



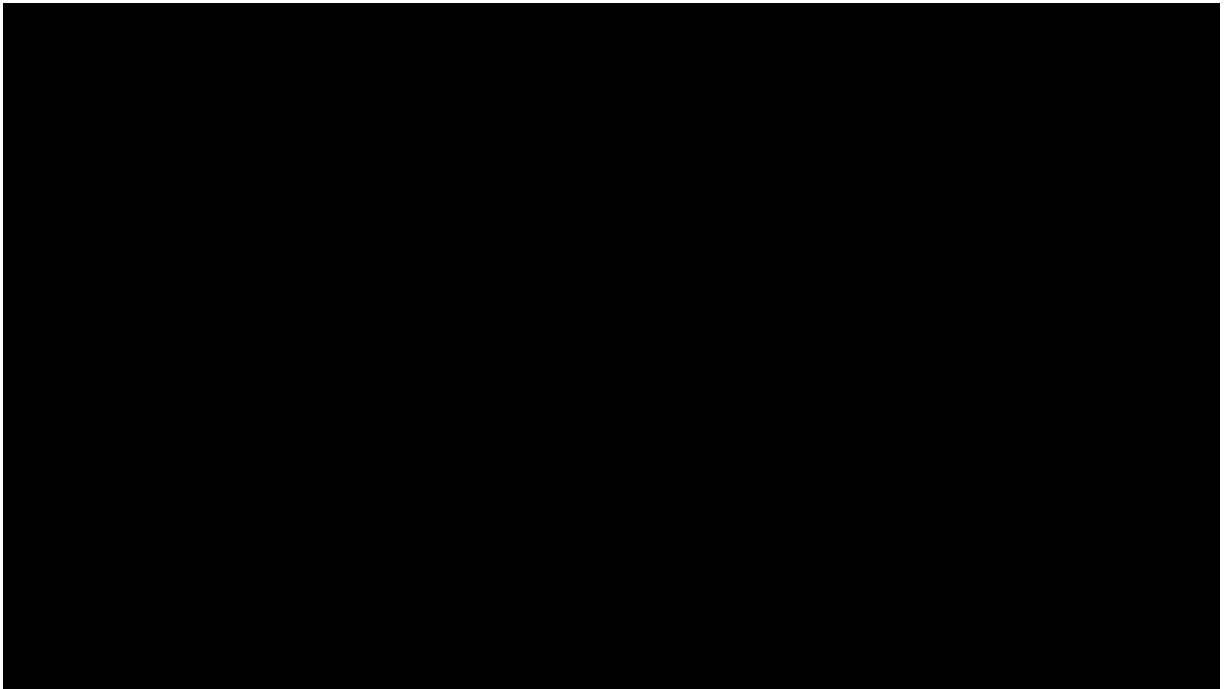
deeplearning.ai

# Face recognition

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What is face  
recognition?

# Face recognition



[Courtesy of Baidu]

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# Face verification vs. face recognition

## → Verification

- Input image, name/ID
- Output whether the input image is that of the claimed person

1:1

99.0%

99.9

## → Recognition

- Has a database of K persons
- Get an input image
- Output ID if the image is any of the K persons (or “not recognized”)

1:K

K=100 ←



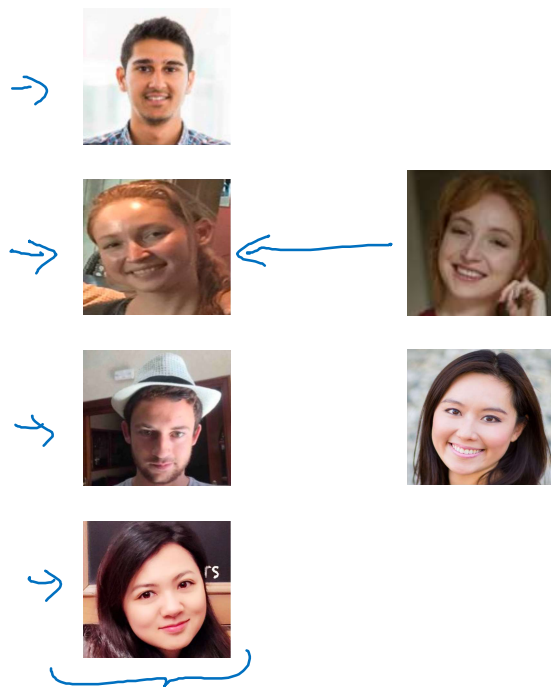
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# Face recognition

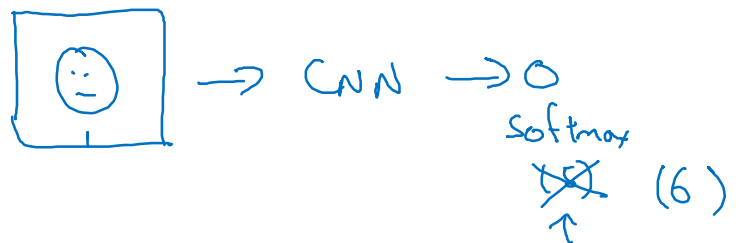
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## One-shot learning

# One-shot learning



Learning from one example to recognize the person again



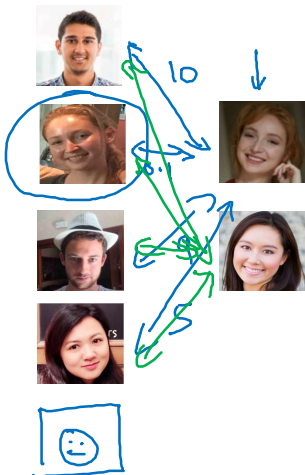
# Learning a “similarity” function

→  $d(\text{img1}, \text{img2})$  = degree of difference between images

If  $d(\text{img1}, \text{img2}) \leq \tau$   
 $> \tau$

“same”  
“different”

} Verification.



$d(\text{img1}, \text{img2})$



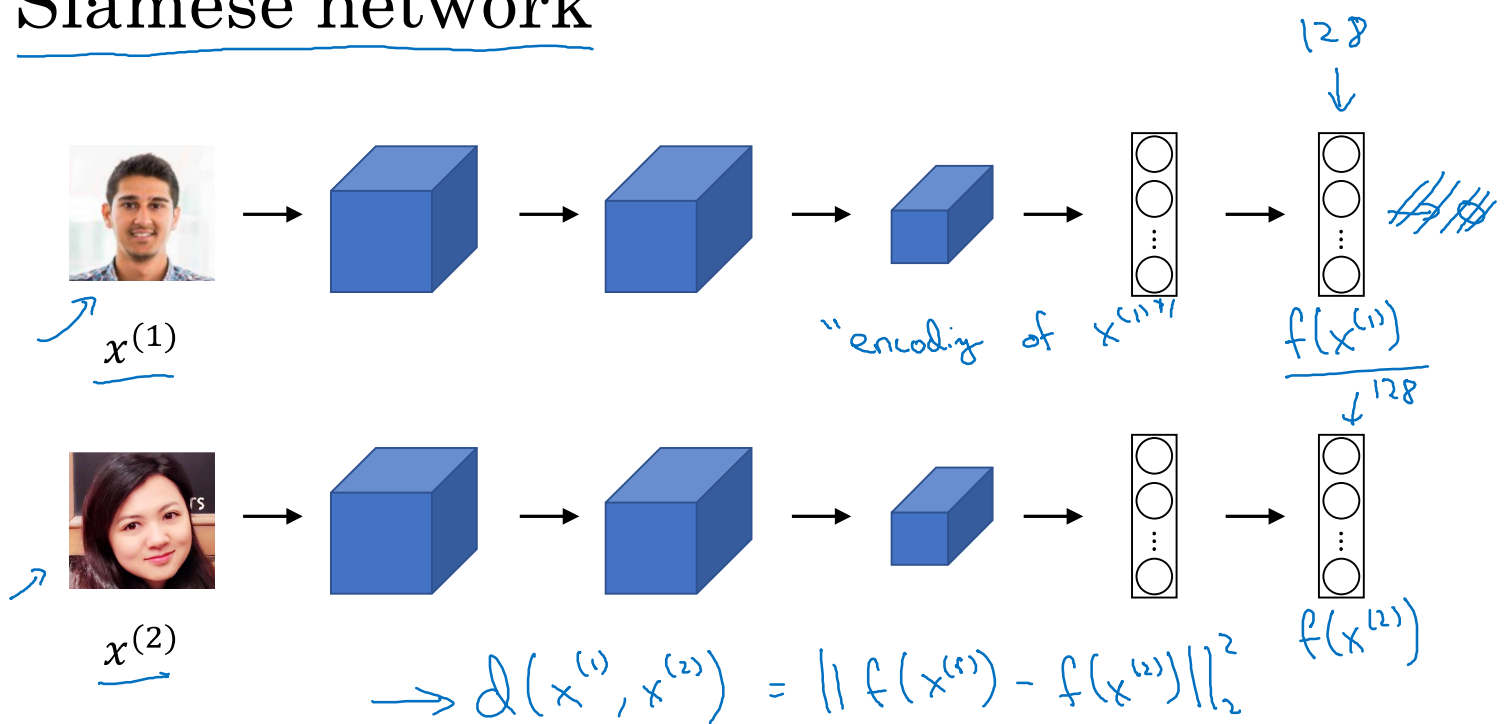
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# Face recognition

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## Siamese network

# Siamese network

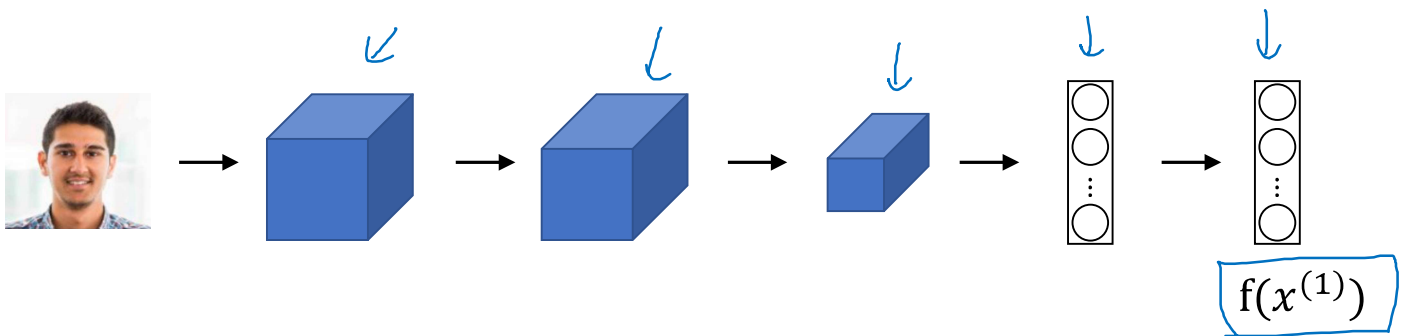


[Taigman et. al., 2014. DeepFace closing the gap to human level performance]

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# Goal of learning



Parameters of NN define an encoding  $f(x^{(i)})$

128

Learn parameters so that:

If  $x^{(i)}, x^{(j)}$  are the same person,  $\|f(x^{(i)}) - f(x^{(j)})\|^2$  is small.

If  $x^{(i)}, x^{(j)}$  are different persons,  $\|f(x^{(i)}) - f(x^{(j)})\|^2$  is large.



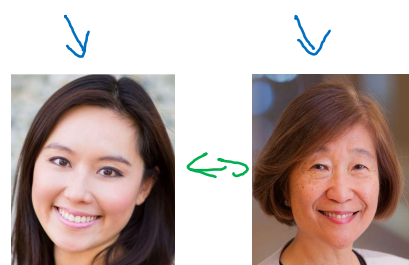
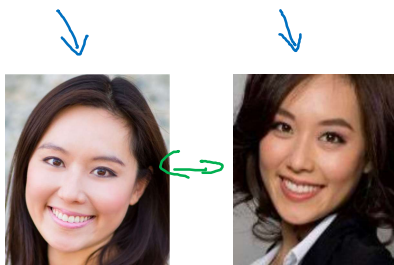
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# Face recognition

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## Triplet loss

# Learning Objective



Anchor

Positive

Anchor

Negative

A

$d(A, P) = 0.5$

Want:  $\underbrace{\|f(A) - f(P)\|^2}_{d(A, P)} + \underline{\alpha} \leq \underline{0.2}$

A

$d(A, N) = 0.7$

$\underbrace{\|f(A) - f(N)\|^2}_{d(A, N)}$

$$\underbrace{\|f(A) - f(P)\|^2}_0 - \underbrace{\|f(A) - f(N)\|^2}_0 + \underline{\alpha} \leq \underline{0} \quad \text{margin}$$

$f(\text{img}) = \vec{0}$

[Schroff et al., 2015, FaceNet: A unified embedding for face recognition and clustering]

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# Loss function

Given 3 images  $A, P, N$ :

$$\underline{\mathcal{L}(A, P, N)} = \max(\underbrace{\|f(A) - f(P)\|^2 - \|f(A) - f(N)\|^2}_{\geq 0} + \alpha, 0)$$

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

$A, P$   
 $\uparrow \quad \uparrow$

Training set: 10k pictures of 1k persons

## Choosing the triplets A,P,N

During training, if A,P,N are chosen randomly,  
 $d(A,P) + \alpha \leq d(A,N)$  is easily satisfied.

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

Choose triplets that're "hard" to train on.

$$\frac{d(A,P) + \alpha}{d(A,P)} \approx \frac{d(A,N)}{d(A,N)}$$

↓                      ↑

Face Net  
Deep Face

↓

[Schroff et al., 2015, FaceNet: A unified embedding for face recognition and clustering]

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# Training set using triplet loss

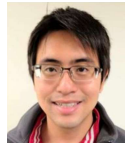
Anchor



⋮



Positive



⋮



Negative



⋮



$$J$$
$$d(x^{(i)}, x^{(j)})$$



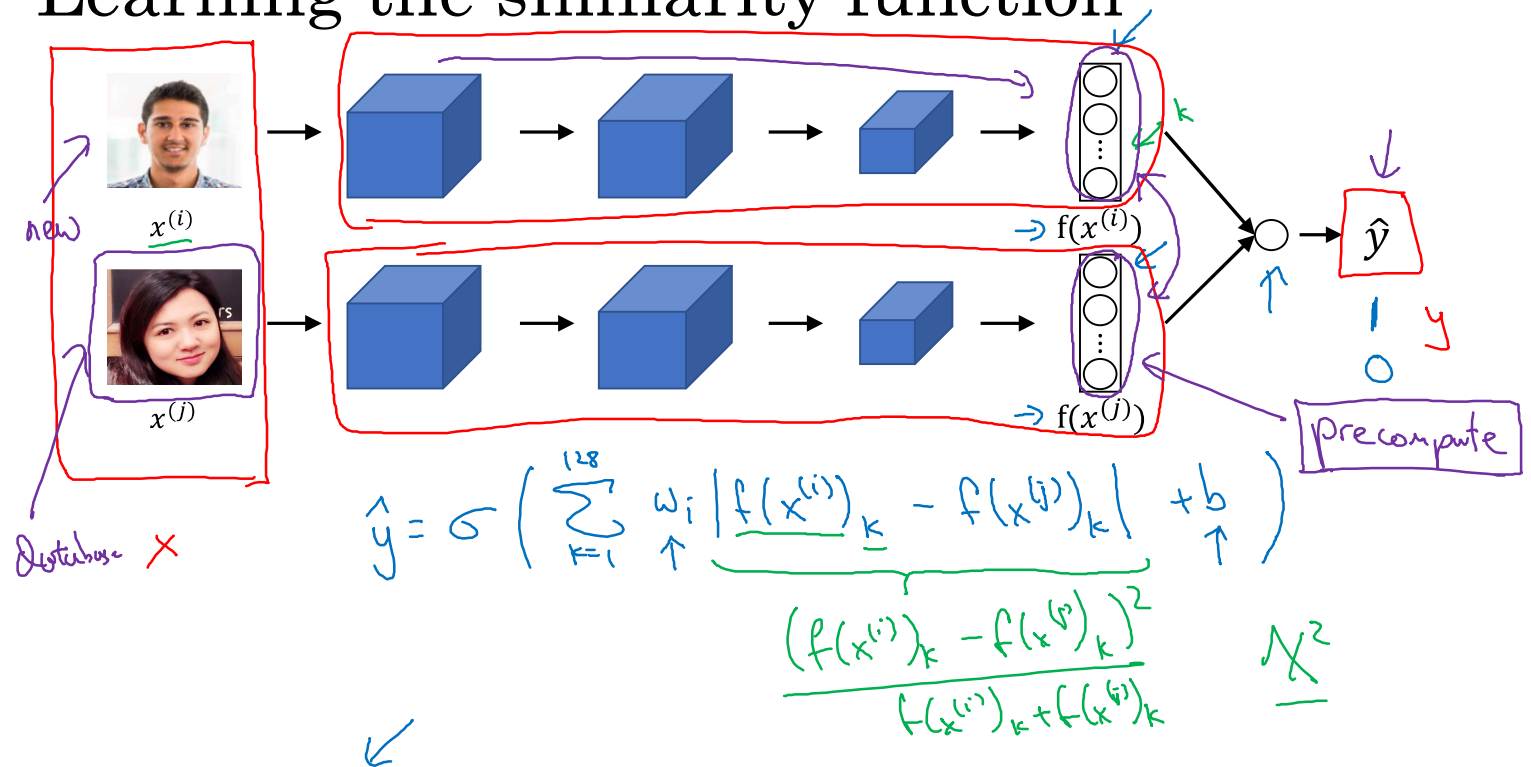
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# Face recognition

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## Face verification and binary classification

# Learning the similarity function











[Taigman et. al., 2014. DeepFace closing the gap to human level performance]

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# Face verification supervised learning

$x$		$y$	
		1	"Same"
		0	"Different"
		0	
		1	

[Taigman et. al., 2014. DeepFace closing the gap to human level performance]

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