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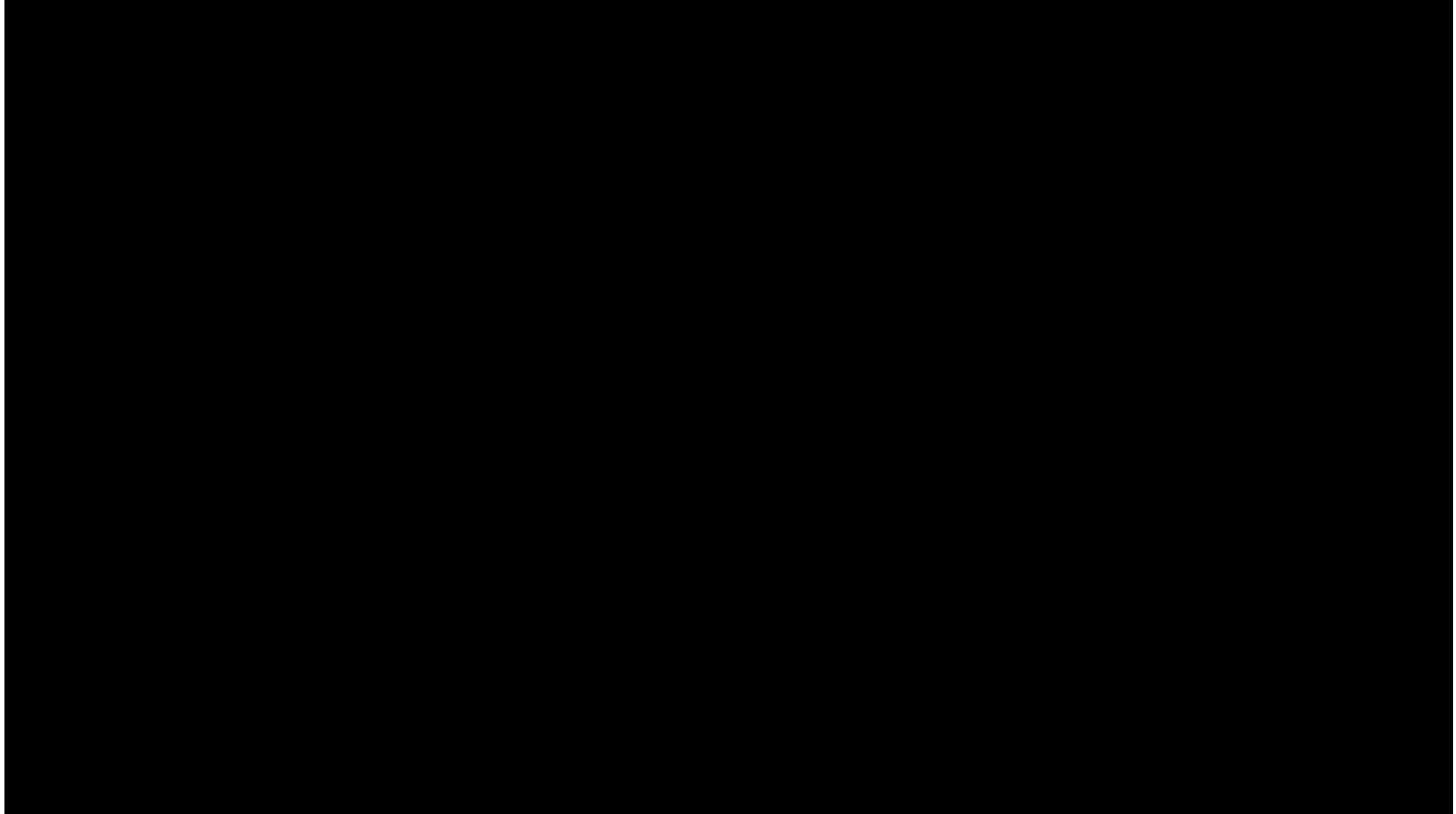
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# Face recognition

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What is face  
recognition?

# Face recognition



# 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 ←



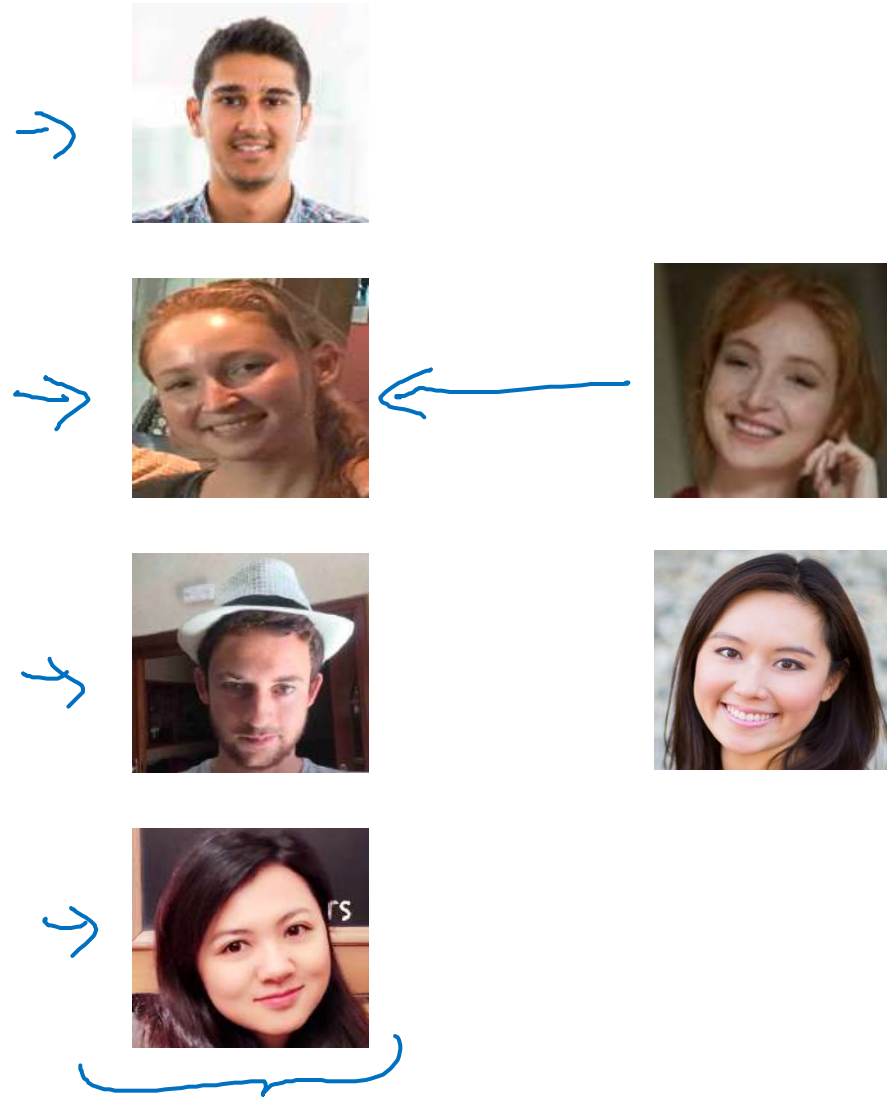
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# Face recognition

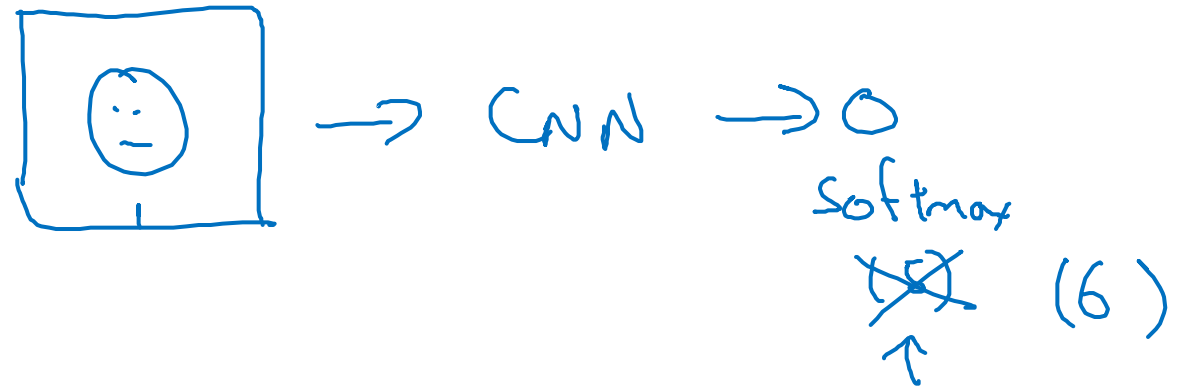
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# One-shot learning

# 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

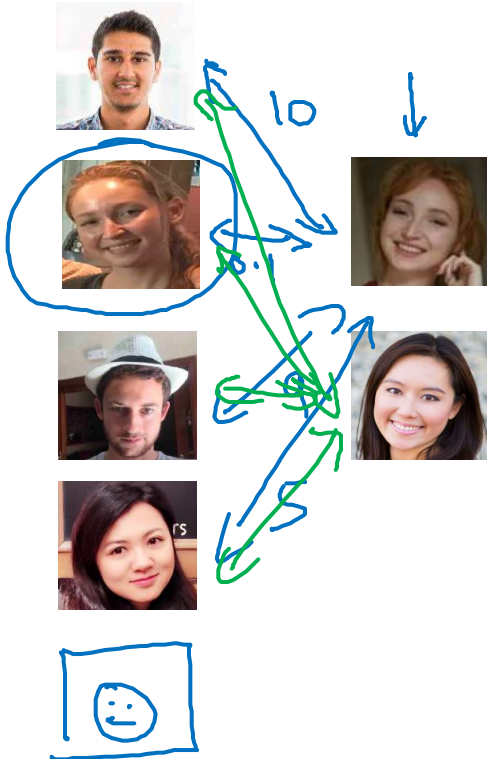
If  $d(\text{img1}, \text{img2}) \leq \tau$

$> \tau$

“same”

“different”

} Verification.



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



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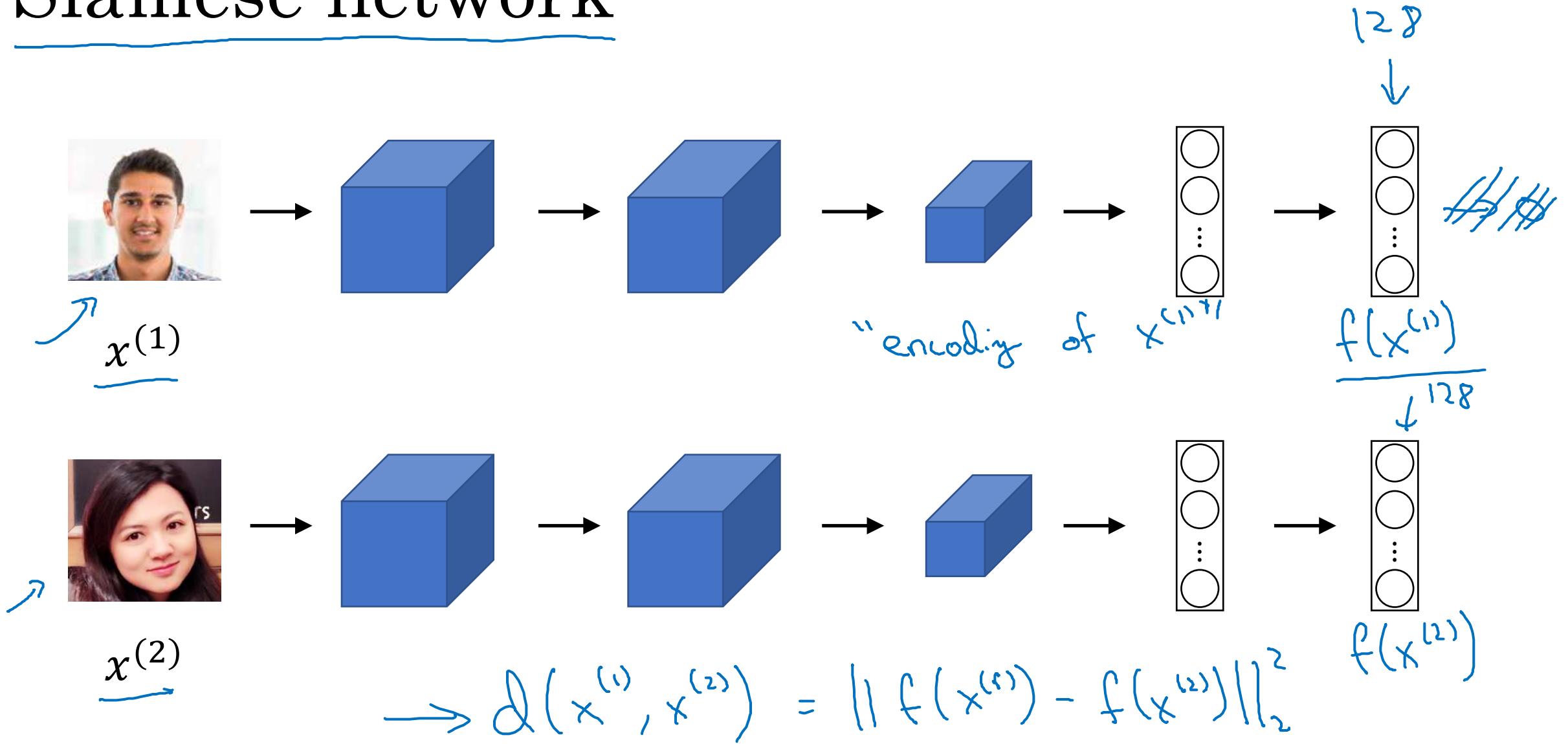
# Face recognition

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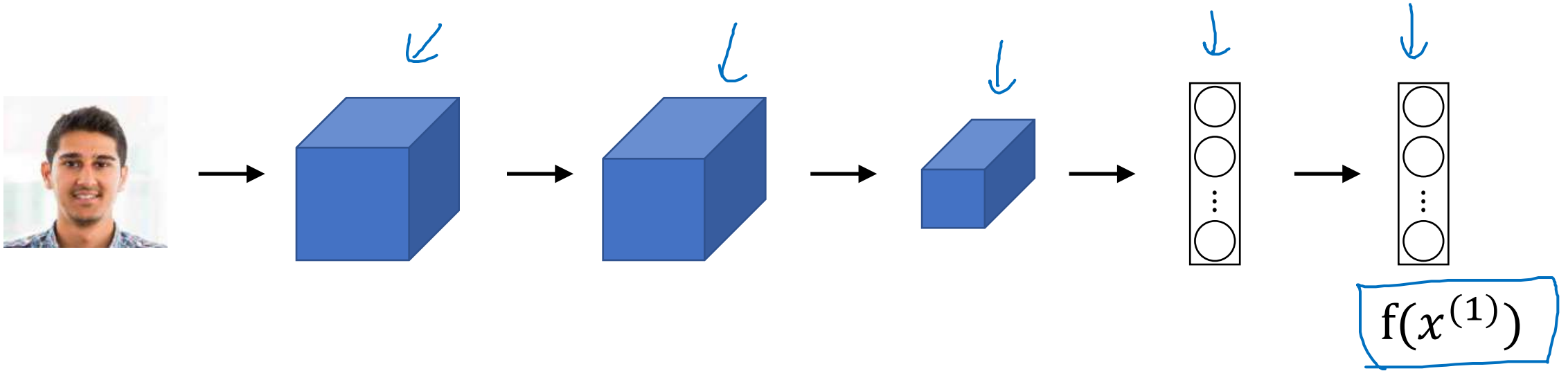
# Siamese network



# Siamese network



# 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.



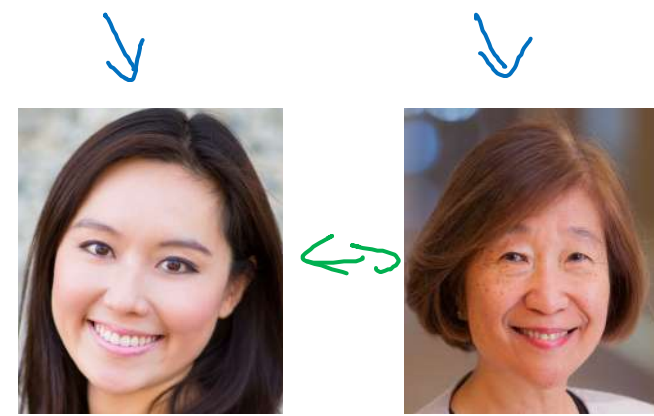
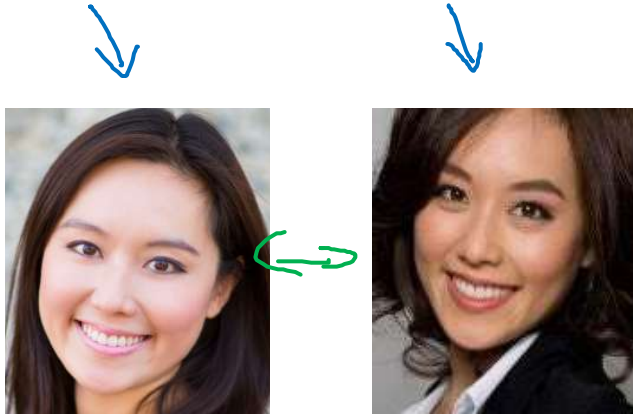
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# Face recognition

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## Triplet loss

# Learning Objective



Anchor

Positive

Anchor

Negative

A

$$d(A, P) = 0.5$$

Want:

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

A

$$d(A, N) = \cancel{0.5} = 0.7$$

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

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

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

# Loss function

Given 3 images

$A, P, N$ :

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

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

$A, P$   
↑ ↑

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

# Training set using triplet loss

Anchor



⋮



Positive



⋮



Negative



⋮



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



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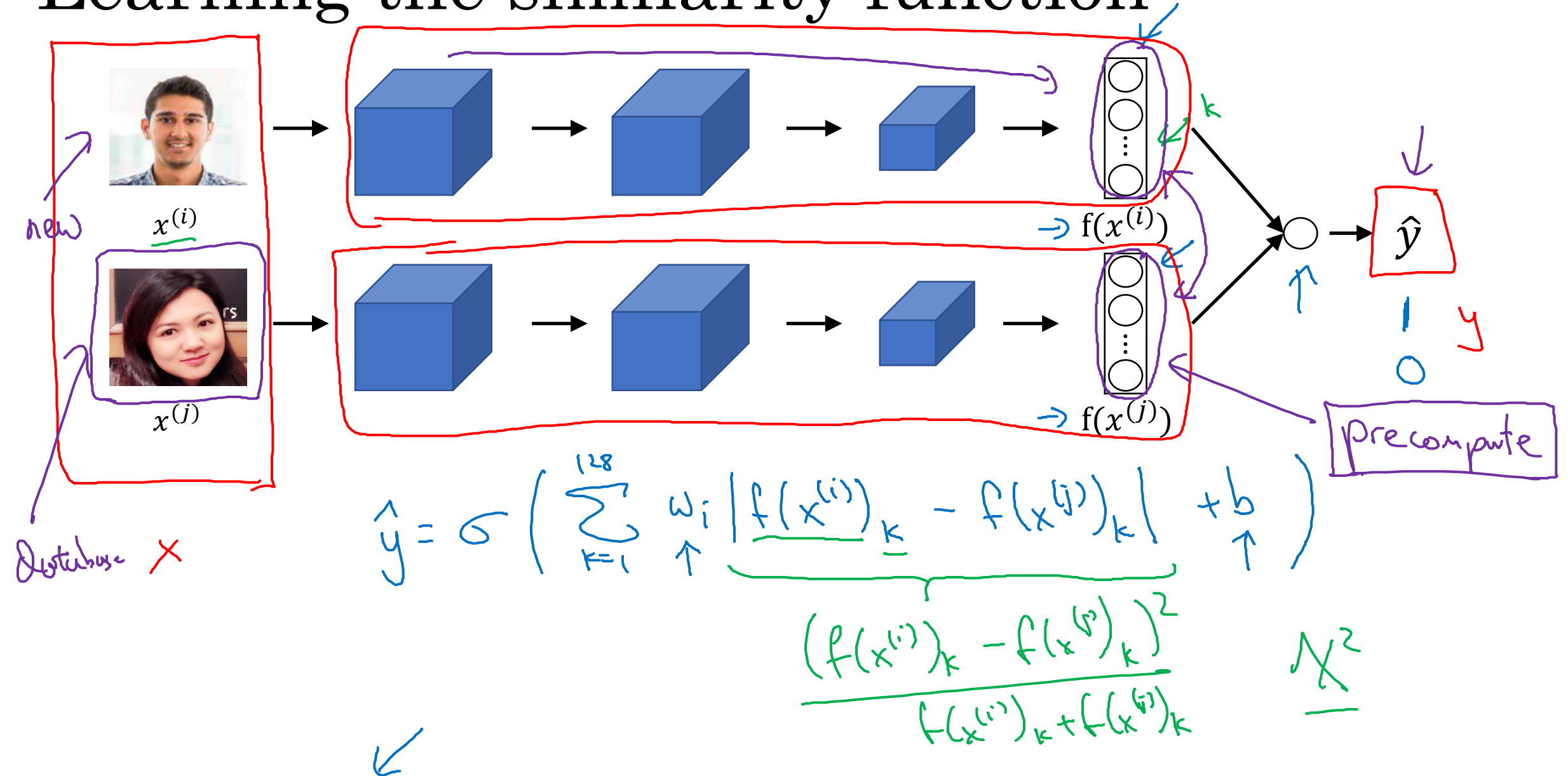
# Face recognition

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## Face verification and binary classification



# Learning the similarity function



# Face verification supervised learning

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