## Link

Paper link - https://arxiv.org/pdf/1609.02489.pdf

## **Brief**

Trying to predict whether the customer will purchase a product or not by supplying the product's data both image and attributes.

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# Input

- 1. Image  $177 \times 256$
- 2. "expert labels" such as color, pattern, fabrics composition etc. The price labels were created by k-mean clustering. Label information was preprocessed into one-hot vectors.

## Label

 $\Pi_{ij} \in \{0,1\}$  for i-th articles, j-th customer. 0 - hasn't bought, 1 - has bought. The data is devided into 4 parts (refer to figure 1).

- 1. training data for both customer and articles  $\Pi^{tt}$
- 2. validation data for articles feature (articles unseen, trained customers).  $\Pi^{vt}$
- 3. validation data for customer parameter (customers unseen, trained articles).  $\Pi^{tv}$
- 4. all validation (new article, customers from  $\Pi^{vv}$ )

### Network

For better understanding, please refer to figure 4

### 1. Attribute network

4-layers connected neural network supplied by one-hot attributes (labels). The target is to extract the features from labels.

### 2. Image network

CNN network, Alexnet according to the paper.

#### 3. Combined network

From 2 models above, concatenate the extracted feature and pass them through FC-256, Relu, dropout-0.2 then the final value is called "fDNA" (simply feature). Let's the input data (image, labels) is  $\phi_i$  and the  $\theta$  is the parameters. Therefore  $f_i$  (i-th fDNA for i-th article) is  $f_i = f(\phi_i, \theta)$ 

### Prediction & Loss

Assume that we extracted the fDNA  $f_i$  from the combined network, we model the probability of purchase the article-ith from customer-jth by

$$p_{ij} = \sigma(f_i \cdot w_j + b_j)$$

where  $w_j, b_j$  is a factor associated with customer j-th (Each customer has their own parameters). The loss is calculated by the mean cross entropy loss.

# Training

#### 1. Training

Use the training data  $\Pi_{tt}$  to update both network weight  $\theta$  and customer weight  $w_i, b_i$ 

#### 2. Article Validation

Straightforwardly pass the article validation data  $\Pi_{vt}$ .

### 3. Customer Validation

Freeze the network parameter  $\theta$  but update the the customers weight  $w_j, b_j$  by passing data  $\Pi_{tv}$ .

### 4. All validation

Similar to part 2, we pass data  $\Pi_{vv}$ . This step is for judging whether that the customer validation from part 3 generalizes well to unseen articles.

# Evaluation

Because the ratio of purchase is very low  $E[P(\Pi_{ij}=1)]=1.14\times 10^{-4}$ , the overall prediction quality should not be determined by 0 or 1 instead it can be expressed by receiver operating characteristic (ROC) analysis (refer to Fig 6) and the score is the area under the ROC curve, called AUC score (refer: https://www.medcalc.org/manual/roc-curves.php)