

Enhancing Sketch-Based Image Retrieval by Re-Ranking and Relevance Feedback

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CS688
Student paper presentation

Contents

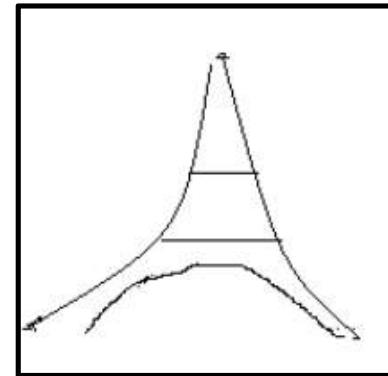
- **Problems & Related work**
- **Solution**
 - Image Grouping
 - Visual Feature Verification
 - Contour-Based Relevance Feedback
- **Experimental Result**
- **Conclusion**

Problems

- Sketch Based Image Retrieval (SBIR)



What a user want to find



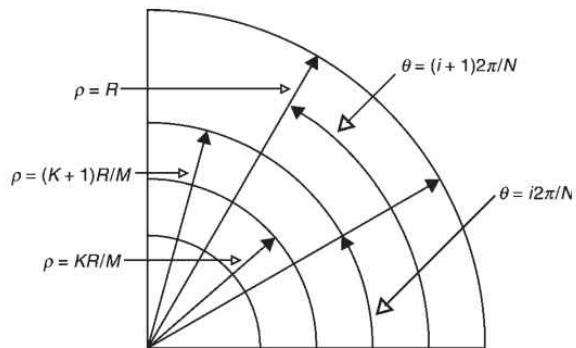
What a user queries

How to measure the relevance of an image and a query sketch?

Problems

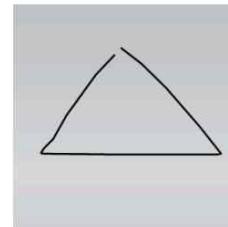
- To solve the problem..
 - Contour matching
 - Local feature matching

Angular Radial Partitioning(ARP)



Edgel index

Sketch Query



Decomposition
→



Edgel index : Cao, Yang, et al. "Edgel index for large-scale sketch-based image search." (2011): 761-768.

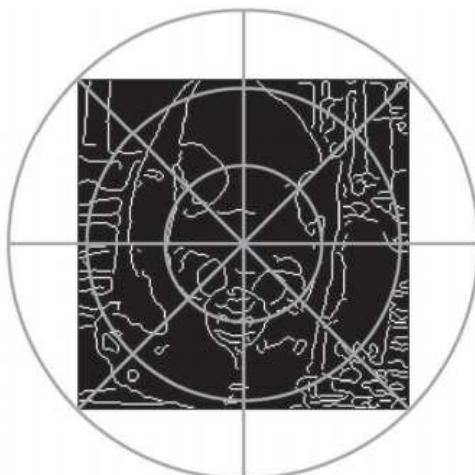
ARP : Chalechale, Abdolah, Alfred Mertins, and G. Naghdy. "Edge image description using angular radial partitioning." *IEE Proceedings-Vision, Image and Signal Processing* 151.2 (2004): 93-101.

Related work

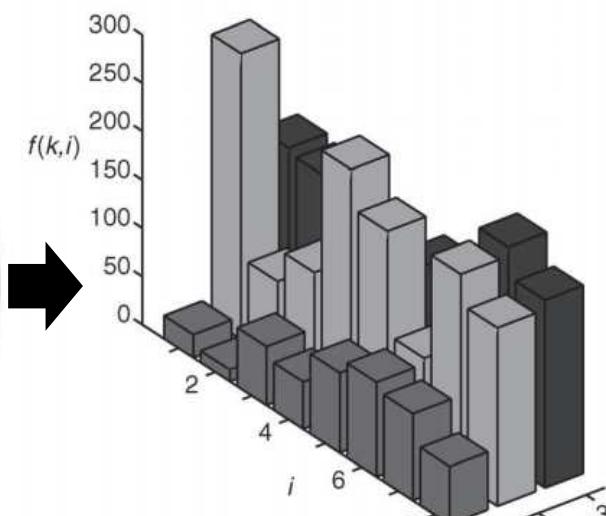
- Angular Radial Partitioning (ARP)



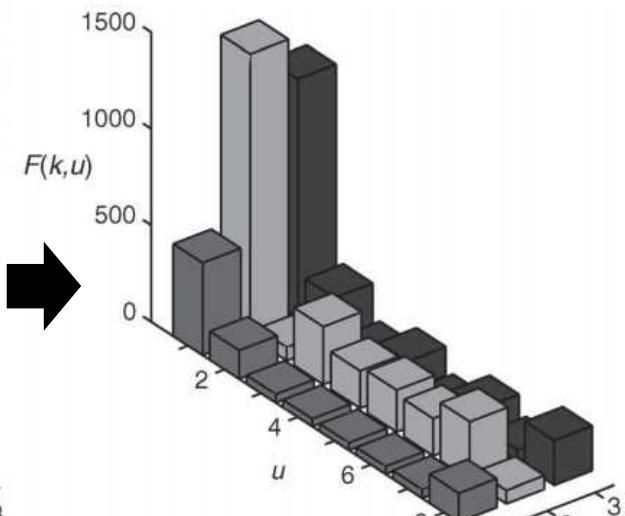
Image



Partitioning



Pixels in each partition

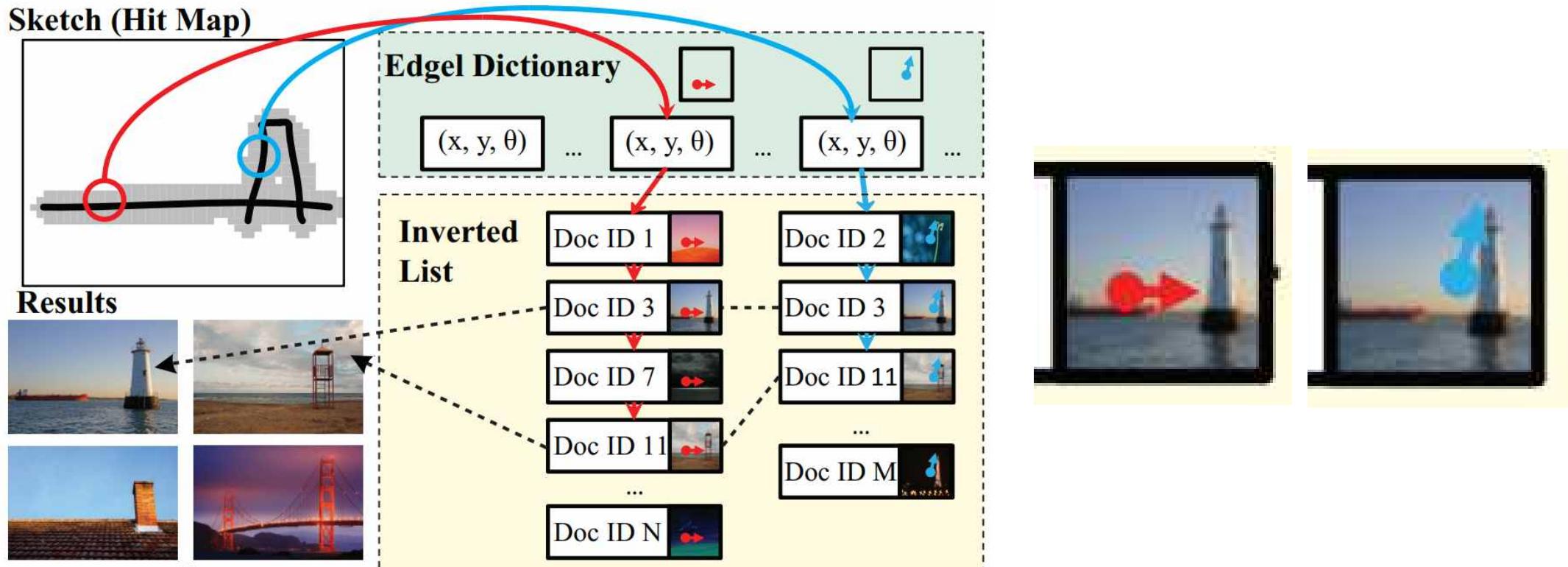


Fourier transformed

Related work

- Edgel index (Edgel : edge pixel)

Sketch (Hit Map)



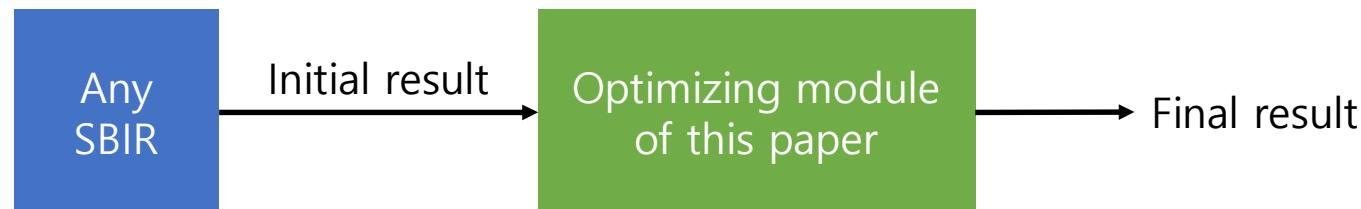
Problems

- Sketch should be fairly close to the image.
- Irrelevant image may be retrieved.

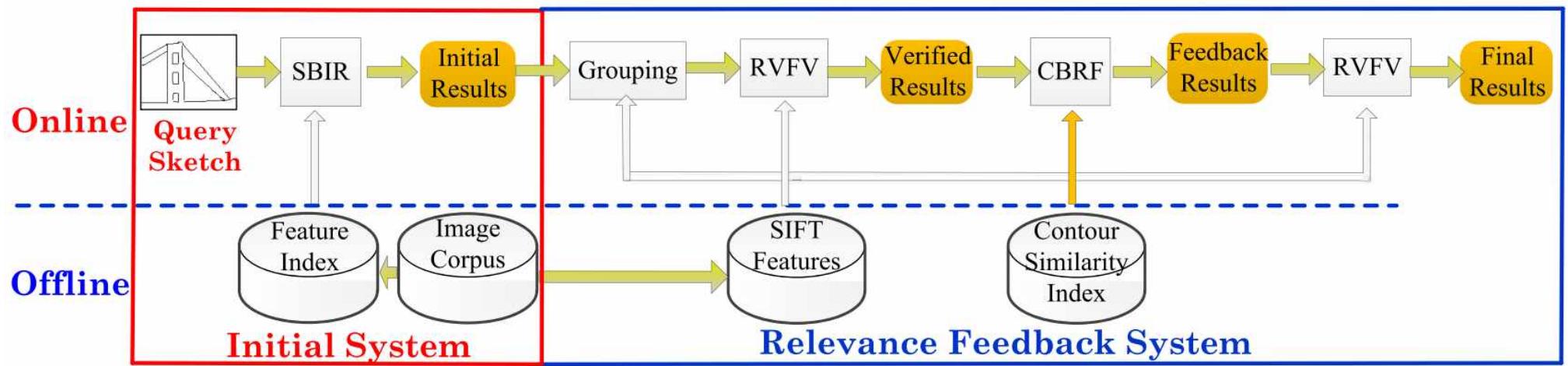
Re-ranking and finding relevant images are important!

Solution

- Contribution
 - **Optimizing module** with the search result of any SBIR framework

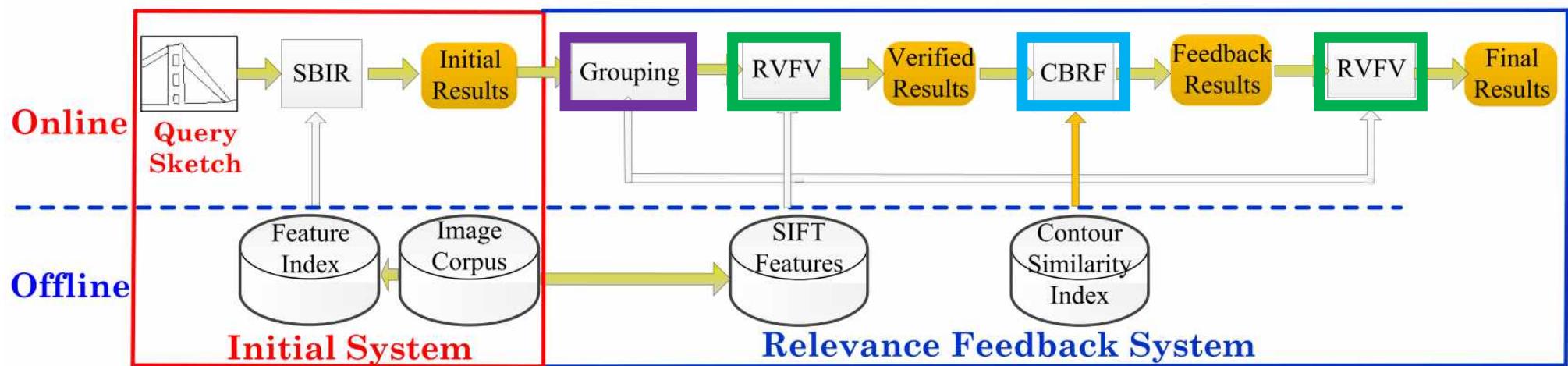


Solution



- **Image Grouping**
 - Fining more relevant images
- **RVFV**
 - Removing irrelevant images
- **CBRF**
 - Making new queries to find relevant images using contours

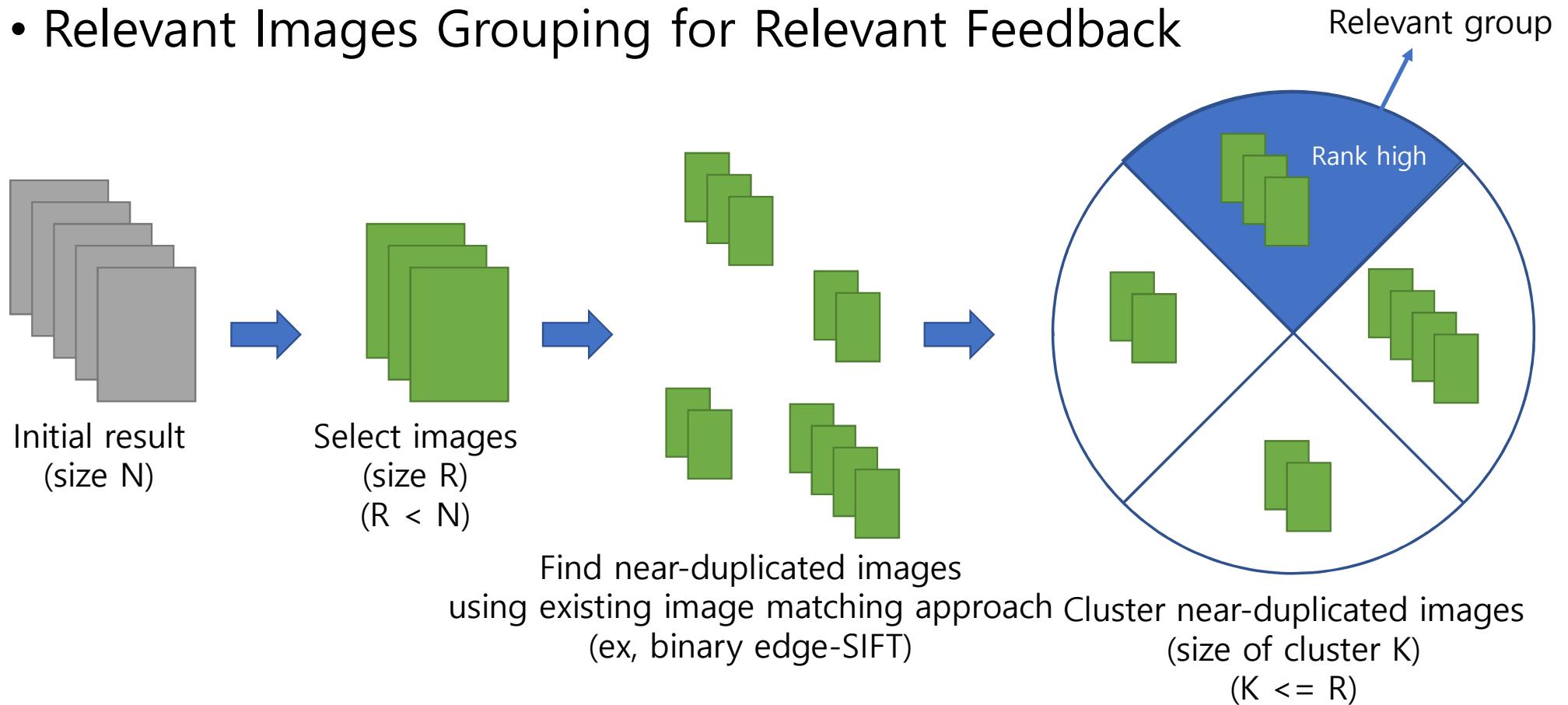
Solution



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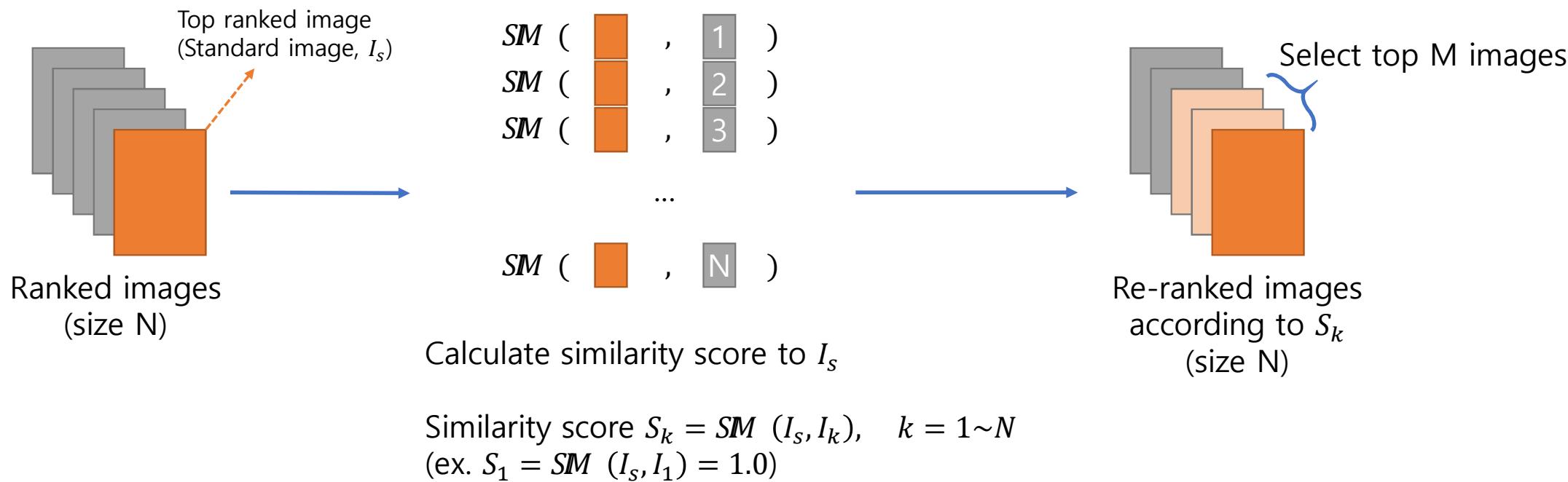
Solution

- Relevant Images Grouping for Relevant Feedback



Solution

- Re-ranking via Visual Feature Verification (RVFV)



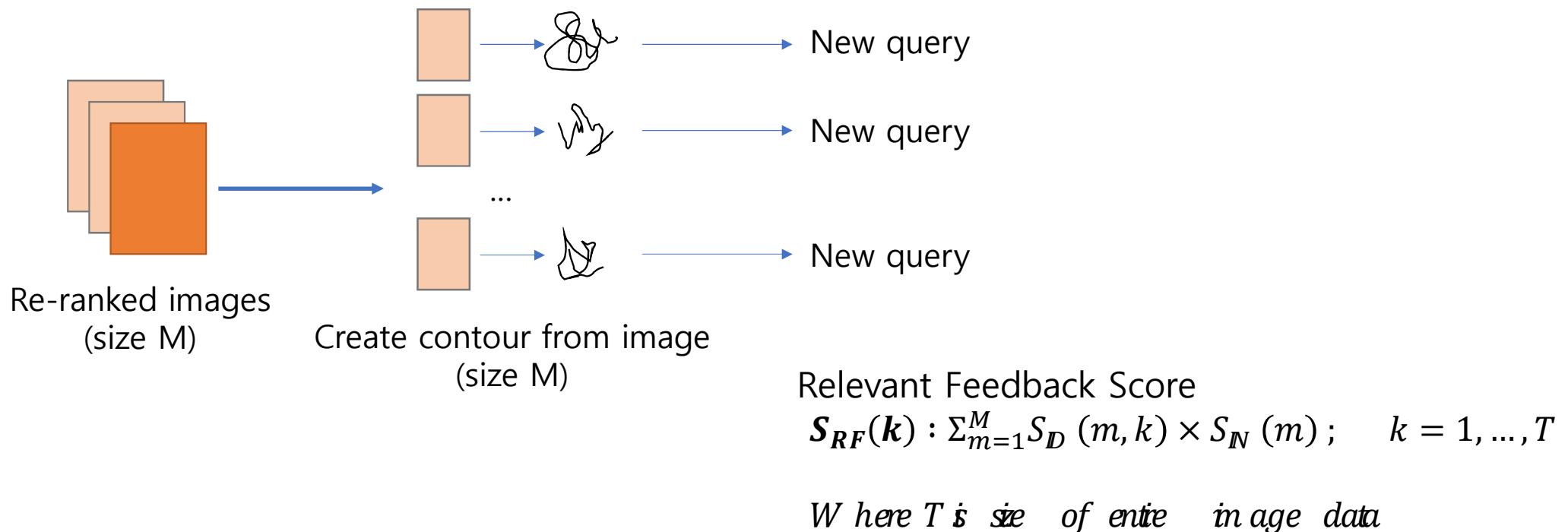
Solution

- Similarity score S_k
 - d_A : SIFT descriptor of image A
 - L2 norm of two descriptor $|d_A - d_B|_2^2 = 2 - \sum_l d_A^l d_B^l$
since $|d_A|_2^2 + |d_B|_2^2 = 2$, $\text{SM } (d_A, d_B) = \sum_l d_A^l d_B^l$
 - $\text{SM } (I_A, I_B) = \sum_m \text{SM } (d_A, d_B) W(m)$

W here,
m is mth SFT pair between I_A and I_B
 $W(m)$ is weight

Solution

- Contour-Based Relevance Feedback



Solution

- Contour-Based Relevance Feedback

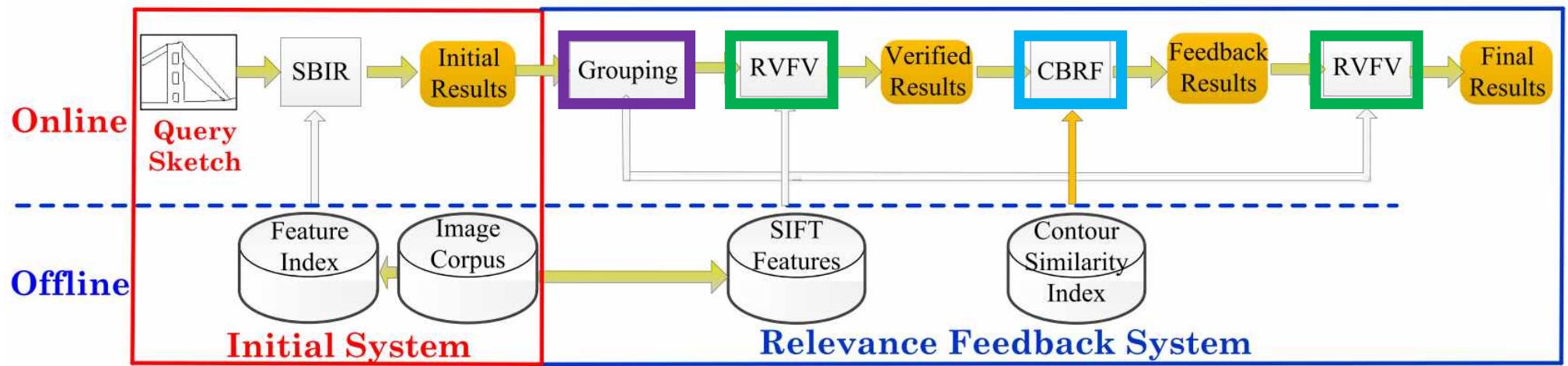
- Relevant Feedback Score

$$S_{RF}(k) : \sum_{m=1}^M S_D(m, k) \times S_N(m); \quad k = 1, \dots, T$$

Where T is size of entire image data

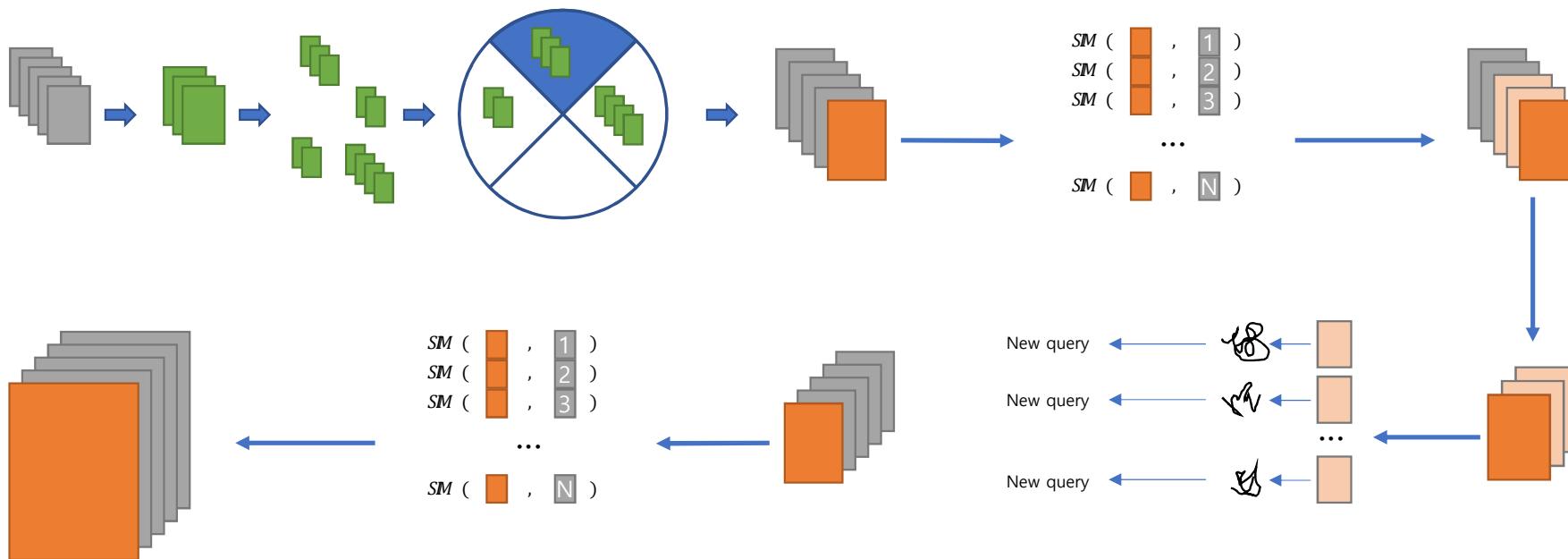
- $S_N(m)$: Initial score of image m
 - $S_D(m, k)$: Score after first RVFV of image k , when a query is contour of image m
- Final score $S(k) = (1 - w) \times S_N(k) + w \times S_{RF}(k); \quad k = 1, \dots, T$
- With $S(k)$, we have new ranked list

Solution



- **Image Grouping**
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Solution

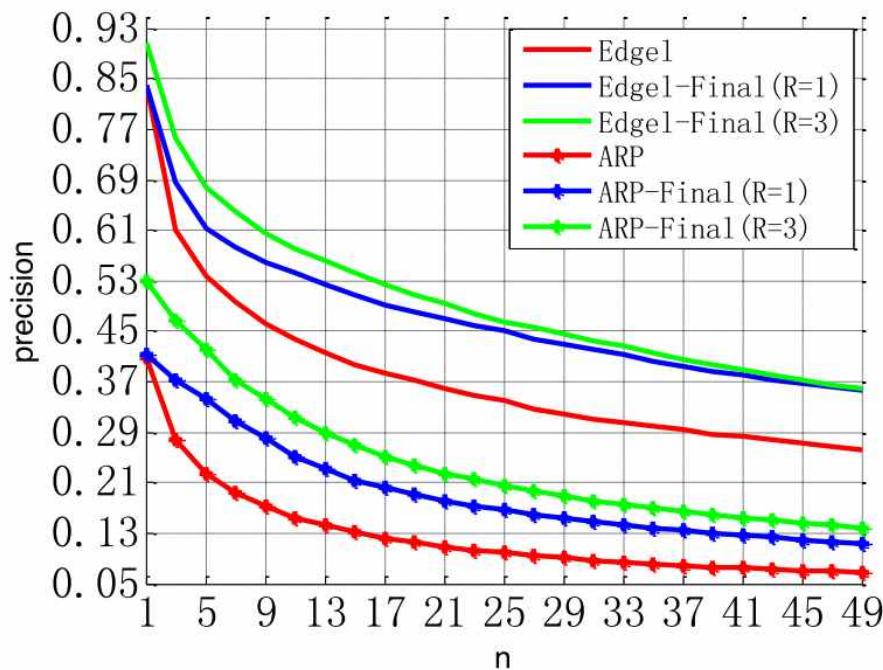


Experimental Result

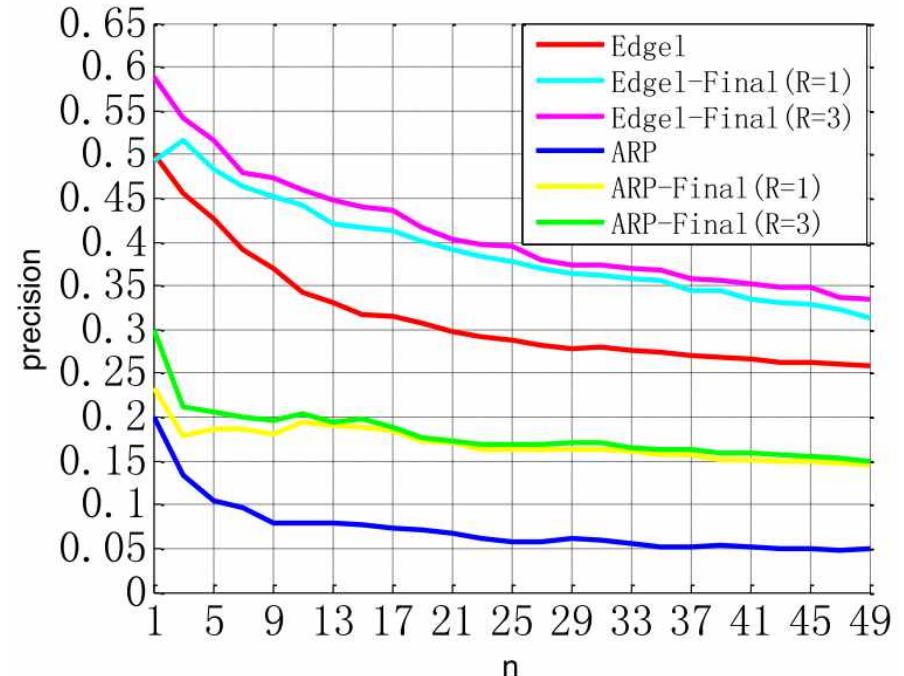
- Experimental setting
 - Dataset
 - SBIR_100K Dataset : 1,240 images for 31 sketches and 100,000 noise images
 - Authors' own Dataset : from Google keyword search 296,562 images with 68,647 sketch-describable images + 523 sketches

Experimental Result

- Result 1. Performance Evaluation



Result of authors' dataset



Result of SBIR_100K dataset

Experimental Result

- Result 2. Computational cost

	Initial SBIR	ours				
		Clustering	RVFV1	CBRF	RVFV2	Total
Edgel	9.77	0.017	0.73	0.14	0.41	11.06
ARP	0.64	0.015	0.53	0.10	0.26	1.55

+1.28s

+0.91s

Conclusion

- Image Grouping
 - Find which images are more relevant
- Re-ranking via Visual Feature Verification (RVFV)
 - Filter out irrelevant images
- Contour-Based Relevance Feedback (CBRF)
 - Explore deeply to retrieve what does not be found with original SBIR
- Improved result with low time cost