# The Third Micro-Expression Grand Challenge (MEGC2020) Guidelines

The goal of this challenge is to spot macro- and micro-expressions interval in long video sequences. For this challenge, we focus on 98 long videos of CAS(ME)<sup>2</sup> database (300 macro-expressions and 57 micro-expressions) and 147 long videos of SAMM Long Videos dataset (343 macro-expressions and 159 micro-expressions). The details of the databases as follow:

# 1 CAS(ME)<sup>2</sup> [2] spotting task

In the part A of CAS(ME)<sup>2</sup> database, there are 22 subjects and 98 long videos. The average duration is 148s. The facial movements are classified as macroand micro-expressions. The video samples may contain multiple macro or micro facial expressions. The onset, apex, offset index for these expressions are given in the excel file. In addition, the eye blinks are labeled with onset and offset time. To download the dataset, please visit:

http://fu.psych.ac.cn/CASME/cas(me)2-en.php

Download and fill in the license agreement form, email to fuxl@psych.ac.cn. (Please use all the video samples in 'rawpic' folder.)

# 2 SAMM Long Videos [1] spotting task

The original SAMM dataset [1] with 159 micro-expressions. In SAMM Long Videos dataset [3], there are 147 videos. The index of onset, apex and offset frames of micro-movements are outlined in the ground truth excel file. The micro-movements interval is from onset frame to offset frame. In this database, all the micro-movements are labeled. Thus, the spotted frames can indicate not only micro-expression but also other facial movements, such as eye blinks. To download the dataset, please visit:

http://www2.docm.mmu.ac.uk/STAFF/M.Yap/dataset.php

Download and fill in the license agreement form, email to M.Yap@mmu.ac.uk with email subject: SAMM long videos.

# References

- [1] Adrian K Davison, Cliff Lansley, Nicholas Costen, Kevin Tan, and Moi Hoon Yap. Samm: A spontaneous micro-facial movement dataset. *IEEE Transactions on Affective Computing*, 9(1):116–129, 2018.
- [2] Fangbing Qu, Su-Jing Wang, Wen-Jing Yan, He Li, Shuhang Wu, and Xi-aolan Fu. Cas (me)<sup>2</sup>: a database for spontaneous macro-expression and micro-expression spotting and recognition. *IEEE Transactions on Affective Computing*, 2017.
- [3] Chuin Hong Yap, Connah Kendrick, and Moi Hoon Yap. Samm long videos: A spontaneous facial micro- and macro-expressions dataset. arXiv preprint arXiv:1911.01519, 2019.

# Result Evaluation Standard

## 1. True positive in one video definition

Supposing there are m micro-expressions in the video, and n intervals are spotted. The result of this spotted interval  $W_{spotted}$  is considered as true positive (TP) if it fits the following condition:

$$\frac{W_{spotted} \cap W_{groundTruth}}{W_{spotted} \cup W_{groundTruth}} \ge k \tag{1}$$

where k is set to 0.5,  $W_{groundTruth}$  represents the micro-expression interval (inset-offset).

If not, the spotted interval is regarded as false positive (FP).

#### 2. Result evaluation in one video

Supposing the number of TP in one video is a ( $a \le m$  and  $a \le n$ ), then FP = n - a, FN = m - a, the values of following metrics could be calculated:

$$Recall = \frac{a}{m}, \ Precision = \frac{a}{n}$$
 (2)

$$F - score = \frac{2TP}{2TP + FP + FN} = \frac{2a}{m+n} \tag{3}$$

Yet, in the real life, there would be following situations for single video:

- The test video do not have micro-expression sequences, thus, m = 0, the denominator of recall would be zeros.
- The spotting method does not spot any intervals. The denominator of precision would be zeros since n = 0.

• If there are two spotting methods, Method<sub>1</sub> spots p intervals and Method<sub>2</sub> spots q intervals, and  $p \leq q$ . Supposing for both methods, the number of true positive is 0, thus the metric (recall, precision or F1-score) values both equal to zeros. However, in fact, the Method<sub>1</sub> performs better than Method<sub>2</sub>.

Considering these real situations, we propose for single video, we just note the result of TP, FP and FN without calculation of other metrics.

#### 3. Evaluation for entire database

Supposing in the entire database, there are V videos and M micro-expression sequences, and the method spot N intervals in total. The database could be considered as one long video, thus, the metrics for entire database can be calculated by following formulas:

$$Recall_D = \frac{\sum_{i=1}^{V} a_i}{\sum_{i=1}^{V} m_i} = \frac{A}{M}$$
 (4)

$$Precision_D = \frac{\sum_{i=1}^{V} a_i}{\sum_{i=1}^{V} n_i} = \frac{A}{N}$$
 (5)

$$F1 - score_D = \frac{2 \times (Recall_D \times Precision_D)}{Recall_D + Precision_D}$$
 (6)

#### Baseline Method and Result

The Baseline method and the result will be available in the middle of December.

## Submission

For the purpose of result verification and to encourage reproducibility and transparency, all entries must submit the following:

- An evaluation log file (.txt, or .csv)indicating the databases, the video id, the ground truth interval range, and the predicted interval range. This is to ensure that all submissions are fairly and correctly evaluated for comparisons.
- A paper highlighting the contribution of the submission, but not limited
  to, the method, experimental results and analysis, prepared according to
  the format stipulated by IEEE FG 2020. For detailed instructions on this,
  please refer to here. All challenge entries should be accompanied by a
  paper submission.

• GitHub repository URL containing codes of your implemented method, and all other relevant files such as feature/parameter data. To help publicize our workshop and domain area, please do mention (or add relevant links on) MEGC Workshop 2020 and FG 2020. You may provide this URL in a simple text file while submitting.

For all files except for the paper, please submit in a single zip file and upload to the submission system as supplementary material.

# Sample log

Header consists of the database labels ('1' for SAMM Long Videos, '2' for CAS(ME)<sup>2</sup>), follow by:

Video\_ID GT\_onset GT\_offset Predicted\_onset Predicted\_offset Result. In one video, results are sorted by Predicted\_onset.

1					
006_1	_	_	1000	1050	FP
006_1	2324	2403	2310	2395	TP
006_1	3912	3988	-	-	FN
006_1	-	-	4500	4575	FP
006_1	5343	5424	5349	5360	FP
006_2	-	-	100	150	FP
006_2	180	274	190	250	FP
2					
15_0101	-	-	100	150	FP
15_0102	699	707	700	710	TP
15_0102	-	-	780	789	FP

The submission portal will be open at MEGC2020 Website Deadline of Challenge: 24 January 2020, 2359 PST (UTC -8)

## Rules

The organizers reserve the right to disqualify submissions with on the basis of

- Challenge results that are likely to be suspicious, i.e. out-of-norm from the distribution of scores from submitters.
- Non-submission of accompanying paper.
- Submission of an accompanying paper that has a substantial overlap with any other paper already submitted or published, or to be submitted during the review period

# For further enquiries, please contact:

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