# Multi-label Co-regularization for Semi-supervised Facial Action Unit Recognition





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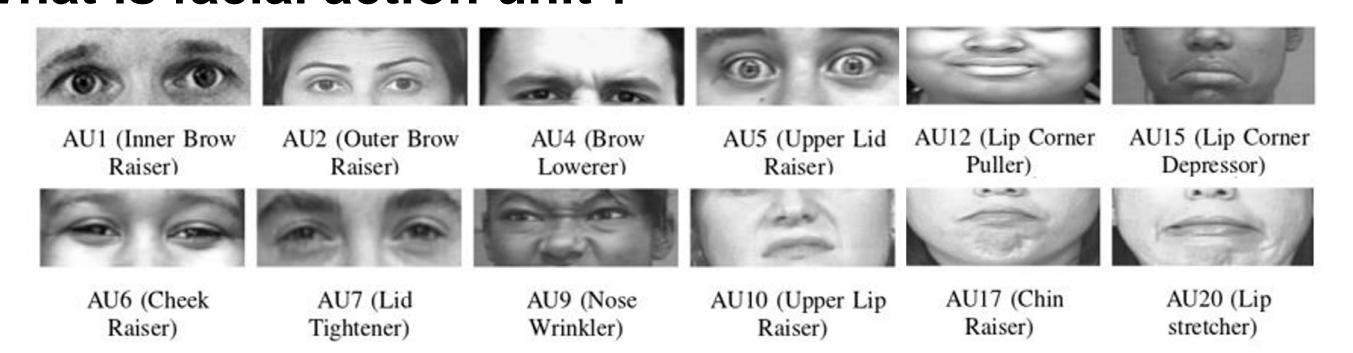
Code:





# Problem

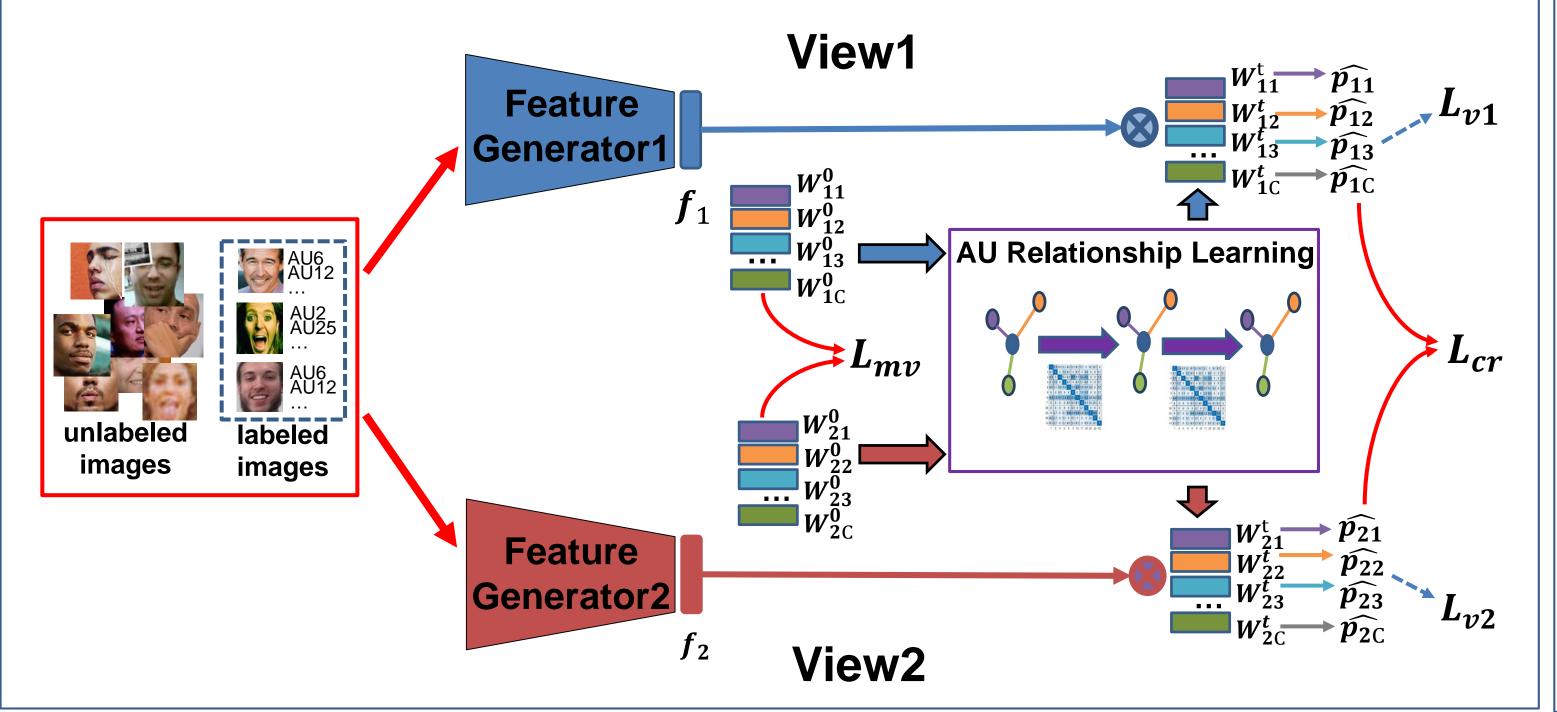
What is facial action unit?



- ➤ Facial action units refer to a set of facial muscle movements coded by their appearance on the face, which can be used for coding nearly any anatomically possible facial expression.
- ➤ Since AUs are subtle, local and have significant subject-dependent variations, qualified FACS experts are required to annotate facial AUs. In addition, labeling AUs is labor-intensive and expensive, making it impractical to manually annotate a large set of face images.

# Overview

• Semi-supervised AU recognition using multi-view co-training and AU relationship modeling to leverage massive labeled face images and a small set of labeled face images.



# **Method Details**

#### Multi-view Loss

$$L_{mv} = \frac{1}{C} \sum_{j=1}^{C} \frac{W_{1j}^{T} W_{2j}}{\|W_{1j}\| \|W_{2j}\|}$$

A multi-view loss by orthogonalizing the weights of the AU classifiers of different views to encourage the two feature generators to get conditional independent features.

### Co-regularization Loss

A co-regularization loss to encourage the classifiers from different views to generate similar predictions in order to make the two views to learn from each other.

$$L_{cr} = \frac{1}{C} \sum_{j=1}^{C} \left( H(\frac{\hat{p}_{1j} + \hat{p}_{2j}}{2}) - \frac{H(\hat{p}_{1j}) + H(\hat{p}_{2j})}{2} \right)$$
$$H(p) = -(p \log p + (1 - p) \log(1 - p))$$

# AU Relationship Learning

Two layer graph convolution layer to model the relationship between different AUs.

$$W_i^t = \hat{A} \ ReLU(\hat{A}W_i^0 H^{(0)})H^{(1)}$$

>Adjacency matrix defined by the dependency matrix.

$$P_{dep} = \frac{1}{2} ([P(L_i = 1 | L_j = 1)]_{C \times C} + [P(L_i = -1 | L_j = -1)]_{C \times C})$$

$$\hat{A} = ABS((P_{dep} - 0.5) \times 2)$$

#### Overall Loss Functions

➤ Binary cross-entropy loss for AU recognition

$$L_{vi} = -\frac{1}{C} \sum_{i=1}^{C} a_c [p_j \log \hat{p}_{ij} + (1 - p_j) \log(1 - \hat{p}_{ij})]$$

**≻Overall loss function** 

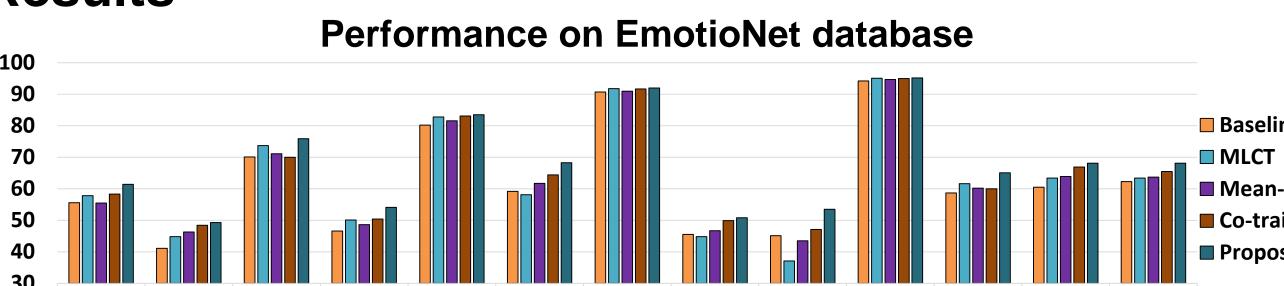
$$L = \frac{1}{2} \sum_{i=1}^{2} L_{vi} + \lambda_{mv} L_{mv} + \lambda_{cr} L_{cr}$$

# **Experimental Results**

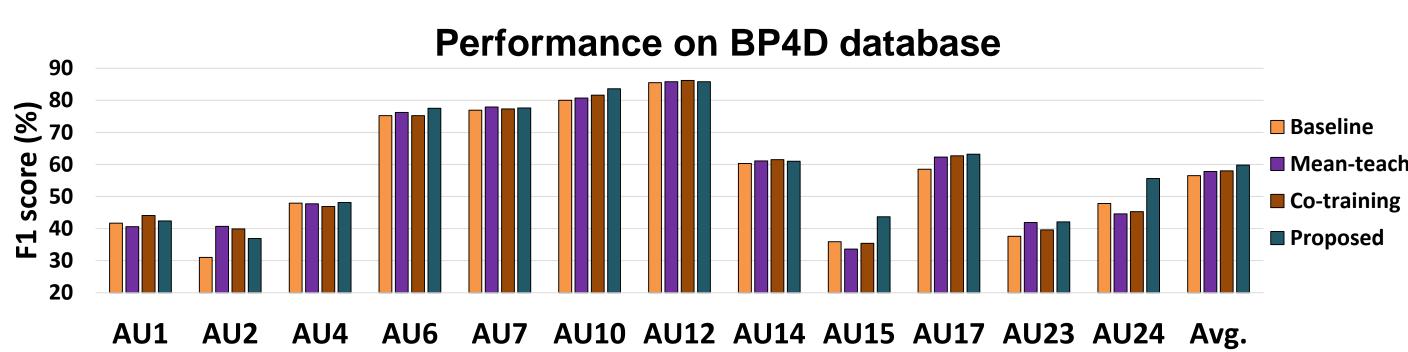
### Databases

	Туре	No. labeled lmgs.	No. labeled lmgs.	No. AU	Protocol
EmotioNet	In-the-wild	20,722	50,000	12	Avg. of three random tests
BP4D	Spontaneous	~140,000	100,000	12	Subject-exclusive three-fold

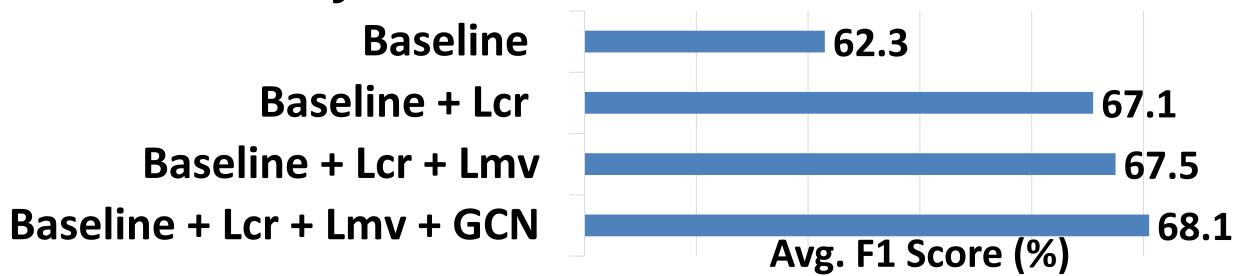
### Results



AU1 AU2 AU4 AU5 AU6 AU9 AU12 AU17 AU20 AU25 AU26 AU43 Avg.



## Ablation Study on EmotioNet



### Extension to Facial Attributes Analysis

