

# Recognizing Fine Facial Micro-Expressions using Two-Dimensional Landmark Feature

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## Introduction

### Micro-expression recognition in computer vision

- Most of the existing emotion recognition methods are studied only on large facial expression
- In actual situations, there are many cases where the expression does not change significantly even though there is emotion.



General-expression example(SFEM: 1.83)      Micro-expression example(SFEM: 0.20)

- In the image sequence, the amount of information change is small → Recognition difficulty ↑

### Motivation and our approach

- Landmark is not influenced by personal characteristics and environment, and is similar in movement pattern on each emotion
- Convert to 2-D feature of Image type and apply to convolutional neural network (CNN)

## Related Works

### Deep temporal appearance-geometry network(DTAGN)[1]

- A neural network-based algorithm that uses image and landmark simultaneously
- Perform emotion recognition through joint fine-tuning number of each network

### Preprocessing method using video magnification[2]

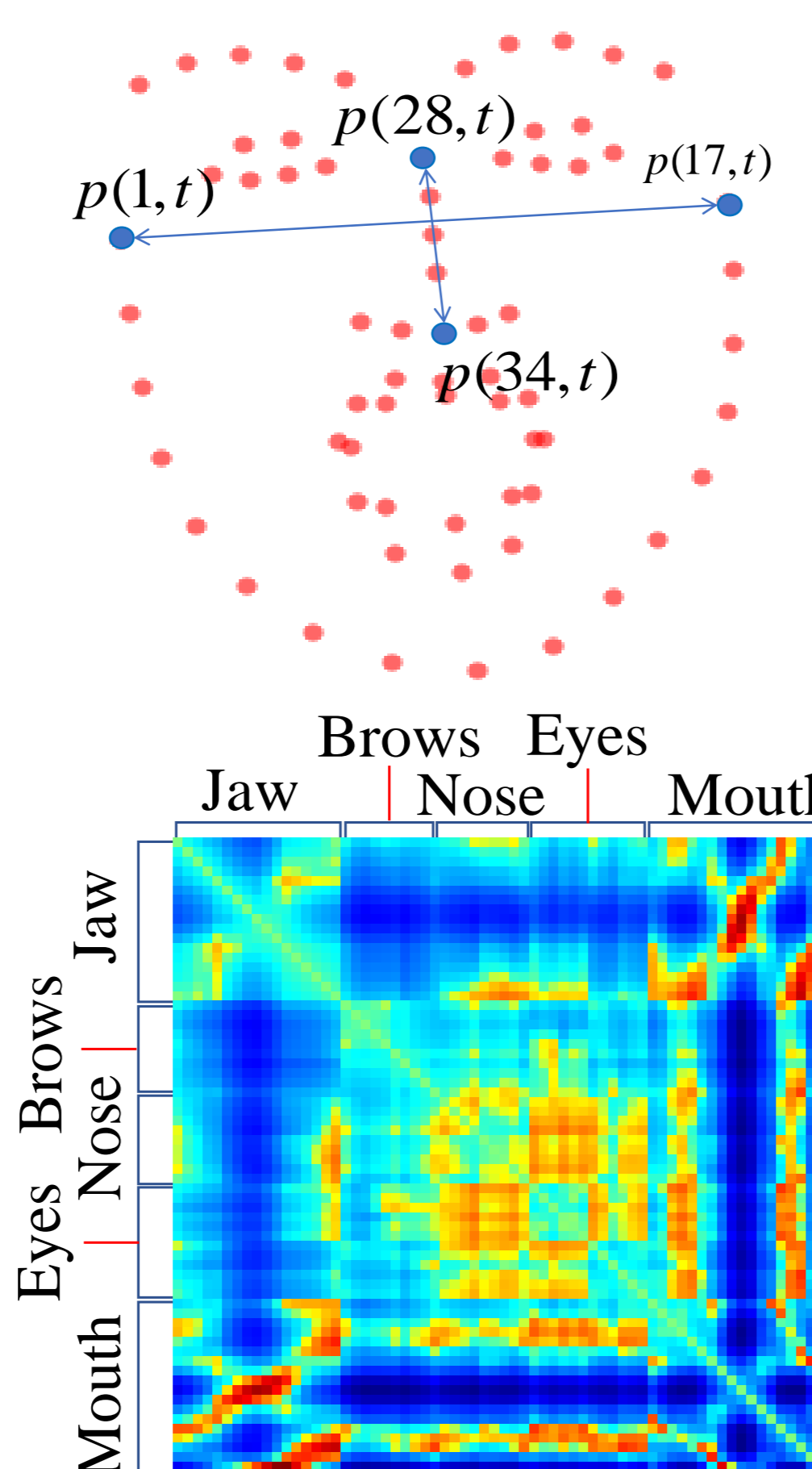
- Apply the existing recognition algorithm after applying the video motion magnification

## 2-D Landmark Feature

### Generation of frame-based landmark feature(LMF)

- The variation of the distance between each landmark is represented by a two-dimensional matrix.
  - Unique pattern can be generated according to emotion
$$LMF_t(i, j) = \|p(i, t) - p(j, t)\|_2 - \|p(i, t-1) - p(j, t-1)\|_2$$

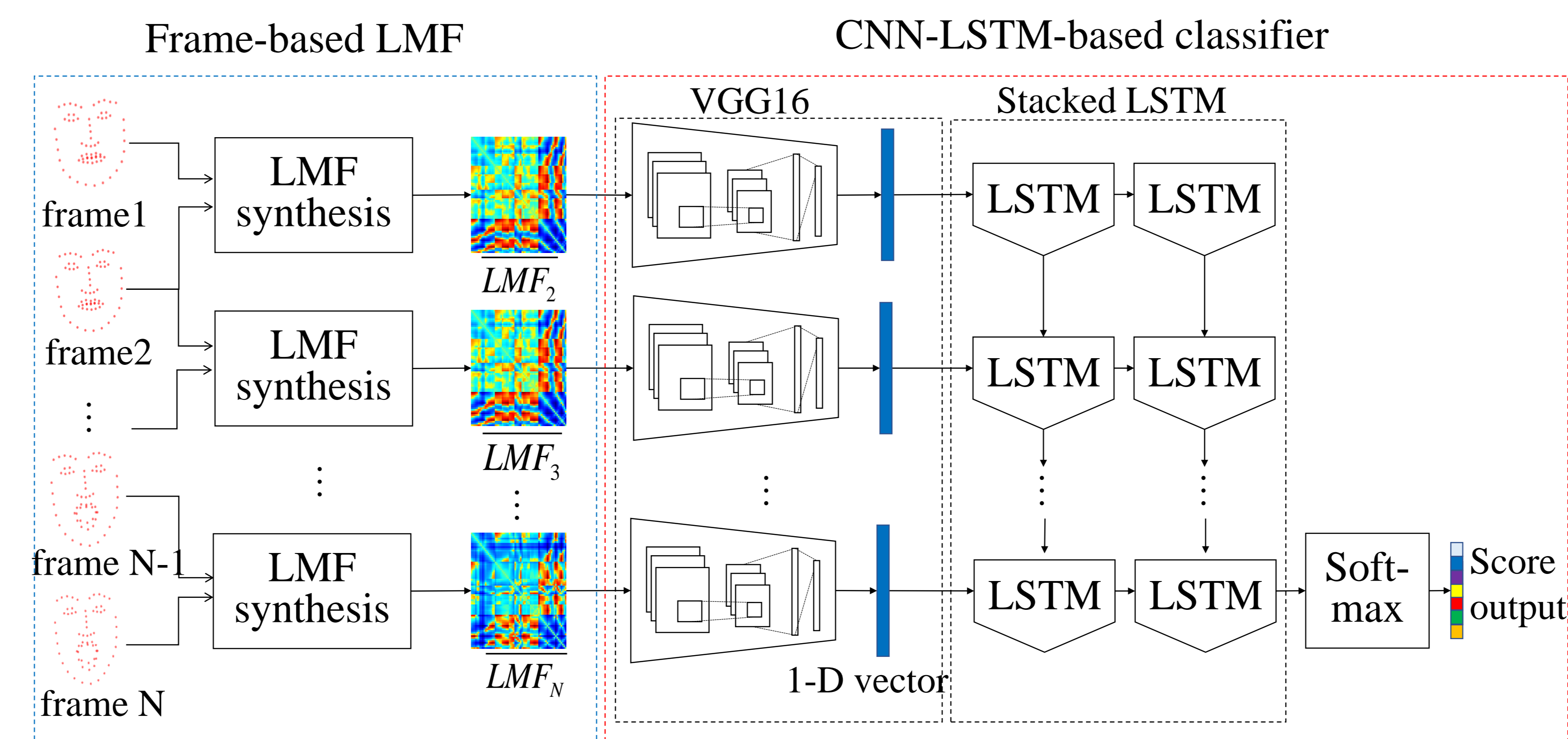
*p*: landmark position
- Perform normalization using the maximum value of distance variation
  - Unique pattern can be generated robust to the size of expression
$$\overline{LMF}_t(i, j) = \begin{cases} 128 \cdot \left\{ \frac{LMF_t(i, j)}{\max(LMF_t)} \right\}^{0.5} + 127 & LMF_t(i, j) > 0 \\ -128 \cdot \left\{ \frac{LMF_t(i, j)}{\max(LMF_t)} \right\}^{0.5} + 127 & LMF_t(i, j) < 0 \\ 127 & else \end{cases}$$



## Landmark Feature-Based Face Expression Recognition algorithm

### CNN-LSTM-based classifier

- Encode 2D feature of each frame through VGG16 based CNN
- Perform classification through stacked LSTM using encoded feature sequence as input

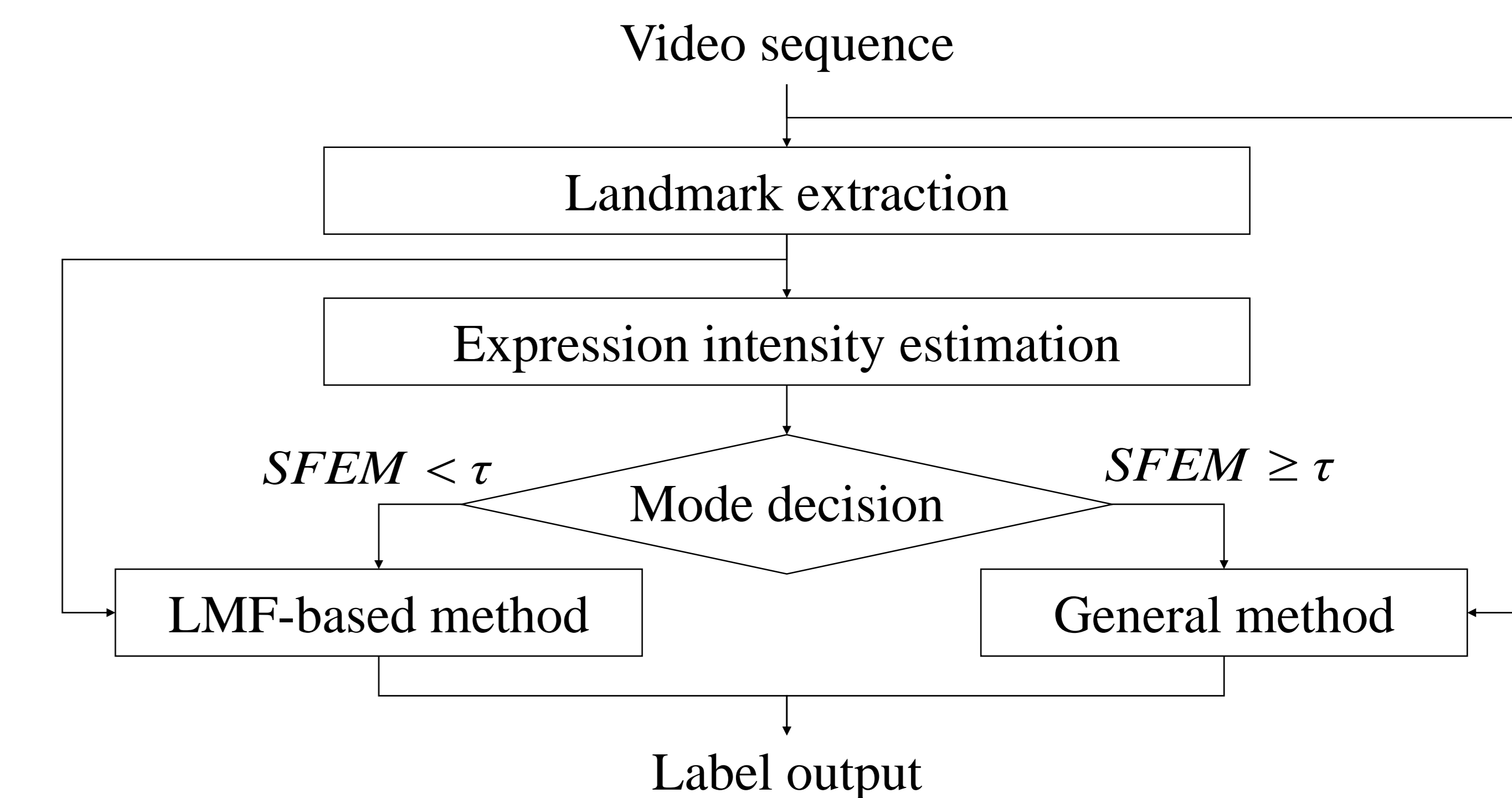


### Jointly FER scheme using image and LMK

- A simple facial expression metric intensity (SFEM): Measure the size of the expression using the sum of the position change of the landmark.
- Algorithm can be selectively processed according to expression size

$$SFEM = \frac{\frac{1}{N} \sum_{t=1}^N \sum_{i=1}^M \sum_{j=1}^M |LMF_t(i, j)|}{\|p(1, t) - p(17, t)\|_2 \cdot \|p(28, t) - p(34, t)\|_2}$$

*N*: # of frames  
*M*: # of landmark



## Experimental Results

### Dataset

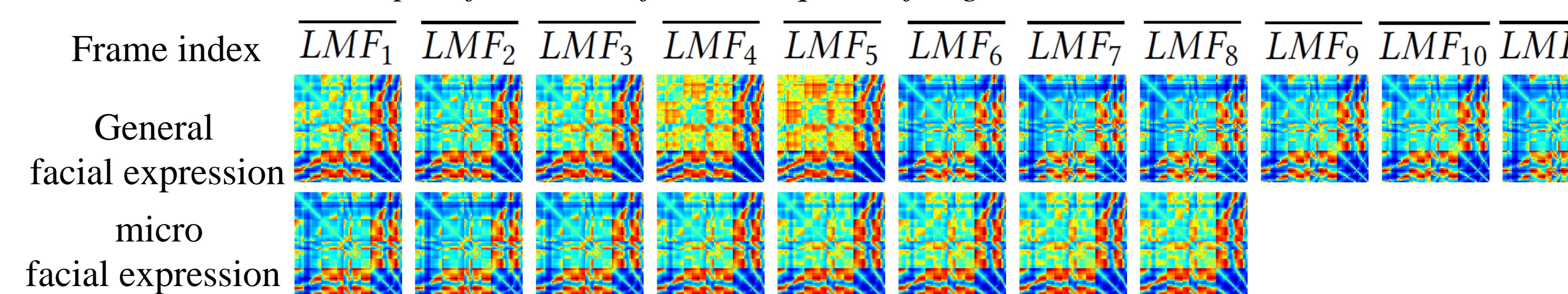
- CK+ dataset(327 sequences), seven emotion categories
- In the case of small motion data, synthesis is performed using only the first three frames of the sequence → Create nine-frame small motion sequence through video frame interpolation[3]
- Landmark information is extracted using activate appearance model(AAM)

### Visual analysis of landmark feature(LMF)

Example of 2D landmark feature based on seven emotions



Example of landmark feature sequence for general case and micro case



### Recognition accuracy

- In training process, used only general expression data
- In test process, used synthesized micro expression data and mixed dataset

	Test dataset[%]		
	General facial expression	Micro facial expression	Mixed
DTAGN[1]	93.88(97.25*)	43.34	70.94
LMF-FER	92.66	<b>77.98</b>	87.46
Joint framework	N/A	N/A	<b>88.69</b>

\* Results stated in [1]

## Conclusion

- Unique pattern can be generated robust to the size of expression
- Through SFEM and joint framework, general and micro motion can be processed

[1] H. Jung et al., "Joint fine-tuning in deep neural networks for facial expression recognition," ICCV2015  
 [2] A. Ngo et al., "Eulerian motion magnification for subtle expression recognition," ICASSP 2016  
 [3] S. Niklaus et al., "Video frame interpolation via adaptive separable convolution," ICCV 2017