

College of Computing **School of Interactive Computing**

Motivation

Many real-world applications can benefit from action representation that is robust to variations in viewpoint, scaling, and appearance (clothing, lighting, etc.).



Intuition

Johansson's moving light display \rightarrow Dynamics of an action provide a powerful cue for discrimination.



Decompose an action into movement primitives (roughly corresponds to body parts) and encode the temporal dynamics of each primitives.



Contributions

- The movement pattern histogram (MPH), a multi-channel temporal distribution of movement primitives.
- Novel optimization approach to simultaneously match and align MPH across videos.
- Stereotypy dataset: annotated video corpus of children with autism in a classroom setting.

MOVEMENT PATTERN HISTOGRAM FOR ACTION RECOGNITION AND RETRIEVAL

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Given a single example video clip containing an action of interest, retrieve all matching instances of that action from an unstructured video collection.

Stereotypy Dataset

- Real-world recording of children with autism in a classroom setting.
- A collection of 10-20 minute classroom sessions with total length of 2 hours.
- 3 behaviors of interest annotated by psychologist with autism expertise.



Action Recognition

IXMAS (11 actions, 5 viewpoints, 1650 total clips)

Method	Test View					
	c0	c1	c2	c3	c4	Avg.
MPH	83.9	81.8	87.6	83.0	73.6	82.0
[1]	66.4	73.5	71.0	75.4	66.4	70.5
[2]	62.0	65.5	64.5	69.5	57.9	63.9
[3]	77.0	78.8	80.0	73.9	63.6	74.6

UCF50 (50 actions from Youtube, 6618 total clips)

MPH + FV of Den DenseTrajectories MBH + scen

GIST3D

M

References: [1] Zhang, Z., Wang, C., Xiao, B., Zhou, W., Liu, S., Shi, C.: Cross-view action recognition via a continuous virtual path. CVPR (2013). [2] Li, R., Zickler, T.: Discriminative Virtual Views for Cross-View Action Recognition. CVPR (2012). [3] Junejo, I., Dexter, E., Laptev, I., P'erez, P.: View-Independent Action Recognition from Temporal Self-Similarities. PAMI (2010).

[4] Wang, H., Kl["]aser, A., Schmid, C., Liu, C.L.: Dense trajectories and motion boundary descriptors for action recognition. IJCV (2013). [5] Reddy, K.K., Shah, M.: Recognizing 50 human action categories of web videos. Machine Vision and Applications (2012) [6] Solmaz, B., Assari, S.M., Shah, M.: Classifying web videos using a global video descriptor. Machine Vision and Applications (2012)

[7] Kliper-Gross, O., Gurovich, Y., Hassner, T., Wolf, L.: Motion interchange patterns for action recognition in unconstrained videos. ECCV (2012)

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Results **Action Retrieval**





thod	Accuracy (LoGo)
nseTrajectories [4]	90.5
[4] w/ FV encoding	88.9
e context [5]	76.9
+ STIP [6]	73.7
P [7]	72.7