

Facial Attribute Classification

Image Processing, CNN with Transfer Learning, and Evaluation Methods

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Data Processing

200.000 images cropped to 109 x 89 of celebrity faces.

Initially:

- Loaded each image
- Standardized around training mean
- Reshaped to 1D array

Final model:

- Used three ImageDataGenerators for train, validation, and test
- Rescaled from 255 to 0-1
- These generated batches of images with specified attributes and labels





Building the Model

Model from Scratch

- Conv2D, Flatten, Dense, and Dropout Layers
- Fully trained: test accuracy of 0.0257, weighted average precision of 0.40, recall of 0.34, and f-1 score of 0.36.

Pros: More customizable, predictions centered around a similar threshold for each attribute

Transfer Learning

- Transfer Learning using ResNet50
- Froze the ResNet50 layers
- Ran for five epochs, fine-tuned for five epochs
- test accuracy of 0.009, weighted average precision of 0.63, recall of 0.30, and f-1 score of 0.39 (however, not descriptive because of variety in thresholds)

Pros: Less training time, higher accuracy, less data necessary

Evaluating the Models

Sklearn.classification_report

 We were able to see and compare the precision, f-1, and recall, as well as the difference between the weighted and unweighted averages – pretrained had more 0s.

Confusion matrices

We generated confusion matrices for each of the 40 attributes being looked at by our model. This allowed us to determine one of the issues with our accuracy was a tendency to predict all positive or all negative based on the attribute's threshold

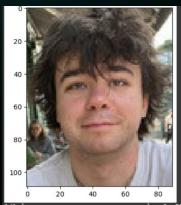
nuscaciie		0.01	0.01	112	
Narrow_Eyes	0.00	0.00	0.00	2968	
No_Beard	0.85	0.81	0.83	17041	
Oval_Face	0.28	0.02	0.04	5901	
Pale_Skir	0.00	0.00	0.00	840	
Pointy_Nose	0.26	0.03	0.05	5704	
Receding_Hairline	0.00	0.00	0.00	1694	
Rosy_Cheeks	0.06	0.00	0.00	1432	
Sideburns	0.04	0.01	0.02	926	
Smiling	0.51	0.44	0.47	9987	
Straight_Hair	0.00	0.00	0.00	4190	
Wavy_Hair	0.36	0.17	0.24	7267	
Wearing_Earrings	0.22	0.06	0.09	4125	
Wearing_Hat	0.03	0.00	0.01	839	
Wearing_Lipstick	0.52	0.46	0.49	10418	
Wearing_Necklace	1.00	0.00	0.00	2753	
Wearing_Necktie	0.07	0.02	0.03	1399	
Young	0.76	0.84	0.80	15114	
micro avo	0.52	0.32	0.40	184653	
macro avo	0.26	0.14	0.16	184653	
weighted av	0.41	0.32	0.34	184653	
samples avo	0.54	0.33	0.38	184653	

Mustache	0.00	0.00	0.00	39	
Narrow_Eyes	0.00	0.00	0.00	146	
No_Beard	0.87	1.00	0.93	867	
Oval_Face	0.00	0.00	0.00	286	
Pale_Skin	0.00	0.00	0.00	36	
Pointy_Nose	0.28	1.00	0.43	277	
Receding_Hairline	0.00	0.00	0.00	83	
Rosy_Cheeks	0.00	0.00	0.00	83	
Sideburns	0.00	0.00	0.00	39	
Smiling	0.52	0.95	0.67	512	
Straight_Hair	0.12	0.00	0.01	213	
Wavy_Hair	0.00	0.00	0.00	362	
Wearing_Earrings	0.00	0.00	0.00	214	
Wearing_Hat	0.00	0.00	0.00	42	
Wearing_Lipstick	0.51	1.00	0.67	506	
Wearing_Necklace	0.00	0.00	0.00	131	
Wearing_Necktie	0.00	0.00	0.00	61	
Young	0.74	0.45	0.56	761	
micro avg	0.35	0.32	0.33	9328	
macro avg	0.12	0.21	0.10	9328	
weighted avg	0.29	0.32	0.24	9328	
samples avg	0.35	0.34	0.33	9328	

Evaluating the Model (Part 2)

Custom Images:

By using custom images, we were able to get a better visual sense of how our model was performing, as well as the different thresholds. For example, aspects such as hair color always had low probability, whereas aspects like male always had high probability.



```
1/1 [======] - 0s 22ms/step
Probability of 5 o Clock Shadow: 0.1488
Probability of Arched_Eyebrows: 0.0209
Probability of Attractive: 0.1067
Probability of Bags Under Eyes: 0.1949
Probability of Bald: 0.0317
Probability of Bangs: 0.2567
Probability of Big Lips: 0.1308
Probability of Big Nose: 0.2999
Probability of Black Hair: 0.0905
Probability of Blond Hair: 0.0083
Probability of Blurry: 0.1874
Probability of Brown Hair: 0.1113
Probability of Bushy Evebrows: 0.0716
Probability of Chubby: 0.1009
Probability of Double Chin: 0.0845
Probability of Evenlasses: 0.1330
Probability of Goatee: 0.2729
Probability of Gray_Hair: 0.0344
Probability of Heavy_Makeup: 0.0143
Probability of High Cheekbones: 0.0742
Probability of Male: 0.7424
Probability of Mouth Slightly Open: 0.1935
Probability of Mustache: 0.1341
```

Probability of Narrow Eyes: 0.1074



```
1/1 [============= ] - 0s 317ms/step
Probability of 5 o Clock Shadow: 0.1109
Probability of Arched Eyebrows: 0.0796
Probability of Attractive: 0.2918
Probability of Bags Under Eyes: 0.1873
Probability of Bald: 0.0209
Probability of Bangs: 0.4066
Probability of Big Lips: 0.2044
Probability of Big_Nose: 0.2176
Probability of Black Hair: 0.1227
Probability of Blond_Hair: 0.0780
Probability of Blurry: 0.1584
Probability of Brown Hair: 0.2199
Probability of Bushy_Eyebrows: 0.0914
Probability of Chubby: 0.0661
Probability of Double Chin: 0.0447
Probability of Eveglasses: 0.0575
Probability of Goatee: 0.0949
Probability of Gray_Hair: 0.0347
Probability of Heavy_Makeup: 0.1121
Probability of High Cheekbones: 0.2026
Probability of Male: 0.4909
Probability of Mouth_Slightly_Open: 0.2261
Probability of Mustache: 0.0436
```

Discussion

If we had more time and resources:

We would run each of our models to full completion and potentially increase the number of layers.

We'd fine-tune each of our thresholds for each of the attributes to get more helpful f-1 scores during evaluation.

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