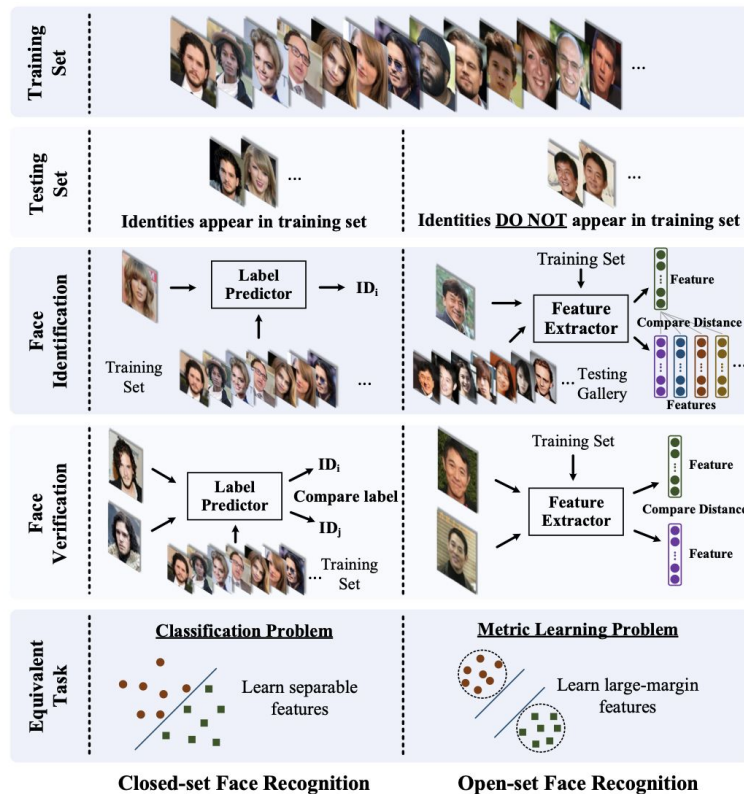
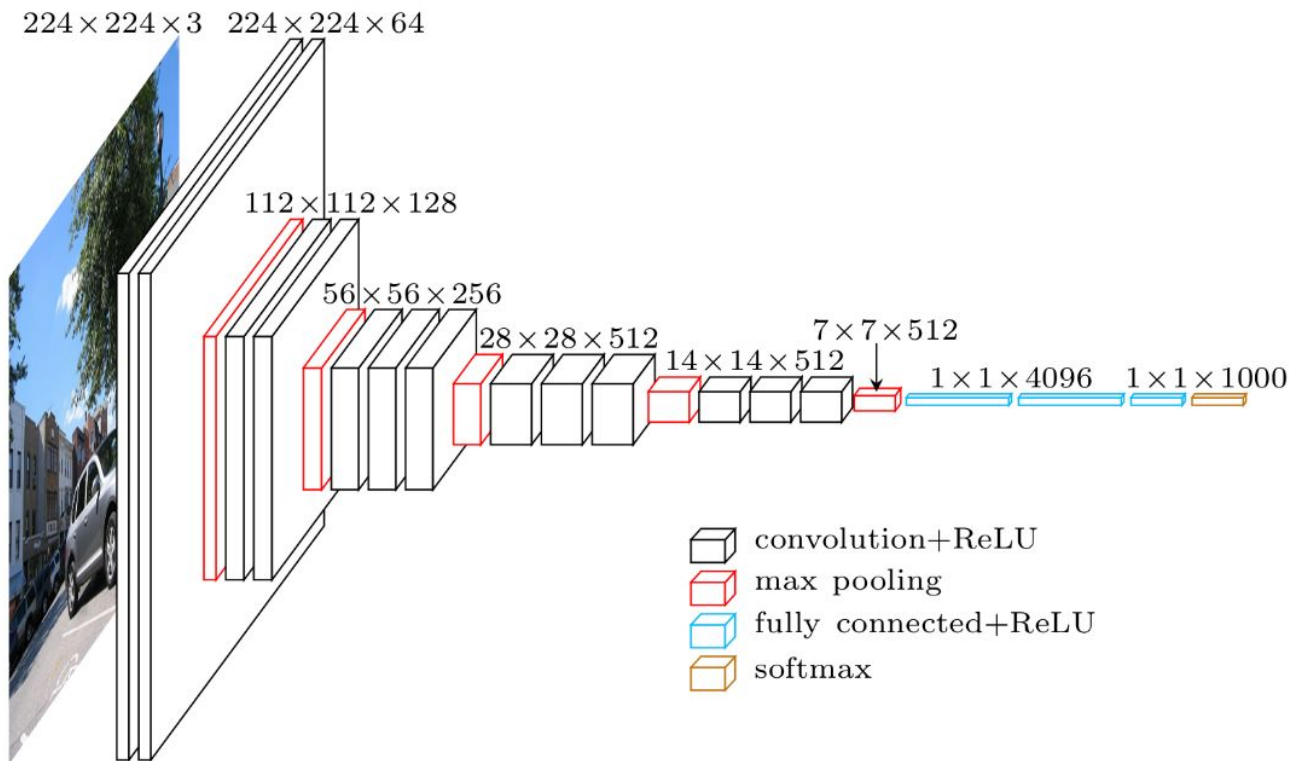


# Deep Learning Metric Learning

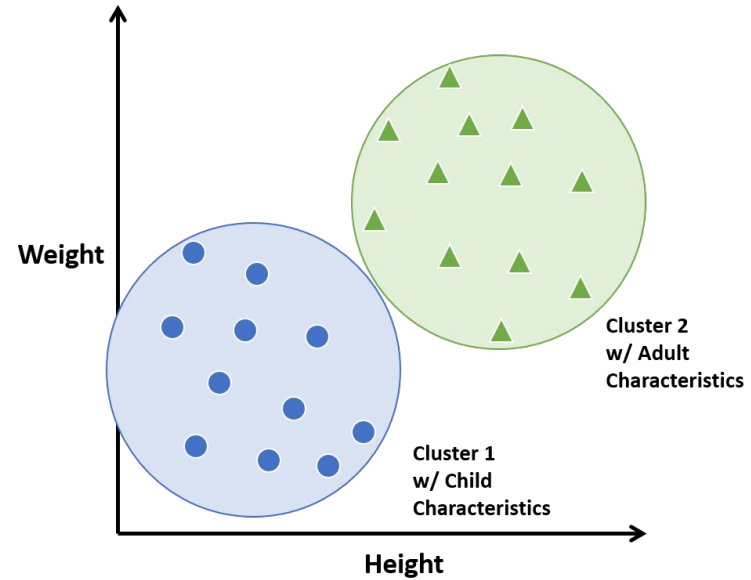
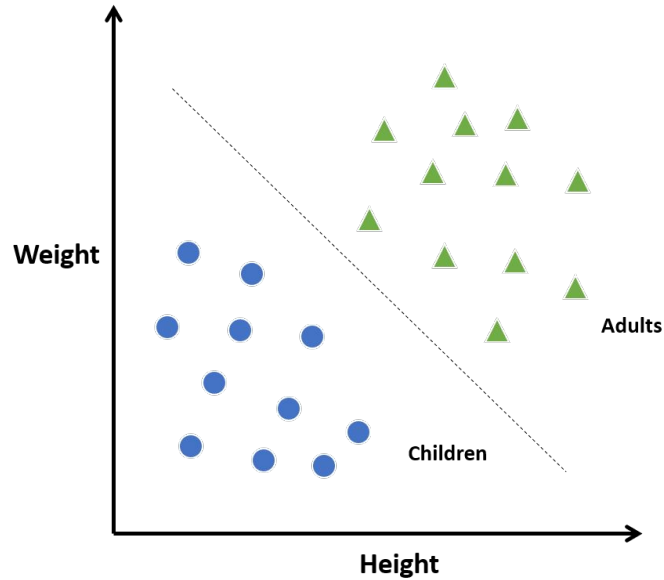
# Person Identification



# Feature Extraction



# Classification vs Clustering



# Classification vs Clustering



# Class is important



# Datasets

Many classes, few objects pre class

- Omniglot
- Person ReID
- LFW



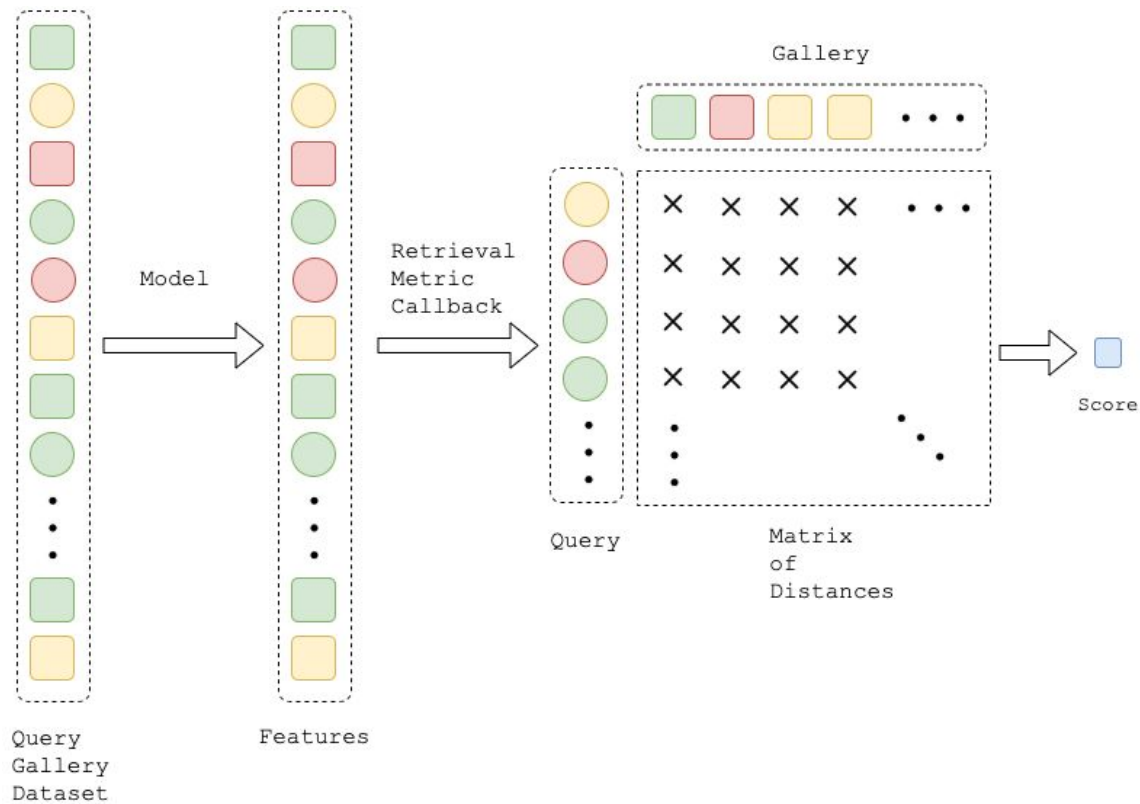


# Datasets. ReID





# Metric. Query Gallery









# Metrics. mAP

$$\text{Recall}_k = \frac{\# \text{ predicted labels that are relevant}}{\# \text{ all labels}}$$

$$\text{Precision}_k = \frac{\# \text{ predicted labels that are relevant}}{\# \text{ all predicted labels}}$$







$$\text{AP}_K = \sum_{k=1}^K (\text{Recall}_k - \text{Recall}_{k-1}) \text{Precision}_k$$

				
	1	2	3	4
	3	4	1	2

# Metrics. Cumulative Matching Characteristics

$$Acc_k = \begin{cases} 1 & \text{if top-}k \text{ ranked gallery samples contain the query identity} \\ 0 & \text{otherwise} \end{cases},$$

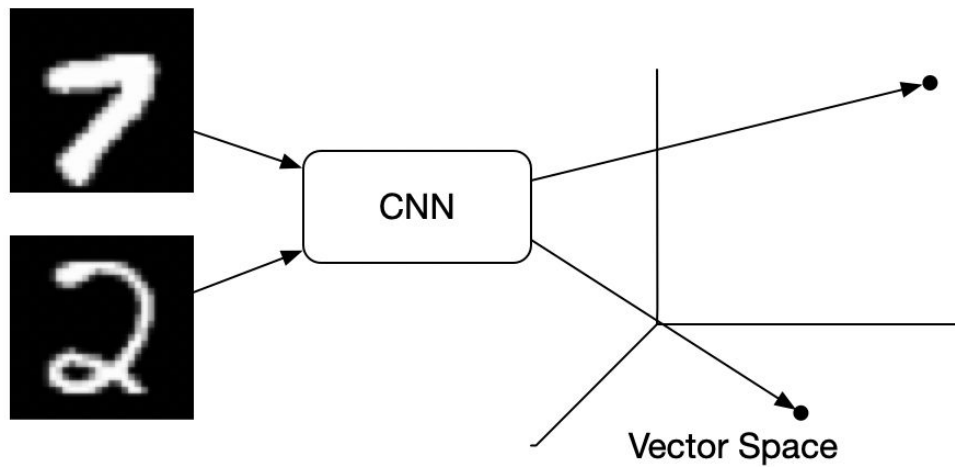
$$CMC = \frac{1}{C} \sum_C Acc_k$$

				
	1	2	3	4
	3	4	1	2

# Metric Learning

- Contrastive Loss
- Triplet Loss
- Cross Entropy
- Siamese Network

# Contrastive Loss



# Contrastive Loss

$$y_{\text{pred}} = \text{dist}(x_i, x_j)$$
$$y_{\text{target}} = \begin{cases} 1, & \text{if } c_i == c_j, \\ 0, & \text{otherwise} \end{cases}$$

# Contrastive Loss

$$y_{\text{pred}} = \text{dist}(x_i, x_j)$$

$$y_{\text{target}} = \begin{cases} 1, & \text{if } c_i == c_j, \\ 0, & \text{otherwise} \end{cases}$$

$$L = y_{\text{target}}y_{\text{pred}} - (1 - y_{\text{target}})y_{\text{pred}}$$



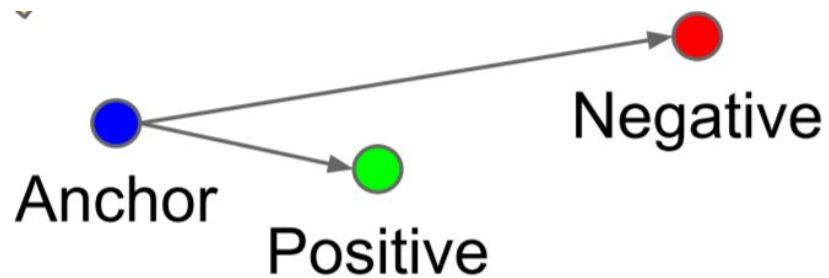
# Contrastive Loss

$$y_{\text{pred}} = \text{dist}(x_i, x_j)$$

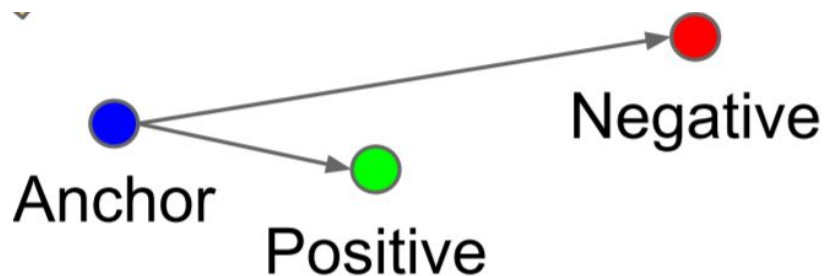
$$y_{\text{target}} = \begin{cases} 1, & \text{if } c_i == c_j, \\ 0, & \text{otherwise} \end{cases}$$

$$L = y_{\text{target}} y_{\text{pred}} + (1 - y_{\text{target}}) \max(0, m - y_{\text{pred}})$$

# Triplet Loss

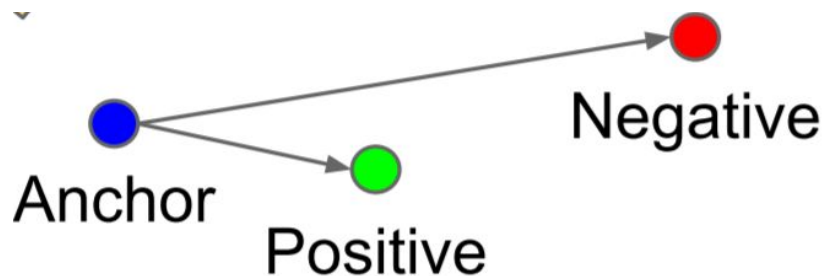


# Triplet Loss



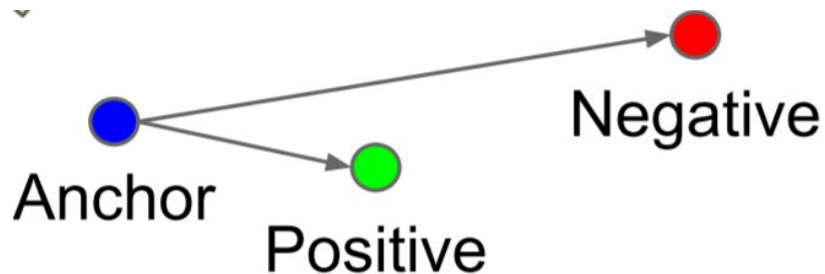
$$L = \text{dist}(x_a, x_p) - \text{dist}(x_a, x_n)$$

# Triplet Loss



$$L = \text{dist}(x_a, x_p) + \max(0, m - \text{dist}(x_a, x_n))$$

# Triplet Loss + Classification



$$L = L_{\text{triplet}} + L_{\text{CE}}$$

# Sampling

- All Triplet
- Hard Triplet
- Hard Cluster Triplet
- Hierarchical Triplet

# All Sampling



# All Sampling

$$\mathcal{L}_{\text{BA}}(\theta; X) = \overbrace{\sum_{i=1}^P \sum_{a=1}^K}^{\text{all anchors}} \overbrace{\sum_{\substack{p=1 \\ p \neq a}}^K}^{\text{all pos.}} \overbrace{\sum_{\substack{j=1 \\ j \neq i}}^P \sum_{n=1}^K}^{\text{all negatives}} \left[ m + d_{j,a,n}^{i,a,p} \right]_+, \quad (6)$$

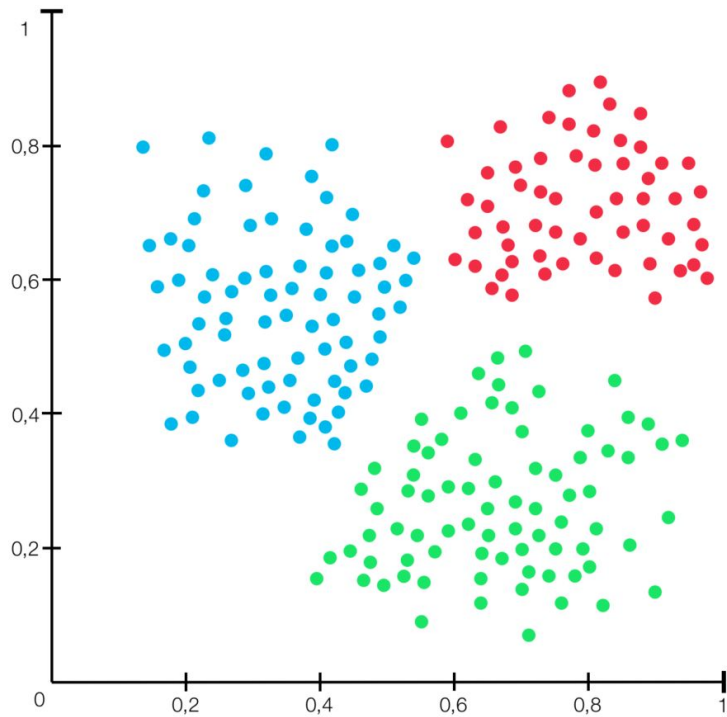
$$d_{j,a,n}^{i,a,p} = D \left( f_{\theta}(x_a^i), f_{\theta}(x_p^i) \right) - D \left( f_{\theta}(x_a^i), f_{\theta}(x_n^j) \right).$$

# Hard Sampling

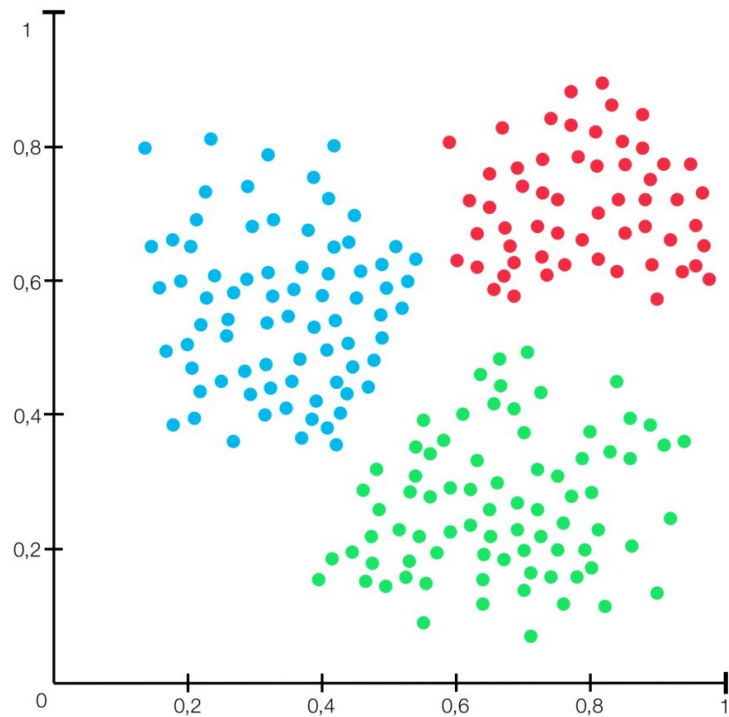
# Hard Sampling

$$\mathcal{L}_{\text{BH}}(\theta; X) = \sum_{i=1}^P \sum_{a=1}^K \left[ m + \overbrace{\max_{p=1 \dots K} D(f_{\theta}(x_a^i), f_{\theta}(x_p^i))}^{\text{hardest positive}} \right. \quad (5) \\
 \left. - \underbrace{\min_{\substack{j=1 \dots P \\ n=1 \dots K \\ j \neq i}} D(f_{\theta}(x_a^i), f_{\theta}(x_n^j))}_{\text{hardest negative}} \right]_+,$$

# Hard Cluster Sampling



# Hard Cluster Sampling

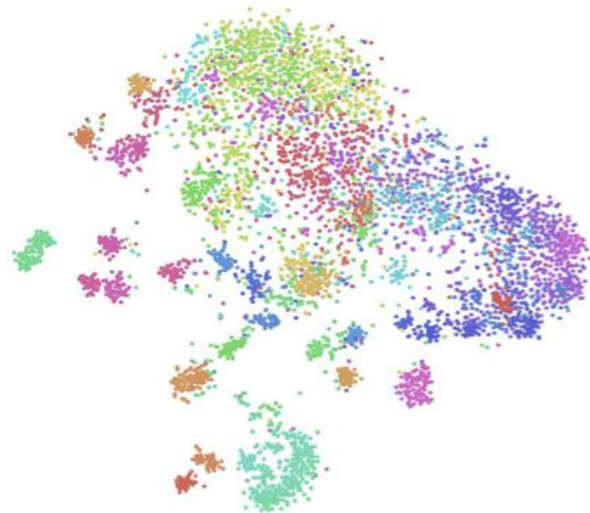
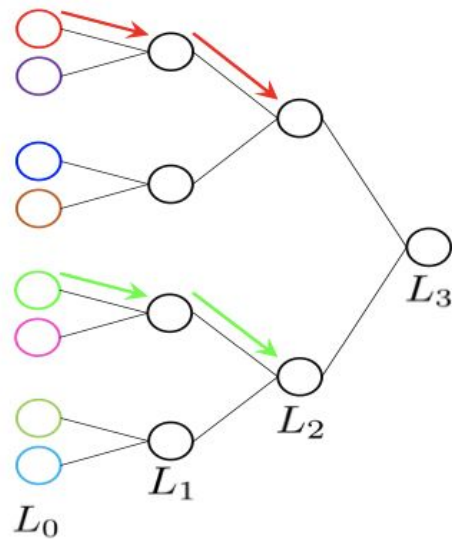


$$d_i^{inter} = \min_{\forall i_d \in P, i_d \neq i} \|f_i^m - f_{i_d}^m\|_2^2$$

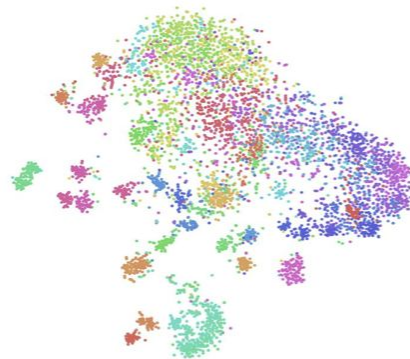
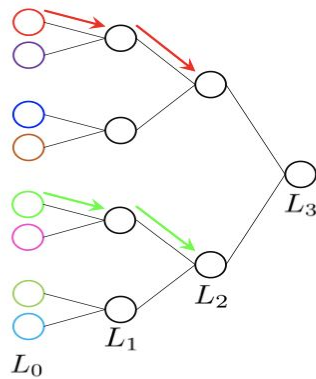
$$d_i^{intra} = \max_K \|f(x) - f_i^m\|_2^2$$

$$Lb_c = \sum_i^P \max((d_i^{intra} - d_i^{inter} + \alpha), 0)$$

# Hierarchical Sampling

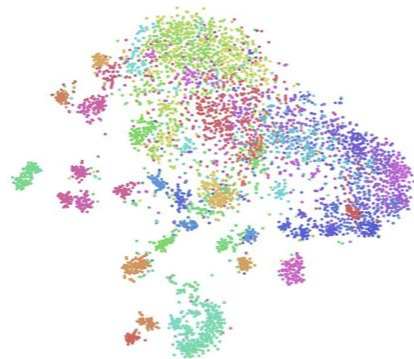
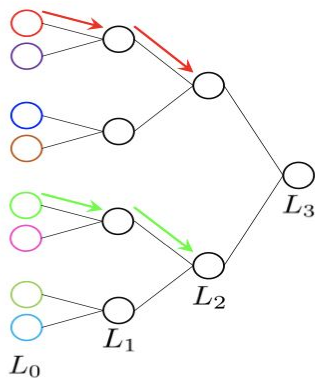


# Hierarchical Sampling





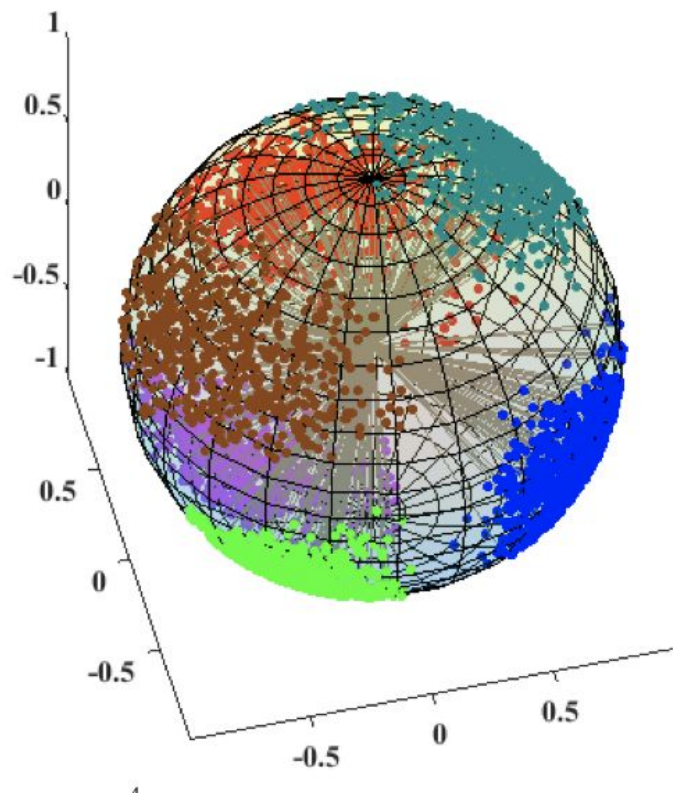
# Hierarchical Sampling



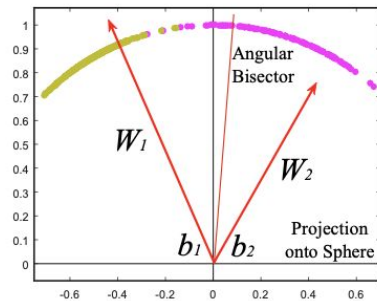
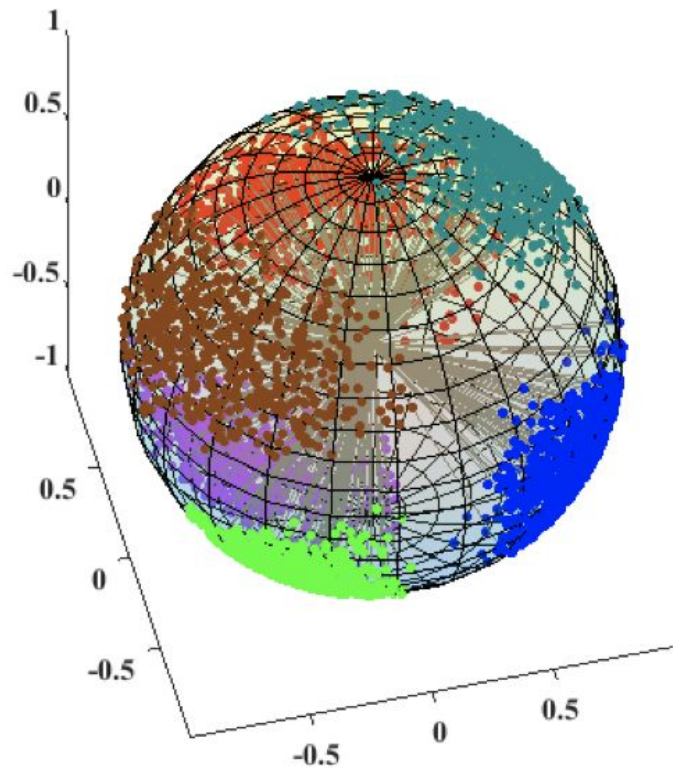
$$\mathcal{L}_{\mathcal{M}} = \frac{1}{2Z_{\mathcal{M}}} \sum_{\mathcal{T}^z \in \mathcal{T}^{\mathcal{M}}} [\|\mathbf{x}_a^z - \mathbf{x}_p^z\| - \|\mathbf{x}_a^z - \mathbf{x}_n^z\| + \alpha_z]_+.$$

$$\alpha_z = \beta + d_{\mathcal{H}}(y_a, y_n) - s_{y_a}$$

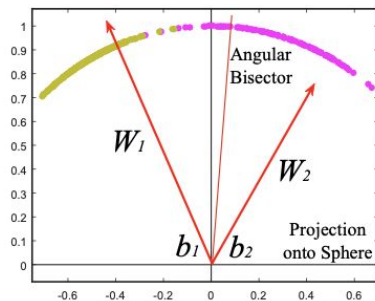
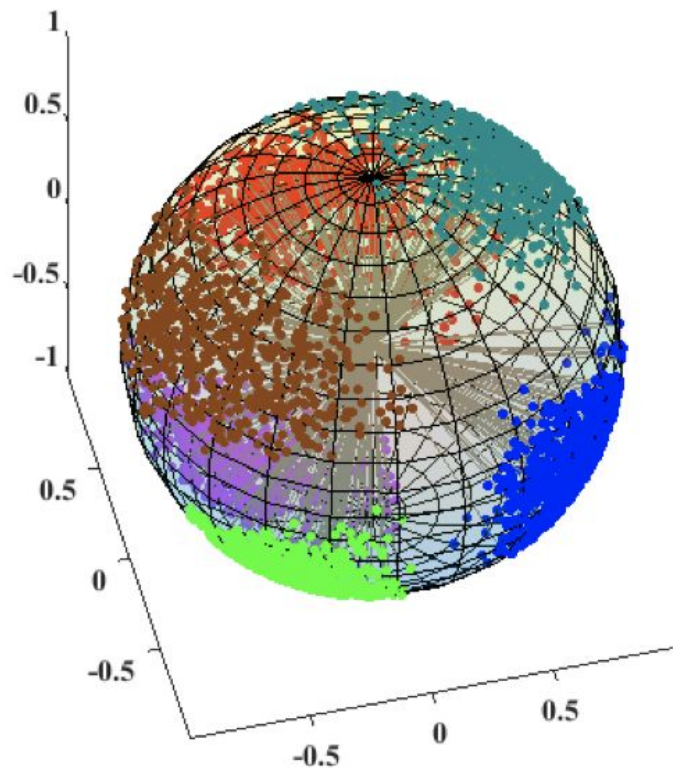
# SphereFace



# SphereFace



# SphereFace

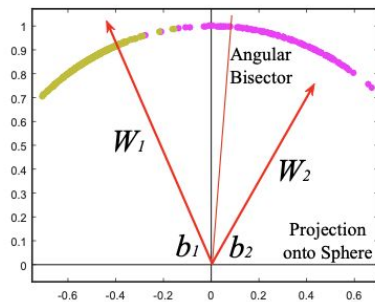
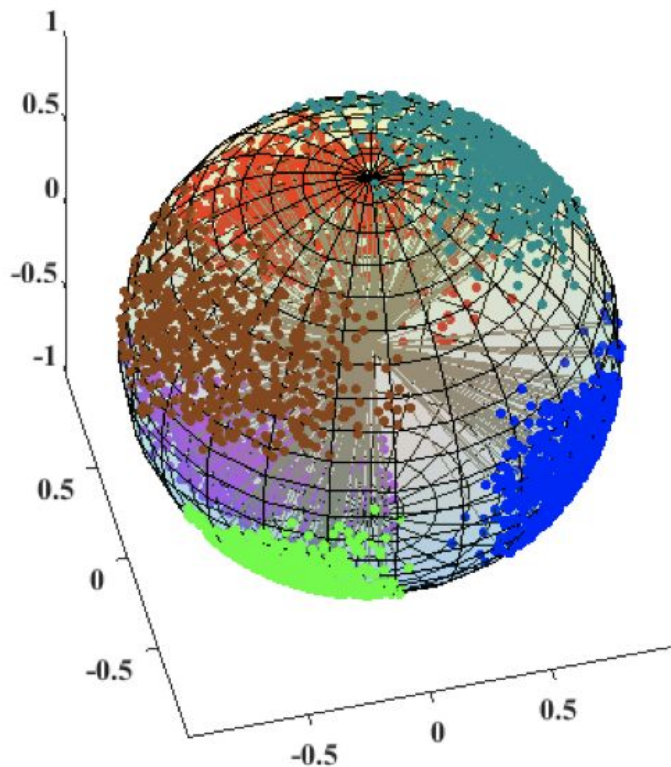


$$f_1 = W_1 x$$

$$f_2 = W_2 x$$

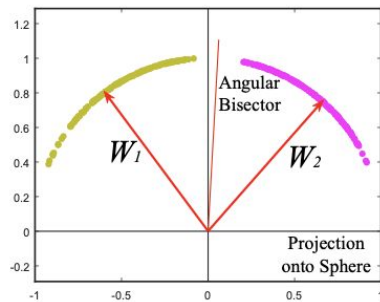
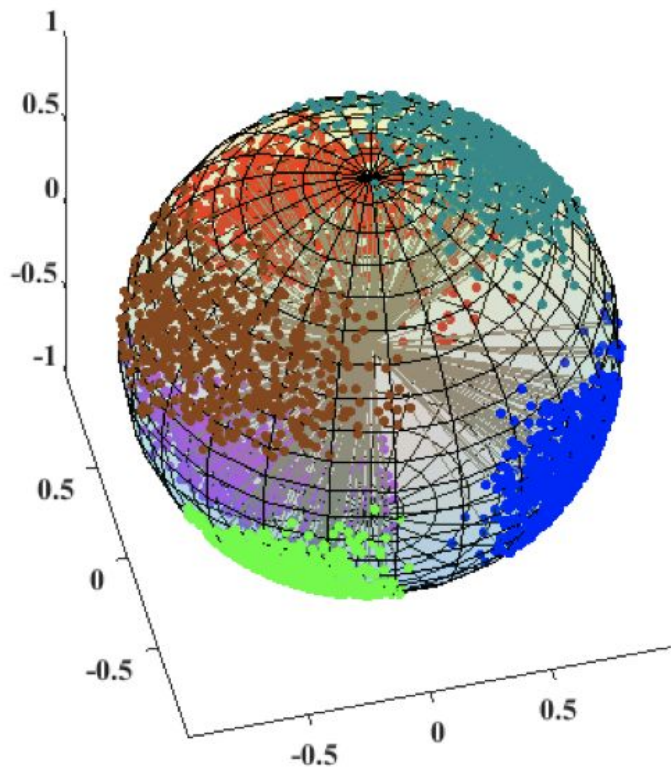
$$\text{dist}(f_1, f_2) = \cos(f_1, f_2)$$

# SphereFace



$$f_1 = W_1 x = \cos(\theta_1) ||W_1|| ||x||$$
$$f_2 = W_2 x = \cos(\theta_2) ||W_2|| ||x||$$
$$\text{dist}(f_1, f_2) = \cos(f_1, f_2)$$

# SphereFace



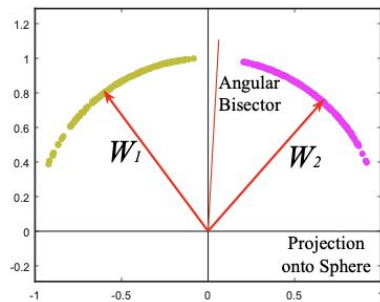
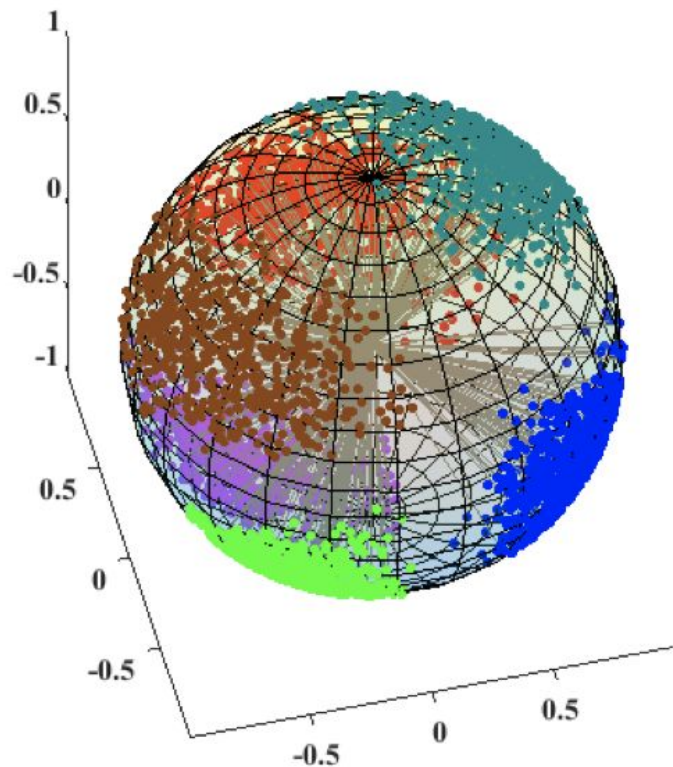
$$f_1 = W_1 x = \cos(\theta_1) ||W_1|| ||x||$$

$$f_2 = W_2 x = \cos(\theta_2) ||W_2|| ||x||$$

$$\text{dist}(f_1, f_2) = ||x|| (\cos(\theta_1) - \cos(\theta_2))$$

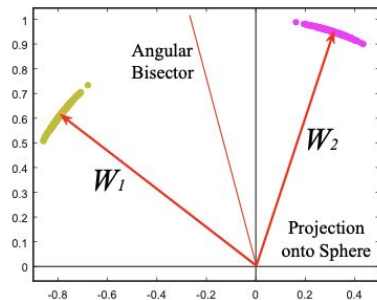
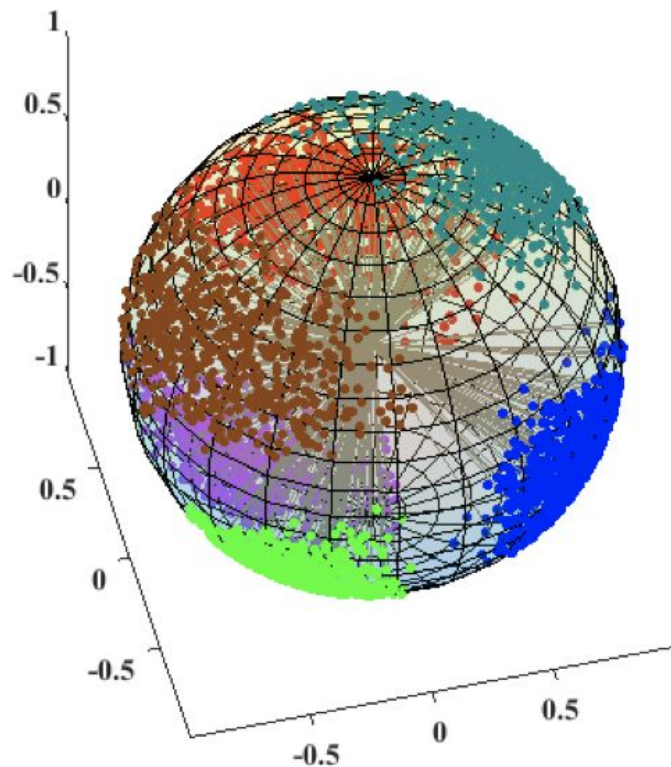


# SphereFace



$$f_1 = W_1 x = \cos(\theta_1) ||W_1|| ||x||$$
$$f_2 = W_2 x = \cos(\theta_2) ||W_2|| ||x||$$

# SphereFace

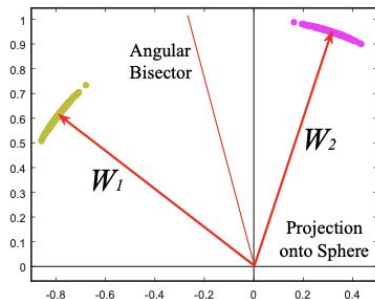
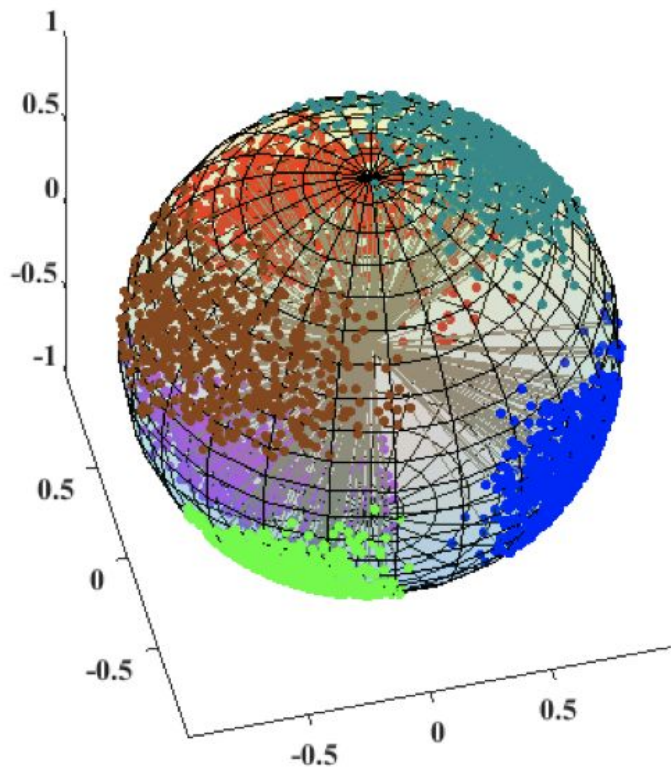


$$f_1 = W_1 x = \cos(\theta_1) \|W_1\| \|x\|$$

$$f_2 = W_2 x = \cos(\theta_2) \|W_2\| \|x\|$$



# SphereFace



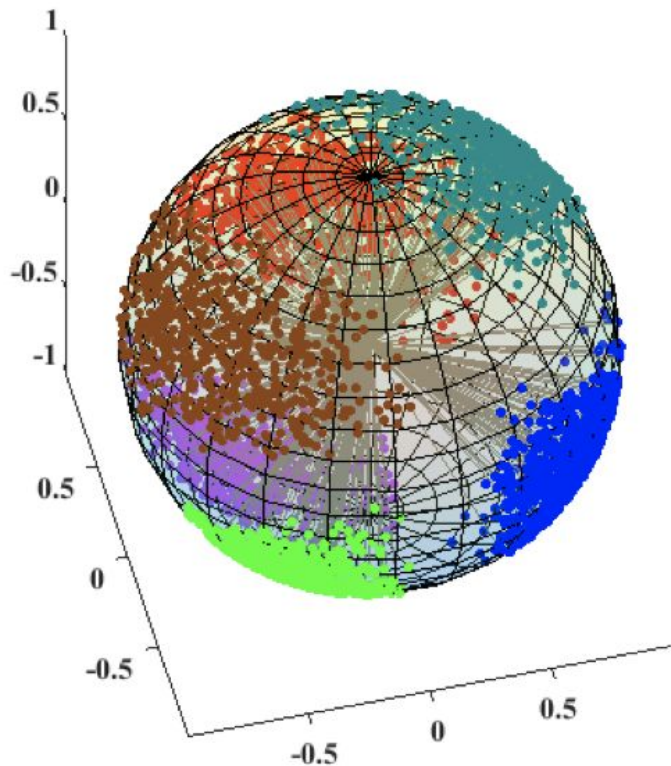
$$f_1 = W_1 x = \cos(\theta_1) \|W_1\| \|x\|$$

$$f_2 = W_2 x = \cos(\theta_2) \|W_2\| \|x\|$$

$$\text{dist}(f_1, f_2) = \|x\| (m \cos(\theta_1) - \cos(\theta_2))$$

$$\text{dist}(f_2, f_1) = \|x\| (\cos(\theta_1) - m \cos(\theta_2))$$

# SphereFace



Method	Models	Data	LFW	YTF
Softmax Loss	1	WebFace	97.88	93.1
Softmax+Contrastive [26]	1	WebFace	98.78	93.5
Triplet Loss [22]	1	WebFace	98.70	93.4
L-Softmax Loss [16]	1	WebFace	99.10	94.0
Softmax+Center Loss [34]	1	WebFace	99.05	94.4
SphereFace	1	WebFace	<b>99.42</b>	<b>95.0</b>

CosFace

# CosFace

$$L_{lmc} = \frac{1}{N} \sum_i -\log \frac{e^{s(\cos(\theta_{y_i,i})-m)}}{e^{s(\cos(\theta_{y_i,i})-m)} + \sum_{j \neq y_i} e^{s \cos(\theta_{j,i})}},$$

# CosFace

Method	LFW	YTF
Softmax Loss [23]	97.88	93.1
Softmax+Contrastive [30]	98.78	93.5
Triplet Loss [29]	98.70	93.4
L-Softmax Loss [24]	99.10	94.0
Softmax+Center Loss [42]	99.05	94.4
A-Softmax [23]	<b>99.42</b>	95.0
A-Softmax-NormFea	99.32	95.4
<b>LMCL</b>	99.33	<b>96.1</b>

# ArcFace

# ArcFace

$$L_4 = -\frac{1}{N} \sum_{i=1}^N \log \frac{e^{s(\cos(m_1 \theta_{y_i} + m_2) - m_3)}}{e^{s(\cos(m_1 \theta_{y_i} + m_2) - m_3)} + \sum_{j=1, j \neq y_i}^n e^{s \cos \theta_j}}.$$

# ArcFace

Loss Functions	LFW	CFP-FP
ArcFace (0.4)	99.53	95.41
ArcFace (0.45)	99.46	95.47
ArcFace (0.5)	<b>99.53</b>	<b>95.56</b>
ArcFace (0.55)	99.41	95.32
SphereFace [18]	99.42	-
SphereFace (1.35)	99.11	94.38
CosFace [37]	99.33	-
CosFace (0.35)	99.51	95.44



# Siamese Network

