

Google Al ML Winter Camp 谷歌 Al 机器学习应用冬令营

I Fashion

Pick up your favourite one!

Burn My TPU Team
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I saw a cool T-shift!







Pick up your favourite one!





But I can't find it in the Internet. So sad..













http://34.80.200.205

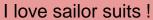
Wow, so cool!







Pick up your favourite one!





Career Dresses!





How to recommend??

Emmmm, I wanna...





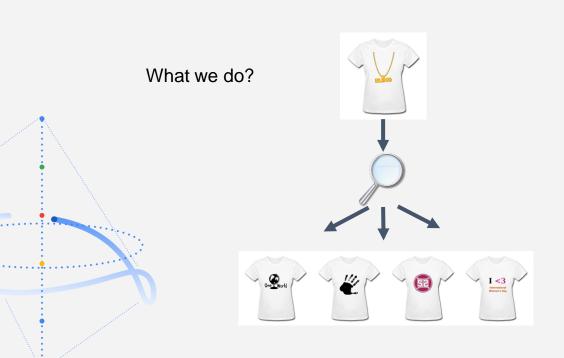




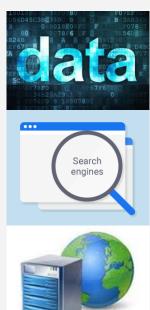
Project Name

Pick up your favourite one!





How to construct it?





Data Insight

Dive into iMaterialist Dataset



- 1 million+ images
 - 1,014,544 train, 9,897 validation, 39,706 test (~100:1:4)
- 8 kinds of labels:
 - category, color, gender, material, neckline, pattern, sleeve, style
- 228 labels in total.



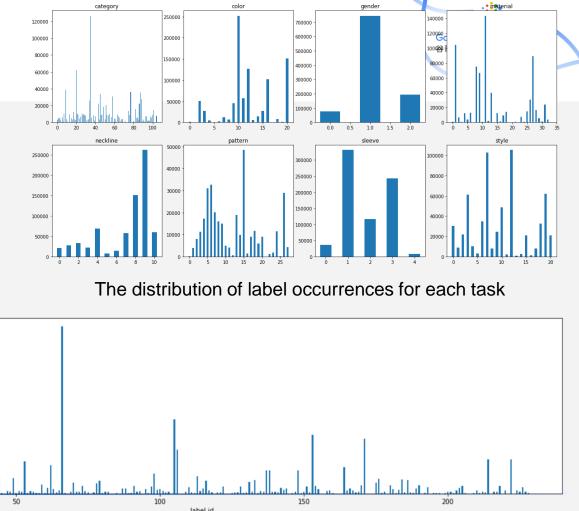


Data Insight

Dive into iMaterialist Dataset

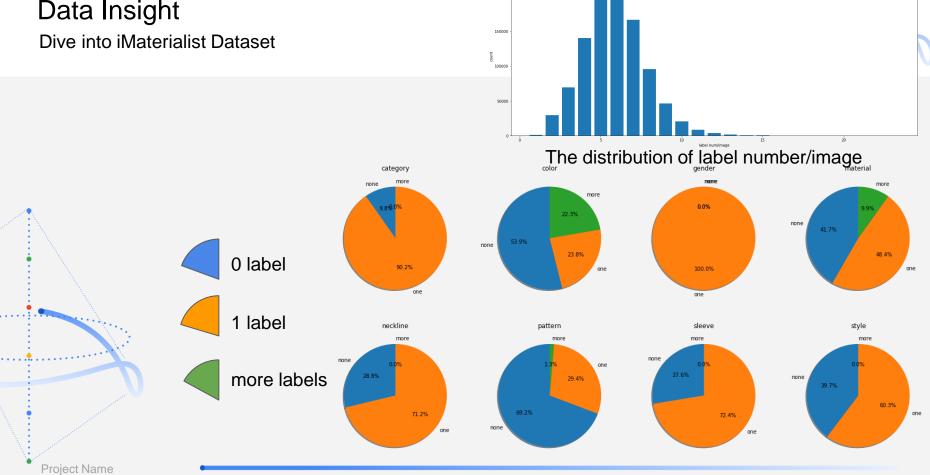
Project Name

 The labels are very imbalanced!



The distribution of label occurrences from a macro view





The distribution of label number/image for each task

Data Insight

Project Name

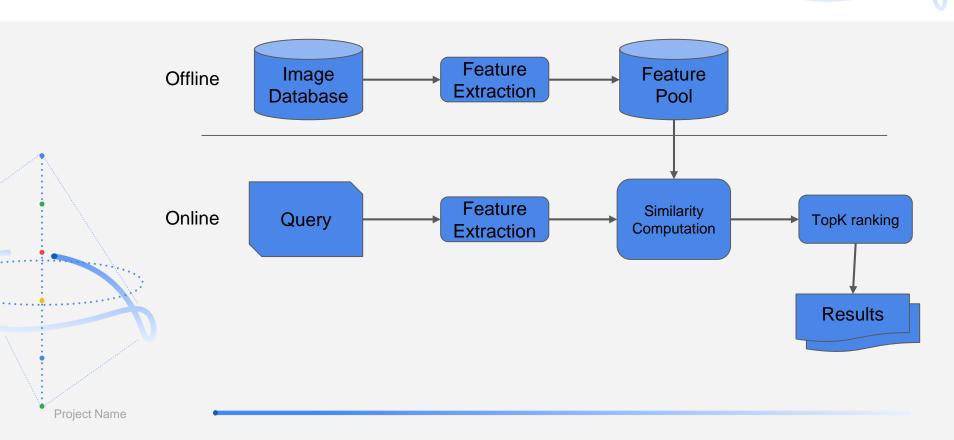
Dive into iMaterialist Dataset





Naive Pipeline

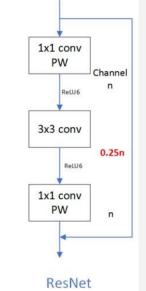


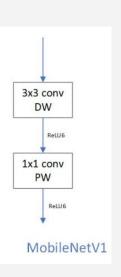


Origin choice: ImageNet pretrained model









	ResNet50	MobileNet
top1 error	22.96	26.61
top5 error	6.85	8.95
Params	25,549K	4,232K
FLOPs	3,875M	580M

Sophisticated choice: finetune on iMaterialist Dataset





micro F1 (average on images)

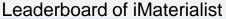
Naive method:

- multi-label learning for all labels
- use Binary Cross Entropy Loss (BCE Loss)
- ResNet50 + 112*112 + 1 epoch: 45.27

Fancy method: learning

- learning on each seperate task
- share the same backbone
- BCE Loss or CE Loss
- ResNet50 + 112*112 + 1 epoch: 48.48
- MobileNet + 112*112 + 1 epoch: 47.71
- MobileNet + 224*224 + 1 epoch: 49.47

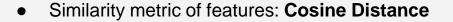




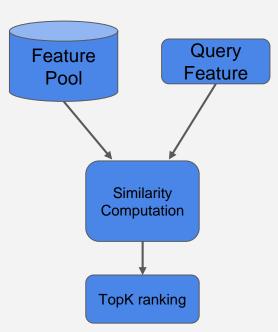


Similarity Computation





- Brutally search:
 - Compare with each image in gallery
- Levelly search:
 - Clustering features using K-means, etc.
 - Compare with cluster center
 - then compare with images in one cluster
- Category search: (used in Alibaba)
 - Clustering images with their attributes
 - o ..





Evaluation Method





gallary: 100000 images query: 1000 images

Similarity Metric for images:

$$IOU(A, B) = |L(A) \cap L(B)| / |L(A) \cup L(B)|$$

Makeup the groundtruth if IOU(A, B) >= 0.5: pair(A, B)

Evaluation metric for search: precision, recall, F1 for topk results

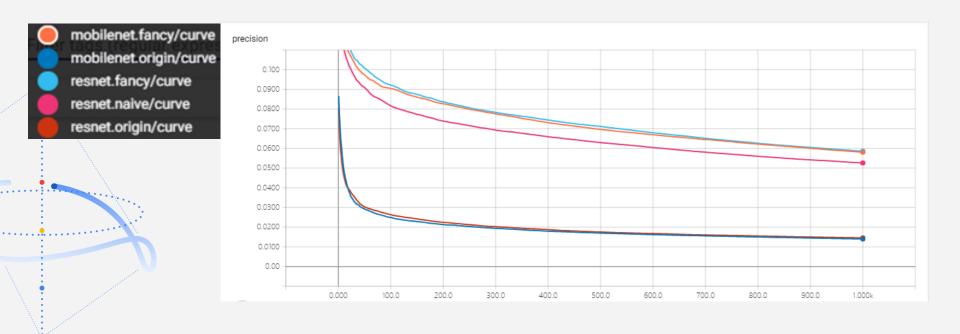




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Evaluation upon precision

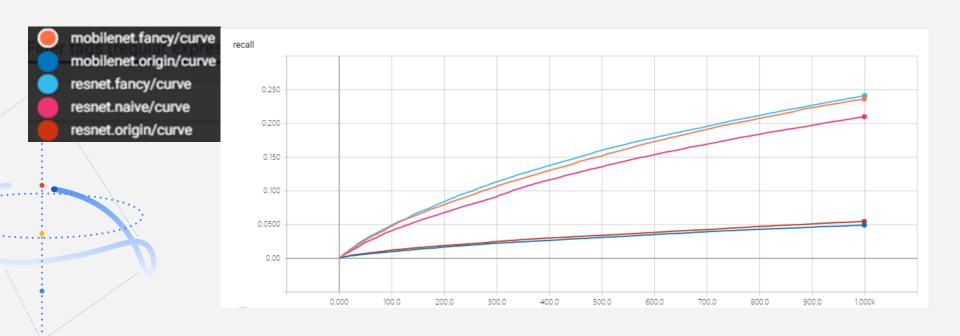




Evaluation upon recall

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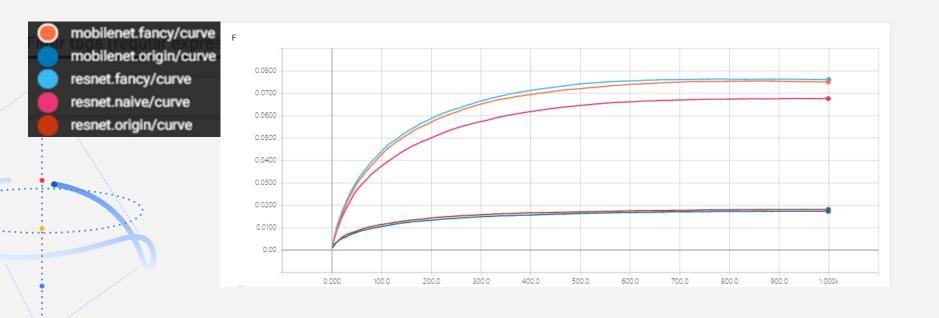




Evaluation upon F1

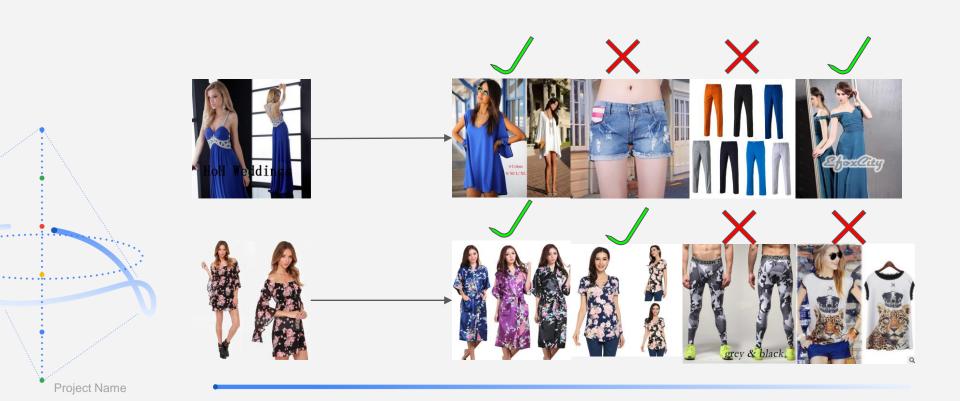
Project Name





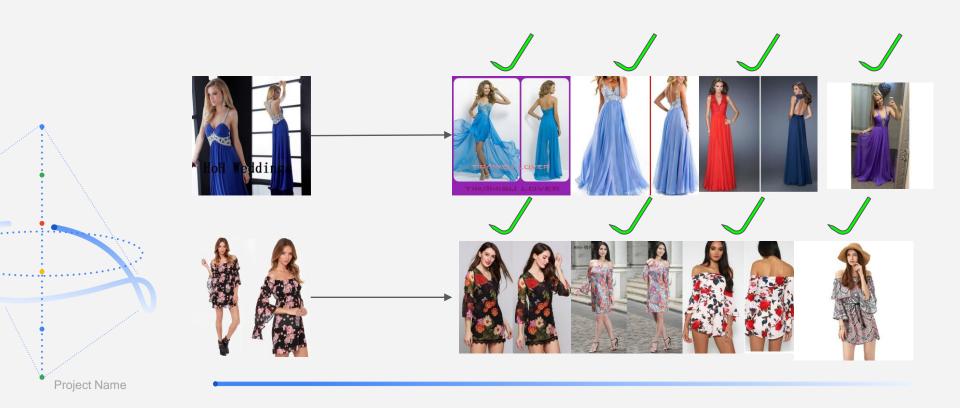
Visual example: mobilenet pretrained (origin)





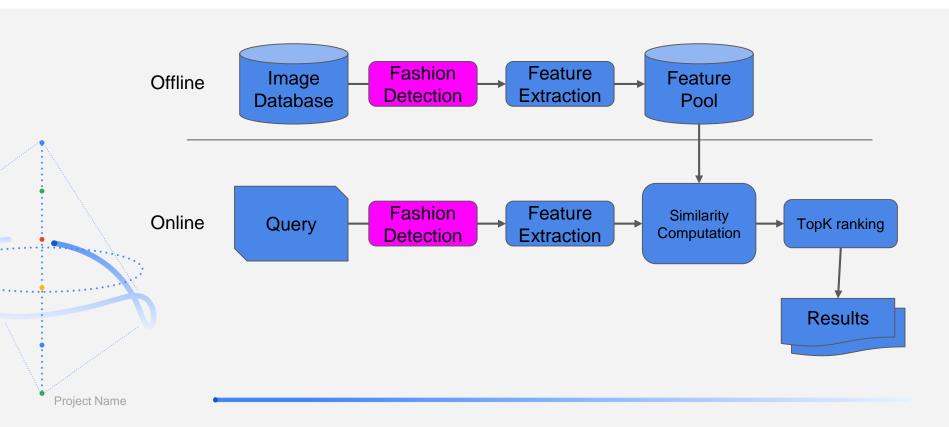
Visual example: mobilenet fine-tuned (fancy)





Sophisticated Pipeline







Mask rcnn-benchmark

https://github.com/facebookresearch/maskrcnn-benchmark

Result on Fashion Dataset



1: Can't detect fashion

2: Mask is not needed





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Yolov3

https://github.com/TencentYoutuResearch/ObjectDetection-OneStageDet

1: In one stage

2: Faster

3: Good performance

4: The work of Our Youtu Lab

Good performance

544x544	VOC2007 Test(mAP)	Time per forward (batch size = 1)
Yolov2	77.6%	11.5ms
Yolov3	79.6%	23.1ms

But ... can't detect fashion too





Training

Dataset

DeepFashion: Fashion Landmark Detection (By CUHK MMLab)

http://mmlab.ie.cuhk.edu.hk/projects/DeepFashion/LandmarkDetection.html





Deploy

- Modify the code support the format from voc to single raw image
- Integrated with Classification

Result





Project Name

Good cases Bad cases





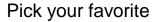
- ❖ Back end
- Framework
 - > Django
- ❖ Inference
 - Load Gallery Features, net and weights to GPU when start up

- Front end
- Framework
 - ➤ AngularJS
- Interaction
 - Send users' data and fetch results from back end

Demo

http://34.80.200.205

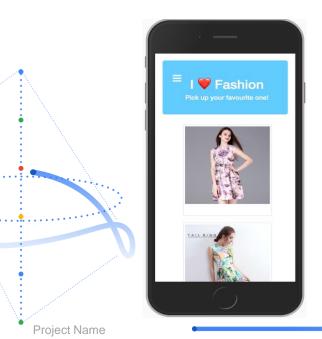




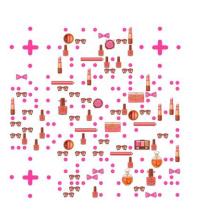


Upload your image

NOW, TRY IT YOURSELF!



skirt.jpg



Team member BurnMyTPU















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张潮鹏

胡杨

