MultiGap: Multi-Pooled Inception Network with Text Augmentation for Aesthetic Prediction of Photographs

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Main contribution

- Deep neural network architecture called MultiGAP that exploits features from multiple inception modules pooled by global average pooling (GAP), evaluated by prediction of (10-class) categorical distribution, before performing score binarization
 The incorporation of textual features trained simultaneously with MultiGAP using a recurrent neural network (RNN) with gated recurrent unit (GRU) layers
- 3. The ability to leverage the GAP block for visualization of activation maps based on the aesthetic category and rating

Related work

Handcrafted low-level features

color, hue, saturation, light exposure, and also other heuristics driven by rule of thumbs used by professional photographers

Generic features

SIFT , feature encoding method such as a Fisher Vector

Deep learning models

CNN, Double Column CNN, Multi-modal CNN

Binary classification

VS



High



NA



Regression (Score)









Rating distribution









Classifier

AVA 250K images

20K testing

Classifier

2012, N. Murray

Label (aesthetically high or low)

$\boldsymbol{\delta}$ used to filter out from noisy images

230K training

20K

testing

low					high				
1	2	3	4	5	6	7	8	9	10

threshold = 5 $\delta = \{2, 1.5, 1, 0.5, 0\}$



threshold = 5

 $\boldsymbol{\delta} = \{0\}$

AVA 250K images

AVA 250K images

DPChallenge

AVA Comments 1.5 million comments

Zhou et al.

Proposed method



Visual features

GoogleNet



Visual features

Inception module



Visual features

Proposed method



GAP (Global Average Pooling) Layer



Textual Features



Proposed method



Results

	Model	Accuracy		
Image	DCNN [2]	73.25		
	RDCNN [2]	74.46		
	Kao et al. [12]	74.51		
	AlexNet [10] – finetuned	75.11		
	DMA [5]	75.41		
	GoogLeNet [16] - finetuned	75.60		
	MultiGAP	75.76		
	SingleGAP	76.31		
	BDN [13]	76.80		
Text	word2vec [19]	78.40		
	1D-CNN [20]	79.48		
	Naive Bayes SVM [14]	80.90		
	RNN (1-layer GRU)	81.09		
	RNN (2-layer GRU)	81.79		
Joint	Multimodal DBM [14]	78.88		
	SingleGAP + RNN (2-layer GRU)	80.54		
	MultiGAP + RNN (2-layer GRU)	82.27		

Results



Confusion matrix



Class Activation Map (CAM)



Class Activation Maps (CAM)



Thank you for your attention

Q&A