

## 1 Introduction

### Problem

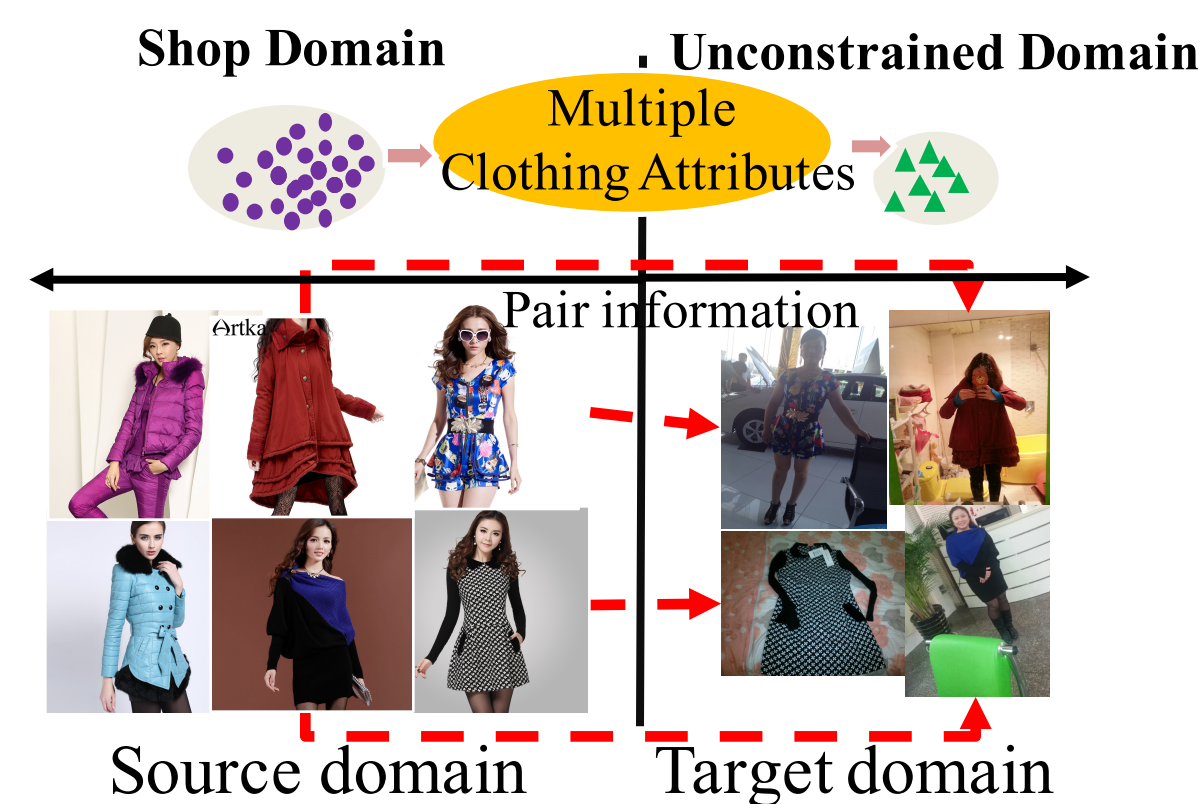
Domain transfer learning for recognising fine-grained multi-label clothing attributes in the street (wild) given limited training data.

### Limitation of Existing Methods

- Hand crafted features.
- Single task deep learning for multi-label recognition.
- Lack of end-to-end cross domain transfer learning.

### Contributions

- Novel **Multi-Task Curriculum Transfer (MTCT)** deep learning strategy.
- Effective **Multi-Task Network (MTN)** for learning from sparse target data.



## 2 Overview of method

**Clothing detection**  
Faster R-CNN[4] for clothing detection



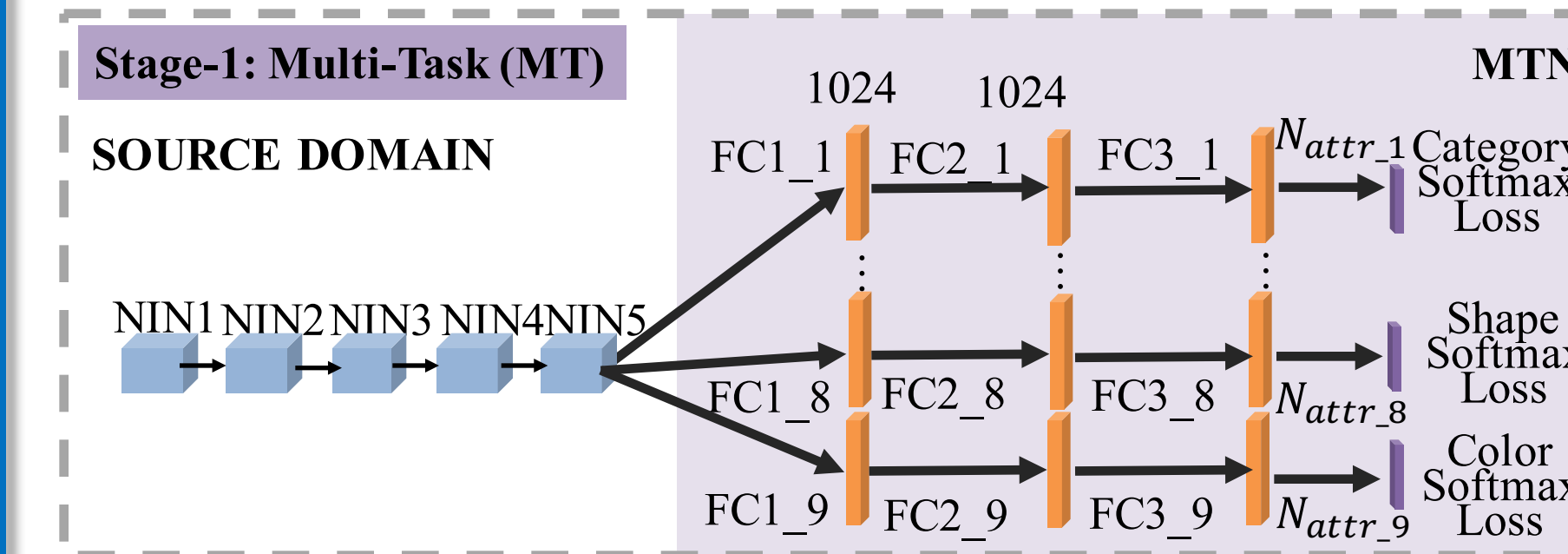
### Stage1: Shop domain (clean)

Pretrain **MTN** on ImageNet and train on shop images.

### Stage 2: Street domain (wild)

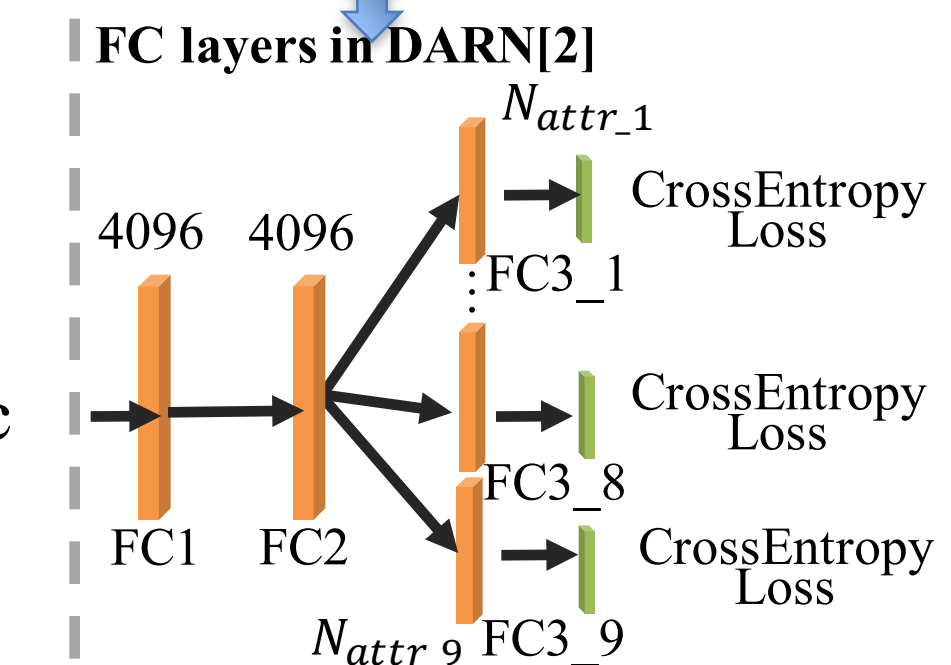
Initialize 3MTN by shop domain images trained model and then fine-tune FC layers using cross-domain triplet information for **transfer learning**

## 3 Multi-task deep learning



### MTN

A three-layer branch for learning specifics of each attribute category, with shared learning of generic features in conv layers.

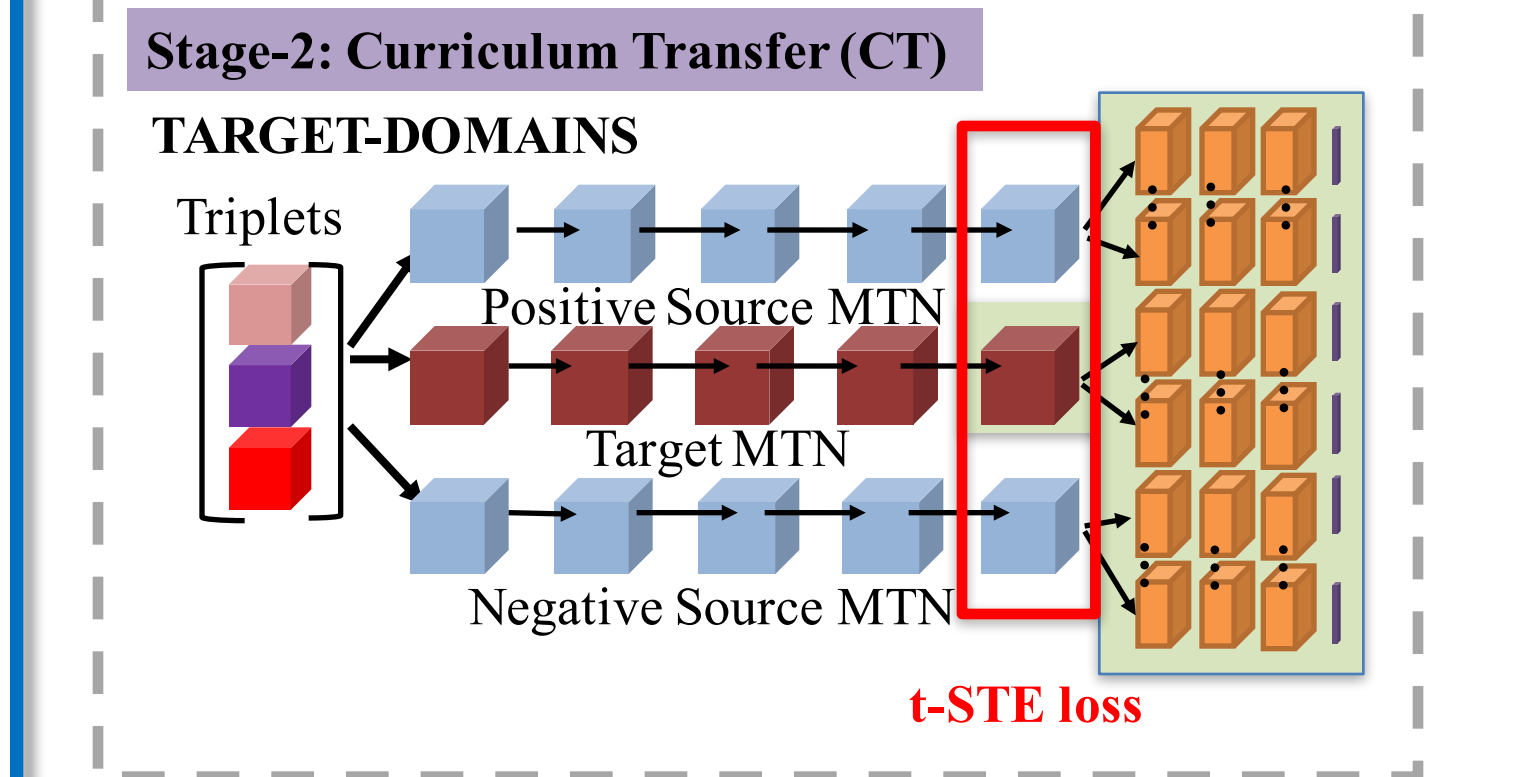


## SHOP DOMAIN

Layer Name	Parameters
NIN1	Conv1 7x7-96-2 Conv1_1 1x1-96-1 Conv1_2 1x1-96-1 Maxpooling-3x3
NIN2	Conv2 5x5-256-2 Conv2_1 1x1-256-1 Conv2_2 1x1-256-1 Maxpooling-3x3
NIN3	Conv3 3x3-512-1 Conv3_1 1x1-512-1 Conv3_2 1x1-512-1
NIN4	Conv4 3x3-1024-1 Conv4_1 1x1-1024-1 Conv4_2 1x1-512-1 Conv4_3 1x1-384-1
NIN5	Conv5 3x3-512-2 Conv5_1 1x1-512-1 Conv5_2 1x1-512-1 Maxpooling-3x3
FC-1024 for any branch	
FC-1024 for any branch	
FC- N_attr_i for any branch	
Softmax for any branch	

## 4 Curriculum transfer learning

### STREET DOMAIN



### Cross-domain loss function

$$l_{t-STE} = \sum_{\{I_s, I_p, I_m\} \in T} \log \frac{(1 + \frac{\|f_s(I_s) - f_s(I_p)\|^2}{\alpha})^\beta}{(1 + \frac{\|f_s(I_s) - f_s(I_p)\|^2}{\alpha})^\beta + (1 + \frac{\|f_s(I_s) - f_s(I_m)\|^2}{\alpha})^\beta}$$

- [1] Q. Chen, J. Huang, R. Feris, L. M. Brown, J. Dong, and S. Yan. Deep domain adaptation for describing people based on fine-grained clothing attributes. CVPR2015.  
 [2] Huang, R. S. Feris, Q. Chen, and S. Yan. Cross-domain image retrieval with a dual attribute-aware ranking network. ICCV2015.  
 [3] Z. Liu, P. Luo, S. Qiu, X. Wang, and X. Tang. Deepfashion: Powering robust clothes recognition and retrieval with rich annotations. CVPR2016.  
 [4] S. Ren, et al. Fasterr-cnn: Towards real-time object detection with region proposal networks. In Advances in Neural Information Processing Systems, pages 91-99, 2015.

## 5 Experiments

### Comparison to the State-of-The-Arts

Methods	Category	Button	Colour	Length	Pattern	Shape	Collar	Slv-Len	Slv-shp	mAP <sup>cls</sup>	mP <sup>ins</sup>	mR <sup>ins</sup>
DDAN[1]	12.56	24.13	20.72	35.91	61.67	47.14	31.17	80.63	73.96	43.10	45.41	52.20
DARN[2]	52.55	37.48	58.24	51.49	67.53	47.70	47.77	82.04	73.72	57.61	57.79	67.29
FashionNet[3]	55.85	39.52	60.33	53.08	68.65	49.79	51.27	83.79	75.34	59.84	59.97	69.74
<b>MTCT</b>	<b>65.96</b>	<b>43.57</b>	<b>66.86</b>	<b>58.27</b>	<b>70.55</b>	<b>51.40</b>	<b>58.79</b>	<b>86.05</b>	<b>77.54</b>	<b>64.35</b>	<b>64.97</b>	<b>75.66</b>

### MTN and Transfer learning

Method	mAP <sup>cls</sup>	mP <sup>ins</sup>	mR <sup>ins</sup>
JAN(No Adpt)	50.46	50.39	58.40
MTN(No Adpt)	51.38	51.82	60.00
MTN(UD)	58.76	60.16	70.00
MTN(FTT)	61.82	62.53	72.76
<b>MTCT</b>	<b>64.35</b>	<b>64.97</b>	<b>75.66</b>

\*No Adpt: without adaptation  
\*UD: United domains \*FTT: Finetuning on the target domain

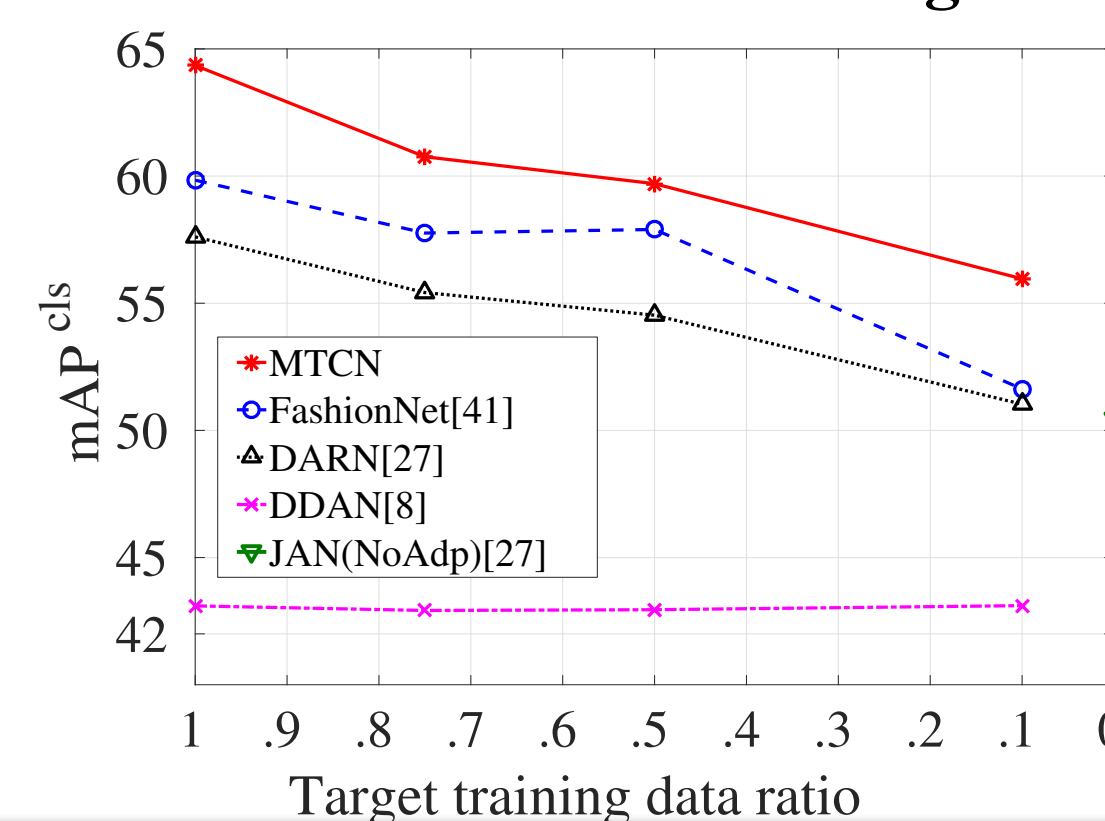
### End-to-End vs. Curriculum Transfer learning

Method	mAP <sup>cls</sup>	mP <sup>ins</sup>	mR <sup>ins</sup>
End-to-End	62.30	63.00	73.37
<b>Curriculum</b>	<b>64.35</b>	<b>64.97</b>	<b>75.66</b>

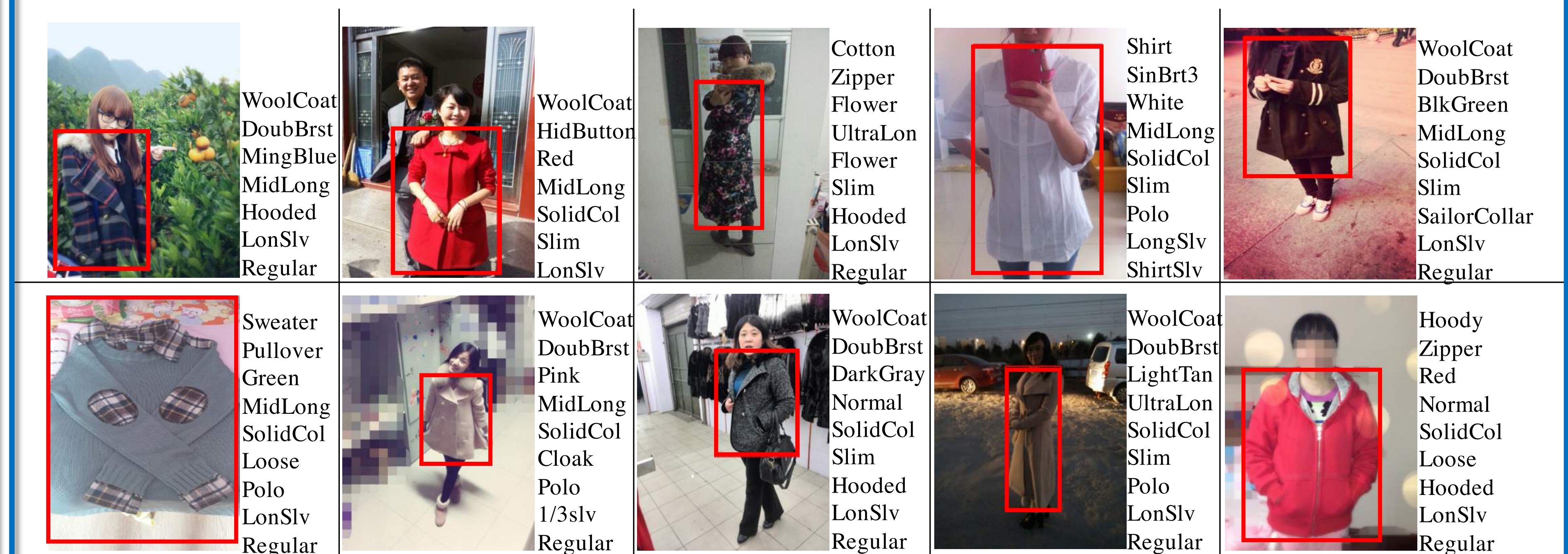
### Different cross-domain loss functions

Method	mAP <sup>cls</sup>	mP <sup>ins</sup>	mR <sup>ins</sup>
TripletRanking	62.60	63.45	73.83
<b>t-STE</b>	<b>64.35</b>	<b>64.97</b>	<b>75.66</b>

### Model Robustness vs. target data size



## 6 A qualitative evaluation of MTCT



Attribute order from top to bottom: Category, Button, Colour, Length, Pattern, Shape, Collar, Slv-Len, Slv-Shape