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## Abstract:

Visual features are vitally important for action recognition in videos. However, traditional features fail to effectively recognize actions for two reasons: on one hand, spatial features are not powerful enough to capture appearance information of complex video actions; on the other hand, important temporal details are always ignored when pooling and encoding. In this paper, we present a new architecture that fuses multiple augmented spatio-temporal features. In order to strengthen spatial features, we conduct crop and horizontal flip on original frame images. Then we feed these processed images into deep Two-Stream network to produce robust spatial representations. To get powerful temporal features, we employ fourier temporal pyramid (FTP) to capture three different levels of video context, including short-term level, medium-range level, and global-range level. At last, we fuse these augmented spatio-temporal features using canonical correlation analysis (CCA) method, which is capable to capture the correlation between these features. Experimental results on UCF101 dataset show that our method can achieve excellent performance for action recognition.

## Conclusions

In this paper, we propose to fuse multiple augmented spatio-temporal features for better action recognition. The enhanced spatial features are extracted by feeding multiple crop images into VGG networks. Then through a three-level FTP, the features are capable to capture different level temporal context information about action. Finally, the method is capable to improve the performance effectively by CCA fusion. Our experimental results show that the model can achieve comparable accuracy to the state of the art methods.

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# MOL2NET, 2016, Vol. 2, J; http://sciforum.net/conference/MOL2NET-02/SUIWML-01

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### Acknowledgements

This work was partially supported by National Natural Science Foundation of China (NSFC Grant No. 61272258, 61170124, 6130129, 61272005), Provincial Natural Science Foundation of Jiangsu (Grant No. BK20151254, BK20151260), Key Laboratory of Symbolic Computation and Knowledge Engineering of Ministry of Education, Jilin University (Grant No. 93K172016K08), and Collaborative Innovation Center of Novel Software Technology and Industrialization.