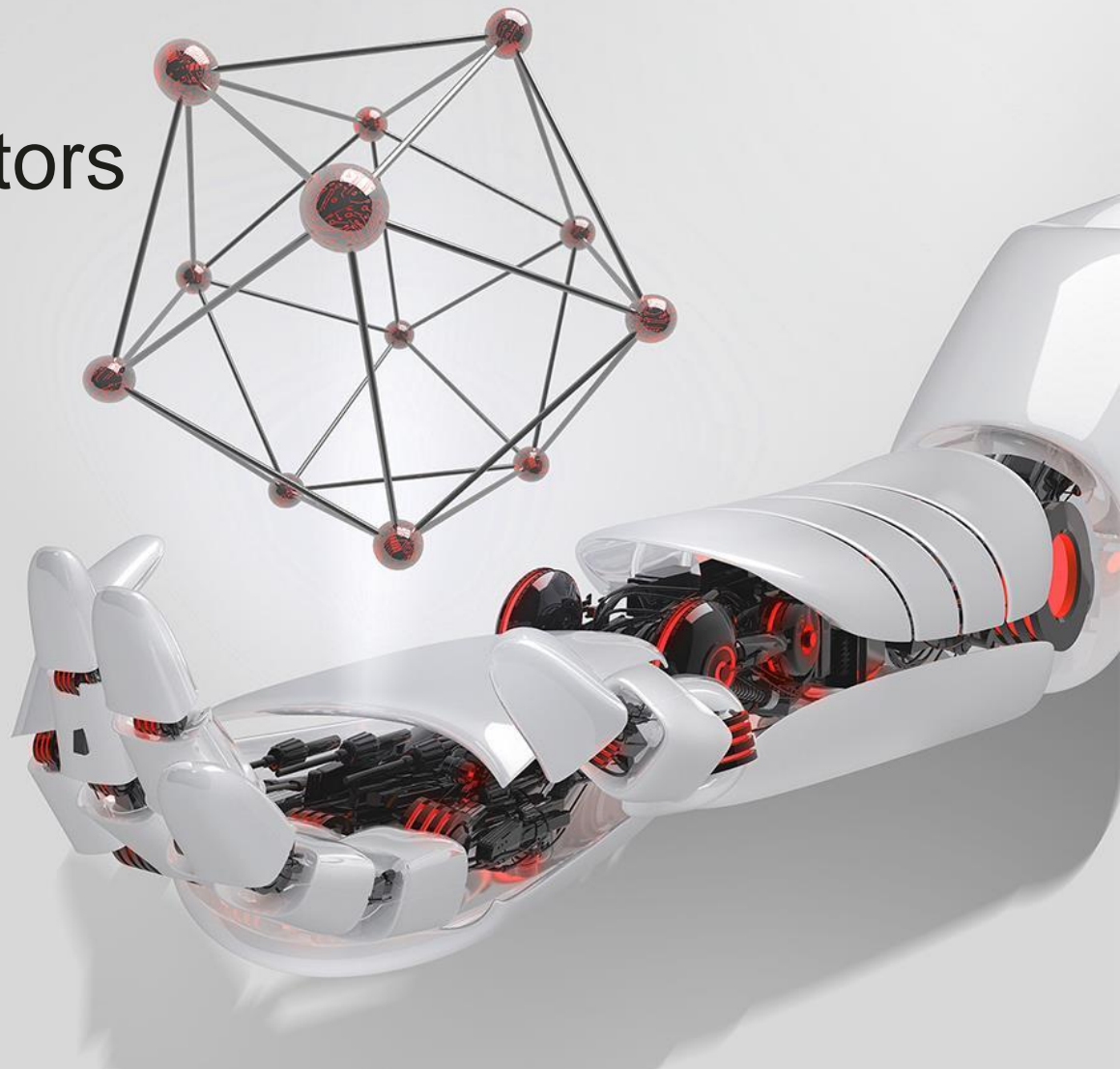


# Metric learning for facial descriptors

Department: name: **Intelligent System Lab**

Author's name: **Petyushko Alexander, Komkov Stepan**

Date: **4<sup>th</sup> of July, 2019**



Security Level:

## RAAI Summer School 2019

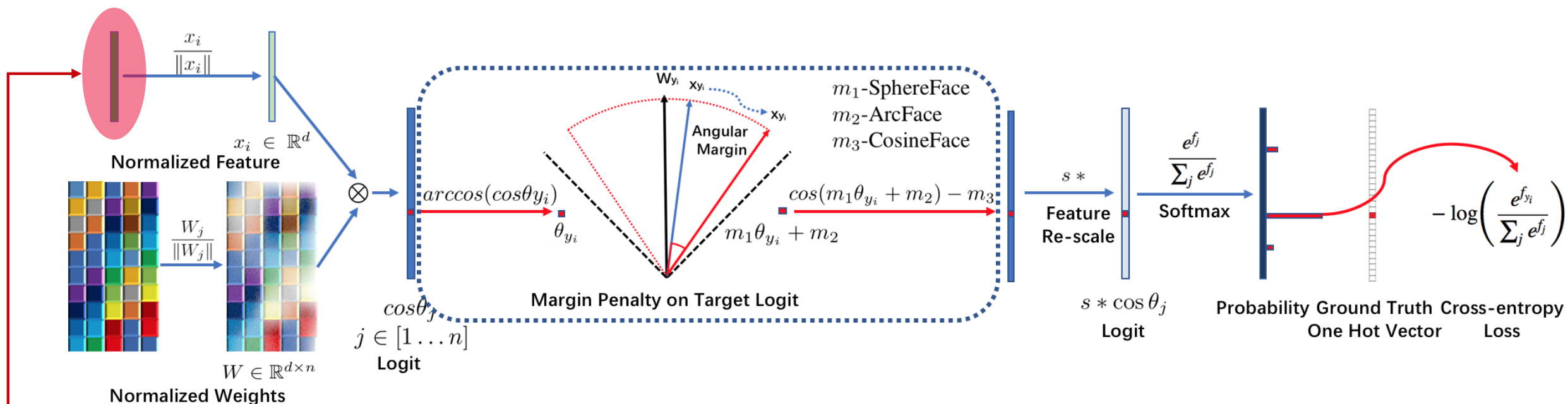
*Developing artificial don't overlook natural*



# ArcFace - Face ID one of the best public solution

Jan 23, 2018: a paper with novel ArcLoss loss function was published

**“ArcFace: Additive Angular Margin Loss for Deep Face Recognition”** by J. Deng, J. Guo, N. Xue and S. Zafeiriou



Don't worry: we are working with this embedding only

# Case Study: Deep learning algorithms for Face Recognition

## Standard pipeline:

- 1) Face detection (rectangular bounding box)
- 2) Landmarks detection (eyes, nose, mouth etc)
- 3) Face normalization (face warping to a predefined position)
- 4) Face feature (embedding) extraction using convolutional DNN
- 5) Usage of the similarity function depending on the specific task

Also, some preprocessing / postprocessing etc...

← Our area of interest as for now



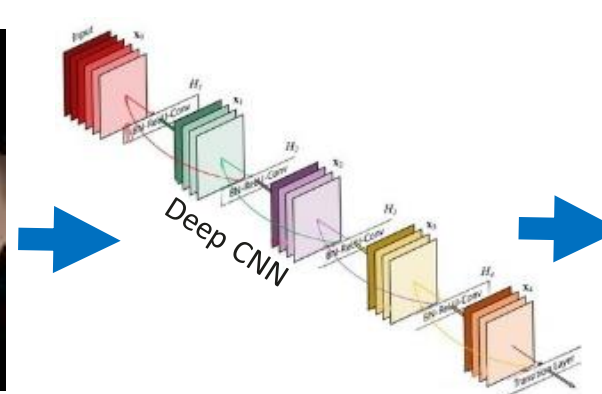
Input face



Face and Key points detection



Face alignment


















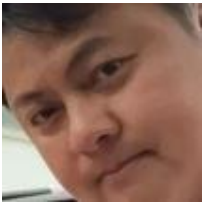




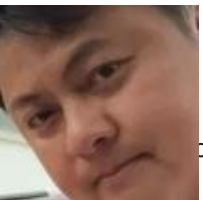






Feature extraction

0.7869
0.7236
0.7232
0.7137
0.3031
0.4667
0.2042
0.7379
0.7137
0.5891
0.7137
0.5967
0.2042
0.3031
...
0.5265

Face features  
[embedding]

# Case Study: Face Verification

Embeddings		VS		Embeddings	$acc_1$	$acc_2$
[0.1 0.3 ... 0.2]		$d=0.1$ 		[0.0 0.29 ... 0.2]		
						$th_1=0.45$
[0.1 0.3 ... 0.2]		$d=0.5$ 		[0.5 0.1 ... 0.2]		
						
						$th_2=0.55$
[0.1 0.3 ... 0.2]		$d=0.7$ 		[0.0 0.4 ... 0.0]		
[0.1 0.3 ... 0.2]		$d=0.8$ 		[0.3 0.4 ... 0.7]		
[0.1 0.3 ... 0.2]		$d=0.9$ 		[0.2 0.0 ... 0.9]		

**Face Verification problem:** design an algorithm to detect whether two input images belong to the same person

# Case Study: Face Verification metric

## Verification metric:

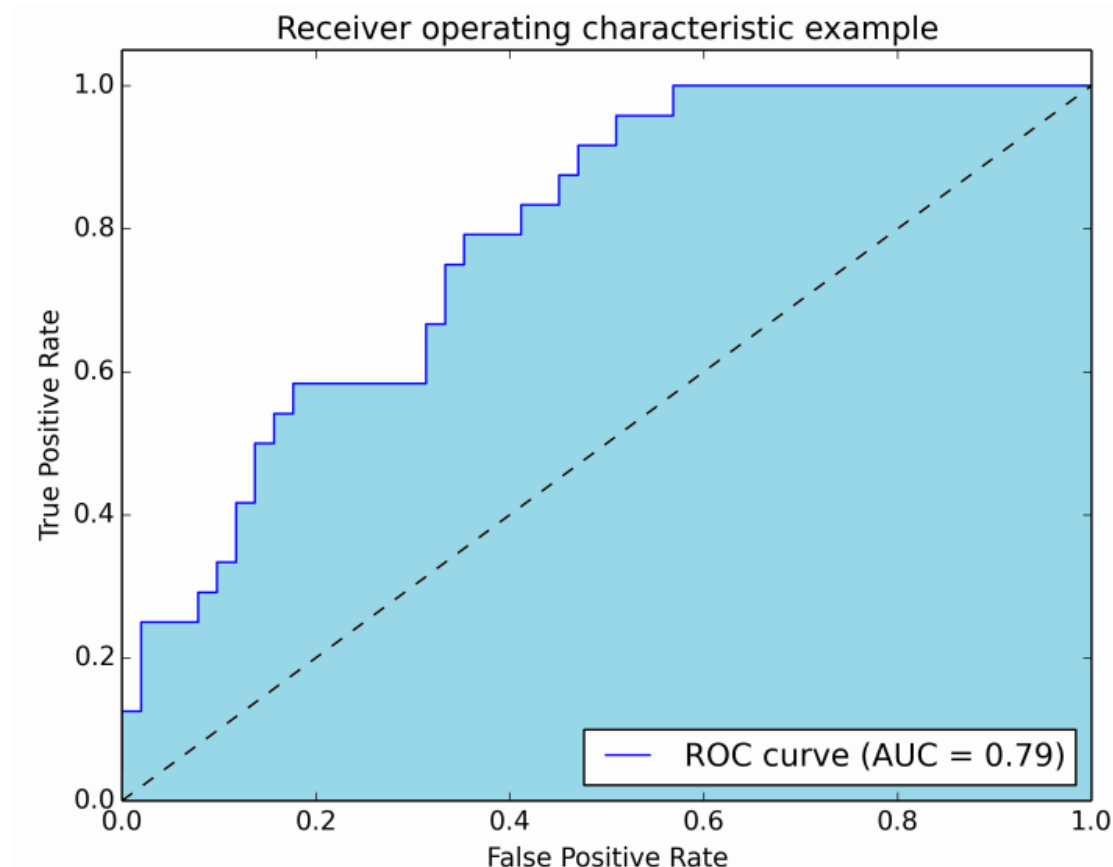
- 1) Receiver Operating Characteristic (ROC)
- 2) Area Under ROC (AUROC)

## ROC:

- 1) X axis: False Positive Rate (FPR) – ratio of “different” pairs recognized as the same person
- 2) Y axis: True Positive Rate (TPR) – ratio of “same” pairs recognized as the same person
- 3) Each point corresponds to some threshold  $th$

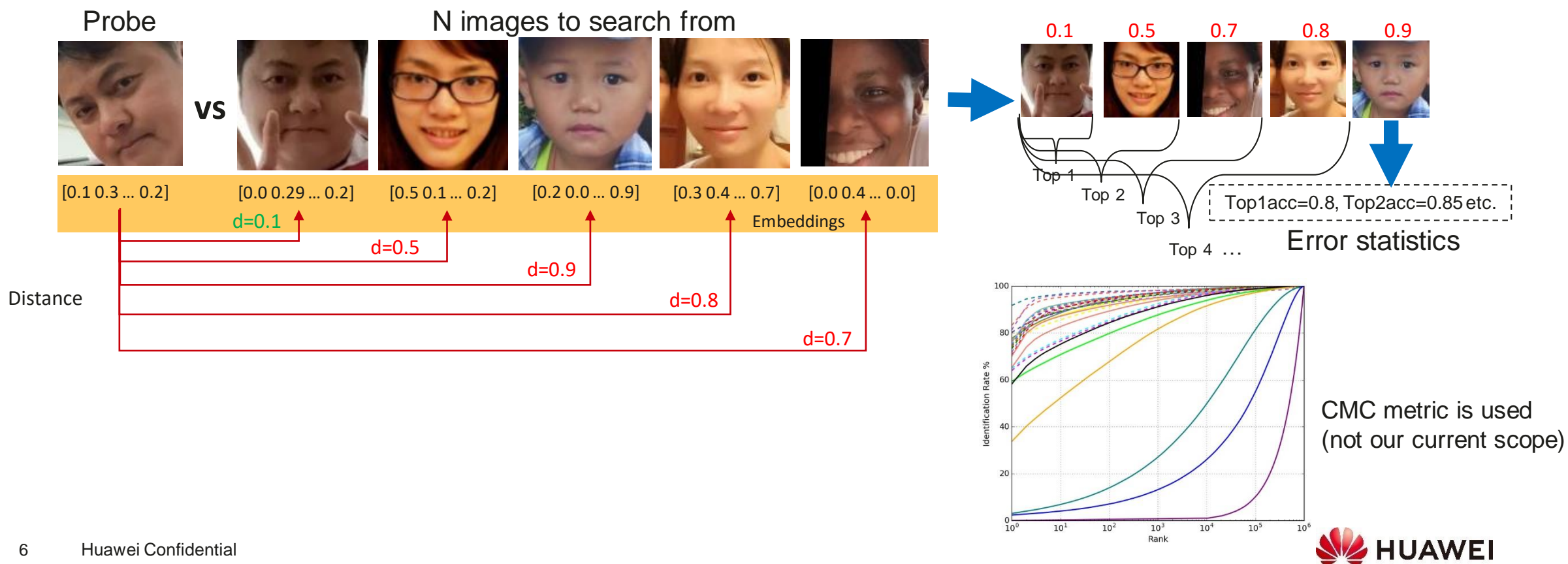
## AUROC:

- 1)  $0 \leq \text{AUROC} \leq 1$
- 2) The more close to 1, the better



# Case Study: Face Identification

**FaceID problem:** design an algorithm to find a face of the given person presented by 1 image among N images





# Metric Learning

- Why **metric learning**?

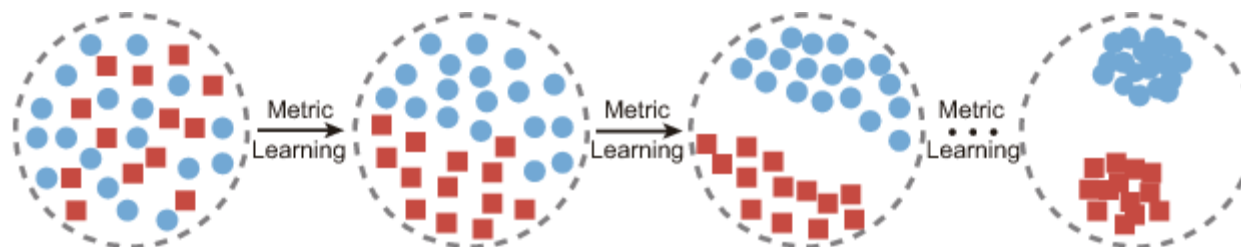
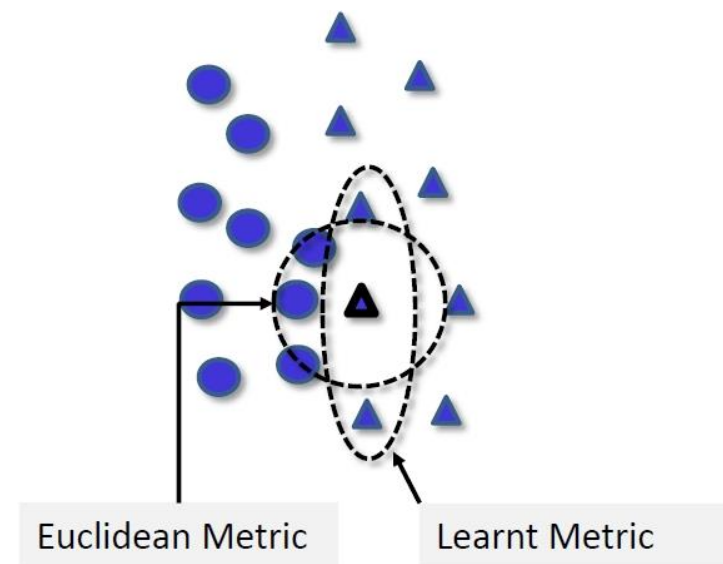
- Sometimes  $L_2$  Euclidian distance is not enough
- No need to retrain the model
- No need to re-dump the embeddings

- What needs to be done?

- Only slightly change the test procedure
- Provide different distance / similarity function
  - Distance function is  $d(x^1, x^2)$ , where  $x^1, x^2$  are input embeddings from  $R^N$
  - The simplest case is  $d(x^1, x^2) = \sum_{k=0}^{N-1} (x_k^1 - x_k^2)^2$

- Main goal:

- Make embeddings from the same person\_id be closer than for different person\_id



## Task description

- **Embedding vector: 512 elements**
- **Train data (if needed):**
  - train\_labels.csv
    - Columns: index, class
    - index: unique counter (0..454713)
    - Class: person\_id (0..10574)
  - train\_set.npy
    - Matrix: 454714 x 512



## Task description

- **Test data:**

- test\_pairs.csv
  - Columns: index1,index2,Id
  - index1,2: rows number from test embedding DB (0..11604)
  - Id: unique counter (0..31443)
- test\_set.npy
  - Matrix: 11605 x 512

# Task statement

- **Task**

- 1) Implement another metric function different from Euclidian one:

$$d(x^1, x^2) \neq \sum_{k=0}^{N-1} (x_k^1 - x_k^2)^2$$

- 2) Dump to csv the distance for every pair listed in test\_pairs.csv (you don't know whether this pair corresponds to the same person or not)
- 3) Register on **kaggle** (if not done already) and submit your solution to it
- 4) Register on **github** (if not done already) and make a pull request with the solutions better than the baseline (Euclidian distance)

## Baseline code (Euclidian metric)

```
import numpy as np

import pandas as pd

pairs = pd.read_csv('test_pairs.csv')

test = np.load('test_set.npy')

distances = []

for i in range(len(pairs)):

    index1 = pairs['index1'][i]

    index2 = pairs['index2'][i]

    dist = np.sum((test[index1]-test[index2])**2)

    distances.append(dist)

baseline = pd.DataFrame({'Predicted':distances})

baseline['Id'] = range(len(baseline))

baseline.to_csv('baseline.csv',index=False)
```

# Challenge notes

- **Important notes**

- Finish time: **6<sup>th</sup> of July, 23:59**
- Public portion: 25% of test
- Private results will be available after “*Finish time*”
- Public/Private portions have the same distribution
- Number of submissions per day: 20 (max for kaggle)
- Scored private submissions: 2
  
- **1-3 places should prepare a couple of slides about their solution and present it on 7<sup>th</sup> of July**
- **Some prizes are expected!**

# Important links

- Kaggle
  - <https://www.kaggle.com/c/metric-learning/overview>
- Github (+ this presentation)
  - <https://github.com/papermsuocode/raai2019hackface>
- Telegram chat
  - <https://t.me/joinchat/B2iAGhBq7EaqwFtmPReIRw>



# Thank you.

Bring digital to every person, home, and organization for a fully connected, intelligent world.

**Copyright©2018 Huawei Technologies Co., Ltd.  
All Rights Reserved.**

The information in this document may contain predictive statements including, without limitation, statements regarding the future financial and operating results, future product portfolio, new technology, etc. There are a number of factors that could cause actual results and developments to differ materially from those expressed or implied in the predictive statements. Therefore, such information is provided for reference purpose only and constitutes neither an offer nor an acceptance. Huawei may change the information at any time without notice.

