# 浏览式阅读

## 1 自己的总结、评价以及应用

## 2 文章的主要问题（abstract、疑问句中）

Supervised re-id存在问题：scalability problem-->引出了unsupervised re-id-->但是unsupervised re-id也存在问题：it is very challenging to learn discriminative information in the absence of pairwise labels across disjoint camera views

Re-id中存在的问题：

可拓展性问题scalability problem

提出的解决方案：

we propose a top-push distance learning model

## 3 结论（abstract以及conclusion中）

## 4 思路脉络（小标题中的关键句）

1. **Introduction**

为了解决supervised learning的calability problem，引入unsupervised re-id by clustering on the target unlabelled data [52, 53, 8] or transfering the knowledge from other labeled source dataset（reference dataset），但是仍然存在很大的问题：

难以识别the identity discriminative information（存在较大的类内差异和较小的类间差异，也就是说，同一目标之间的差异过大，而不同目标之间的差异又很小，这就形成了一种矛盾）

解决方法：we propose a novel soft multilabel learning to mine the potential label information in the unlabeled RE-ID data

如何表示unlabled person的特征？答：借用reference person

要求：Intuitively, a pair of images of the same person should be not only visually similar to each other (i.e. they should have similar absolute visual features), but also equally similar to any other reference person (i.e. they should also have similar relative comparative characteristics with respect to the reference persons).

有一个词： relative soft multilabel representation

hard negative pair

主要贡献 our contributions：

1. a novel soft multilabel reference learning method以mine the potential label information latent in the unlabeled RE-ID data
2. a novel deep model named deep soft multilabel reference learning (MAR)
3. Related Work

Unsupervised re-id的定义：Unsupervised RE-ID refers to that the target dataset is unlabelled but the auxiliary source dataset is not necessarily unlabelled.

先前的工作：

①transfer source label knowledge

②assuming strong prior knowledge

最近的工作：

1. exploiting video tracklet associations for unsupervised RE-ID
2. reducing the labelling effort is to minimize the labelling budget on the target（complementary to the unsupervised RE-ID）
3. 其他方法：use the labeled source dataset by the unsupervised domain adaptation [50, 7, 62, 48] to transfer the discriminative knowledge from the auxiliary source domain

（缺点：do not mine the discriminative information in the unlabeled target domain）

而我们的方法与上述不同：

because the transferred discriminative knowledge might be less effective in the target domain due to the domain shift [28] in discriminative visual clues

简单介绍一下Multilabel classification：

An instance belongs to a set of classes，换言之，一个instance可以有多个标签，比如：红绿灯可以同时具备红、绿、黄三种颜色（标签）

简单介绍Zero-shot learning：

1. Deep Soft Multilabel Reference Learning
   1. Problem formulation and Overview

unlabeled target RE-ID dataset X = {xi}Nui=1

auxiliary RE-ID dataset Z = {zi , wi} Na i=1

一个问题或者叫难点： hard negative mining

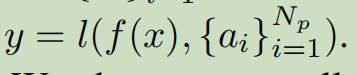
解决办法：Soft multilabel-guided hard negative mining

for visually similar pairs we determine they are positive or hard negative by comparing their soft multilabels

另一个概念：Cross-view consistent soft multilabel learning跨视角一致性软性多标签学习

又一个概念：reference agent learning

最终the soft multilabel function 简化成下面的公式：



**4. Experiment**

1. **Conclusion**

## 5 难理解点

专业术语：