

Feature-Supervised Action Modality Transfer

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Action classification for small datasets



Large source dataset



Small target dataset

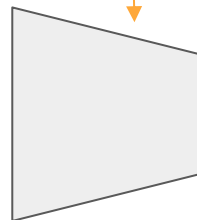
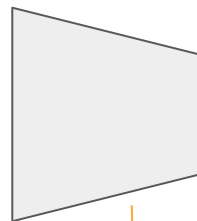
Transfer Learning



Pre-train



Finetune

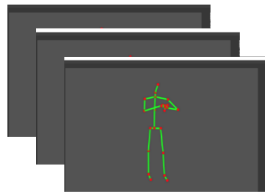
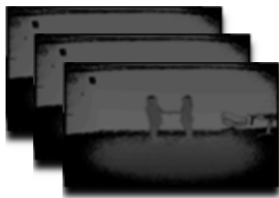


Better accuracy

Problem: pre-training datasets for non-RGB modalities unavailable.

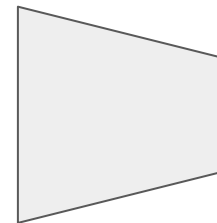
Problem Statement

Depth Maps or 3D-Skeletons



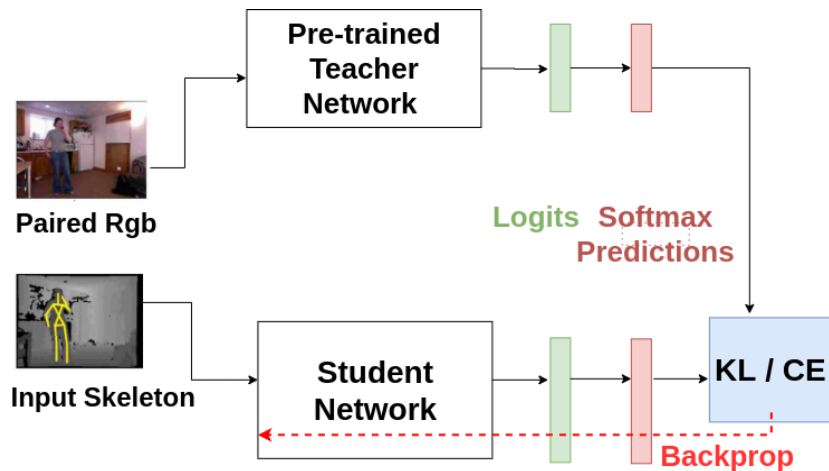
Learn

Action model



Can we transfer RGB action information to non-RGB modalities?

Standard Knowledge Distillation (Hinton et al. NeurIPS wshop, 2015)



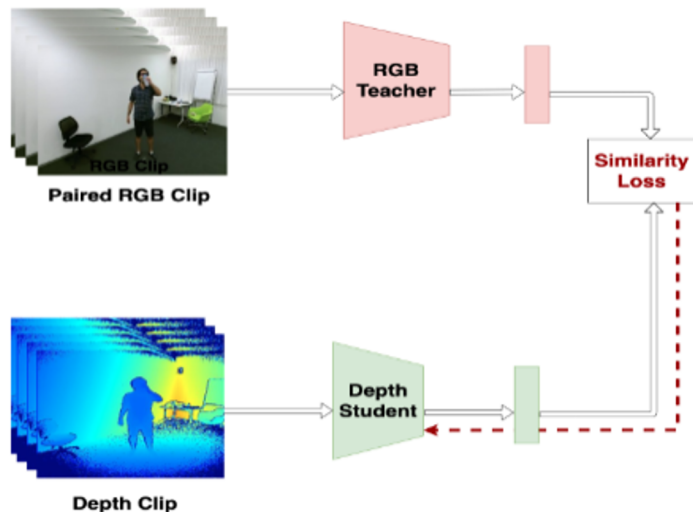
Transfer class-level information from the pretrained RGB teacher.

Match student softmax predictions with those of the teacher.

Requires teacher to be pre-trained for same action classes as the student.

Our Proposal: Feature-Supervised Action Modality Transfer

1. Match action embeddings of modality pairs via **feature-level supervision**.
2. Finetune for new action classes on a small labeled **non-RGB** dataset.



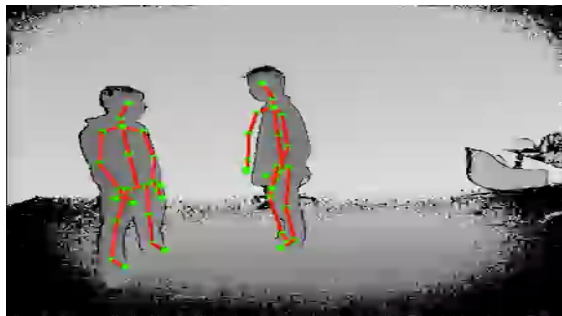
RGB teacher trained on a source dataset with non-overlapping action classes.

Multi-modal paired video data

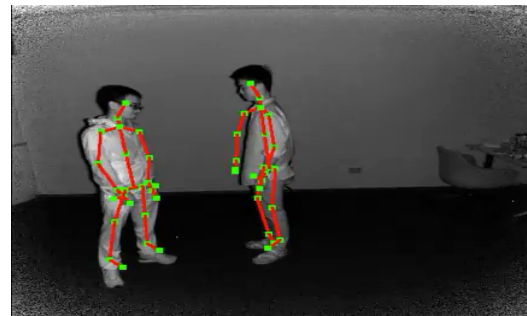
A sample action scene captured in multiple modalities (Handshaking).



RGB



Depth

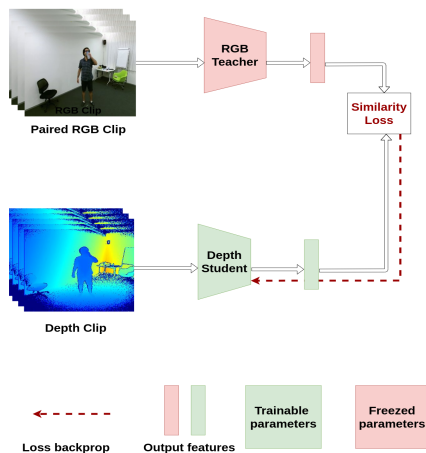


Infrared

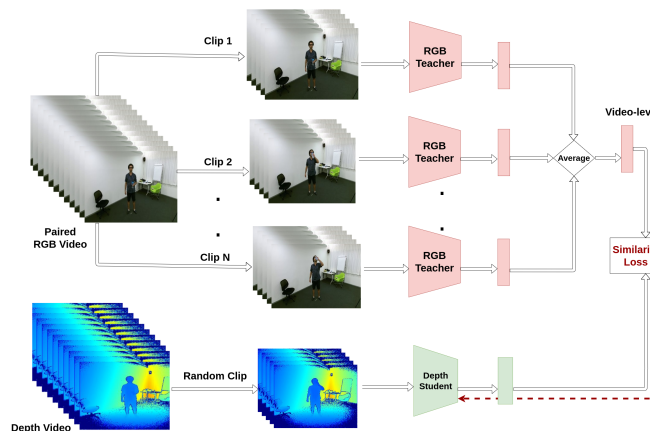
Transfer knowledge from pre-trained RGB models via unlabelled modality pairs.

Three Knowledge Transfer Granularities

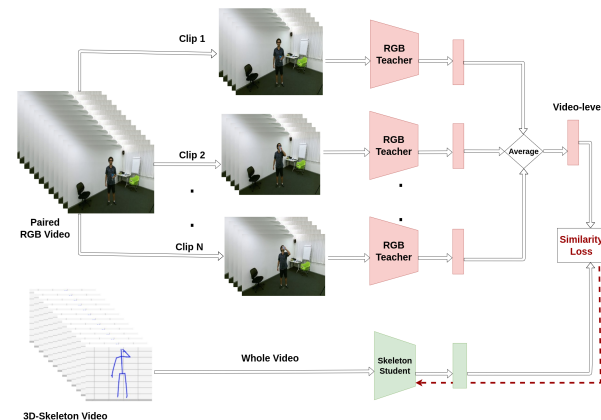
Clip-to-Clip



Video-to-Clip



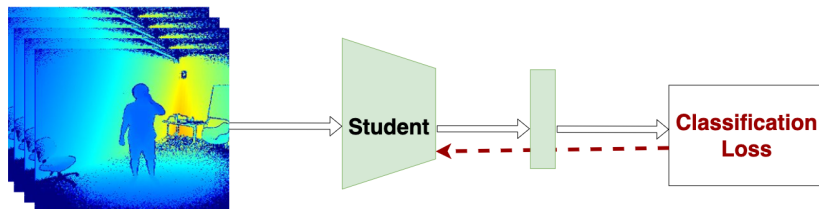
Video-to-Video



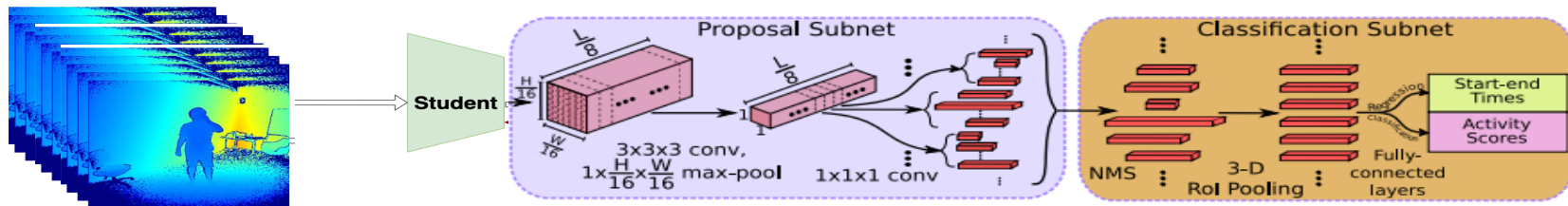
Cosine distance loss is minimized between action embeddings.

Fine-tuning with non-RGB examples

Task I: Action Classification



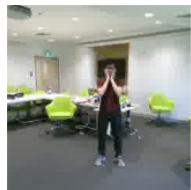
Task II: Action Detection (Xu et al. ICCV 2017)



Finetune pre-trained student with the task specific non-RGB labelled examples.

Experimental Setup

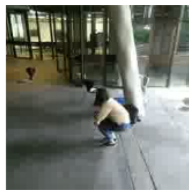
Source (NTU RGB+D 120 minus 60, Kinetics-400)



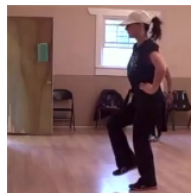
Apply Cream



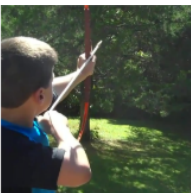
Exchange Things



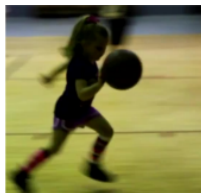
Squat Down



Zumba

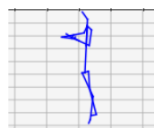
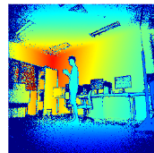


Archery

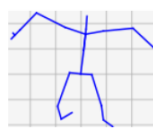
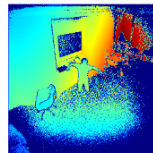


Dribbling basketball

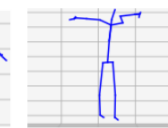
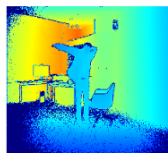
Target (NTU RGB+D 60, PKU-MMD)



Clapping



Falling Down



Wear Jacket

Pretrain teacher on RGB/Flow modality of the source dataset.

Transfer via unlabeled modality pairs of NTU RGB+D 60 training set.

Finetune with labeled examples from NTU RGB+D 60 / PKUMMD training set.

Ablation Studies

Source-Modality	Target-Modality: Depth		
	20 per-class	50 per-class	100 per-class
RGB	62.85±0.5	66.01±0.6	68.64±0.3
Flow	68.43±0.2	71.53±0.1	73.43±0.3

Which Source Modality?

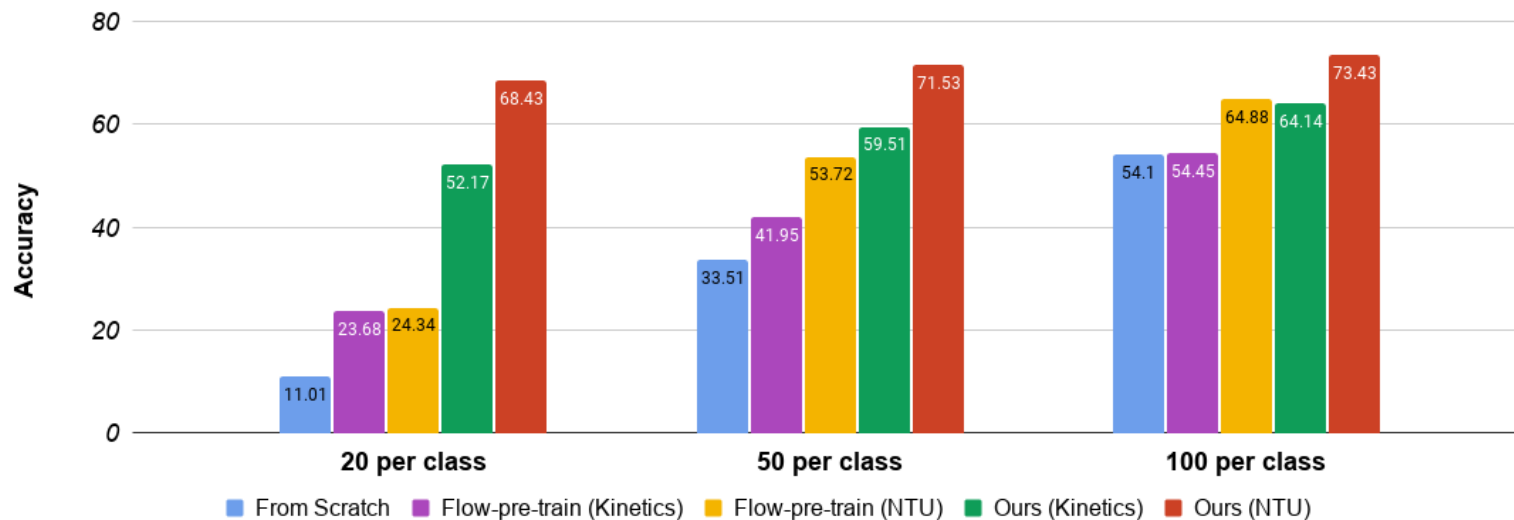
Granularity	Target-Modality: Depth		
	20 per-class	50 per-class	100 per-class
Clip-to-Clip	64.80±1.0	70.30±0.4	72.92±0.5
Video-to-Clip	68.43±0.2	71.53±0.1	73.43±0.3
Video + Clip	69.16±0.2	73.60±0.1	76.24±0.3

Which Granularity?

Optical-flow with video+clip granularity provides best feature-level supervision.

Results

Action classification from depth maps for NTU RGB+D 60 dataset

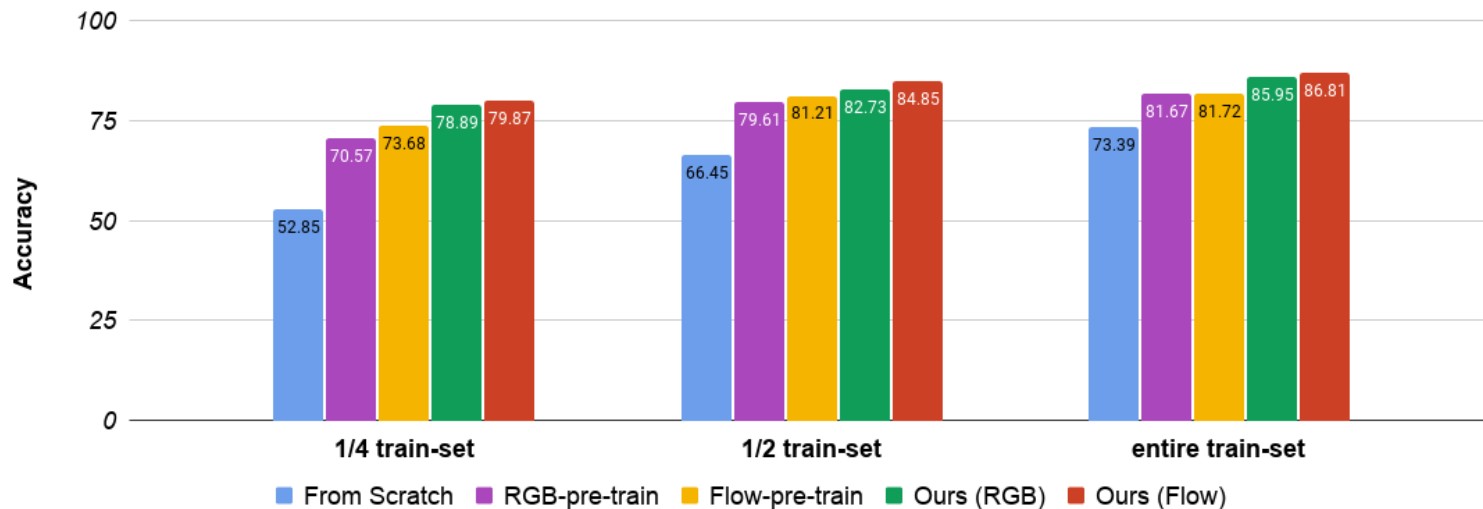


Considerable improvement over training from scratch and simple pretraining.

Source dataset with a similar domain provides better action transfer features.

Results

Action detection from depth maps for PKU-MMD dataset



Transfer results for 3D-skeleton action classification in paper.

Our method generalizes for temporal action detection as well.

Conclusion

RGB action datasets act as pre-training source for non-RGB modalities.

Optical-flow from a similar domain provides best feature-supervision.

Boost non-RGB action classification and detection when labels are scarce.

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