

Face Recognition

One-shot learning: learn from 1 example

- learn a similarity function (verify if same person)

Siamese function

run 2 identical CNNs on 2 example, then use the similarity function to compare the distance of the 2 encodings

$$d(x^{(1)}, x^{(2)}) = \|f(x^{(1)}) - f(x^{(2)})\|_2^2$$

{ small if same person
large if diff person

Triplet loss

Anchor

Positive

Negative

$$\|f(A) - f(P)\|_2^2 \leq \|f(A) - f(N)\|_2^2$$

$$d(A, P)$$

$$d(A, N)$$

$$d(A, P) - d(A, N) + \alpha \leq 0$$

formalize

$$L(A, P, N) =$$

$$\max(\|f(A) - f(P)\|_2^2 - \|f(A) - f(N)\|_2^2 + \alpha, 0)$$

$$J = \sum_{i=1}^M L(A^{(i)}, P^{(i)}, N^{(i)})$$

hyperparam: margin

to prevent d just to output

0 every time

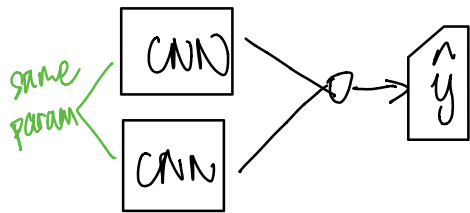
How to choose A, P, N:

choose triplets that are hard to train

improve the efficiency of the training algo.

$$d(A, P) + \alpha \leq d(A, N) \Rightarrow \underbrace{d(A, P)}_{\downarrow} \approx \underbrace{d(A, N)}_{\uparrow}$$

As Binary Classification



$$\hat{y} = \sigma \left(\sum_{k=1}^{K^S} w_k |Ax^{(i)}_k - f(x^{(i)}_k) + b \right)$$

Neural Style Transfer

cost function

$$J(G) = \alpha J_{\text{content}}(C, G) + \beta J_{\text{style}}(S, G)$$

content cost

$$J_{\text{content}}(C, G) = \frac{1}{2} \|a^{[L]}(C) - a^{[L]}(G)\|^2$$

style cost
↓

$$\underline{G_{KK'}}^{[L]} = \sum_{i=1}^{N_H^{[L]}} \sum_{j=1}^{N_W^{[L]}} \underbrace{a_{ijk}^{[L]}}_{\text{channel}} \underbrace{a_{ijk}^{[L]}}_{\text{channel}}$$

"Gram Matrix"

How correlated are
activations across different
channel?

$$J_{\text{style}}^{[L]}(S, G) = \|G^{[L]}(S) - G^{[L]}(G)\|_F^2$$

$$\rightarrow \frac{1}{(2nwc)^2} \sum_k \sum_{k'} (G_{kk'}^{(S)} - G_{kk'}^{(G)})^2$$

Conv 3D : (32, 32, 32, 6) → 32 Filters (3, 3, 3)

output ⇒ (30, 30, 30, 32)