Progressive Teacher-student Learning for Early Action Prediction

Xionghui Wang, Jian-Fang Hu*, Jianhuang Lai, Jianguo Zhang, and Wei-Shi Zheng

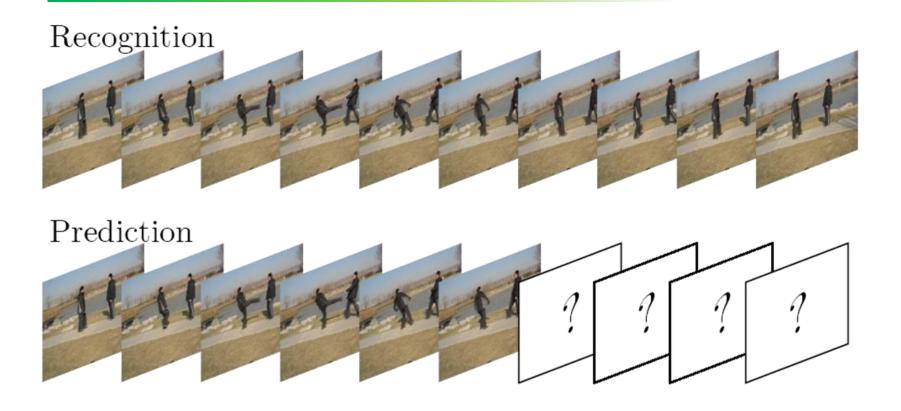
http://isee.sysu.edu.cn/~hujianfang/

SUN YAT-SEN University (中山大学)

OUTLINE

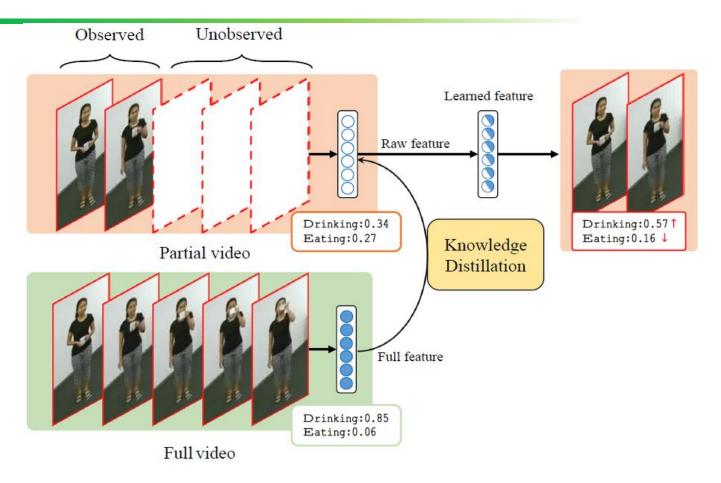
- 1 Introduction & Motivation
- 2 Feature Extraction
- 3 Teacher Student Learning
- 4 Summary

1. Introduction & Motivation



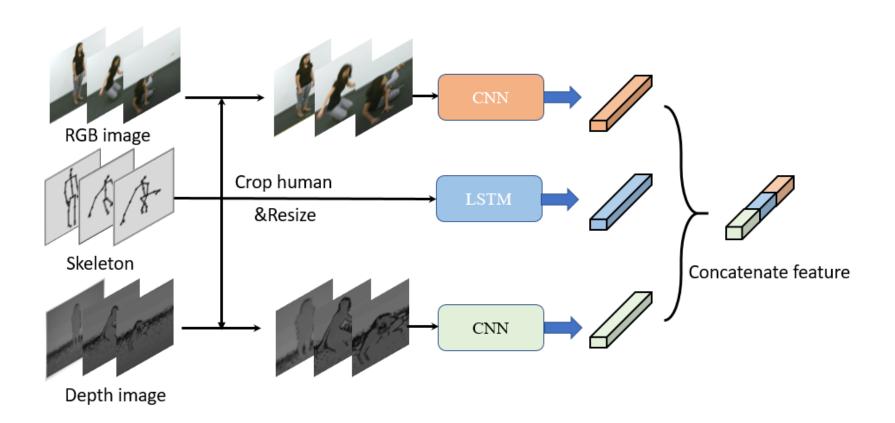
- Recognition: actions are fully executed and captured, after-the-fact analysis.
- Prediction: recognizing actions before they are fully executed, prefact analysis.

1. Introduction & Motivation



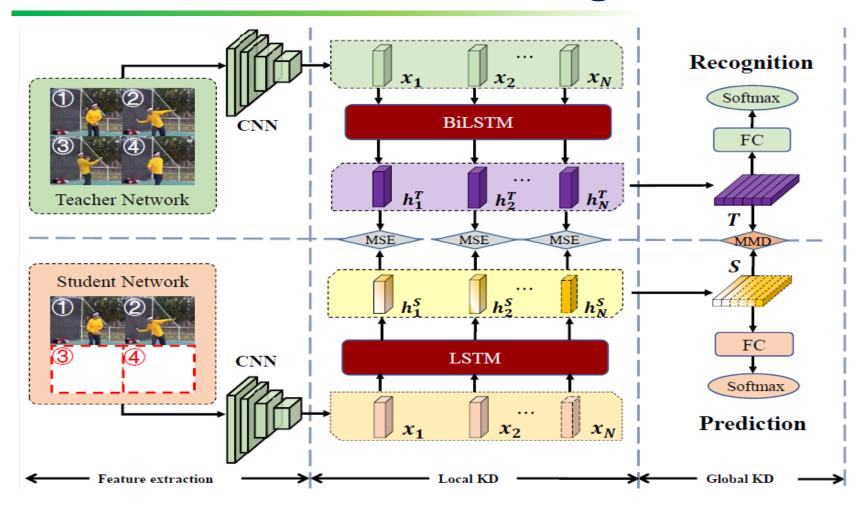
- Full videos contain more action information than partial videos.
- Knowledge distilled from full videos is beneficial for the action prediction with partially observed action videos.

2. Feature Extraction



- Extracting multimodal features (16-channel CNN and LSTM)
- Concatenating multimodal features along the modality dimension
- More features often means more useful action information

3. Teacher Student Learning



- Knowledge distillation cross two different video tasks
- Distilling both local and global knowledge for obtaining better prediction results

3. Teacher Student Learning

Local progressive-wise knowledge:

$$L_{MSE}(S_i, T_i) = ||S_i \odot \omega - T_i \odot \omega||_F^2$$

Mean Square Error loss: Each individual progress level

Global distribution knowledge:

$$L_{MMD}(S_i, T_i) = \left\| \frac{1}{N} \sum_{i=1}^{n} \phi(S_i(:, n) - \frac{1}{N} \sum_{i=1}^{n} \phi(T_i(:, n)) \right\|^2$$

Maximum Mean Discrepancy loss: ALL the progress levels

It can be simplified as:

$$L_{MMD}(S_i, T_i) = \left\| S_i S_i^{\mathsf{T}} - T_i T_i^{\mathsf{T}} \right\|_2^2$$

When ϕ is the second order polynomial kernel function:

$$k(x,y) = \langle \phi(x), \phi(y) \rangle = (x^{T}y)^{2}$$

NTU RGB+D Dataset:

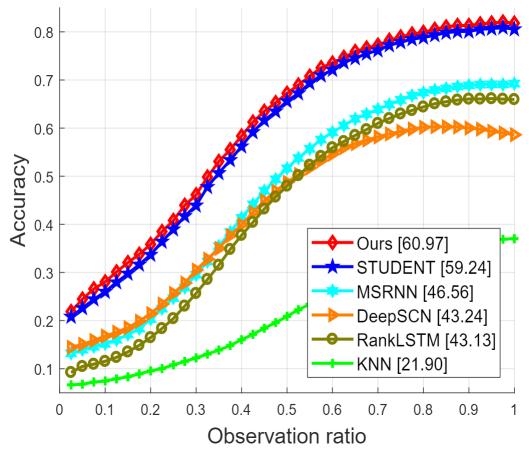
- 60 action classes, >56K action sequences, 3 camera views
- 40 progress levels
- RGB, depth, and skeleton modalities

Observation ratio	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%	AUC
KNN [16]	7.45	9.56	12.25	16.04	20.89	25.97	30.85	34.49	36.15	37.02	21.90
RankLSTM [28]	11.54	16.48	25.66	37.74	47.96	55.94	60.99	64.41	66.05	65.95	43.13
DeepSCN [24]	16.80	21.46	30.51	39.93	48.73	54.61	58.18	60.18	60.01	58.62	43.24
MSRNN [16]	15.17	20.33	29.53	41.37	51.64	59.15	63.91	67.38	68.89	69.24	46.56
STUDENT	25.99	33.68	43.91	56.20	65.59	72.12	76.16	78.82	80.09	80.53	59.24
Ours	27.80	35.85	46.27	58.45	67.40	73.86	77.63	80.06	81.47	82.01	60.97

NTU RGB+D Dataset:

- 60 action classes, >56K action sequences, 3 camera views
- 40 progress levels
- RGB, depth, and ske

Observation ratio	10%	20%	30%
KNN [16]	7.45	9.56	12.25
RankLSTM [28]	11.54	16.48	25.66
DeepSCN [24]	16.80	21.46	30.51
MSRNN [16]	15.17	20.33	29.53
STUDENT	25.99	33.68	43.91
Ours	27.80	35.85	46.27



SYSU 3D HOI set:

- 12 action classes (6 pairs of HOI), 480 action sequences
- RGB, depth, and skeleton modalities
- 40 progress level

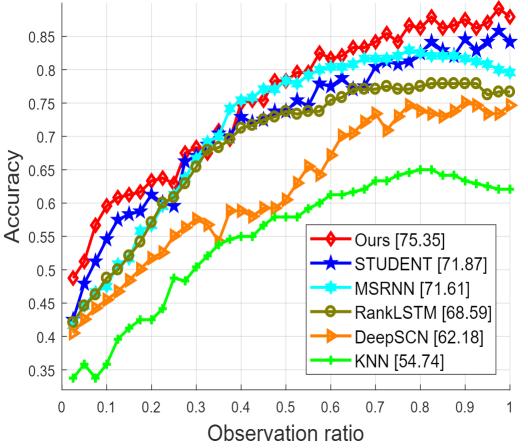
Observation ratio	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%	AUC
KNN [16]	35.83	42.50	50.42	55.00	57.92	61.25	63.33	65.00	63.33	62.08	54.74
RankLSTM [28]	48.75	57.08	65.42	71.25	73.75	75.42	77.08	77.50	77.92	76.67	68.59
DeepSCN [24]	45.50	51.75	57.58	58.83	60.50	67.17	73.42	73.83	75.08	74.67	62.18
MSRNN [16]	47.50	56.67	66.67	75.42	78.33	80.42	81.67	82.50	81.67	79.58	71.61
STUDENT	54.58	61.25	67.08	72.92	73.75	77.50	80.42	82.50	84.58	84.17	71.87
Ours	59.58	63.33	68.33	75.00	78.33	81.67	84.17	86.25	87.50	87.92	75.35

SYSU 3D HOI set:

- 12 action classes (6 pairs of HOI), 480 action sequences
- RGB, depth, and skeleton modalities

40 progress level

Observation ratio	10%	20%	30%	4(
KNN [16]	35.83	42.50	50.42	55
RankLSTM [28]	48.75	57.08	65.42	71
DeepSCN [24]	45.50	51.75	57.58	58
MSRNN [16]	47.50	56.67	66.67	75
STUDENT	54.58	61.25	67.08	72
Ours	59.58	63.33	68.33	75



■ UCF-101 set:

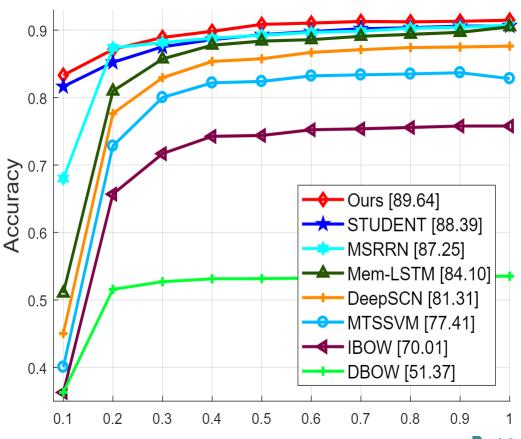
- 101 action classes, >13K action sequences
- Unconstrained video set (RGB)
- 10 progress levels

Observation ratio	10%	20%	30%	40%	50%	60%	70%	80%	90%	100%	AUC
DBOW [32]	36.29	51.57	52.71	53.13	53.16	53.24	53.24	53.34	53.45	53.53	51.37
IBOW [32]	36.29	65.69	71.69	74.25	74.39	75.23	75.36	75.57	75.79	75.79	70.01
MTSSVM [23]	40.05	72.83	80.02	82.18	82.39	83.21	83.37	83.51	83.69	82.82	77.41
DeepSCN [24]	45.02	77.64	82.95	85.36	85.75	86.70	87.10	87.42	87.50	87.63	81.31
Mem-LSTM [22]	51.02	80.97	85.73	87.76	88.37	88.58	89.09	89.38	89.67	90.49	84.10
MSRNN [16]	68.00	87.39	88.16	88.79	89.24	89.67	89.85	90.28	90.43	90.70	87.25
STUDENT	81.64	85.23	87.53	88.59	89.33	89.79	90.20	90.36	90.58	90.63	88.39
Ours	83.32	87.13	88.92	89.82	90.85	91.04	91.28	91.23	91.31	91.47	89.64

UCF-101 set:

- 101 action classes, >13K action sequences
- Unconstrained video set (RGB)
- 10 progress levels

Observation ratio	10%	20%	30%	409
DBOW [32]	36.29	51.57	52.71	53.1
IBOW [32]	36.29	65.69	71.69	74.2
MTSSVM [23]	40.05	72.83	80.02	82.1
DeepSCN [24]	45.02	77.64	82.95	85.3
Mem-LSTM [22]	51.02	80.97	85.73	87.7
MSRNN [16]	68.00	87.39	88.16	88.7
STUDENT	81.64	85.23	87.53	88.5
Ours	83.32	87.13	88.92	89.8



Ablation study:

 with vs. without MSE(L) and MMD(G) losses, S stands for STUDENT only.

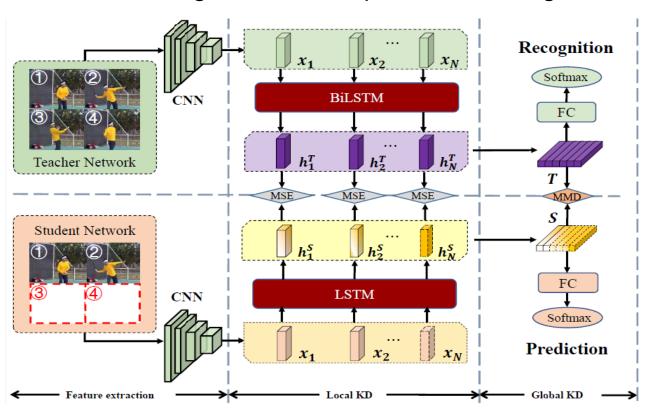
Observati	servation ratio		30%	50%	70%	100%	AUC
	S	54.58	67.08	73.75	80.42	84.17	71.87
SYSU	S+L	57.08	67.08	75.83	80.42	85.83	73.53
3130	S+G	57.50	66.67	76.67	80.42	85.00	73.08
	S+L+G	59.58	68.33	78.33	84.17	87.92	75.35
	S	81.64	87.53	89.33	90.20	90.63	88.39
UCF-101	S+L	83.19	88.43	90.22	91.20	90.98	89.27
UCF-101	S+G	83.57	88.02	90.14	90.63	90.71	89.01
	S+L+G	83.32	88.92	90.85	91.28	91.47	89.64

With vs. without joint learning

Observation ratio		10%	30%	50%	70%	100%	AUC
SYSU	with without			74.58 78.33	81.67 84.17	84.58 87.92	72.49 75.35
UCF-101	with without	83.60 83.32	88.35 88.92	89.82 90.85	90.20 91.28	90.85 91.47	89.07 89.64

4. Summary

- A novel teacher-student learning framework for distilling progressive action knowledge from action recognition model(teacher) to early action prediction model(student);
- An early action prediction system integrating the action prediction task with action recognition in the spirit of knowledge distillation.



5. MORE INFO.

http://isee.sysu.edu.cn/~hujianfang

EMAIL ME: hujianf@mail2.sysu.edu.cn

感谢各位老师和同学!