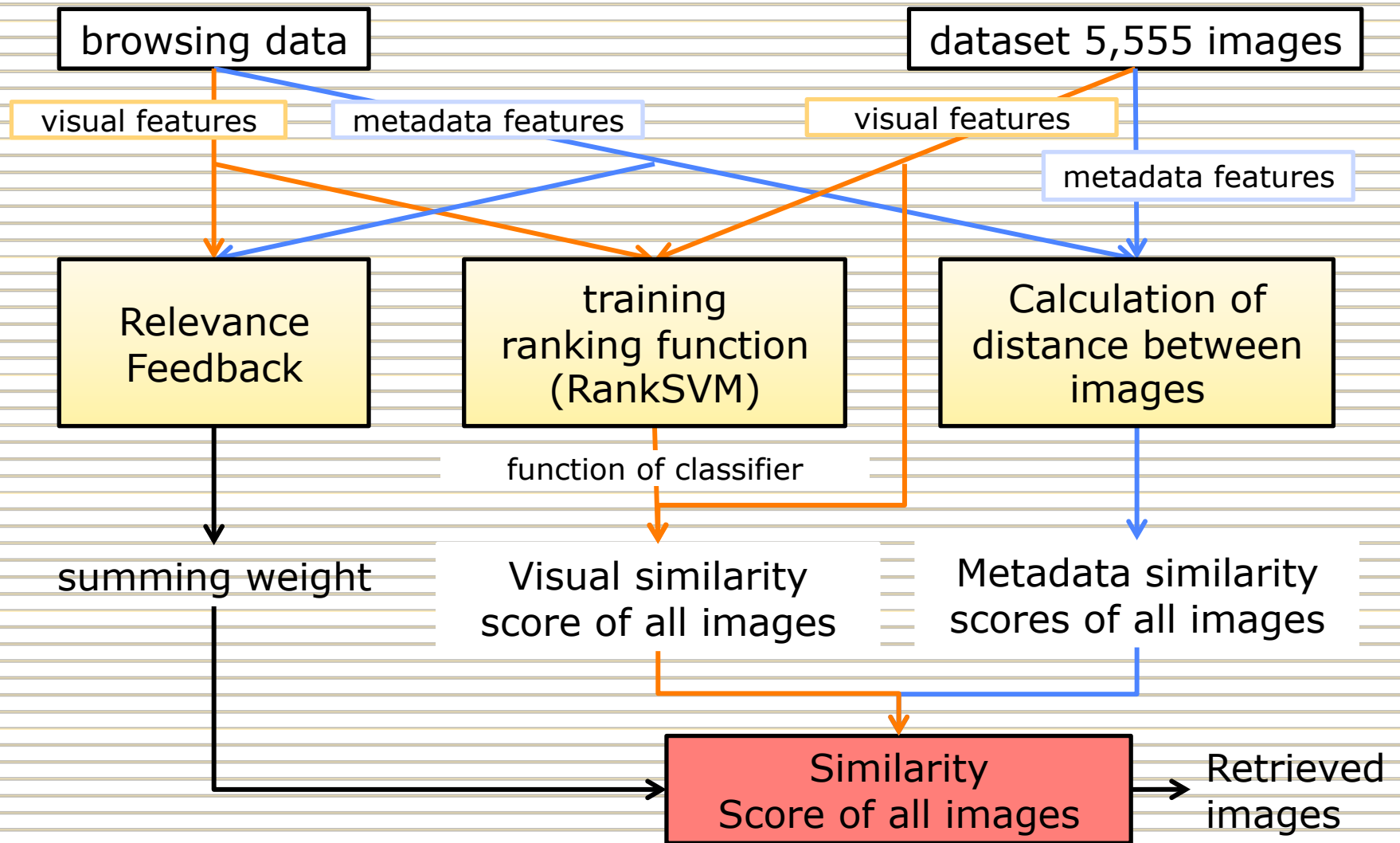
Result 2 Feature combination

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LBP	✓	✓	—	✓	✓	✓
GIST	✓	✓	—	✓	✓	—
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 - LLCs (Locality-constrained Linear Coding) [Lin et al., CVPR 2011]

 - dimension = $1024 * 7 = 7168$

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□ 2. Ranking function comparison (Metadata features)

- RankSVM vs SVM vs Distance metric

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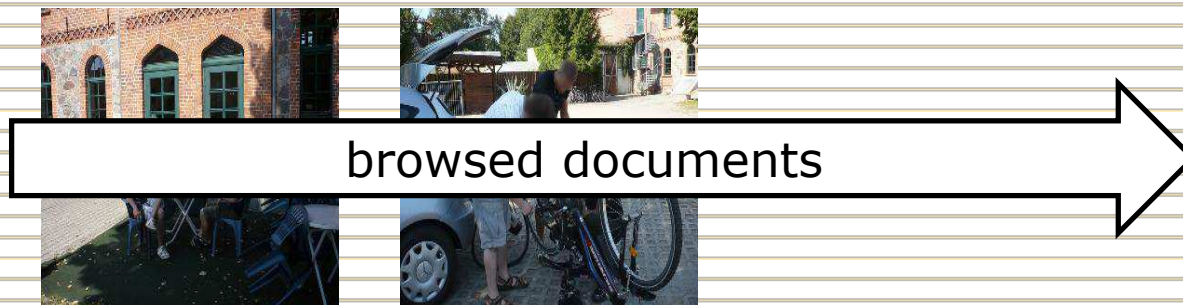
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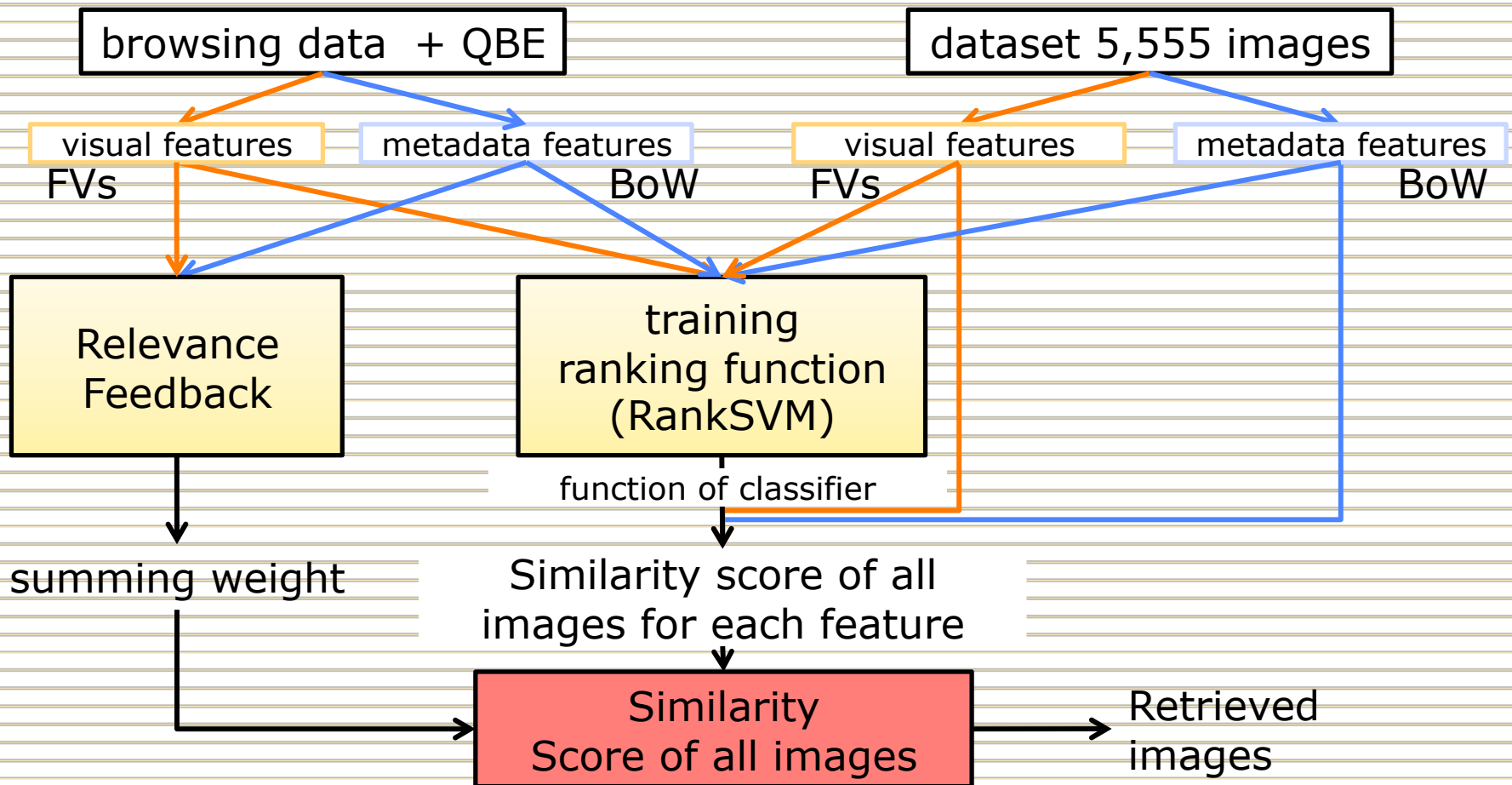
firstly browsed image

[Visual feature]	[Visual 2nd]	⇒ σ^{Visual} large	recalc weights Visual: 0.8 decrease
[DateTime feature]	[DateTime 2nd]	⇒ σ^{DateTime}	DateTime: 1.0
...
[GPS feature]	[GPS 2]	⇒ σ^{GPS} small	GPS: 1.2 increase

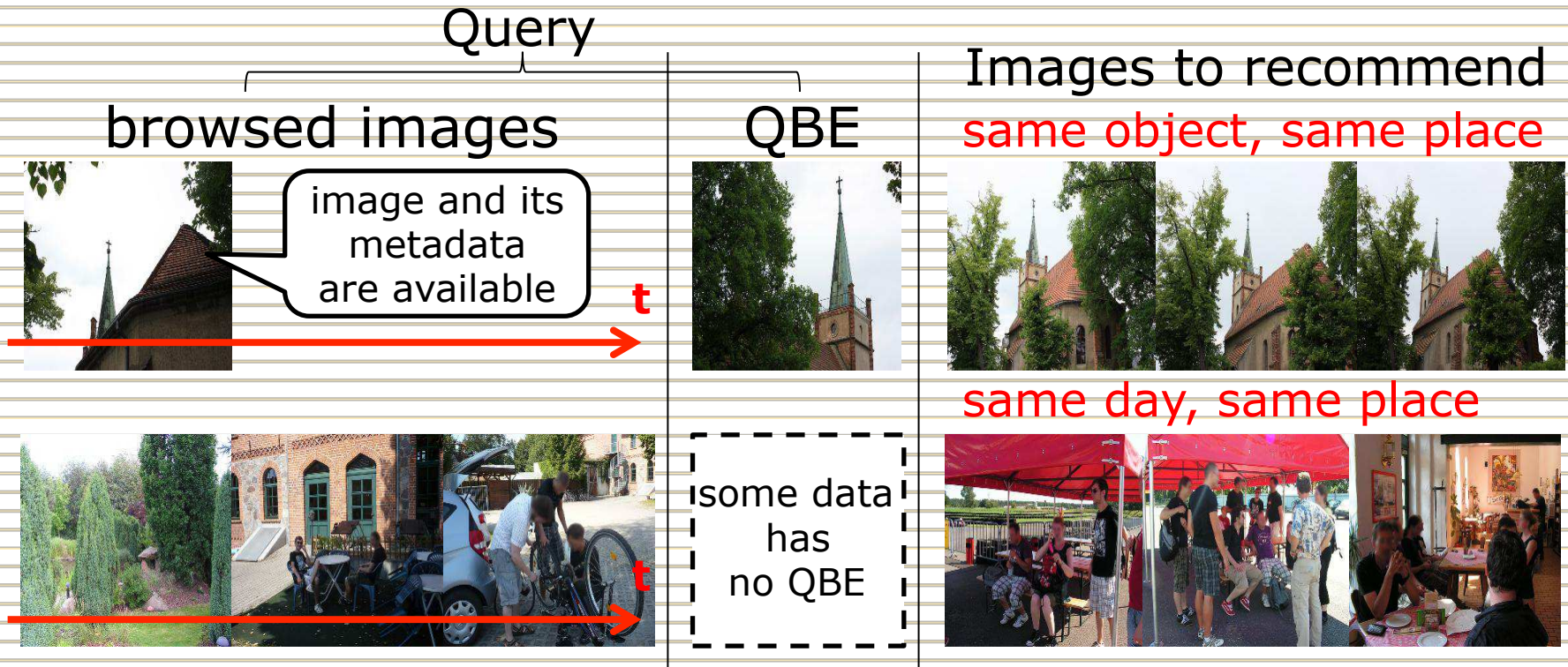
$$\omega_{l,t}^{\text{new}} = \frac{\sigma_l^I}{\sigma_l^{B_t}}$$

$$\omega_{l,t} = \alpha \times \omega_{l,t}^{\text{new}} + (1 - \alpha) \times \omega_{l,t}^{\text{old}}$$

Methodology Outline



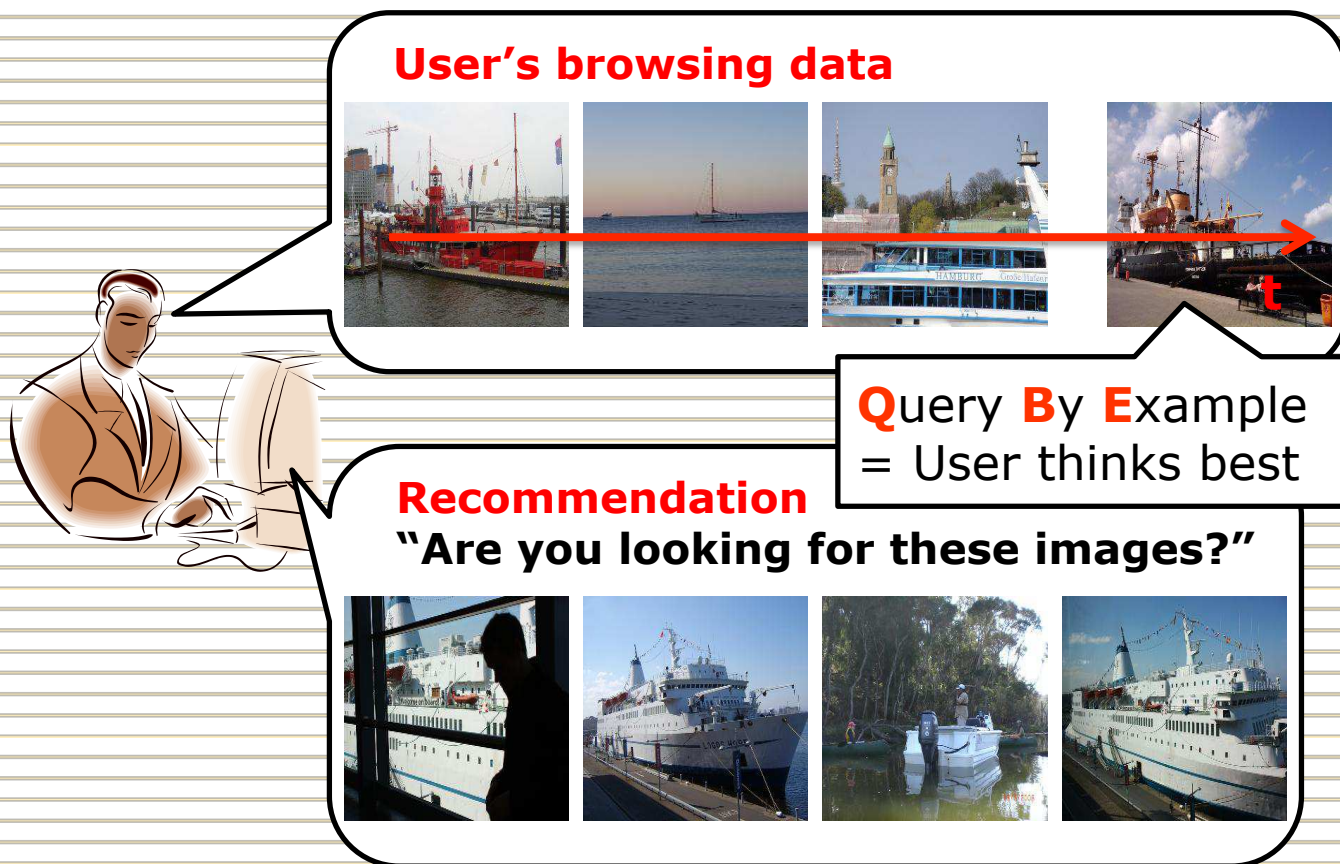
Motivation



- Task: estimating a topic from few query data and retrieve images which have the topic

Subtask2: Personal Photo Retrieval

The system which can help users to retrieve images from a lot of personal photo collections using browsing data.



Visual Feature Extraction

□ FVs (Fisher Vectors)

[F. Perronnin et al., ECCV 2010]

dimension = 262,144

Local descriptors

SIFT, C-SIFT, LBP and GIST

using Global descriptors as

Local one (densely extracted
from five scales of patches on
a regular grid every six pixels)

256 GMM components

Spatial pyramid divided

Machine Intelligence Laboratory
into 1x1, 2x2, and 3x1 cells

Metadata Feature Extraction

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- train the classifier so that QBE gets higher score than browsed images and Later browsed images are regarded as higher ranked than earlier ones.

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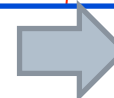


score of browsed 2nd last



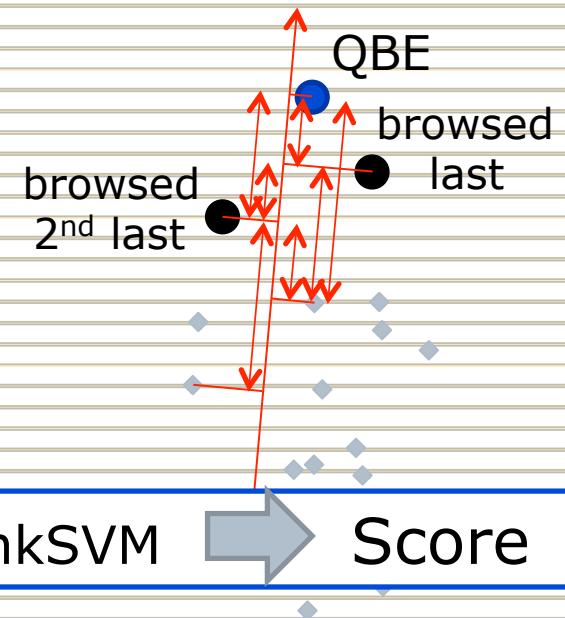
score of the others

Output from RankSVM



Score

RankSVM



Conclusions

□ Motivation

To estimate a topic from
few query data and retrieve
images which have the topic

□ Methodology

visual (**RankSVM** + **FVs** of C-SIFT, LBP, GIST)
+ relevance feedback

metadata (**RankSVM** + **Bow of 10 Exif data**)

□ Result

LLCs < FVs

SVM < NN < RankSVM (Visual)

Distance metric < SVM < RankSVM (Metadata)

Methodology Overview

