



# Salient Color Names for Person Re-Identification

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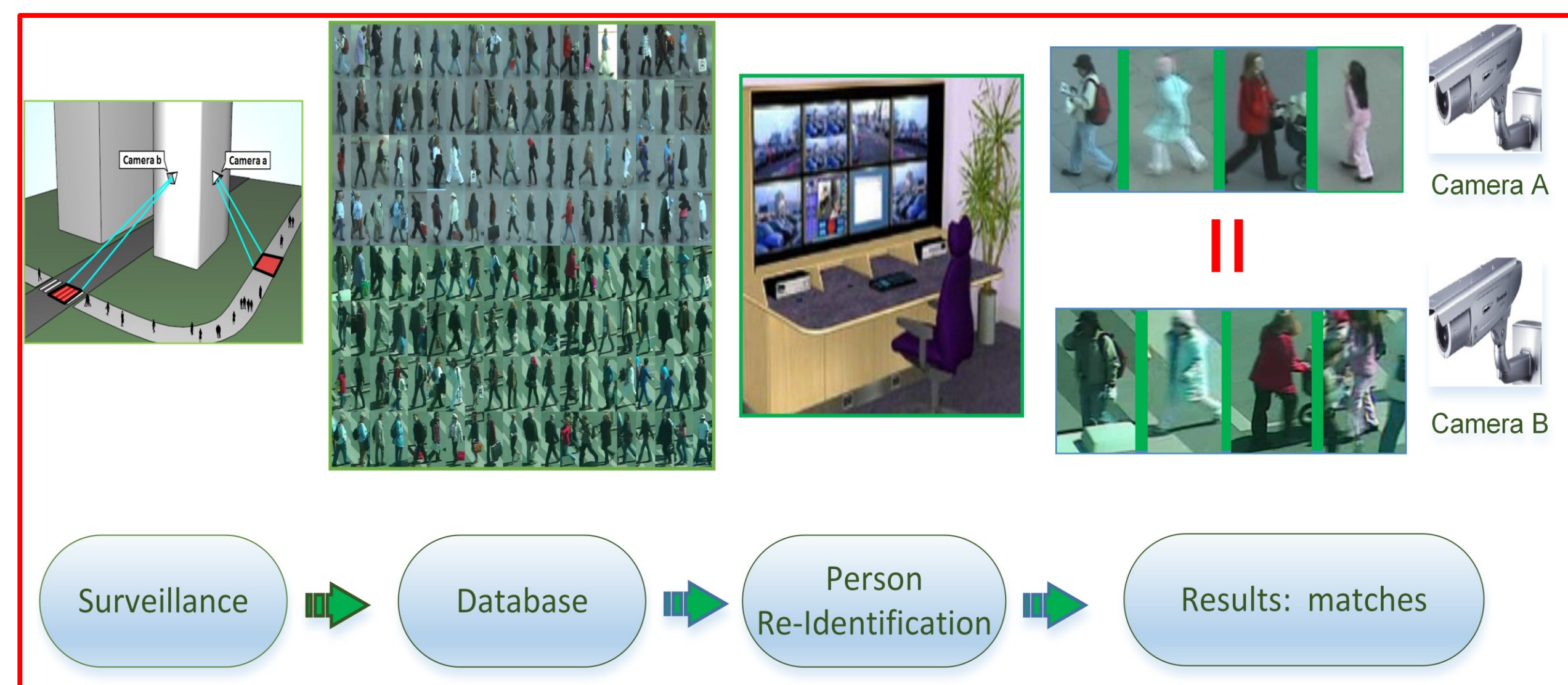
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## Introduction

### Person Re-Identification

Recognize an individual across a network of disjoint cameras.



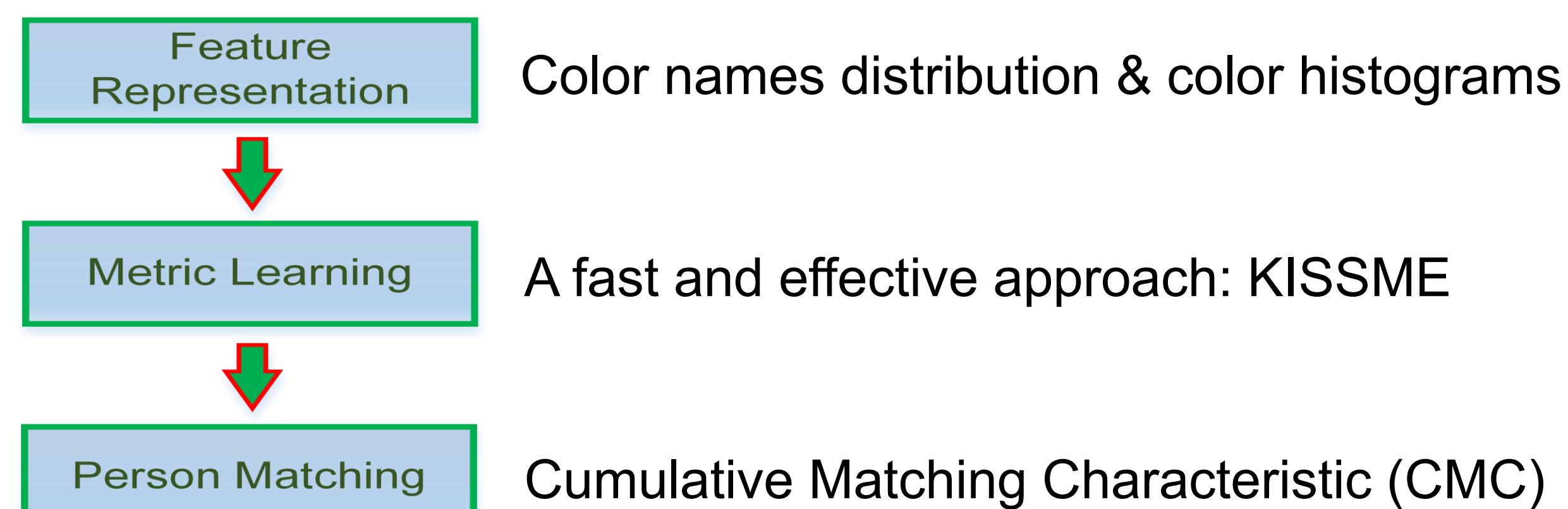
### Challenging

- Illumination
- Viewpoint
- Pose
- Low resolution



### Solution

Fully utilize the information about the colors of clothes.



### Experimental Datasets

#### VIPeR Dataset.

- 632 image pairs of pedestrians
- Challenging in:
  - Pose
  - Illumination conditions
  - Low resolution

#### PRID 450S Dataset.

- 450 image pairs of pedestrians
- Challenging in:
  - background interference
  - Viewpoints
  - Low resolution



VIPeR dataset



PRID 450S dataset

## Contributions

- A novel and effective salient color names (SCNCD) based color descriptor is proposed for person re-identification;
- Background information is exploited to enrich the feature representation which is of good robustness against background interference and partial occlusions;
- To tackle different types of illumination changes, color names distribution and color histograms are fused in four color spaces.

## Feature Representation

### SCNCD

Assume  $Z = [z_1, z_2, \dots, z_{16}]$  denotes a set of 16 color names, the probability of assigning  $d$  to a color name  $z$  is:

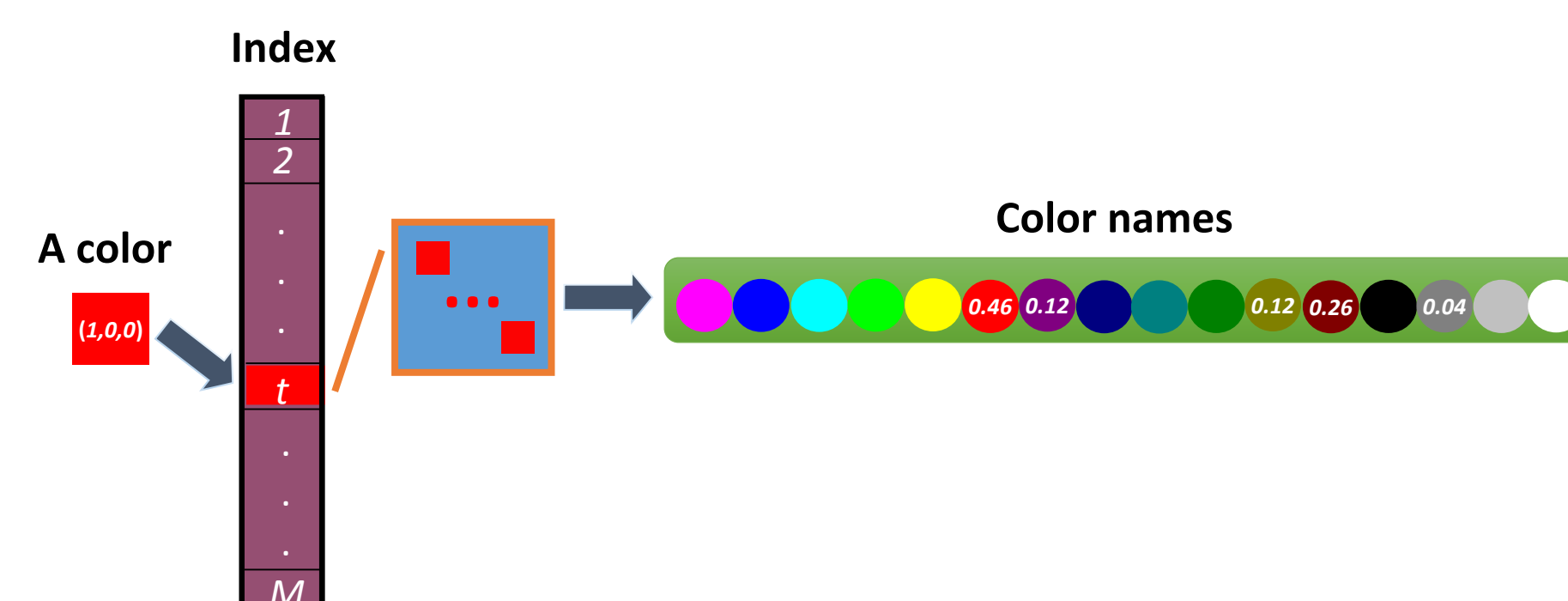
$$p(z|d) = \sum_{n=1}^{512} p(z|w_n)p(w_n|d),$$

with

$$p(z|w_n) = \begin{cases} \frac{\exp(-\|z - w_n\|^2 / \frac{1}{K-1} \sum_{z_l \neq z} \|z_l - w_n\|^2)}{\sum_{p=1}^K \exp(-\|z_p - w_n\|^2 / \frac{1}{K-1} \sum_{z_s \neq z_p} \|z_s - w_n\|^2)} & , \text{ if } z \in KNN(w_n) \\ 0 & , \text{ otherwise} \end{cases}$$

and

$$p(w_n|d) = \frac{\exp(-\alpha \|w_n - \mu\|^2)}{\sum_{l=1}^{512} \exp(-\alpha \|w_l - \mu\|^2)}.$$



### Color Names Distribution

We partition an image into six horizontal stripes of equal size. Then, based on SCNCD, the color names distribution is obtained.

### Image-foreground feature representation

Color names distribution + color histograms of image-only & foreground (in RGB, rgb,  $I_1 I_2 I_3$  and HSV color models) =  $SCNCD_{all}(ImgF)$

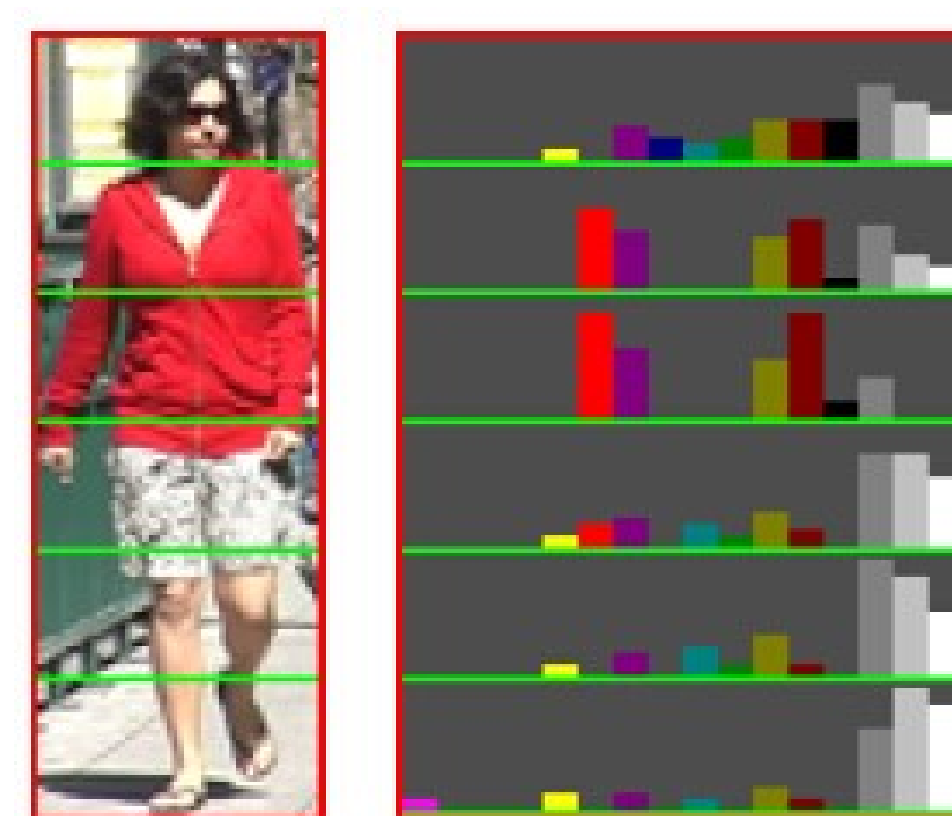
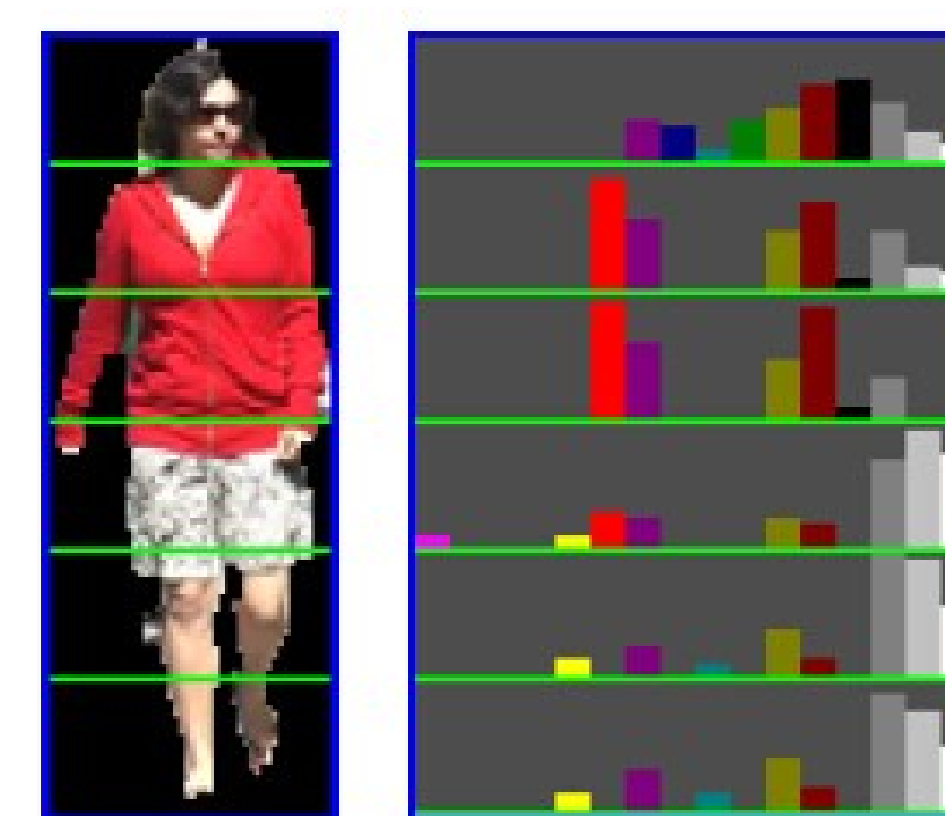


Image-only (RGB space)



Foreground (RGB space)

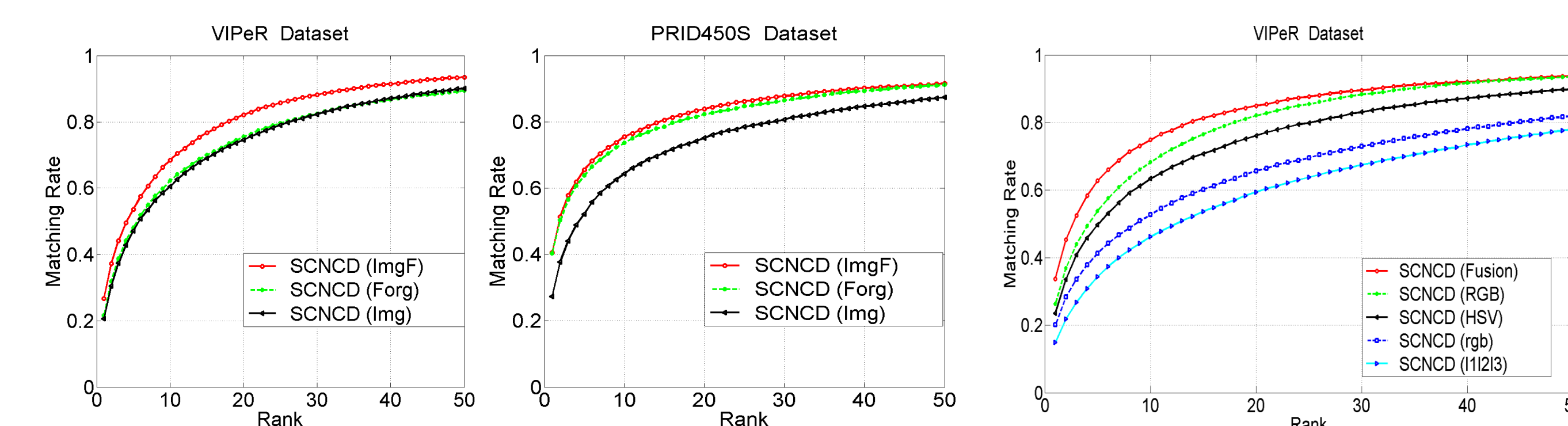
## Experimental Results

### Analysis of SCNCD

- Comparison with Other Color Descriptions

Rank	VIPeR dataset				PRID 450S dataset			
	1	5	10	15	1	5	10	15
Hist(RGB)	6.5	22.8	34.8	43.4	4.9	17.6	28.7	36.7
SCN[17]	11.9	32.3	45.9	55.0	6.6	20.7	31.6	39.3
SCR[8]	12.5	32.9	45.9	54.3	9.6	26.2	37.0	44.4
DD[10]	17.6	40.3	52.4	60.2	17.6	40.3	52.4	60.2
CN[27]	19.6	44.2	58.1	66.3	20.4	42.6	53.3	60.3
SCNCD(Ours)	20.7	47.2	60.6	68.8	26.9	52.9	64.2	70.4

- Color Models and foreground and background

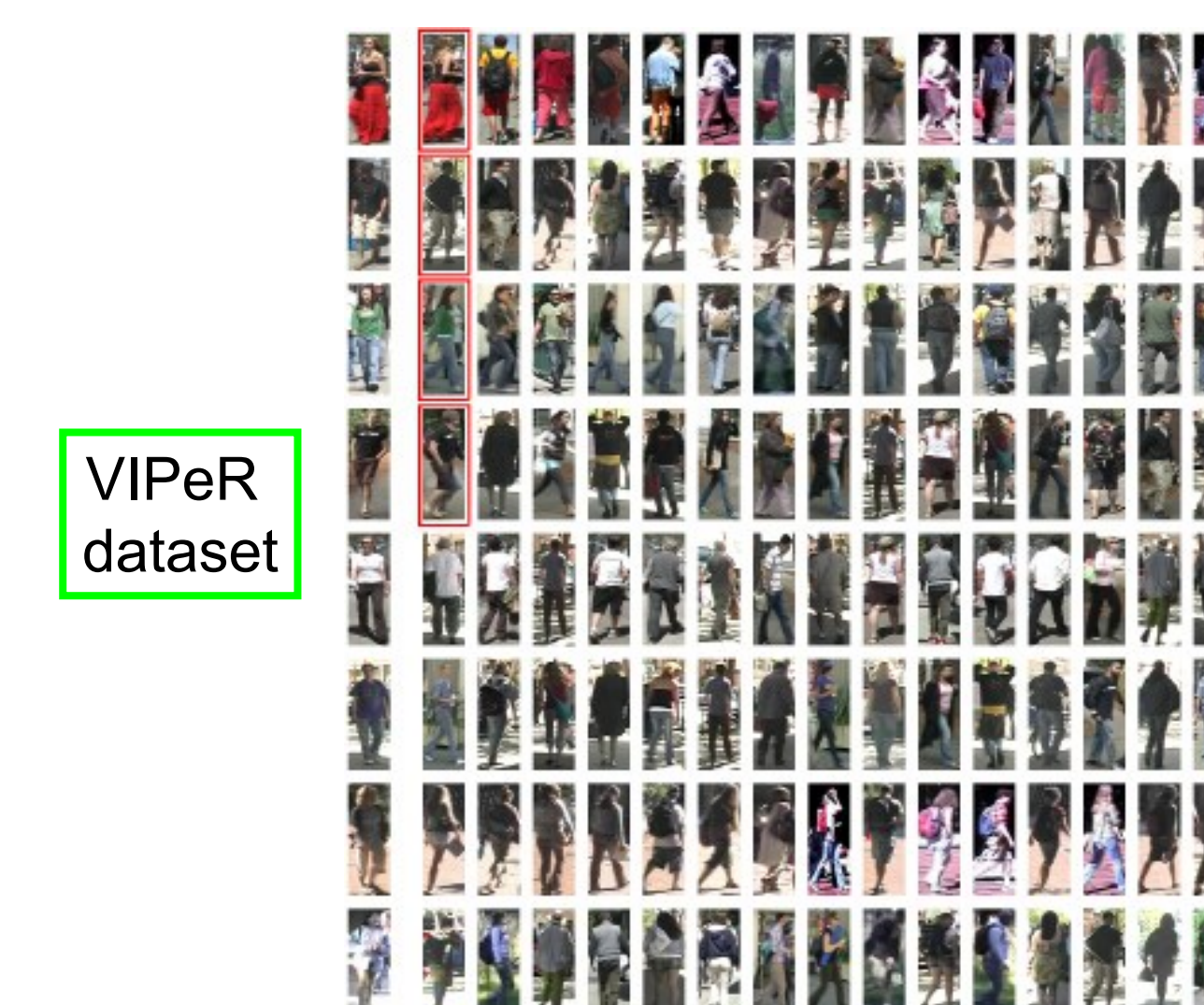


### Comparison with the State-of-the-art Methods

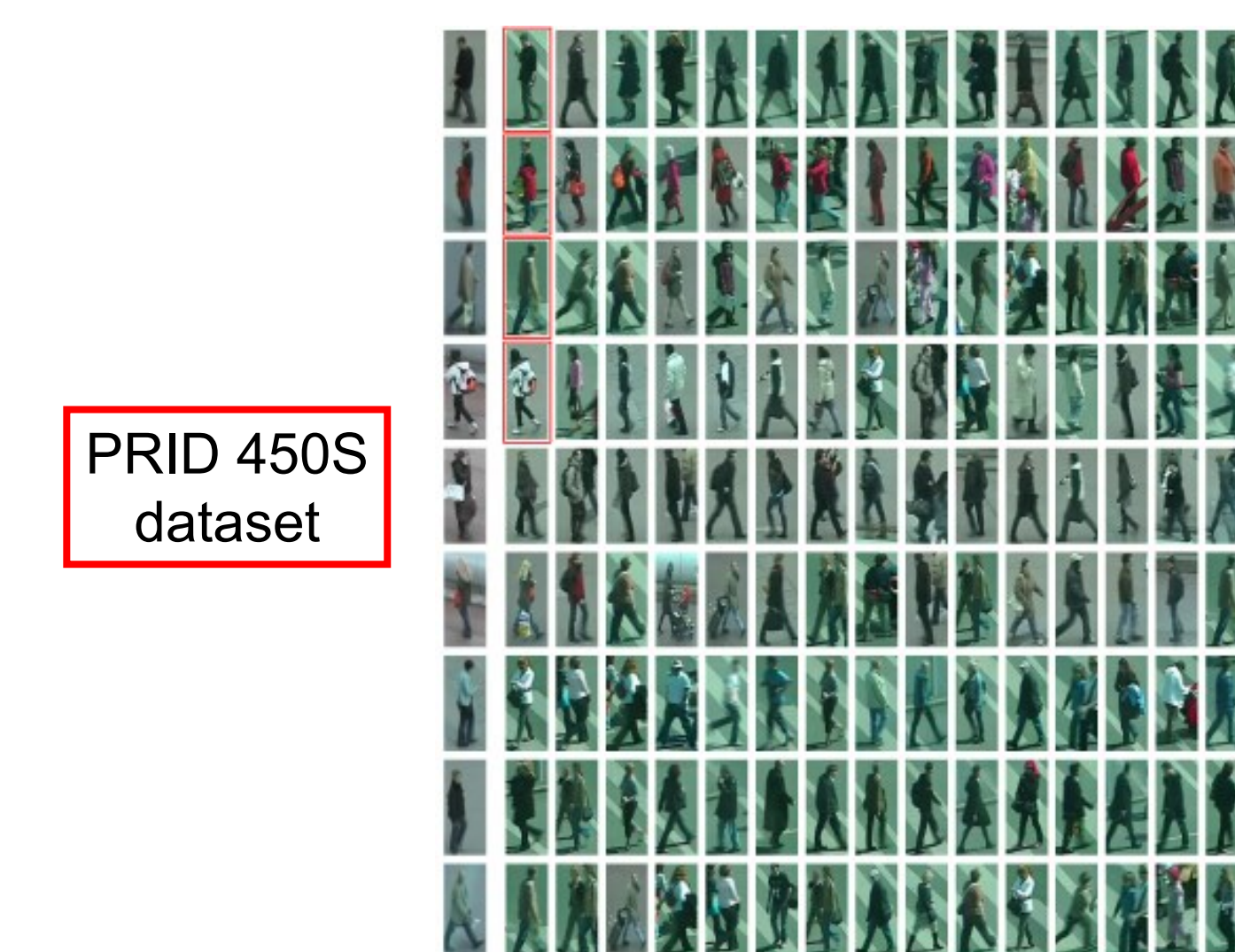
Rank	1	5	10	15	20	25	30	50
PRDC[32]	15.7	38.4	53.9	-	70.1	-	-	-
Fusing+PRDC[15]	16.1	37.7	51.0	-	66.0	-	-	-
RPLM[6]	27	-	69	-	83	-	-	95
EIML[5]	22	-	63	-	78	-	-	93
KISSME[11]	19.6	-	62.2	-	-	80.7	-	91.8
KISSME*[20]	27.0	-	70.0	-	83.0	-	-	95
eSDC-ocsvm[31]	26.7	50.7	62.4	-	76.4	-	-	-
RankBoost[12]	23.9	45.6	56.2	-	68.7	-	-	-
LF[19]	24.2	-	67.1	-	-	85.1	-	94.1
Salience[30]	30.2	52.3	-	-	-	-	-	-
SCNCD <sub>all</sub> (ImgF)	33.7	62.7	74.8	81.3	85.0	87.7	89.6	93.8
Final(ImgF)	37.8	68.5	81.2	87.0	90.4	92.7	94.2	97.0

Rank	1	5	10	15	20	25	30	50
KISSME*[20]	33.0	-	71.0	-	79.0	-	-	90.0
EIML[5]	35	-	68	-	77	-	-	90
SCNCD <sub>all</sub> (ImgF)	41.5	66.6	75.9	81.1	84.4	86.7	88.4	92.4
Final(ImgF)	41.6	68.9	79.4	84.9	87.8	90.0	91.8	95.4

### Success and Failure Cases



VIPeR dataset



PRID 450S dataset