

Action Recognition In RGB-D Egocentric Videos Yansong Tang*, Yi Tian, Jiwen Lu, Jianjiang Feng and Jie Zhou

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1. OVERVIEW

Motivation

- Egocentric videos: self-generated and embodied, providing richer semantic cues, growing dramatically in recent years.
- RGB-D videos: containing both appearance information and 3D structure of the scenes.
- Exploring the complementary information in egocentric RGB-D videos.

Main Contributions

- An RGB-D egocentric action dataset with diversity and scale.
- A tri-stream convolutional network (TCNet) to take advantage of both the RGB and depth inputs.

3. APPROACH



- RGB images, optical flows and depth images are sent into the appearance stream, motion stream and depth stream respectively.
- We finally make decision level fusion to predict the action label.

Project Page: http://ivg.au.tsinghua.edu.cn/dataset/THU_READ.php *My Website: http://ivg.au.tsinghua.edu.cn/people/Yansong_Tang









Data collection

- Mounting the RGB-D sensor on a helmet, which was placed on the subject's head.
- Keeping the camera in the same direction with the subject's eyesight.

4. EXPERIMENTAL RESULTS

The confusion matrix of the TCNet on our dataset

cut_paper draw_paper fetch_water insert_tube knock_door lift_weights manicure open_door open_drawei open_laptop open_umbrella push_button read_book queeze_toothpaste sweep_floor tear_paper hrow_paperplane tie_shoelaces twist_towel use_chopstick use_mobilephone use_mouse use_stapler wash_fruit wash_hand water_plant wave_hand wear_glove wear_watch write



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2. PROPOSED DATASET



Dataset	Task	Camera	Frames
GTEA[16]	Action	RGB	31,253
GTEA gaze[15]	Action	RGB	
GTEA gaze+[15]	Action	RGB	
UCI ADL[20]	Activity	RGB	93,293
WCVS[19]	Activity	RGB-D	
GUN-71 [21]	Grasp Understanding	RGB-D	12,000
Ours	Action	RGB-D	343,626

Our dataset shares the advantages on modality, scale and diversity compared with related databases.

Method Accuracy dimensional encoding (RGB) 46.9% dimensional encoding (Depth) 52.6% ial stream ConvNet 68.4% 40.9% oral stream ConvNet th stream ConvNet 52.7% ream ConvNet (RGB) 73.3% Net (RGB & Depth) 76.5%

We test some hand-crafted features, deep learning methods and TCNet on our dataset. The result shows the complementary information between different modalities, it also demonstrates the efficiency of our TCNet model in comparisons with the state-

5. FUTURE WORKS

- We will explore to share more semantic information between the RGB and depth modalities, e.g., hand mask and head motion) for action recognition.
- It is desirable to perform more visual tasks like hand-segmentation and human-object interaction on our dataset.

Reference

- Simonyan K, Zisserman A. Two-Stream **Convolutional Networks for Action Recognition** in Videos. In NIPS2014, pp .568-576.
- Wang H, Schmid C. Action Recognition with Improved Trajectories. In ICCV2013, pp. 3551-3558.















