

QuRe: Query-Relevant Retrieval through Hard Negative Sampling in Composed Image Retrieval



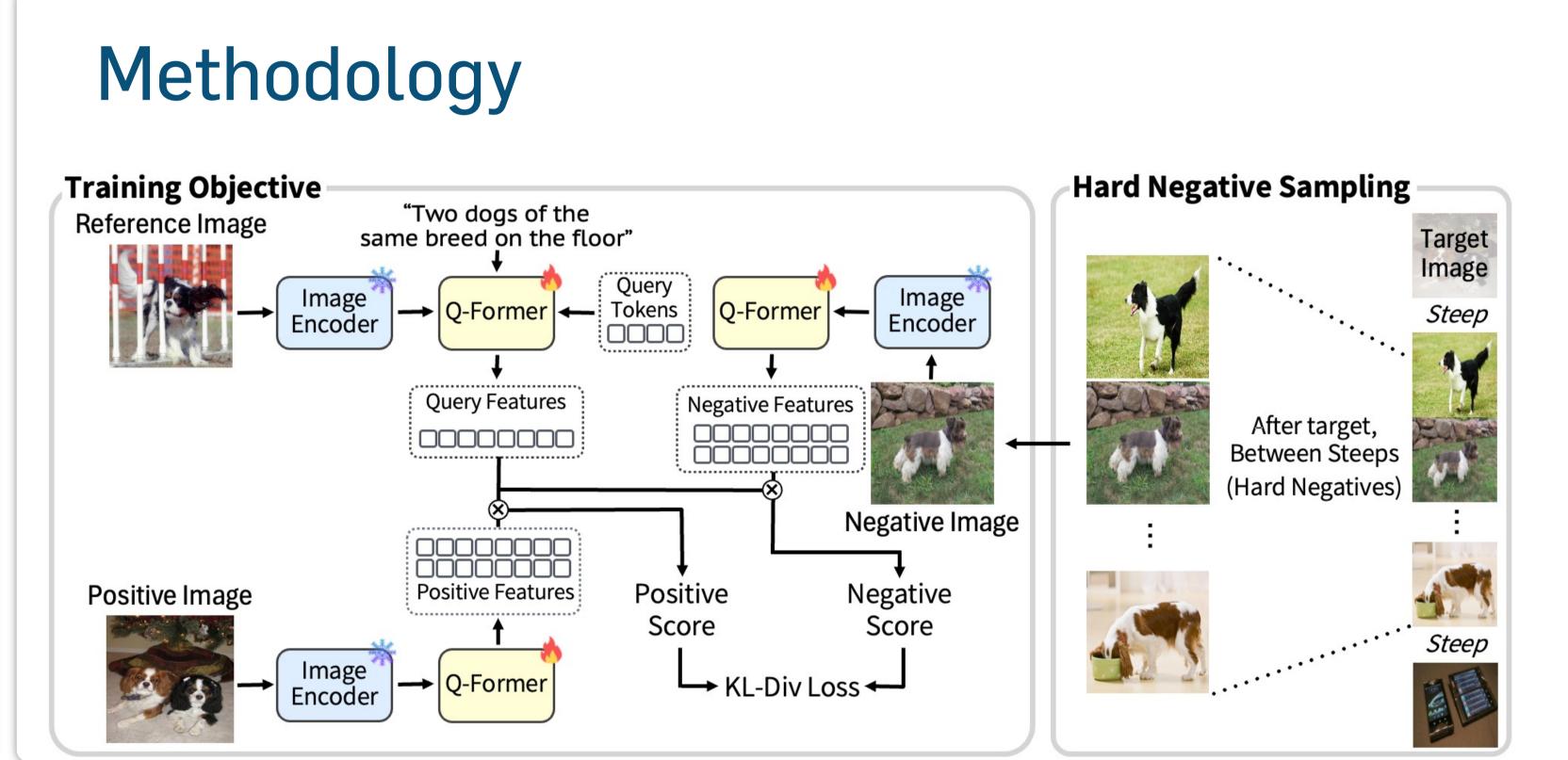
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TL;DR

- ✓ QuRe: CIR model that ranks both the target and other relevant images at the top
- ✓ *HP-FashionIQ*: Human-preference CIR dataset for assessing how well models align with user judgments

"Blue t-shirt with short sleeves" Image Corpus Image Corpus Contrastive Learning Target Irrelevant Existing CIR methods Sort Preference Optimization Target Relevant

- Existing CIR methods retrieve the target correctly yet still place irrelevant images among top results
- ✓ QuRe ranks both the target and other images highly

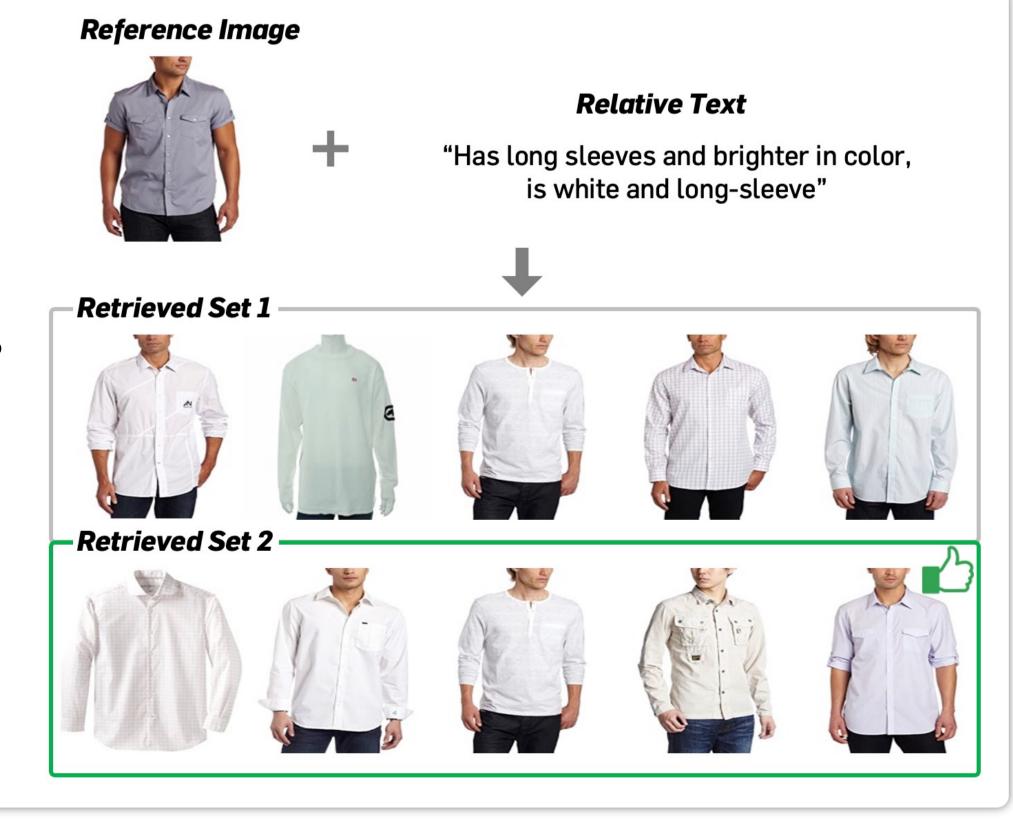


- ✓ Preference Optimization
 Assign one negative per query,
 exclude false negatives
- ✓ Hard-negative Mining
 Update hard negative set by
 selecting images between the two
 largest post-target score drops

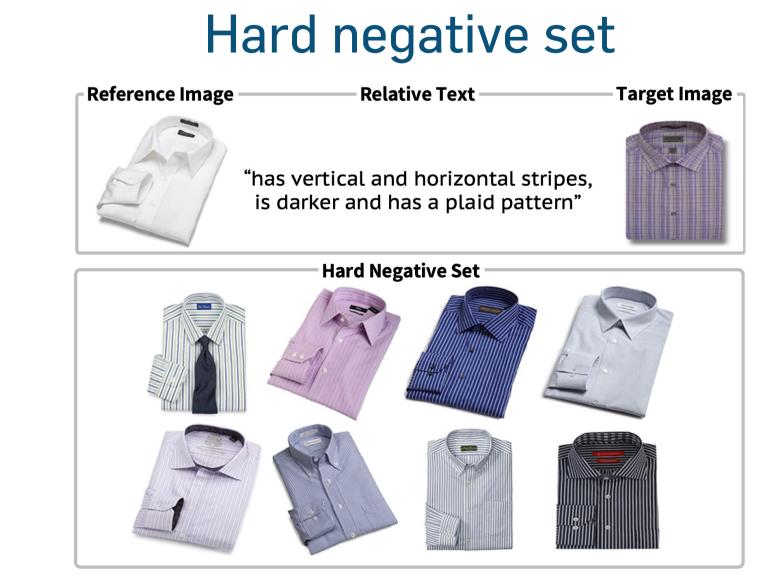
HP-FashionIQ dataset

- ✓ HP-FashionIQ provides human relevance judgements for each retrieved image, annotated by 61 participants on the FashionIQ validation set
- ✓ HP-FashionIQ supports the Preference Ratio metric, quantifying how often a model's higher-scoring set aligns with human preference:

$$P(A > B \mid s_{rel}(A) > s_{rel}(B))$$



Experiment Results



FashionIQ / CIRR

Method	Dress		Shirt		Toptee		Average		
	R@10	R@50	R@10	R@50	R@10	R@50	R@10	R@50	Avg.
CoVR-BLIP (Ventura et al., 2024b)	44.55	69.03	48.43	67.42	52.60	74.31	48.53	70.25	60.24
SPRC (Bai et al., 2024)	<u>45.71</u>	70.00	51.37	<u>72.77</u>	<u>55.48</u>	<u>77.46</u>	<u>50.86</u>	73.41	62.13
QURE	46.80	69.81	53.53	72.87	57.47	77.77	52.60	73.48	63.0
Method	K=1	Reca	K=10	K=50	K=1	Recall _s @	K=3	Ave.	rage · R _s @1
CoVR-BLIP (Ventura et al., 2024b)	39.76	70.15	80.89	95.01	72.46	87.86	94.77	71	.30
CoVR-BLIP (Ventura et al., 2024b) SPRC (Bai et al., 2024)	39.76 <u>50.75</u>	70.15 80.58	80.89 88.72	95.01 <u>97.59</u>	72.46 79.57	87.86 91.76	94.77 96.70	10.00	.30 <u>0.07</u>

HP-FashionIQ

Method	Preference Rate (%)
CosMo (Lee et al., 2021)	72.96
MGUR (Chen et al., 2024)	73.99
CLIP4CIR (Baldrati et al., 2023)	<u>74.45</u>
Bi-BLIP4CIR (Liu et al., 2024)	67.33
CoVR-BLIP (Ventura et al., 2024b)	73.15
SPRC (Bai et al., 2024)	73.82
QURE	74.55

- ✓ QuRe's hard negatives lies close to the target but at lower relevance
- ✓ QuRe achieves the highest Recall@k and best aligns with human preference