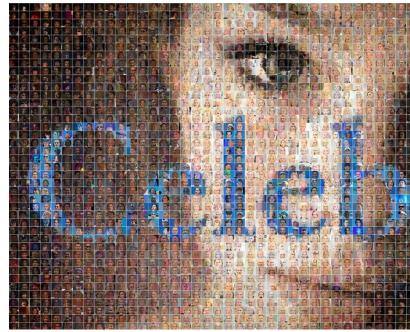
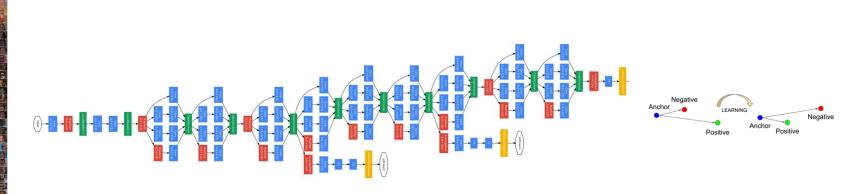


Introduction

Deep Face Recognition: Data, Network, and Loss.





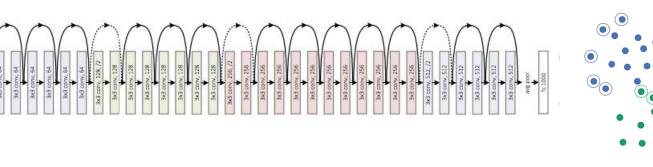
Network Data

Loss

Overview







MS-Celeb-1M (80K ids, 4M images) ResNet (BN) -27

Marginal-loss

Marginal Loss

Optimization Objective: minimizing the intra-class variations while

keeping the distances of inter-classes

$$L_{m} = \frac{1}{m^{2} - m} \sum_{i,j,i \neq j}^{m} \left(\xi - y_{ij} \left(\theta - \left\| \frac{x_{i}}{\|x_{i}\|} - \frac{x_{j}}{\|x_{j}\|} \right\|_{2}^{2} \right) \right)_{+}$$

Decrease intra-class distance

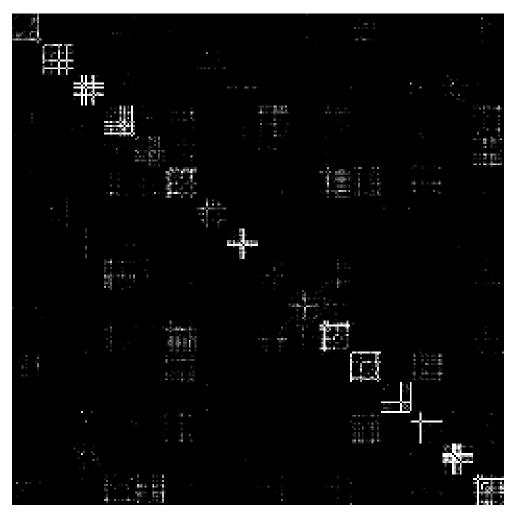
Increase inter-class distance

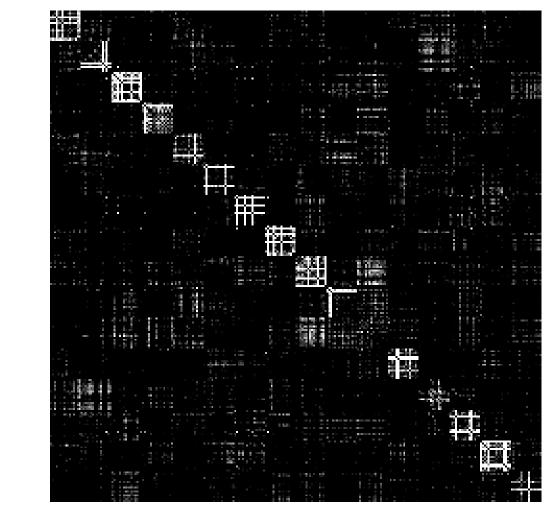
Marginal pairs

Marginal Loss for Deep Face Recognition Jiankang Deng, Yuxiang Zhou, Stefanos Zafeiriou Imperial College London, UK

Experiments

Marginal Loss Enhancement: Off-line class centre ranking





Random

Nearest neighbour

Labelled Face in the Wild (LFW) & Youtube Face (YTF)

Methods	Images	LFW (%)	YTF (%)
DeepID [SCWT14]		99.47	93.20
VGG Face [PVZ15]	2.6M	98.95	97.30
Deep Face [TYRW14]	4M	97.35	91.40
Fusion [TYRW15]	500M	98.37	
FaceNet [SKP15]	200M	99.63	95.10
Baidu [LDB+15]	1.3M	99.13	
Center Loss [WZLQ16]	0.7M	99.28	94.9
Range Loss [ZFW ⁺ 16]	1.5M	99.52	93.70
Multibatch [TRSS+16]	2.6M	98.8	
Aug [MTH ⁺ 16]	0.5M	98.06	
Softmax Loss	4M	98.87	94.16
Marginal Loss	4M	99.48	95.98

Cross-Age Celebrity Dataset (CACD) and AgeDB

Methods	Acc (%)
High-Dimensional LBP [CCWS13]	81.6
Hidden Factor Analysis [GLL+13]	84.4
Cross-Age Reference Coding [CCH14]	87.6
LF-CNNs [WLQ16]	98.5
Human Average	85.7
Human Voting	94.2
Centre Loss [WZLQ16]	97.475
Marginal Loss	98.95



VGG Face [PVZ15] 93.15 92.18 89.15 85.08

5 Yr | 10 Yr | 20 Yr | 30 Yr

95.93 | 95.15 | 93.07 | 90.72

10 Yr 20 Yr

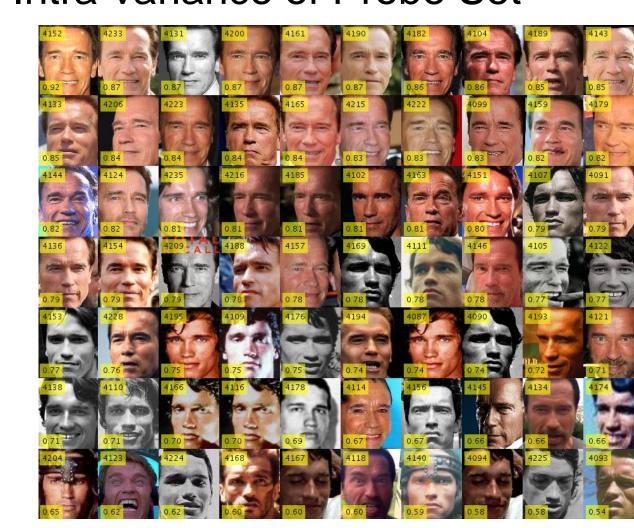
IEEE 2017 Conference on Computer Vision and Pattern Recognition



Mega Face

(a) FaceScrub

Intra-variance of Probe Set

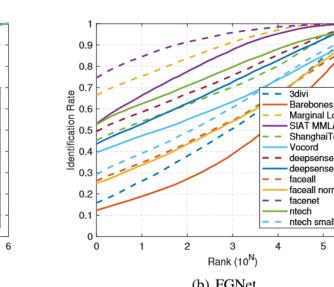


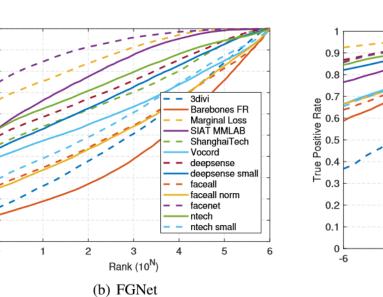


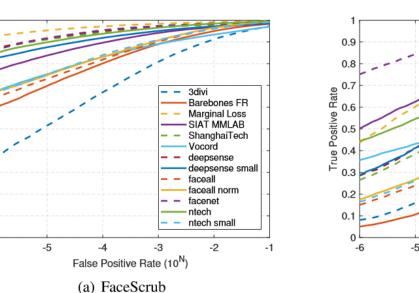
Challenging Age Variations

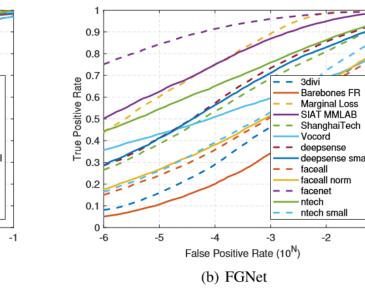
FaceScrub FG-Net

• CMC&ROC curves of different methods with 1M distractors on Set1.









Identification rates and verification TAR at 10e-6 FAR

Methods	protocol	Id (%)	Ver(%)
YouTu Lab	Large	83.290	91.340
DeepSense V2	Large	81.298	95.993
Vocord deepVo1.2	Large	80.258	77.143
Google FaceNet v8	Large	70.496	86.473
GRCCV	Small	77.677	74.887
SphereFace	Small	75.766	90.045
DeepSense	Small	70.983	82.851
Centre Loss [WZLQ16]	Small	65.234	76.516
Marginal Loss	Large	80.278	92.640

	Methods	protocoi	10 (%)	ver(%)
	Google FaceNet v8	Large	74.594	75.550
:	SIATMMLAB	Large	71.247	67.954
	DeepSense V2	Large	63.632	56.767
	SIAT MMLAB	Small	55.304	50.144
	SphereFace	Small	47.555	40.094
	DeepSense	Small	43.540	29.610
	Marginal Loss	Large	66.432	43.703

FaceScrub

FG-Net

Reference

- > Wen Y, Zhang K, Li Z, et al. A discriminative feature learning approach for deep face recognition. ECCV, 2016
- > Tadmor O, Wexler Y, Rosenwein T, et al. Learning a metric embedding for face recognition using the multi-batch method. NIPS, 2016.
- > Zhang X, Fang Z, Wen Y, et al. Range Loss for Deep Face Recognition with Long-tail. arXiv:1611.08976, 2016.