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HyperSpaceX: Radial and Angular Exploration of HyperSpherical Dimensions

Chiranjeev Chiranjeev, Muskan Dosi, Kartik Thakral, Mayank Vatsa, Richa Singh

Project page: <u>https://github.com/IAB-IITJ/HyperSpaceX</u>

Paper: https://arxiv.org/abs/2408.02494





HyperSpaceX Framework: Overview



□ Introducing HyperSpaceX, a novel discriminative feature representation and arrangement learning framework that explores both radial and angular dimensions in multi-hyperspherical space.

□ HyperSpaceX shows its efficacy over multiple small and large-scale image and face recognition datasets.

Framework Comparison



Metric-based Learning

Angular Learning

Radial-Angular Learning

Preliminaries - Existing Loss Functions

$$L_{\rm Con} = y \times D^2 + (1-y) \times max(\eta - D, 0)^2$$

Contrastive Loss

$$L_{\rm CE} = -\frac{1}{N} \sum_{i=1}^{N} \log \frac{e^{\omega_{y_i}^T x_i + b_{y_i}}}{\sum_{j=1}^{K} e^{\omega_j^T x_j + b_j}}$$

Cross-Entropy Loss

$$L_{\text{ArcFace}} = -\frac{1}{N} \sum_{i=1}^{N} \log \frac{e^{\cos(\theta_{y_i}+m)}}{e^{\cos(\theta_{y_i}+m)} + \sum_{j=1, j \neq y_i}^{K} e^{\cos\theta_j}}$$

ArcFace Loss

Proposed DistArc Loss

$$L_{\text{DistArc}} = -\frac{1}{N} \sum_{i=1}^{N} \log \frac{e^{\cos(\theta_{y_i} + m) + \cos(\phi_{y_i}) - \lambda \delta_{y_i}}}{e^{\cos(\theta_{y_i} + m)} + \sum_{j=1, j \neq y_i}^{K} e^{\cos(\theta_j) - \lambda \delta_j}}$$

Predictive Measure

Resultant computation in terms of θ and ϕ :

$$||R_i||_2 = ||x||_2 \cos \phi_i + ||\omega_{r_i}||_2 \cos(\pi - (\theta_i + \phi_i)) \quad \forall \in 1, 2, 3, \dots$$

A favourable class determining predictive measure for the HyperSpaceX framework with its radial-angular based formulation:

$$\hat{y} = argmin_{R_m} \{ R_m \in \mathbb{R}^K : R_m \}$$

DistArc Loss Components



Latent Space Visualization



Tasks Improved with HyperSpaceX

Image Classification



CIFAR-100

TinyImageNet

CUB-200

ImageNet-1K

Small

Scale

MNIST

Fashion-MNIST

CIFAR-10

Face Recognition



CASIA-WebFace

MS1Mv2

D-LORD

LFW

CFP-FP

AgeB-30

CA-LFW

CP-LFW

D-LORD

Loss **Functions**

Cross-Entropy SpherFace CosFace ArcFace SphereFace2 Center Triplet DistArc

Backbones

iResNet50 RN101 ViT-B ViT-L ResNet18

Performance on Image Classification Tasks



Visualization of Feature arrangement learning in Latent space on CIFAR-100 Dataset



Experimental and Visual Ablation



Visual geometric ablation showing significance of (a) $\cos(\theta)$ and δ , and (b) after incorporation of $\cos(\phi)$.

Loss Components	MNIST	Fashion MNIST	CIFAR-10
cos(θ)	98.97	93.69	94.81
cos(θ) & cos(φ)	98.93	93.72	95.16
cos(θ) & δ	99.01	94.83	95.31
cos(θ) & cos(φ) & δ	99.19	95.04	96.03



Ablation based on varying radial gap between consecutive hyper-spheres.



hyper-spheres.

Performance on Face Recognition Tasks





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Thank You!!

For more details visit our poster Thu 3 Oct 2024, Poster #26, Poster Session 6

Chiranjeev Chiranjeev, Muskan Dosi, Kartik Thakral, Mayank Vatsa, Richa Singh

{chiranjeev.1, dosi.1, thakral.1, mvatsa, richa}@iitj.ac.in

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