SizeNet: Weakly Supervised Learning of Visual Size and Fit in Fashion Images Zalando Nour Karessli Romain Guigourès Reza Shirvany Zalando SE - Berlin, Germany {nour.karessli, romain.guigoures, reza.shirvany}@zalando.de

Motivation

Finding clothes that fit is the biggest problem for customers shopping fashion online and offline. Supporting customers on their size and fit Others purchase decision in e-commerce context is particularly challenging:

- Thousands of new articles get activated everyday with short lifetime
- Return process takes from few days to few weeks resulting in zero or a few sales and returns data points for new articles

Contributions

In this paper we introduce a novel teacher-student approach on fashion images to:

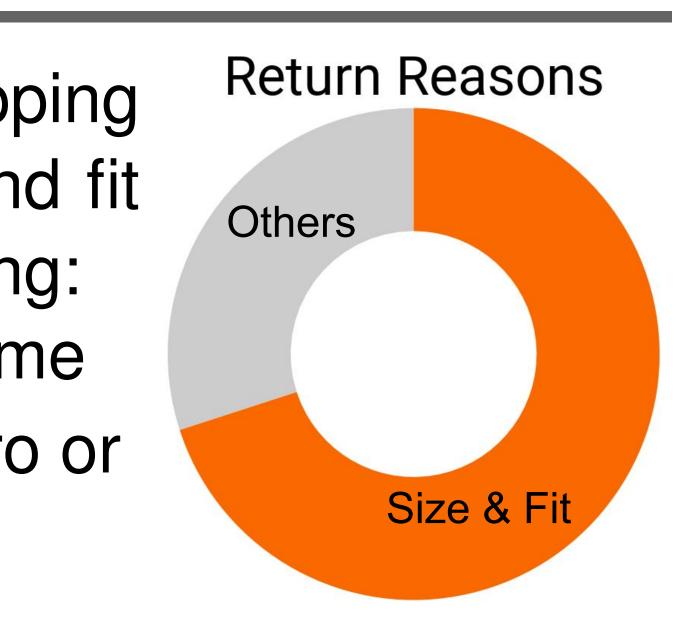
- Investigate and demonstrates the rich value of fashion images in inferring size characteristics of fashion apparel
- Effectively tackle the challenging cold start problem of providing size advice for new articles with zero/few return data
- Generate large scale confidence-weighted weak annotations from crowd's subjective feedback- enabling us to control the influence of weak annotations on the final model

Related Work

Teacher-Student Transfer Learning

Transferring knowledge from privileged information space to decision space [1]:

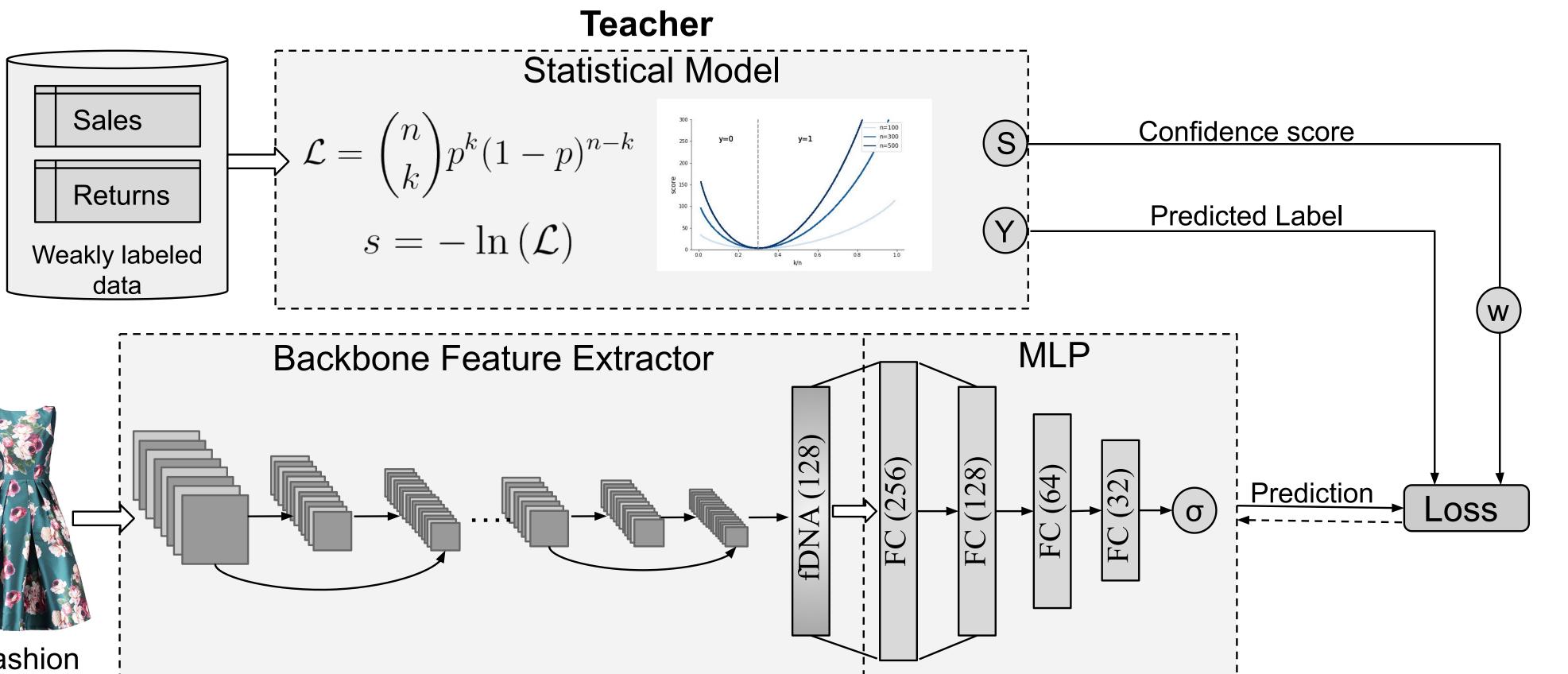
- Teacher leverages privileged historical weakly annotated data of sales and returns
- Student uses this knowledge to learn from *images* in decision space



TEACHER

STUDENT

SizeNet: Learning Visual Size Cues



Student

We formulate the sizing issue as a binary classification considering that: Article categories have different size related return rate • Article sales period influences the return rate Statistical Model: $\mathcal{L} = {n \choose k} p^k (1-p)^{n-k}$. \mathcal{L} : binomial likelihood, p: expected size return rate of article category, k: number of article size returns, n: number of article sales. The estimator score is define as $s = -\ln(\mathcal{L})$

CNN Backbone Feature Extractor

 ResNet [2] pre-trained on ImageNet [3] dataset MLP: Multi-Layer Perceptron

- Transfer knowledge using bottleneck features of pre-trained network
- FashionDNA [4] Resnet-like pre-trained on rich in-house fashion dataset

4 fully connected layers with nonlinear activations. Use weighted binary cross entropy loss based on estimator confidence score $w = \ln (1 + s)$; the logarithmic transformation reduces the score skewness

Evaluation

Dataset Women textile including 12 categories such as dresses, blouses, jeans, skirts, etc.



Attributes Baseline Replace article images with sparse k-hot encoding of human annotated binary fashion attributes \rightarrow fashion images achieve comparable results

Weights Importance

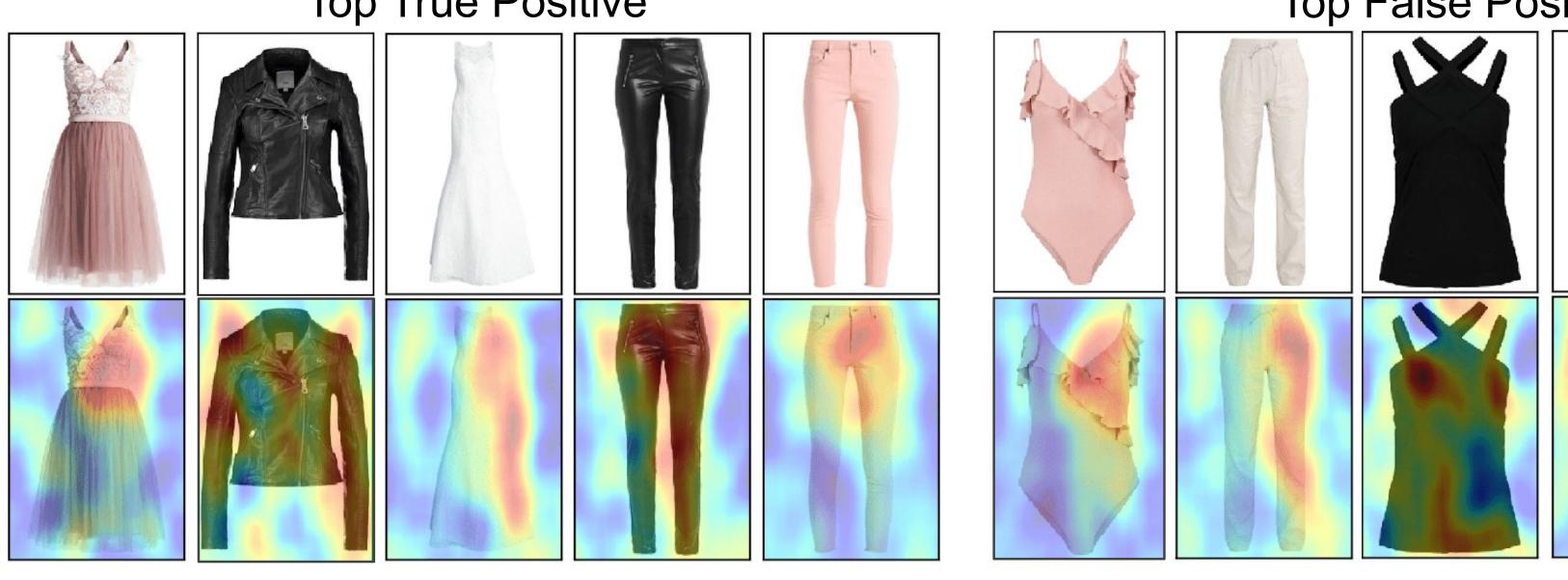
Exploiting weights offers better generalization capacity

Student Prediction vs. Teacher Confidence

BOTTOM RIGHT almost no samples are misclassified by SizeNet when § Teacher is certain of no size issue; TOP LEFT high density of correctly predicted samples by SizeNet where Teacher is unsure; TOP RIGHT samples show that SizeNet has learned accurately from Teacher; **BOTTOM LEFT** SizeNet misclassifies fewer samples where Teacher is unsure

Size Issue Explanations using RISE [5]

True Positives show localized heatmaps where False Positives are affected by article design Top True Positive Top False Positive

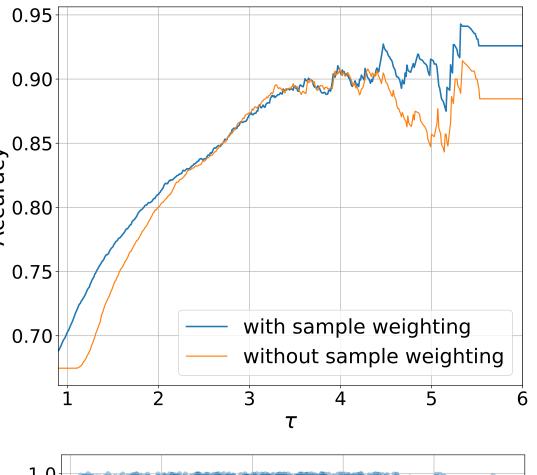




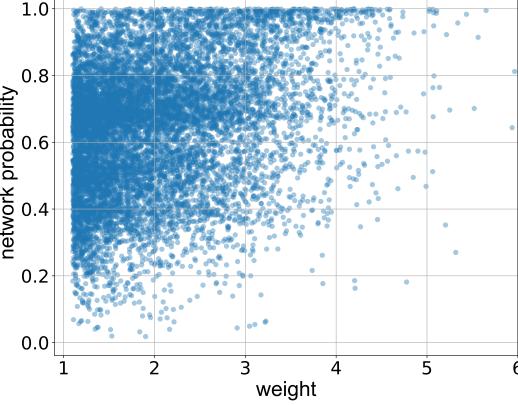
Class	#Articles	# Images
size issue	68,892	69,064
no size issue	58,152	58,321
total	127,044	127,385

Attributes-AUC = 0.7

0.4 0.6 0.8 False Positive Bate



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^[1] V Vapnik and R Izmailov. Learning using privileged information: Similarity control and knowledge transfer. In *JMLR15*. [2] K He, X Zhang, S Ren, and J Sun. Deep residual learning for image recognition. In CVPR16.

^[3] J. Deng, W. Dong, R. Socher, L.-J. Li, K. Li, and L. Fei-Fei. Imagenet: A large-scale hierarchical image database. In CVPR09.

^[4] C Bracher, S Heinz, and R Vollgraf. Fashion dna: Merging content and sales data for recommendation and article mapping. In KDD16. [5] V Petsiuk, A Das, and K Saenko. Rise: Randomized input sampling for explanation of black-box models. In BMVC18.