

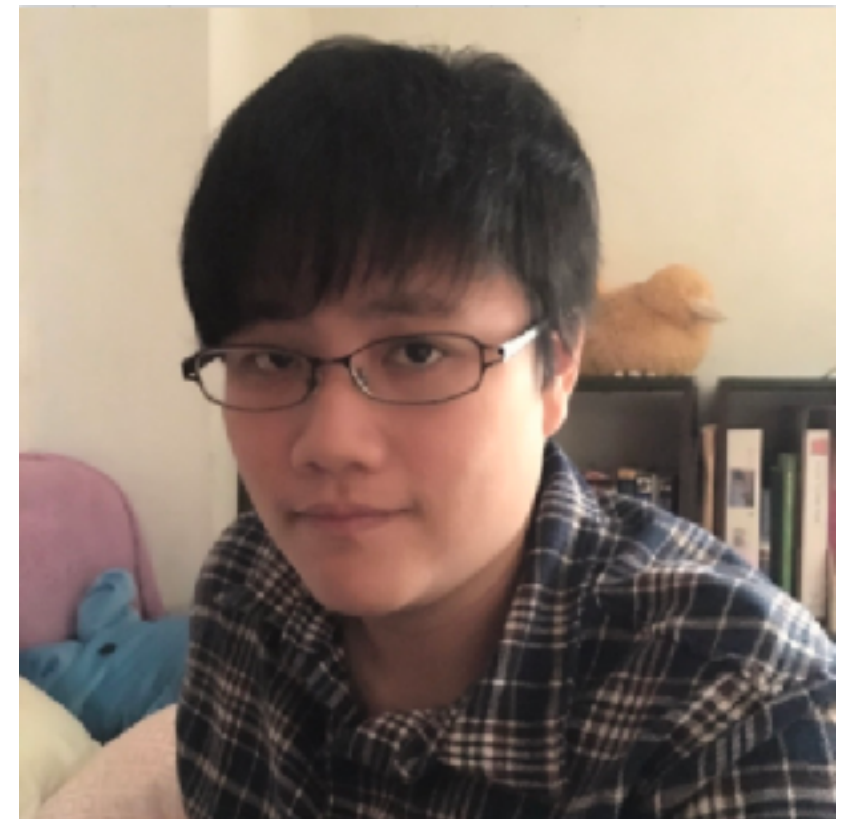
# 筆跡判定

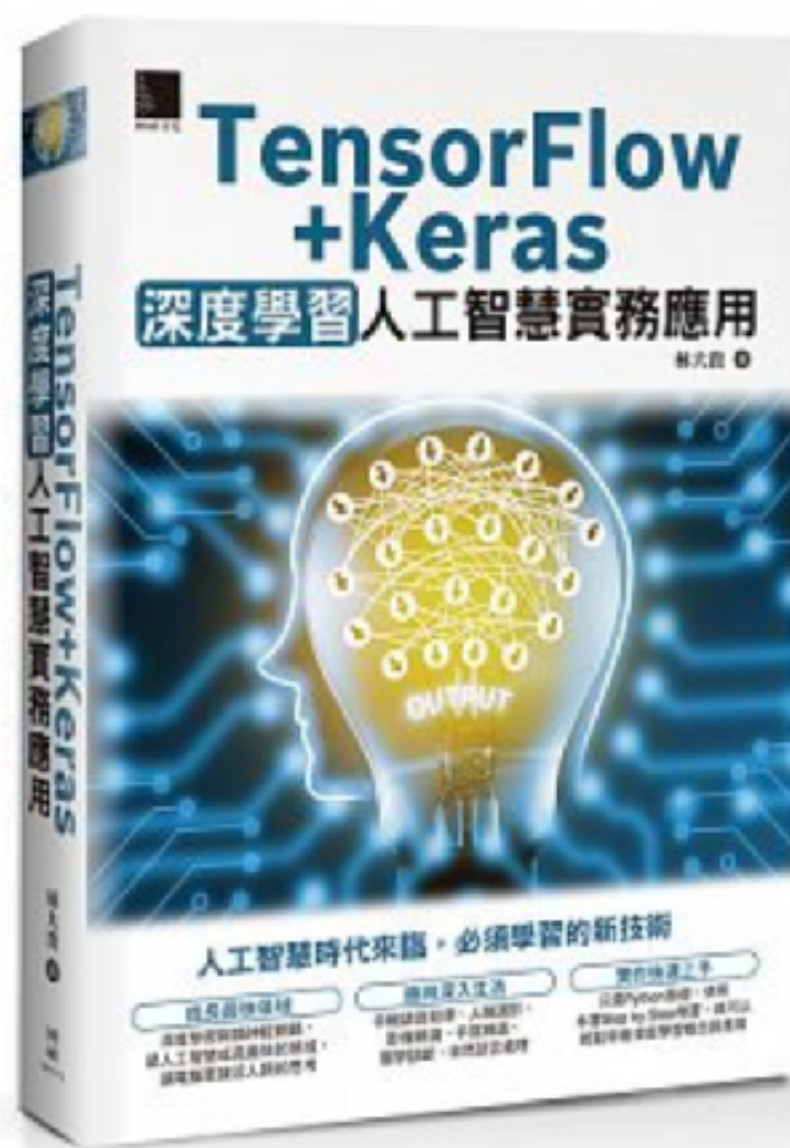
學員 陳祐嘉

# 簽名比對

# About Me

- 政大資管所(2017.07)
- 富邦人壽



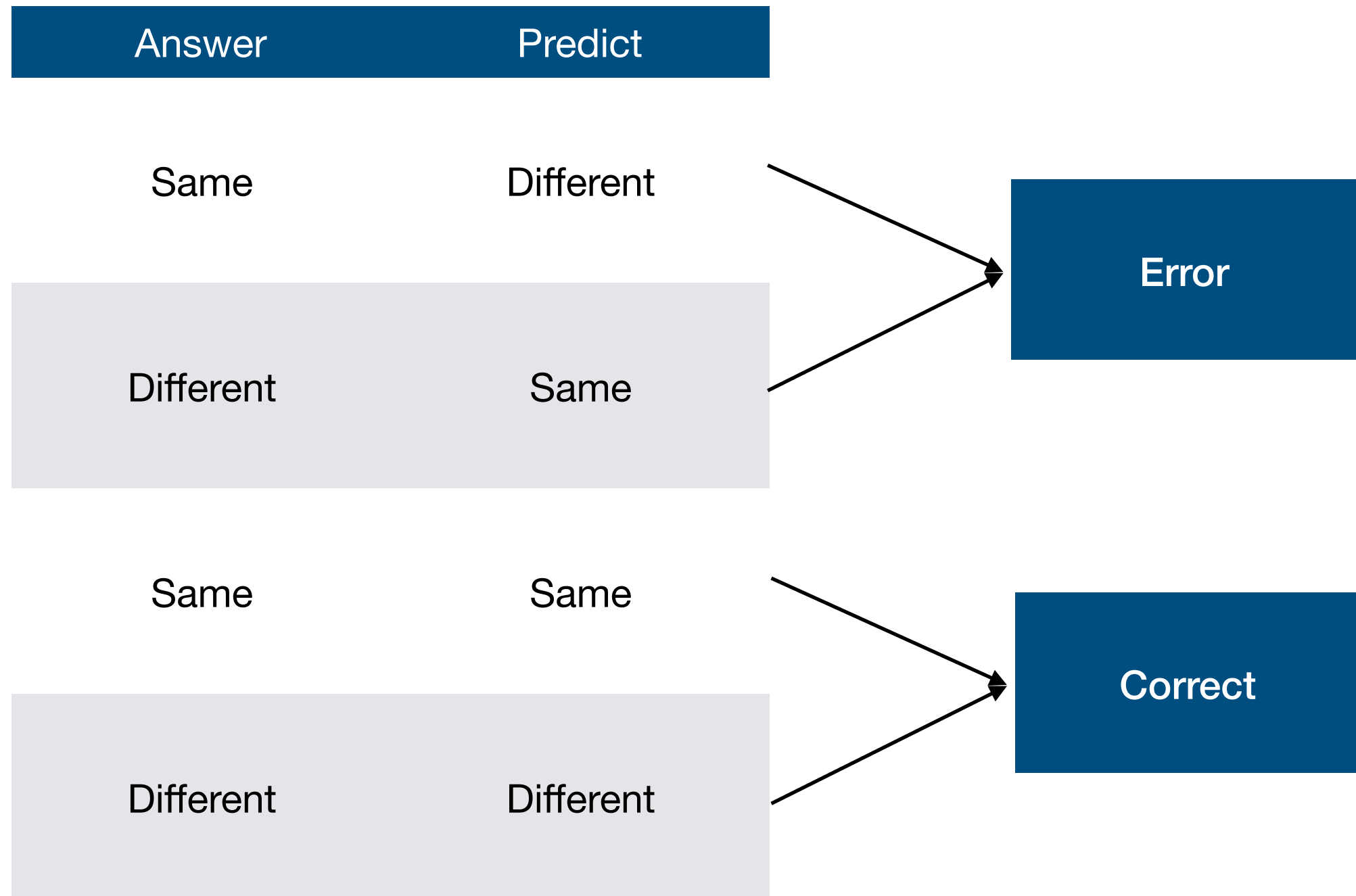


在開始之前...

- 簽名比對不是單純的「分類問題」
- 簽名有分成「線上」及「線下」簽名
- 今天針對的主要在「線下」的部分

成果？

# 準確率





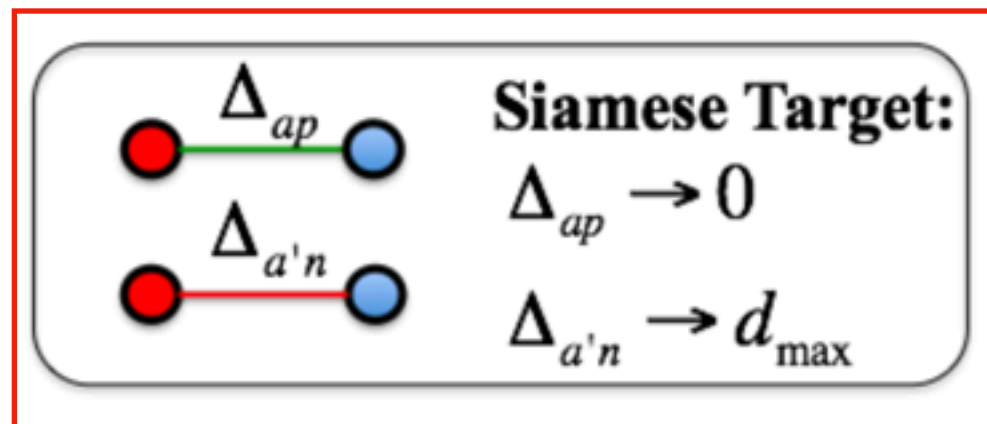
**87%**

**26% ~ 100%**

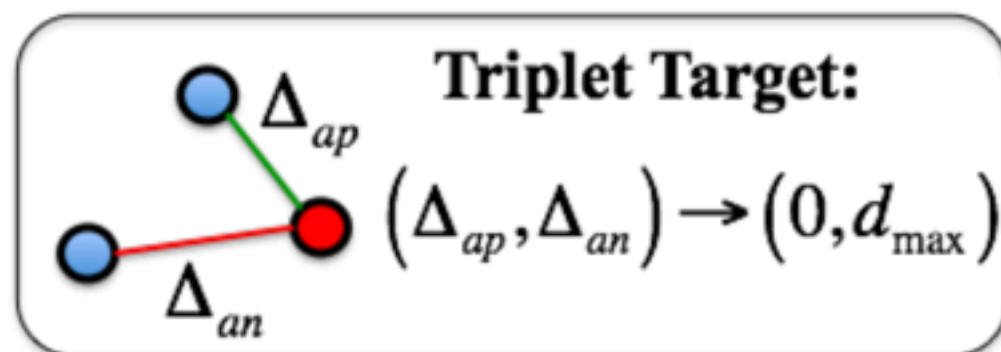
# Siamese Network

# 事情發生在上個禮拜四...

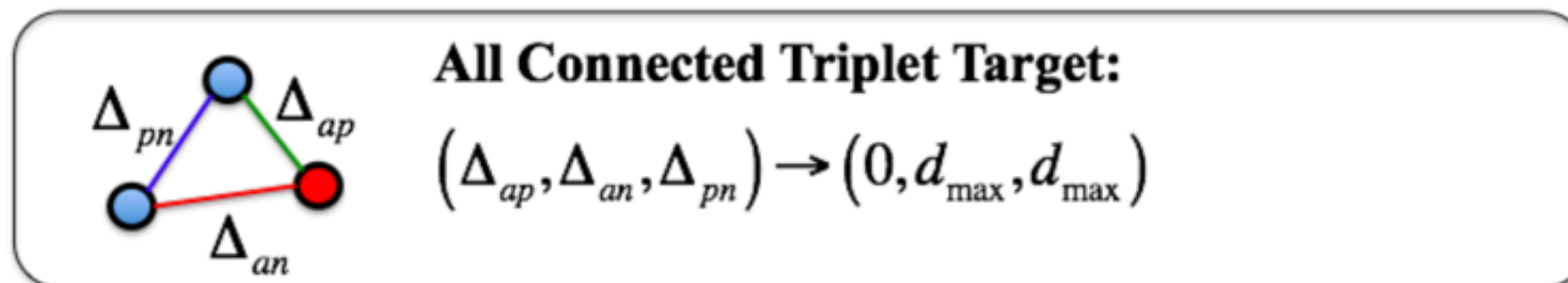
## Siamese Target and Triplet Target



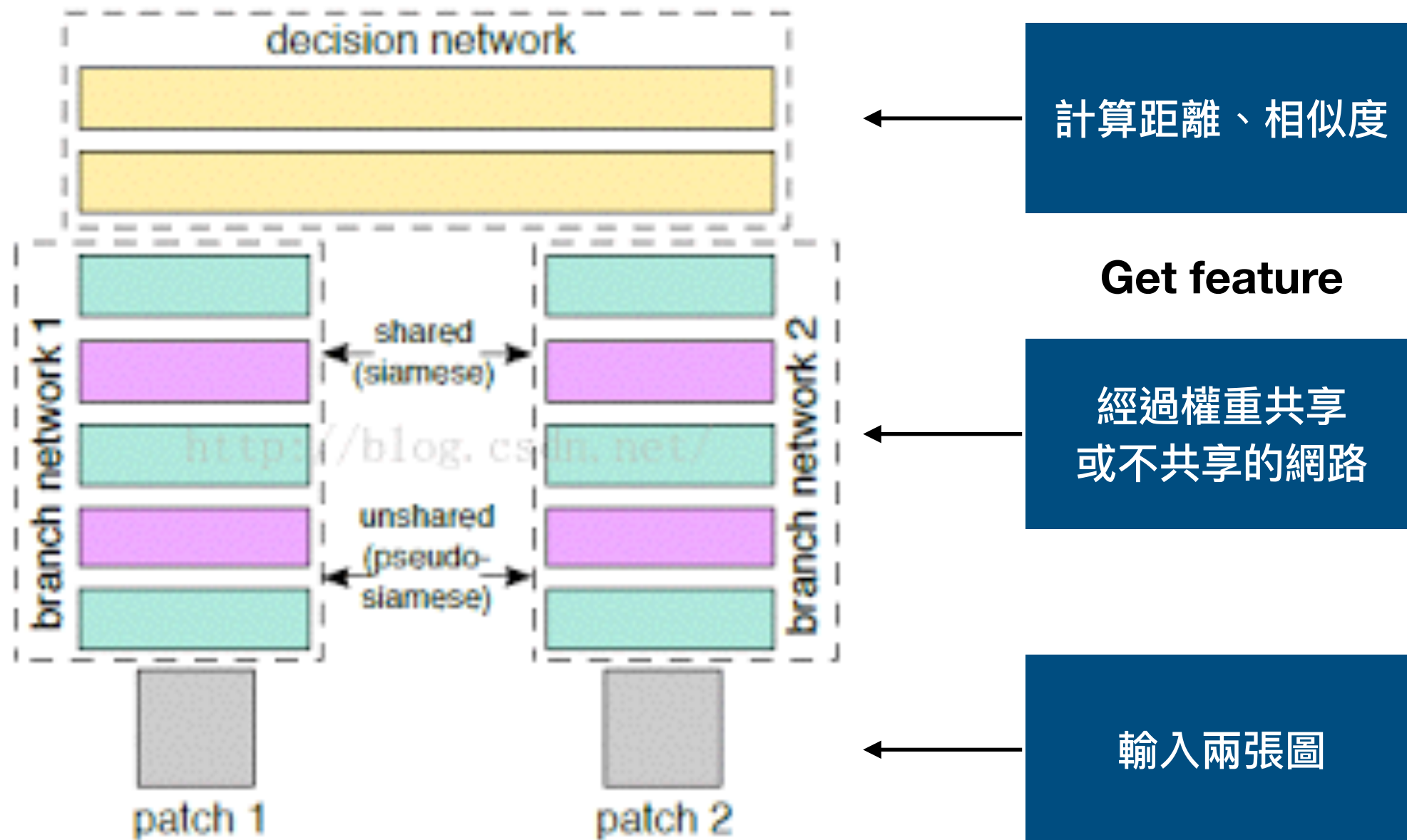
positive pair (a,p) and negative pair (a',n) are separated



positive pair (a,p) and negative pair (a,n) are linked



# Siamese Network



# Siamese Network

**Euclidean distance**

$$\begin{aligned}d(\mathbf{p}, \mathbf{q}) &= d(\mathbf{q}, \mathbf{p}) = \sqrt{(q_1 - p_1)^2 + (q_2 - p_2)^2 + \cdots + (q_n - p_n)^2} \\&= \sqrt{\sum_{i=1}^n (q_i - p_i)^2}.\end{aligned}$$

**From wikipedia**

# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1 - Y)D(F_1, F_2)^2 + \frac{1}{2}Y\max\{0, m - D(F_1, F_2)\}^2$$

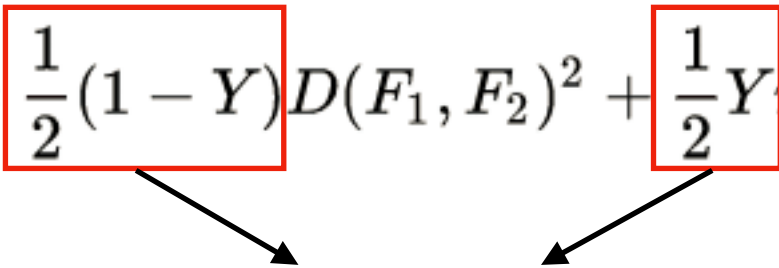
- Training data 中，我們給出成對的圖片
  - 若兩圖相似則給 label = 0
  - 若兩圖不相似則給 label = 1

# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1 - Y)D(F_1, F_2)^2 + \frac{1}{2}Y \max\{0, m - D(F_1, F_2)\}^2$$

Label

The diagram shows the contrastive loss formula. Two terms in the formula are enclosed in red boxes:  $\frac{1}{2}(1 - Y)$  and  $\frac{1}{2}Y$ . Arrows point from these two boxes down to the word "Label" centered below the equation.

- Training data 中，我們給出成對的圖片
  - 若兩圖相似則給 label = 0
  - 若兩圖不相似則給 label = 1



# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1 - Y)D(F_1, F_2)^2 + \frac{1}{2}Y\max\{0, m - D(F_1, F_2)\}^2$$

- $Y = 0$ ，兩圖相似
- $D(F_1, F_2)$  為距離，距離越小，loss 越小

# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1-Y)D(F_1, F_2)^2 + \frac{1}{2}Y\max\{0, m - D(F_1, F_2)\}^2$$

- $Y = 1$ ，兩圖不相似
- $D(F_1, F_2)$  為距離，距離越大，loss 越小

# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1-Y)D(F_1, F_2)^2 + \frac{1}{2}Y\max\{0, m - D(F_1, F_2)\}^2$$

- $Y = 1$ ，兩圖不相似
- $D(F_1, F_2)$  為距離，距離越大，loss 越小



# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1-Y)D(F_1, F_2)^2 + \frac{1}{2}Y \max\{0, m - D(F_1, F_2)\}^2$$

- $Y = 1$ ，兩圖不相似
  - $D(F_1, F_2)$  為距離，距離越大，loss 越小
  - $m$  為一個固定值，當  $D(F_1, F_2) < m$  就會產生損失

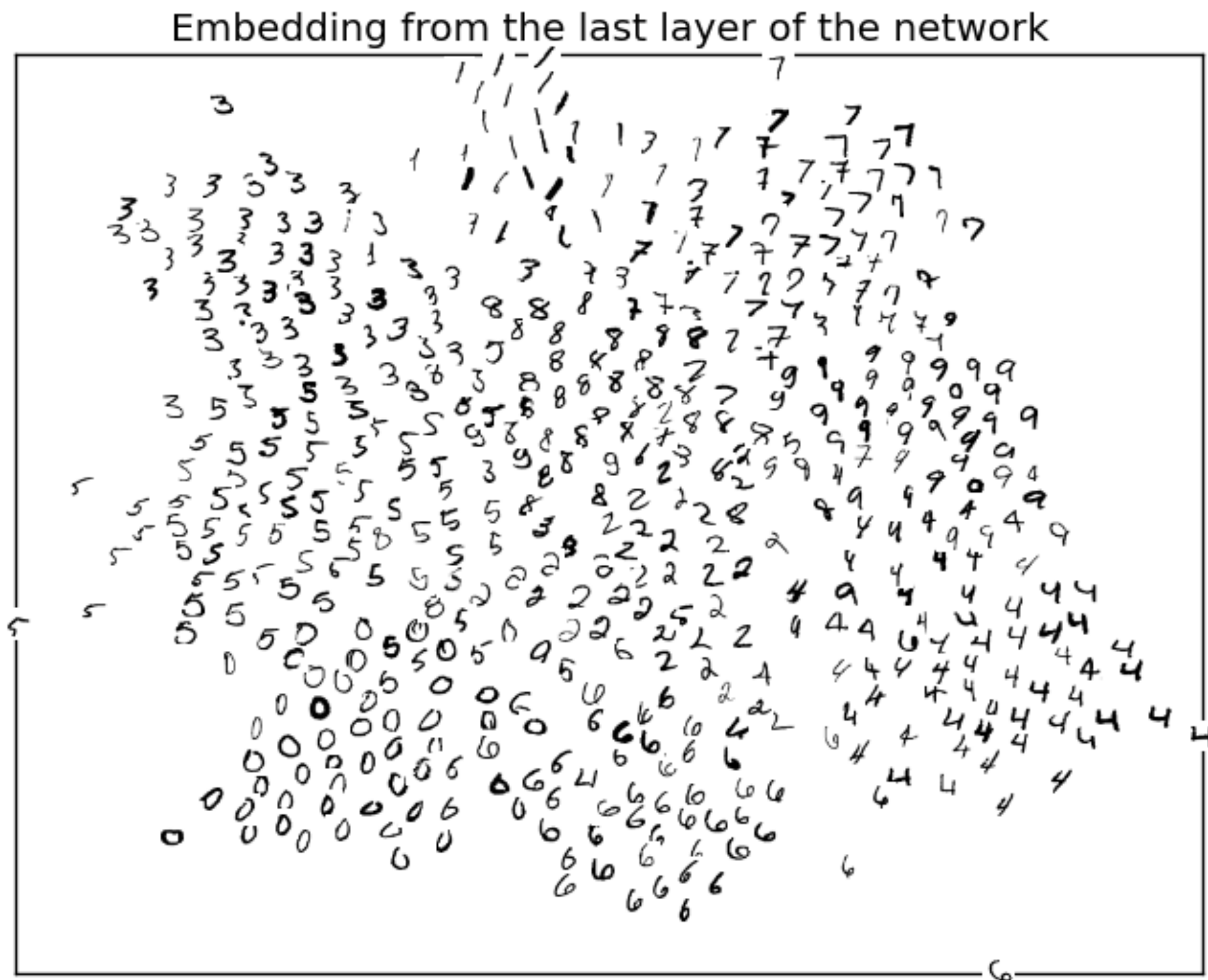
# Siamese Network

## Contrastive loss

$$L(F_1, F_2, Y) = \frac{1}{2}(1 - Y)D(F_1, F_2)^2 + \frac{1}{2}Y\max\{0, m - D(F_1, F_2)\}^2$$

- 這裡的 loss function 要能判別，對兩張圖片
  - 相似性越大，距離越近，loss就越小
  - 相似性越小，距離越遠，loss就越小

# Siamese Network



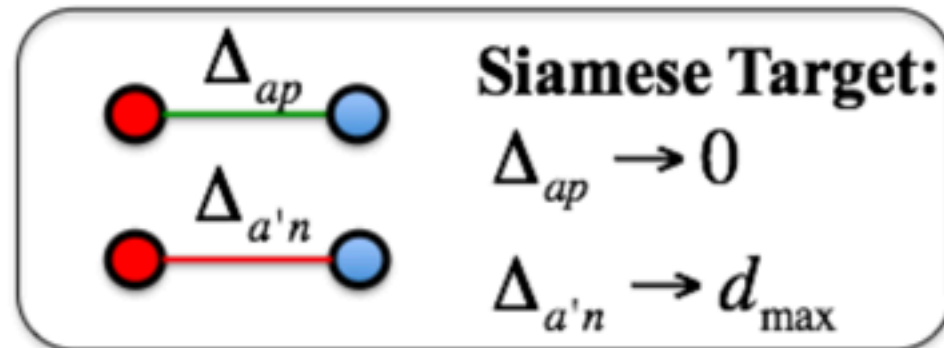
但這不是唯一的方法...

# Triplet loss

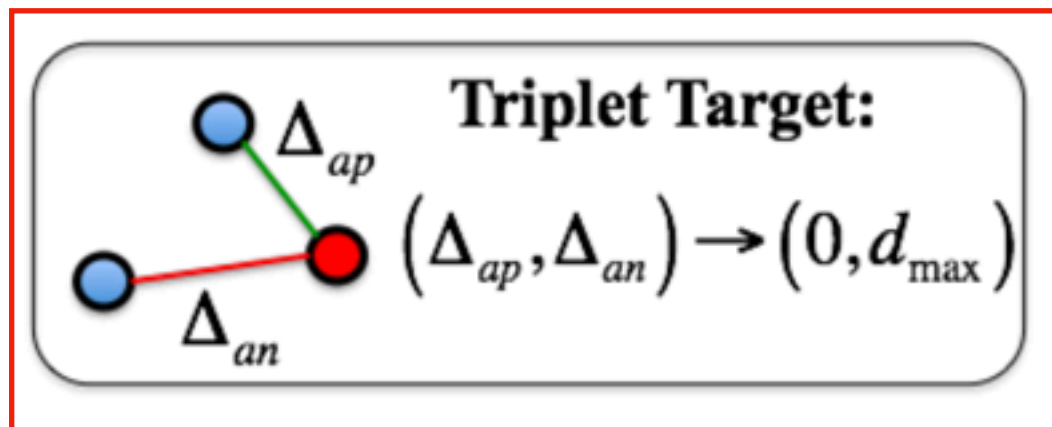


# 還是發生在上個禮拜四...

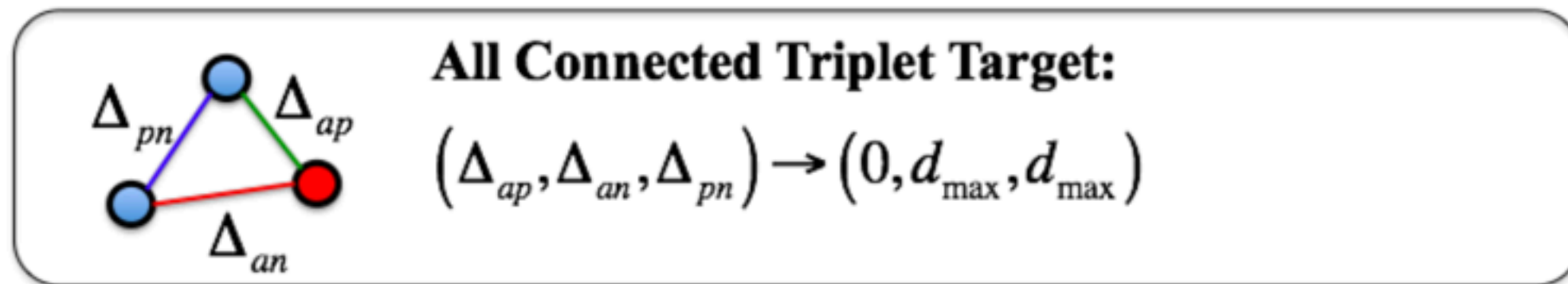
## Siamese Target and Triplet Target



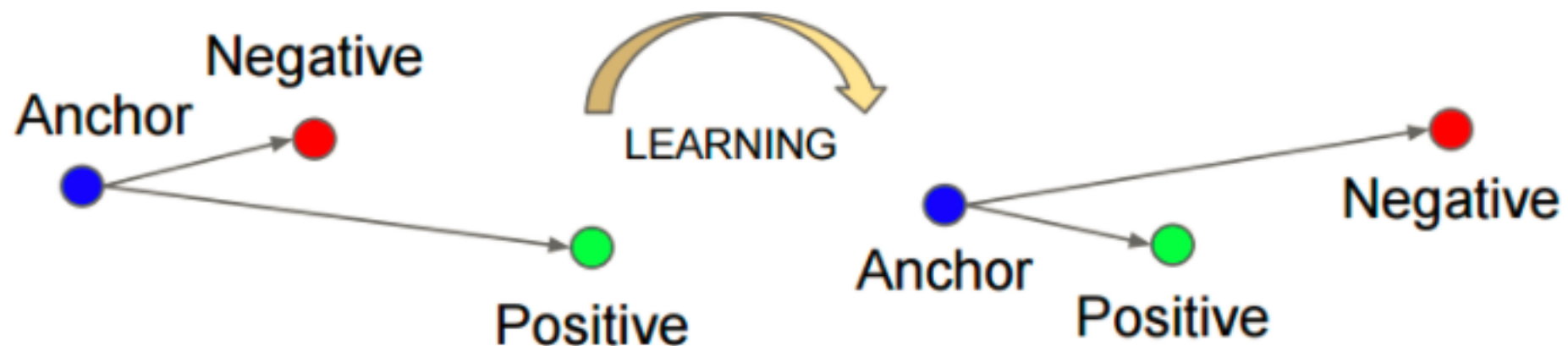
positive pair (a,p) and negative pair (a',n) are separated



positive pair (a,p) and negative pair (a,n) are linked



# Triplet loss

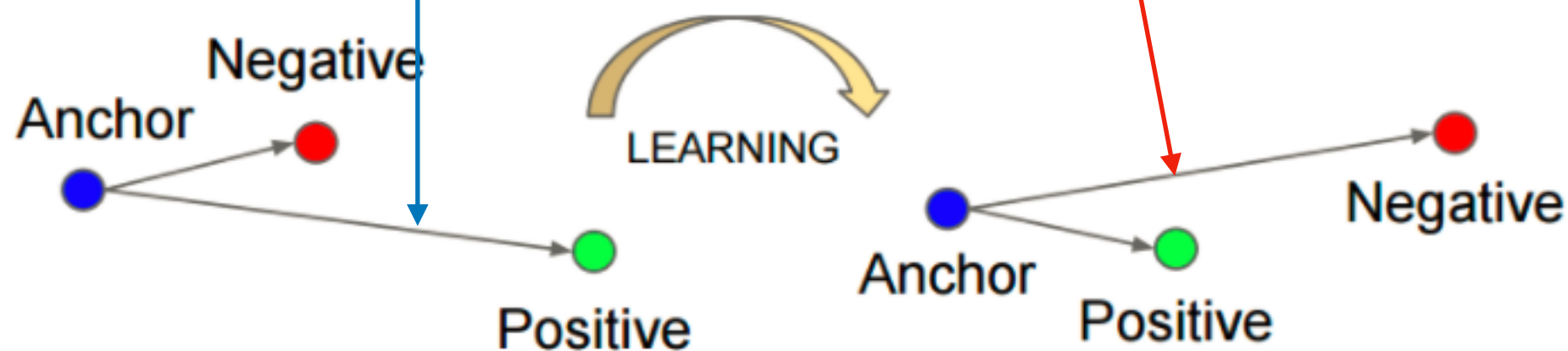


- Triplet 是一個三元組：Anchor, Negative, Positive
- 訓練一個**參數共享或者不共享**的網絡，得到三個元素的特徵表達

# Triplet loss

目標函數：

$$\sum_i^N \left[ \underbrace{\|f(x_i^a) - f(x_i^p)\|_2^2}_{\text{blue line}} - \underbrace{\|f(x_i^a) - f(x_i^n)\|_2^2}_{\text{red line}} + \alpha \right]_+$$

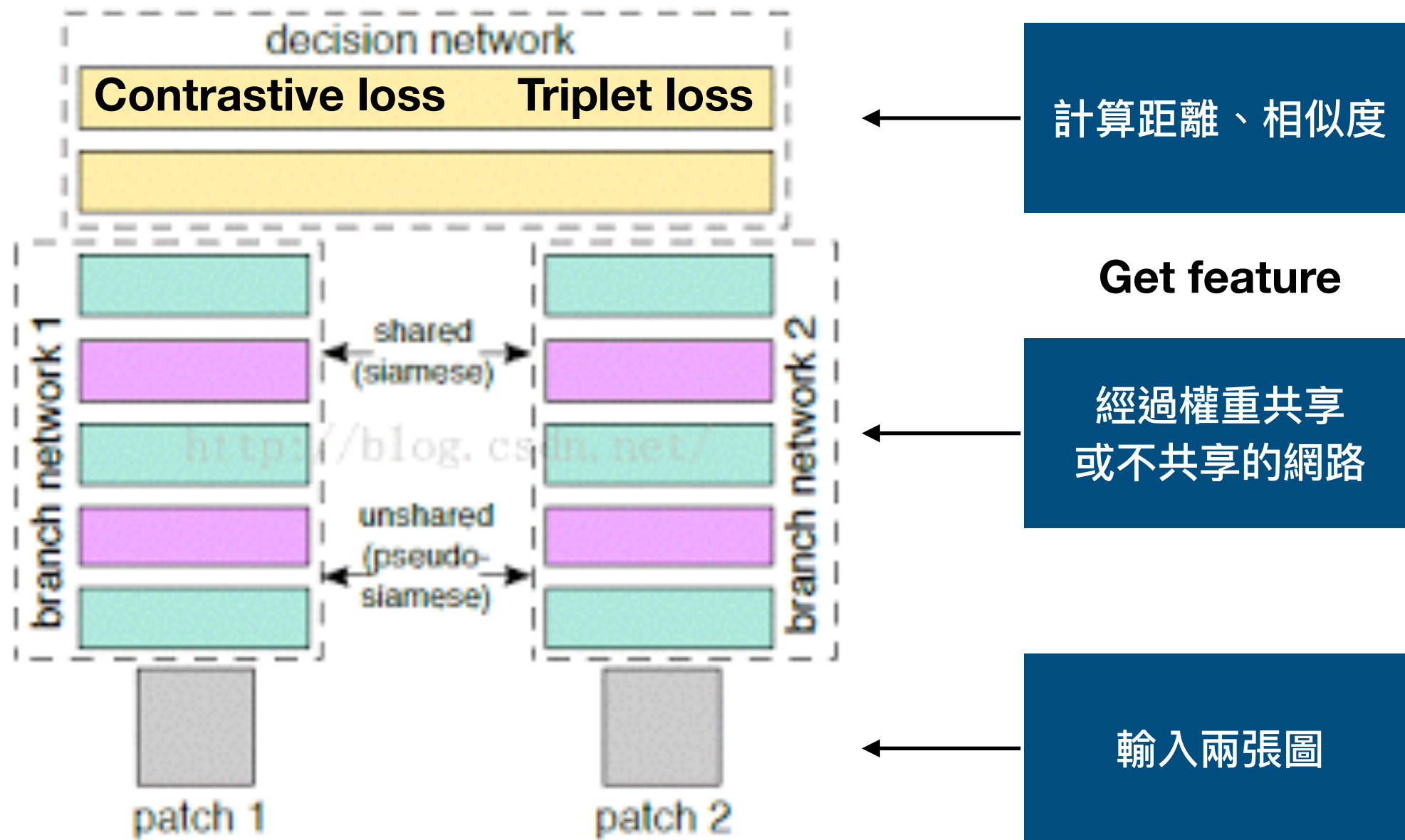


所以我說...Label 呢？



好像...真的沒有

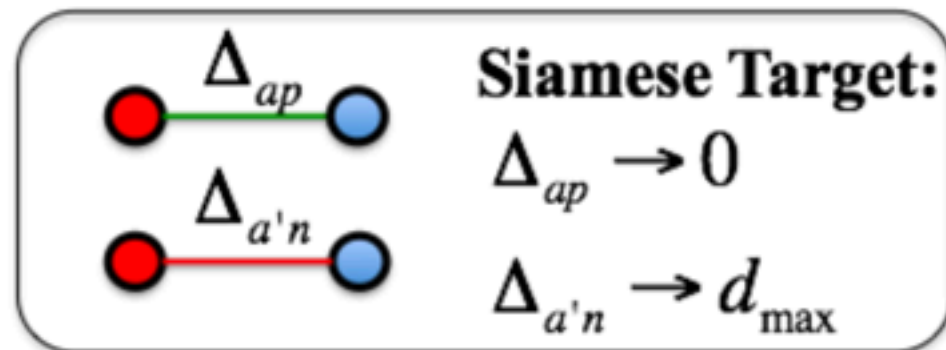
# Conclusion



有人還記得我今天的主題是筆跡判定嗎... ?

# 我保證這是最後一次出現...

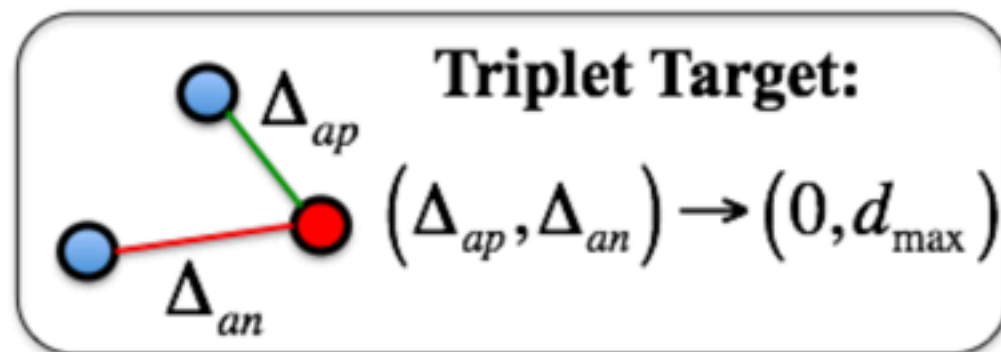
## Siamese Target and Triplet Target



$$\Delta_{ap} \rightarrow 0$$

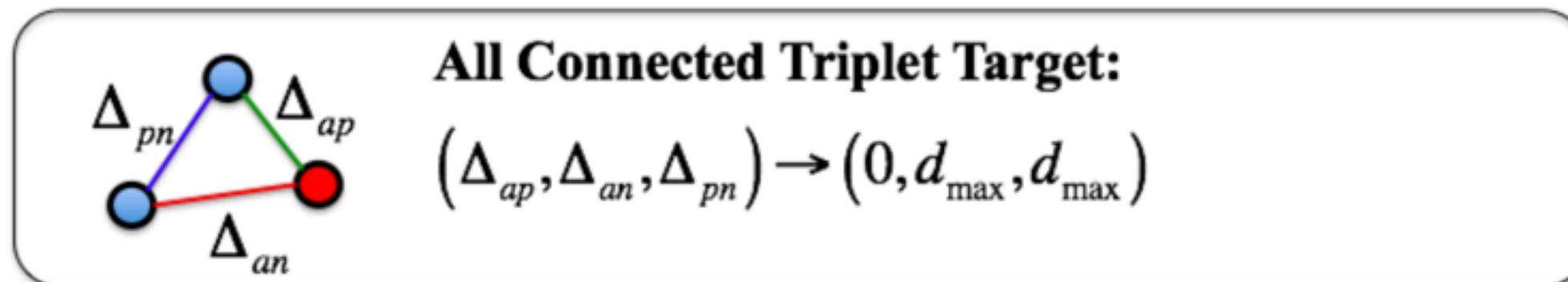
$$\Delta_{a'n} \rightarrow d_{\max}$$

positive pair (a,p) and  
negative pair (a',n) are  
separated



$$(\Delta_{ap}, \Delta_{an}) \rightarrow (0, d_{\max})$$

positive pair (a,p) and  
negative pair (a,n) are  
linked



$$(\Delta_{ap}, \Delta_{an}, \Delta_{pn}) \rightarrow (0, d_{\max}, d_{\max})$$

# 參考資料

- Siamese paper
  - <http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.698.717&rep=rep1&type=pdf> (Signature Verification Using a Siamese Time Delay Neural Network)
- Contrastive loss
  - <http://yann.lecun.com/exdb/publis/pdf/hadsell-chopra-lecun-06.pdf> (Dimensionality Reduction by Learning an Invariant Mapping)
- Triplet loss
  - <http://blog.csdn.net/tangwei2014/article/details/46788025> (triplet loss 原理以及梯度推导)



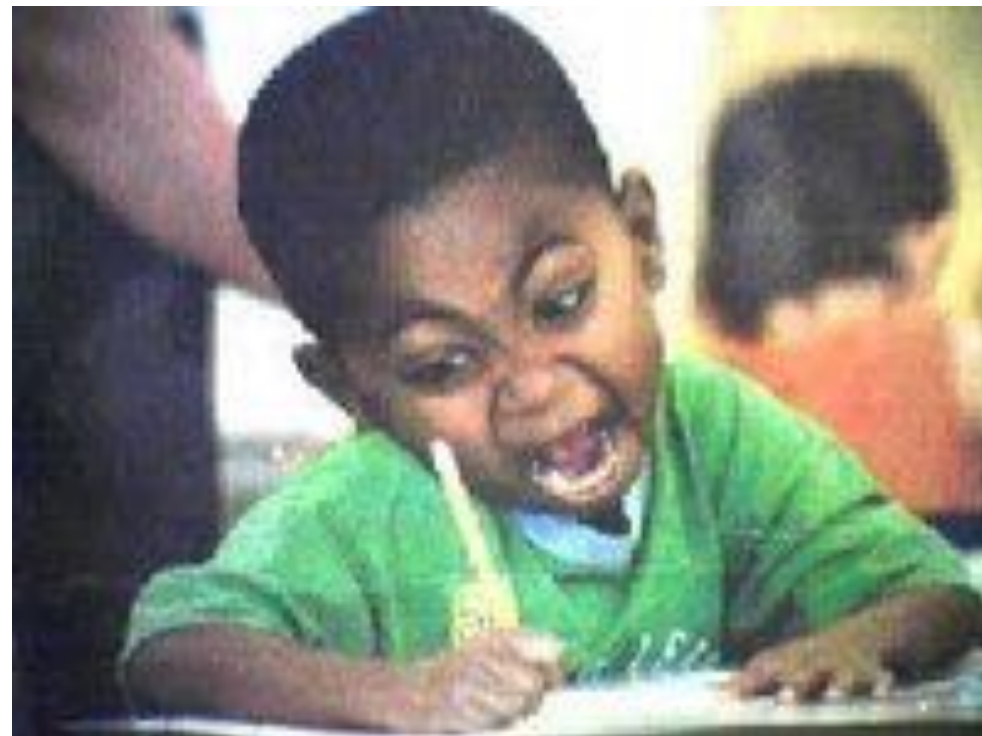
**你們以為結束了嗎...**

敲~碗~



**Q&A**

# 你說資料？







自己的簽名 x 30



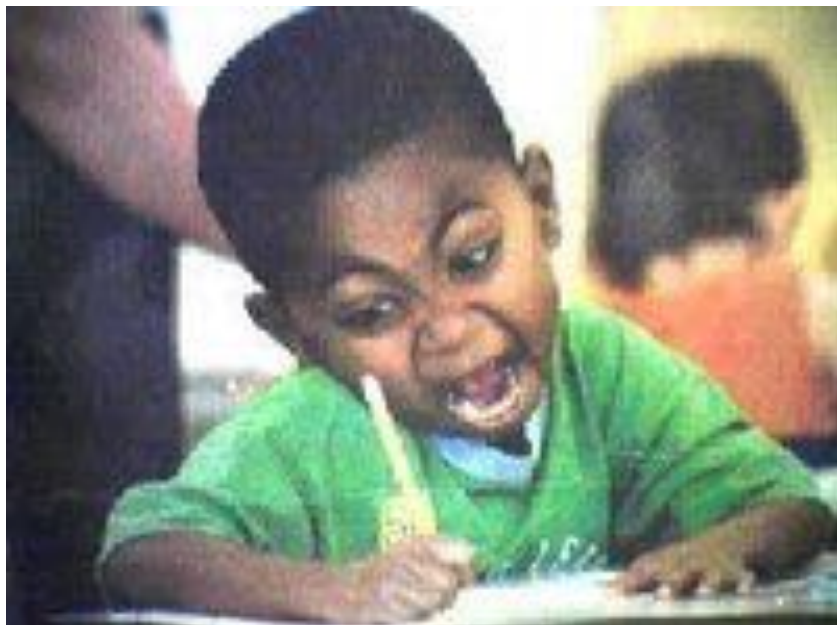
自己的簽名 x 30



模仿別人的簽名 x 30



模仿別人的簽名 x 30





自己的簽名 x 30



自己的簽名 x 30



模仿別人的簽名 x 30



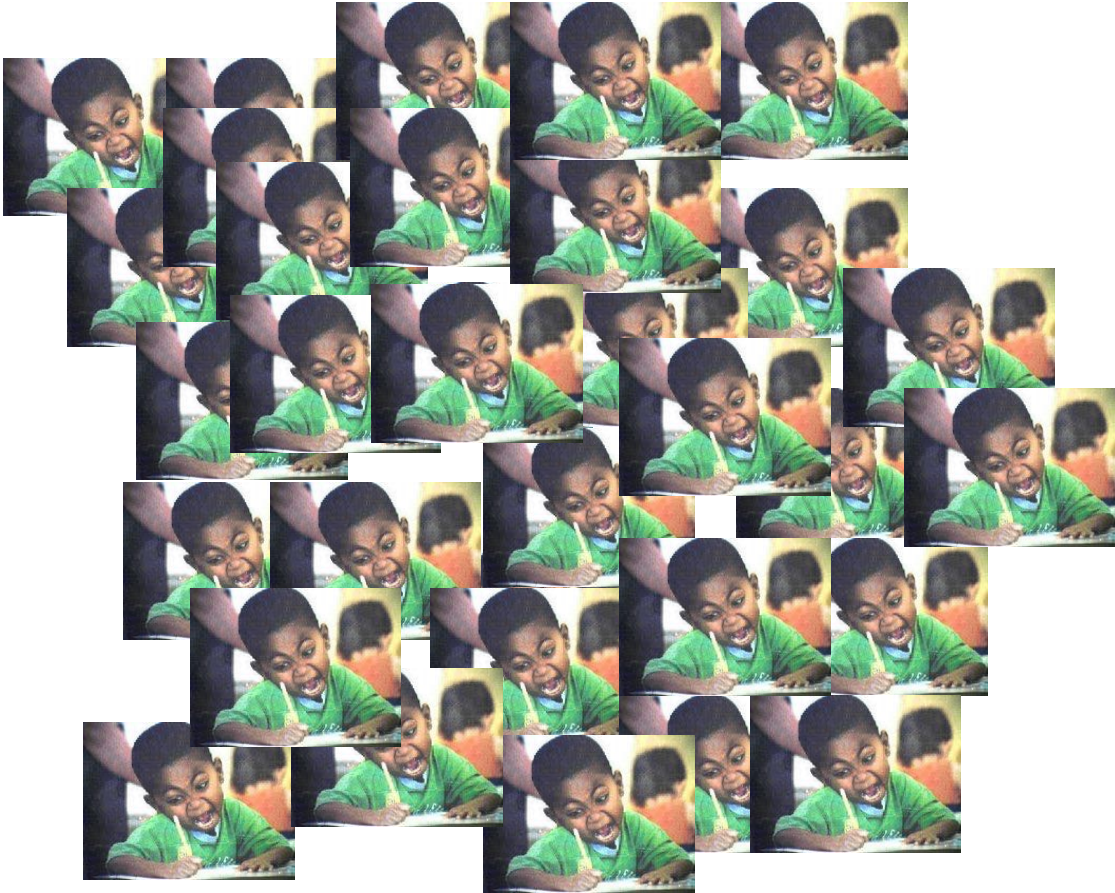
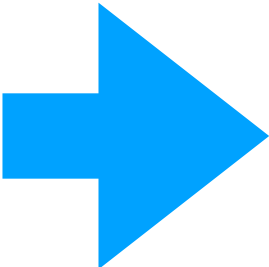
模仿別人的簽名 x 30

For training

For testing









自己的簽名 x 30



自己的簽名 x 30

**900 Positive pairs**



模仿別人的簽名 x 30

**900 negative pairs**



模仿別人的簽名 x 30

謝謝大家