

Generic to Specific Recognition Models for Membership Analysis in Group Videos

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Outline

- Introduction
- The proposed framework
- Experiments and results
- Analysis and discussions
- Conclusions and future work

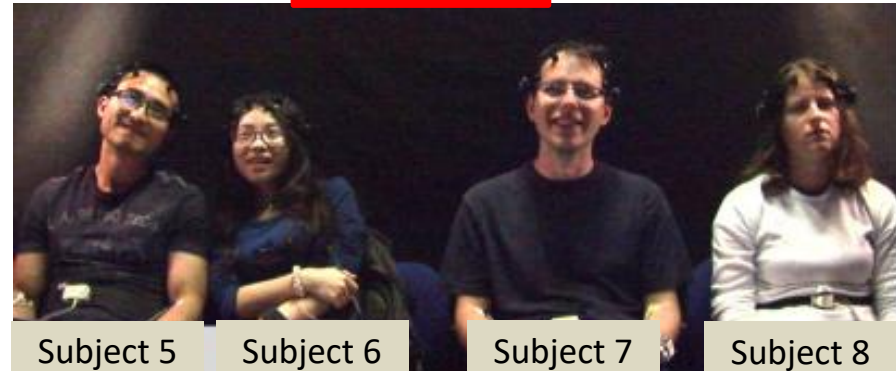
Introduction

- Definitions of group membership recognition
 - Recognize which group each individual is part of

Group 1



Group 2



Group 3



Introduction

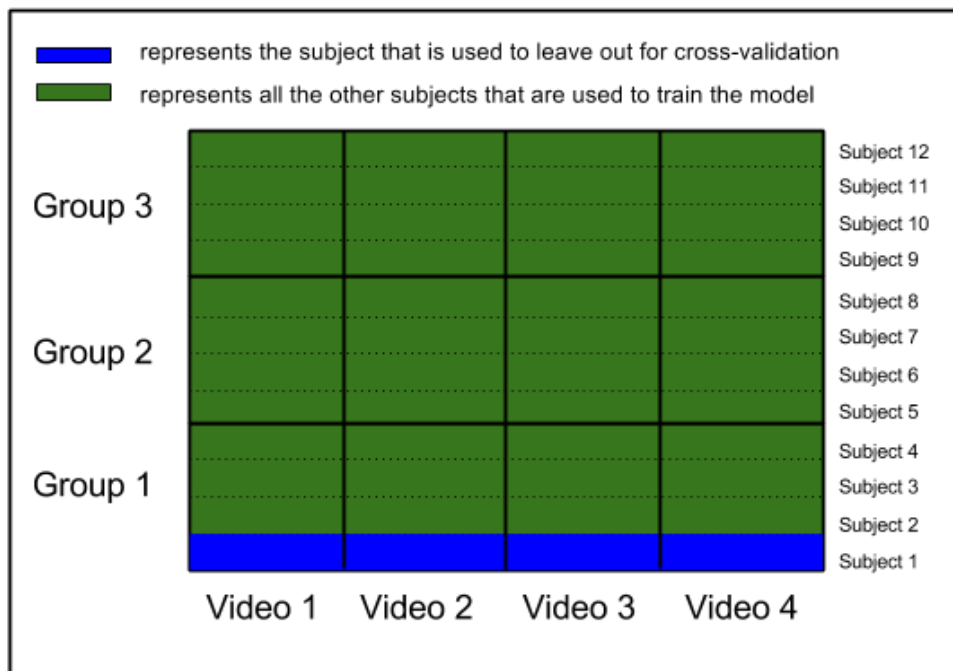
- Motivation
 - People in a group share similar behaviors and emotions [1]
 - Body features work better than the face features in predicting group membership [2]

[1] S. G. Barsade, "The ripple effect: Emotional contagion and its influence on group behavior," *Administrative Science Quarterly*, 2002.

[2] W. Mou, H. Gunes, and I. Patras, "Automatic recognition of emotions and membership in group videos," in *CVPRW*, 2016.

The Proposed Framework

- Group membership recognition is modelled as a classification problem



$$X \xrightarrow{f} Y$$

X : Body behaviors

Y : Group membership
(Group 1 ? 2 ? 3 ?)

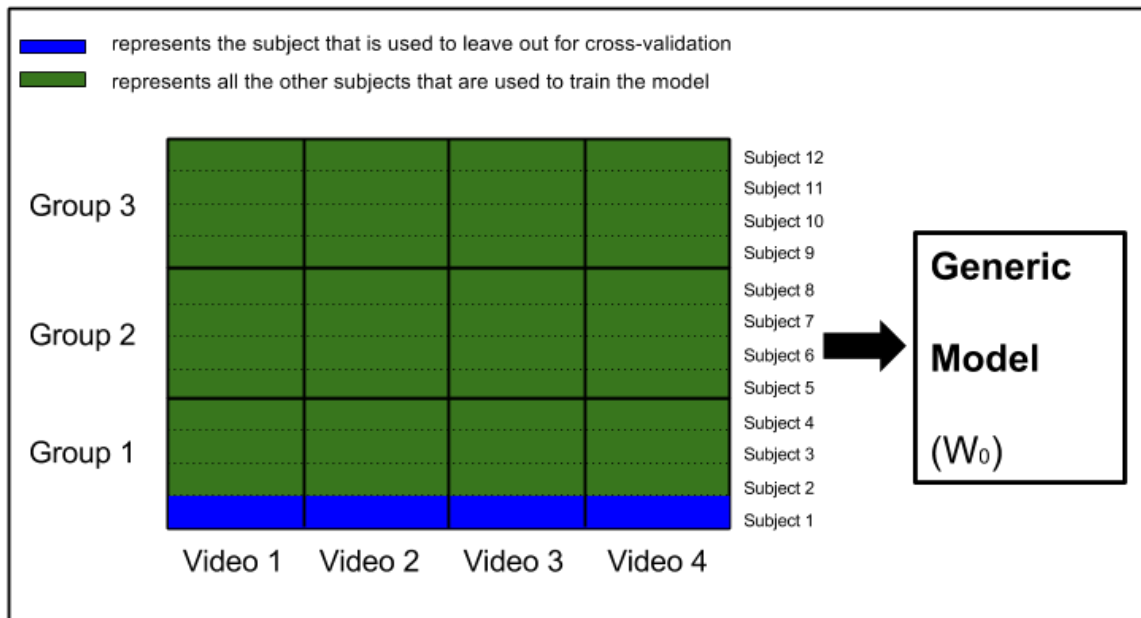
f : Linear-SVM

12 subjects from 3 groups watching 4 videos

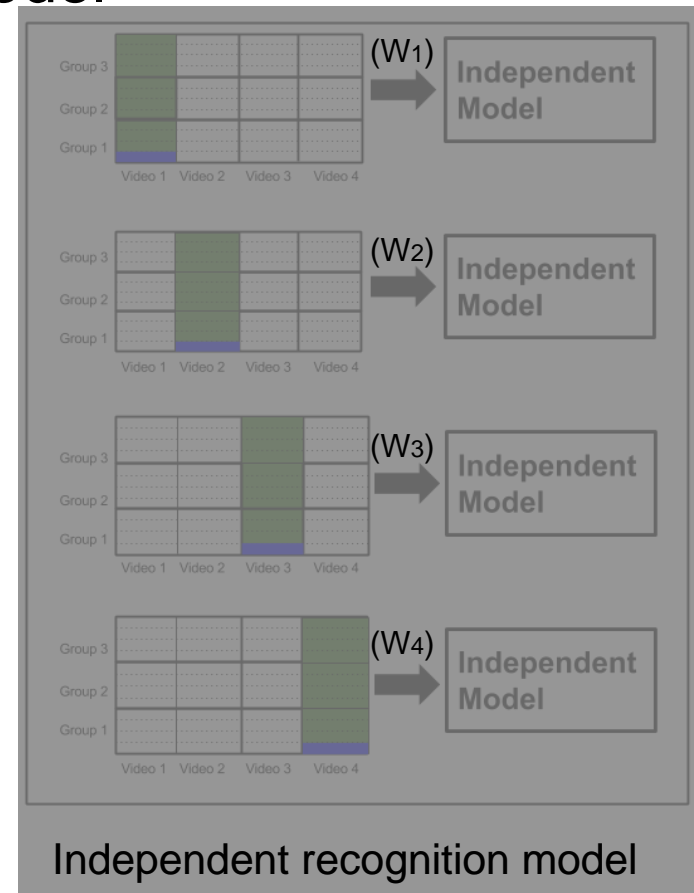
Leave one subject out (blue) cross-validation

The Proposed Framework

- Generic model and independent model



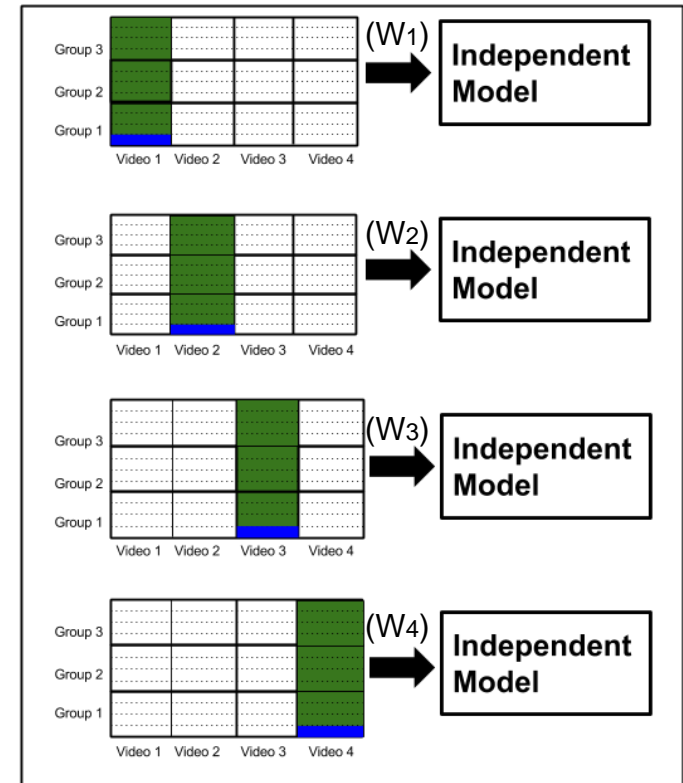
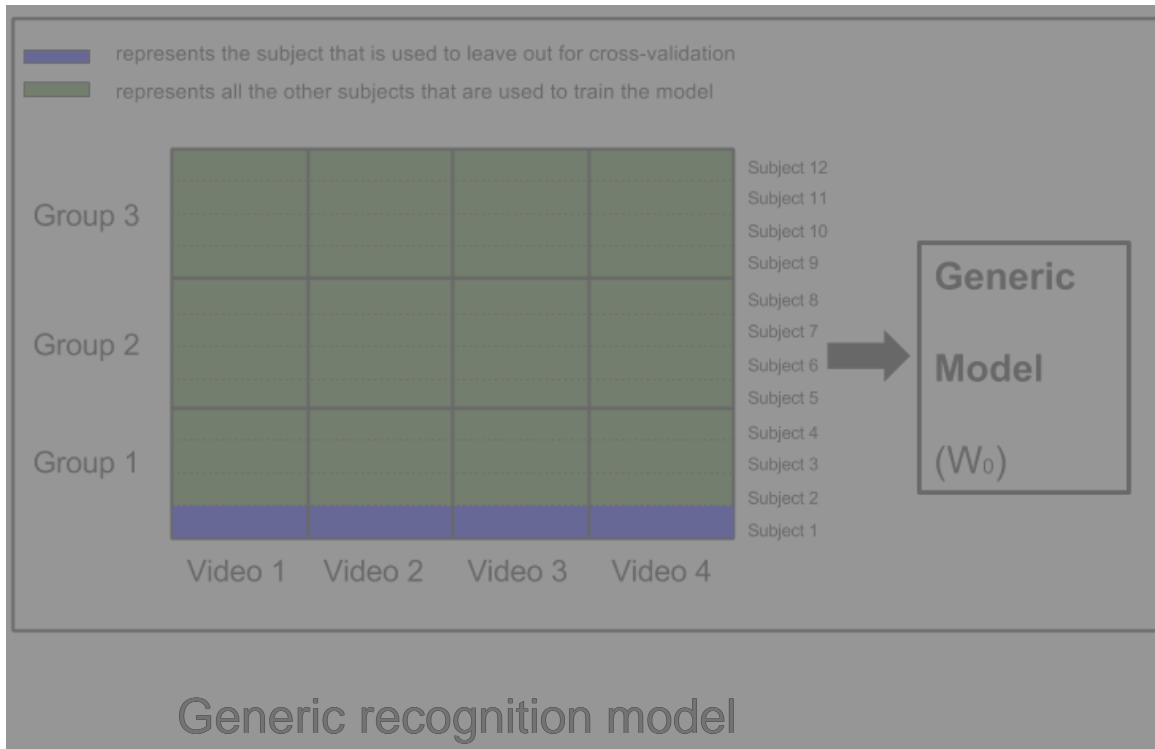
Generic recognition model



Independent recognition model

The Proposed Framework

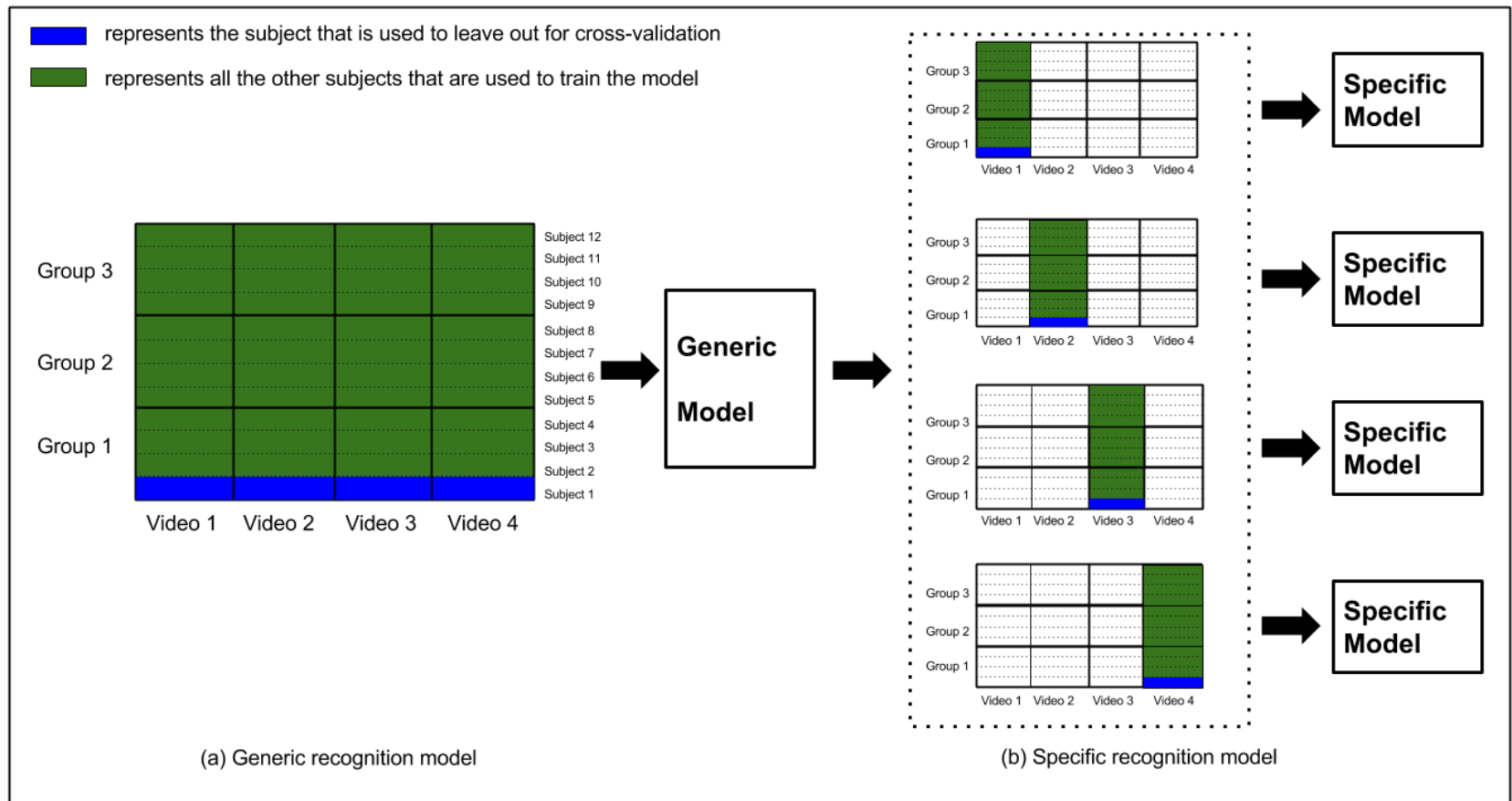
- Generic model and independent model



Independent recognition model

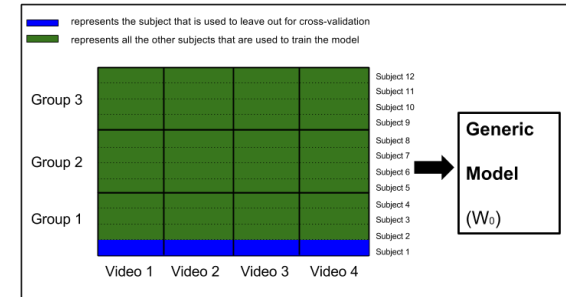
The Proposed Framework

- The proposed two-phase learning model



The Proposed Framework

- Generic model



$$P_{generic}: \min_{w_0, b_0} \frac{\lambda}{2} \|w_0\|^2 + \frac{1}{l} \sum_{i=1}^l L(w_0, b_0; (x_i, z_i))$$

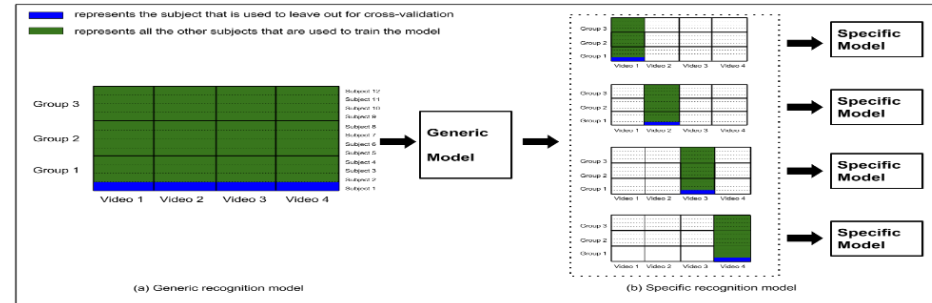
λ : the regularization parameter

w_0, b_0 : the optimization parameters

L : the loss function and is given by the hinge-loss

The Proposed Framework

- Specific model



$$P_{specific}: \min_{w,b} \frac{\mu}{2} \|w\|^2 + \frac{\nu}{2} \|w - w_0\|^2 + \frac{1}{|X_t|} \sum L(w, b; (x_i, z_i))$$

X_t : a subset of the original training set

μ and ν : regularization parameters

$\frac{\nu}{2} \|w - w_0\|^2$: is used to bias w to be close to w_0

The Proposed Framework

- Optimization: stochastic gradient descent (SGD)

Objective function of $P_{specific}$:

$$J(w, b) = \frac{\mu}{2} \|w\|^2 + \frac{\nu}{2} \|w - w_0\|^2 + \frac{1}{k} \sum_{(x_i, z_i) \in X_t} L(w, b; (x_i, z_i))$$

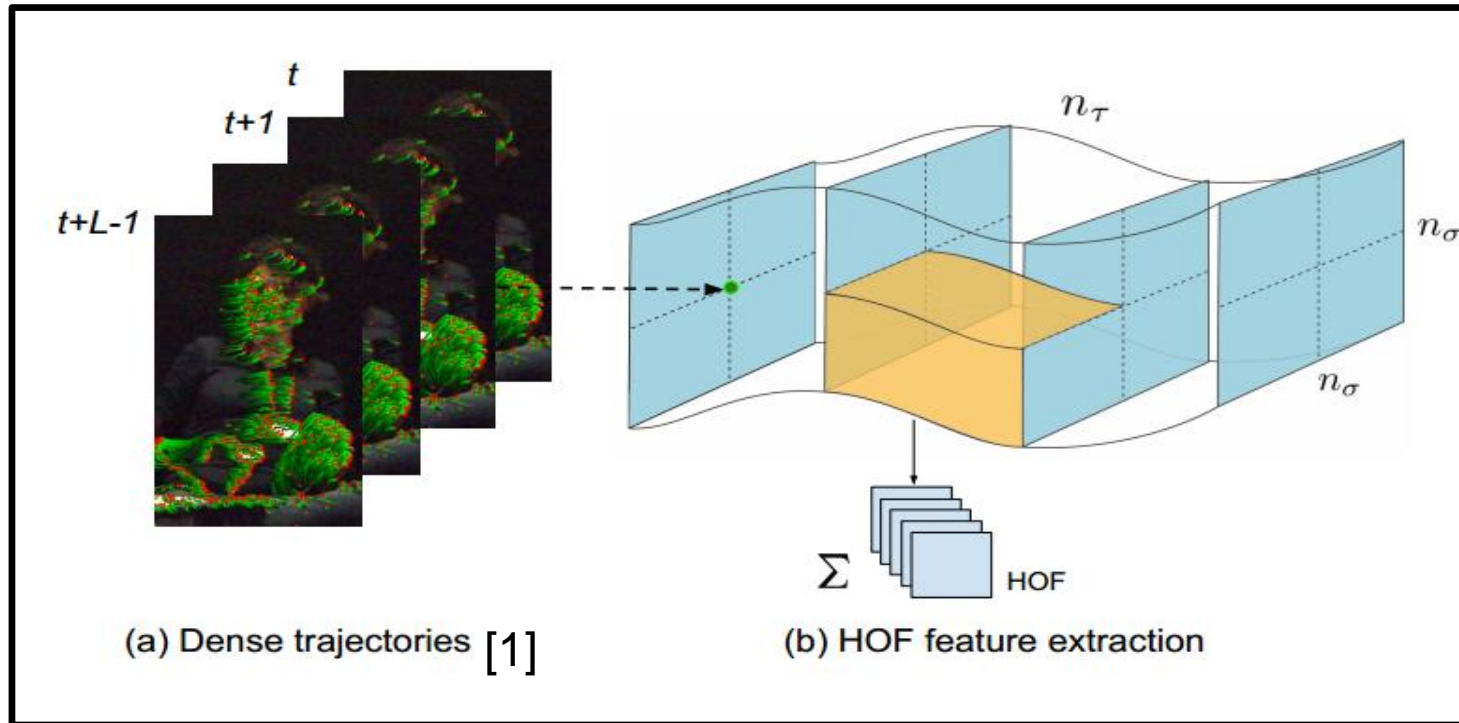
$$\text{Update } w \text{ and } b : w^{(t+1)} = w^t - \frac{\delta_t}{k} \frac{\partial J}{\partial w} \quad b^{(t+1)} = b^t - \frac{\delta_t}{k} \frac{\partial J}{\partial b}$$

The first derivatives of J with respect to w and b :

$$\frac{\partial J}{\partial w} = \mu w - \nu(w - w_0) + \sum_{(x_i, z_i) \in X_t} \frac{\partial J}{\partial w} \quad \frac{\partial J}{\partial b} = \frac{1}{k} \sum_{(x_i, z_i) \in X_t} \frac{\partial J}{\partial b}$$

The Proposed Framework

- Feature extraction



[1] H. Wang et al, "Dense trajectories and motion boundary descriptors for action recognition," IJCV, 2013.

Experiments and Results

- Experiment setup
 - Data: 3 groups of data, 12 participants
 - Feature: Body HOF feature
 - Classifier: linear-SVM
- Experimental results

Different Models	Average Accuracy (p-value)
Specific recognition model	52% (p<0.01)
Generic recognition model	35% (p=0.41)
Independent recognition model	33% (p=0.42)
Chance level	33%

Conclusions and Future Work

- Conclusions
 - The proposed two-phase learning model outperforms the other models
 - Group membership recognition results also indicates that individuals influence each others behaviours within a group and their nonverbal behaviours share commonalities

Conclusions and Future Work

- Future work
 - Testing with a larger database
 - Applying the two-phase learning model to other applications, such as affect recognition

Thank you!