

MultiOOD: Scaling Out-of-Distribution Detection for Multiple Modalities

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NeurIPS 2024

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EPFL

Motivation

- Real world is dynamic



Motivation

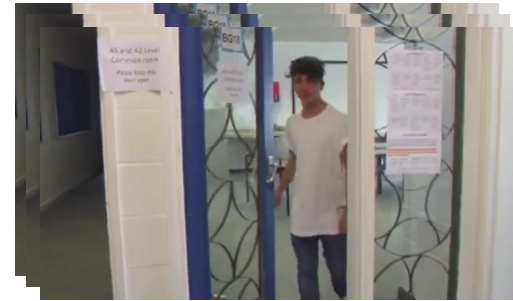
- Real world is multimodal



RGB images



LiDAR



Video



Optical Flow

Out-of-Distribution Detection

Train



dog

cat

fish

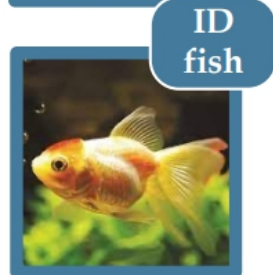
Test



ID
dog



ID
cat

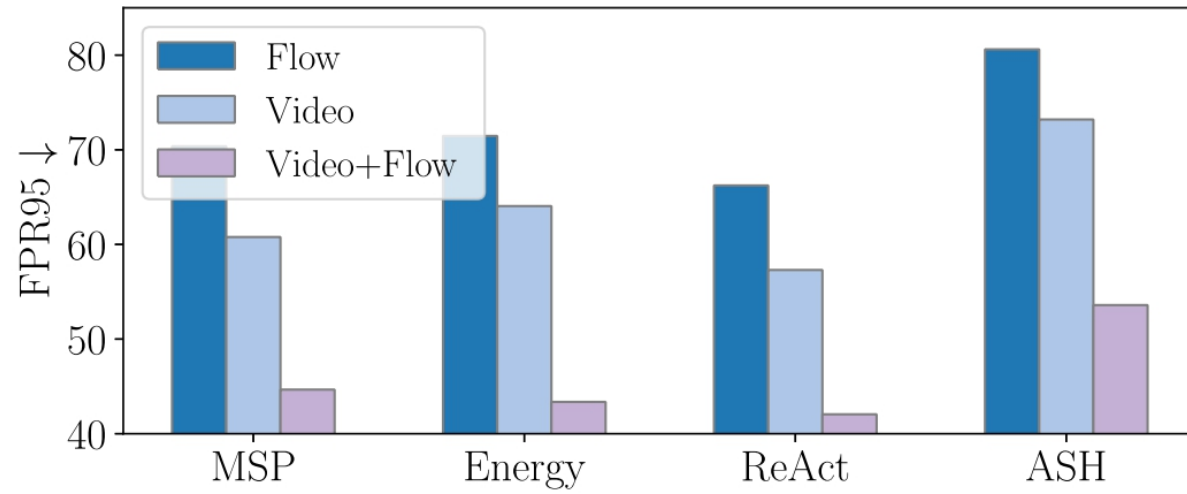


ID
fish



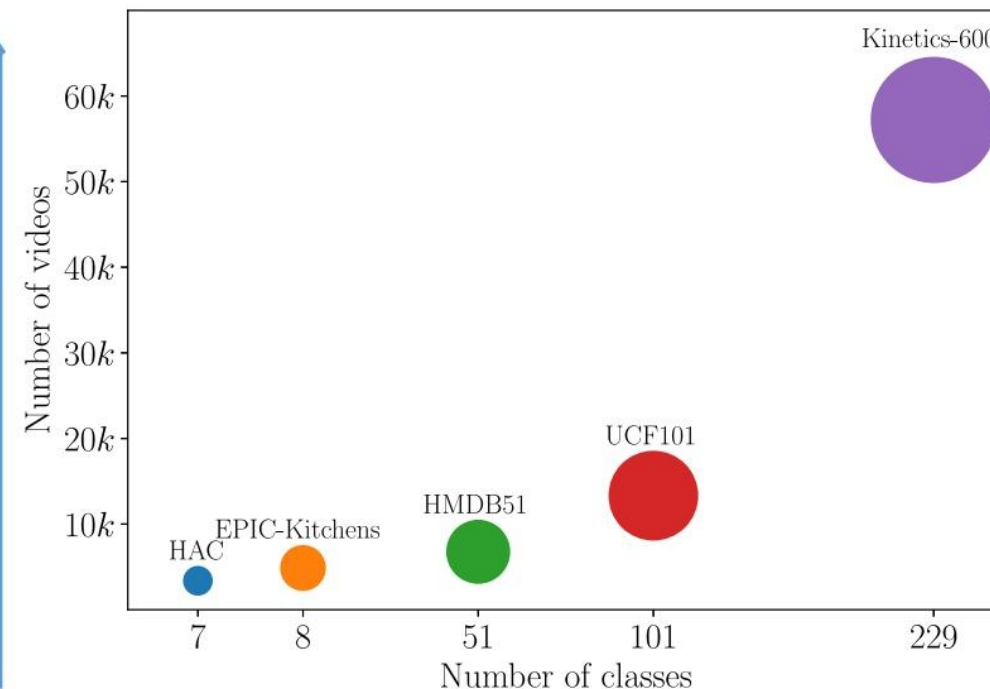
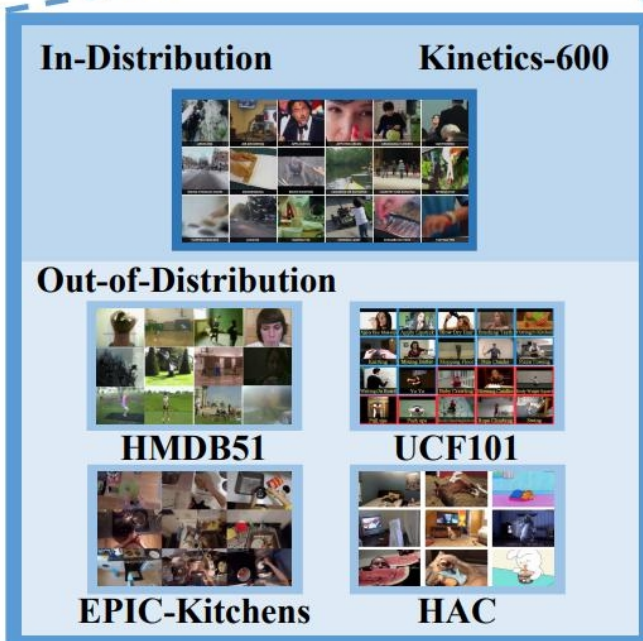
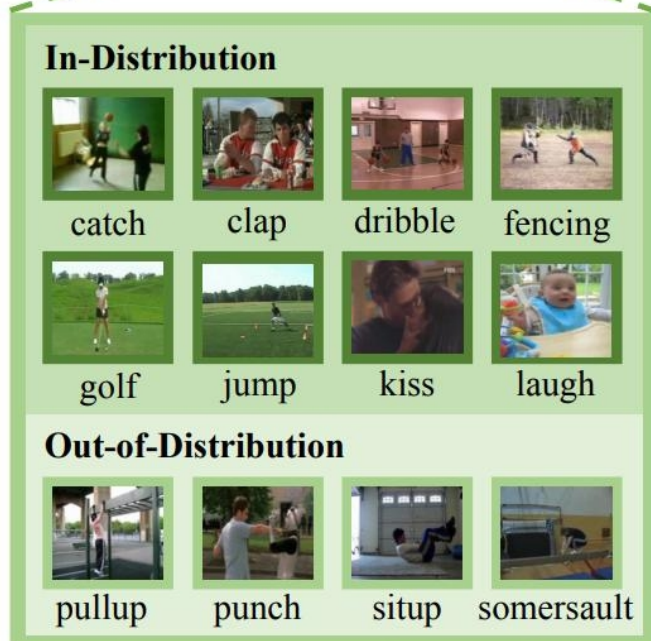
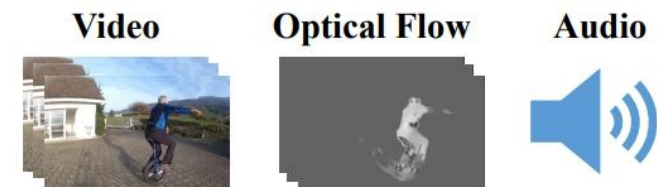
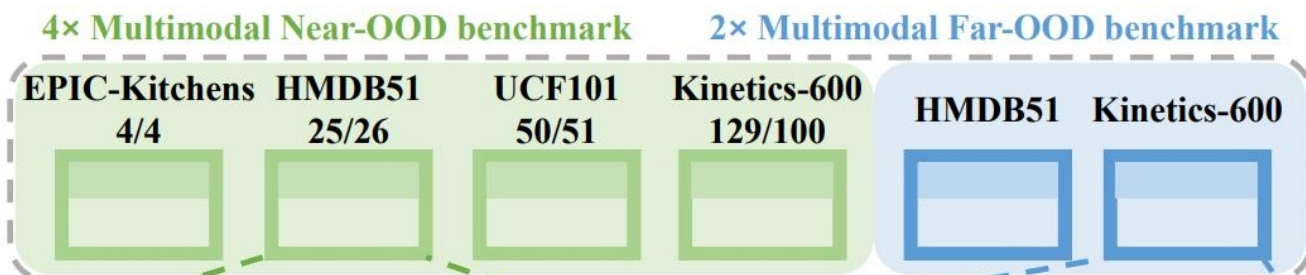
OOD

Why Multimodal

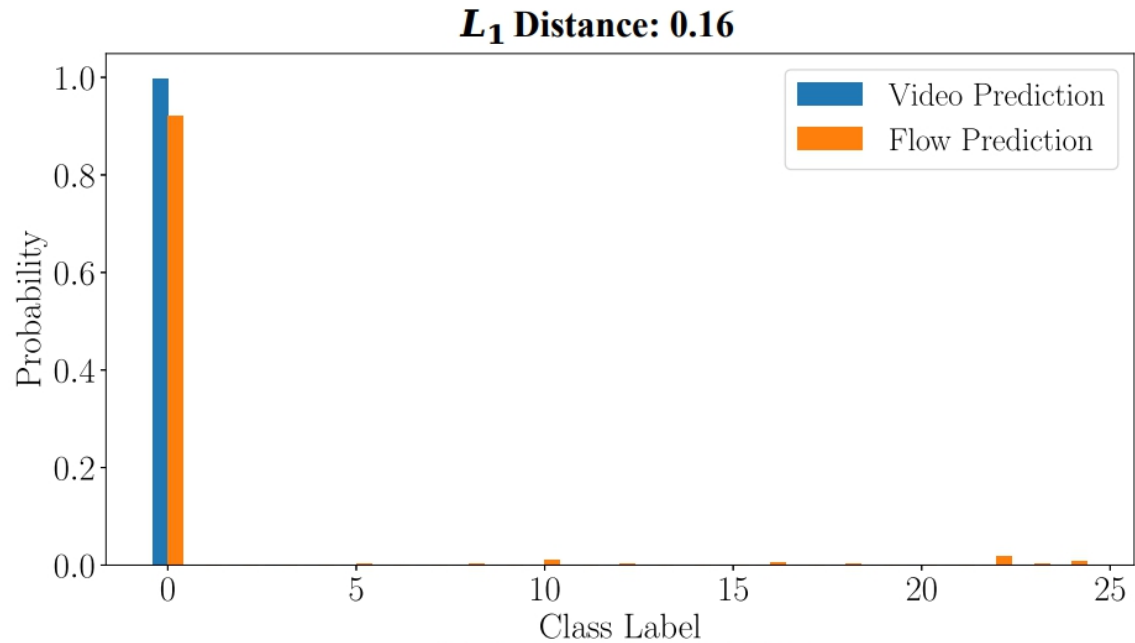


MultiOOD Benchmark

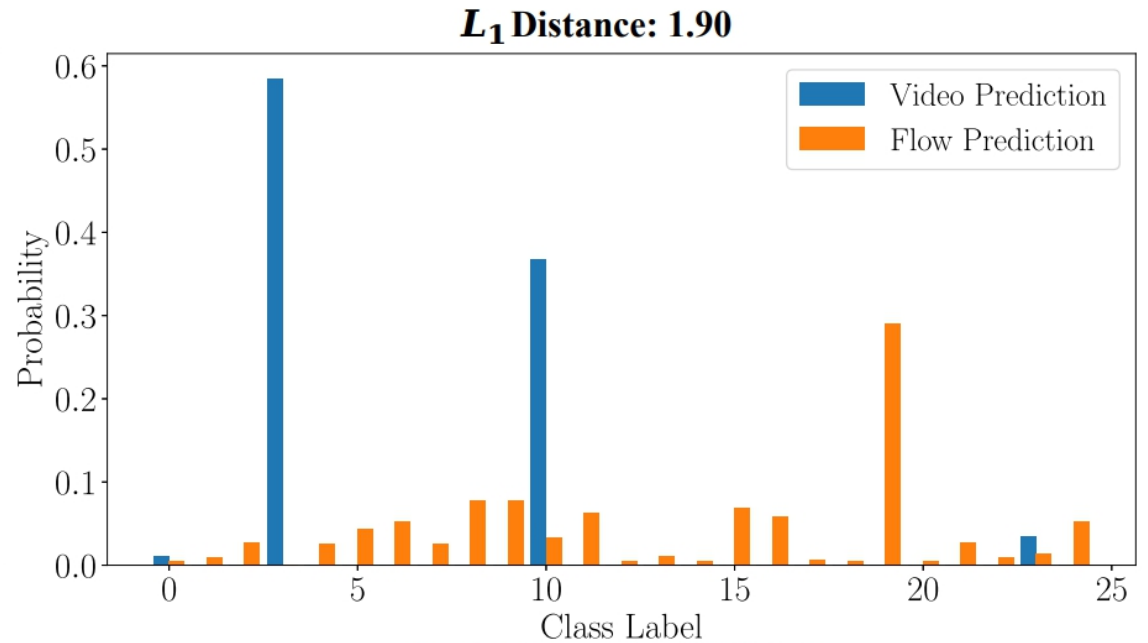
- The *first* benchmark for Multimodal OOD Detection



Modality Prediction Discrepancy

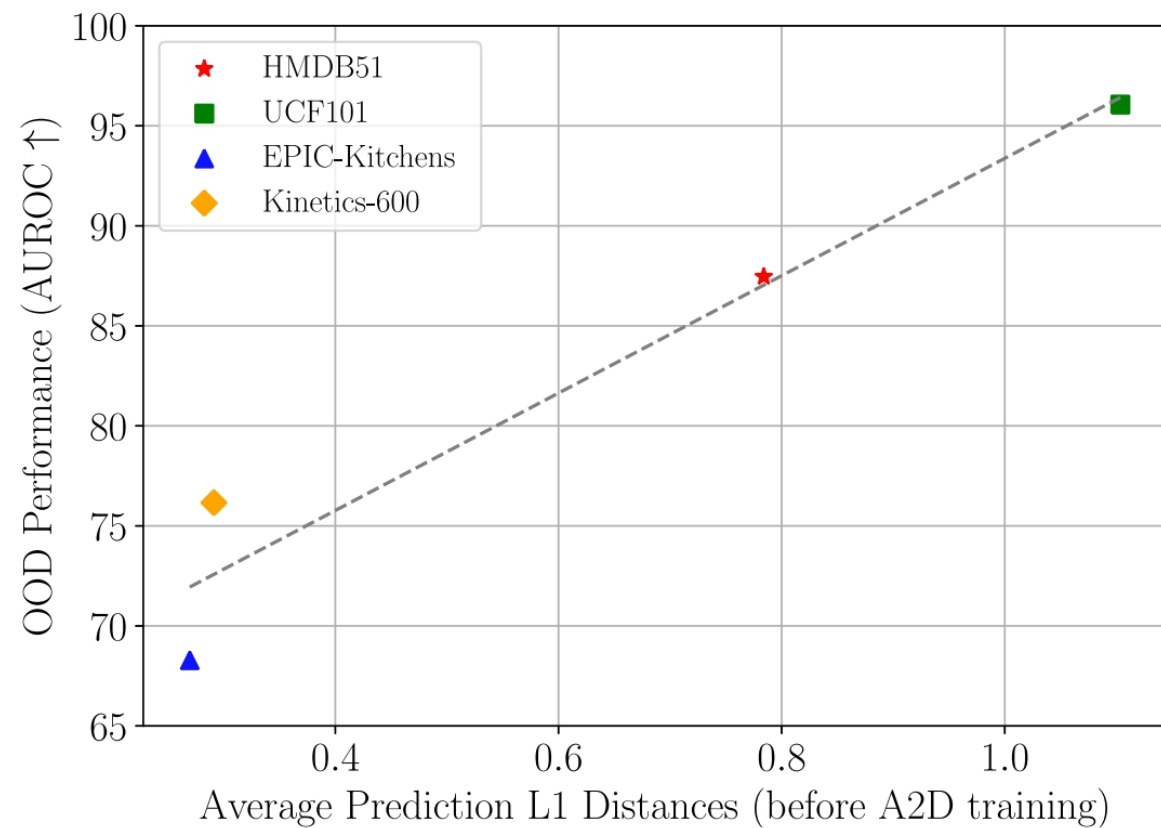


(a) Prediction for ID data

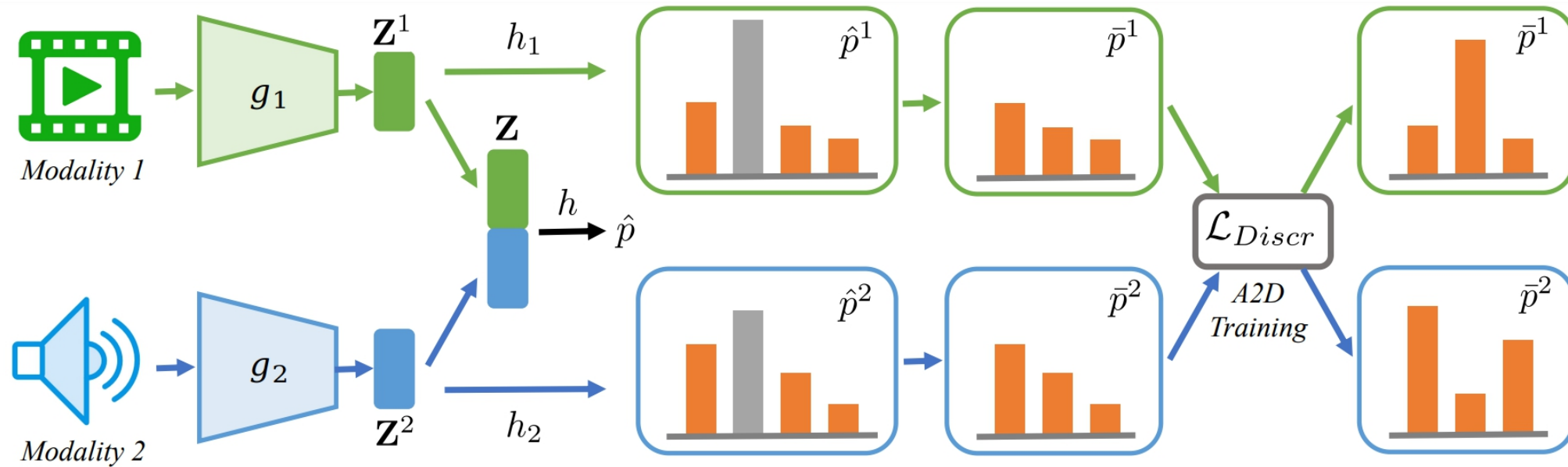


(b) Prediction for OOD data

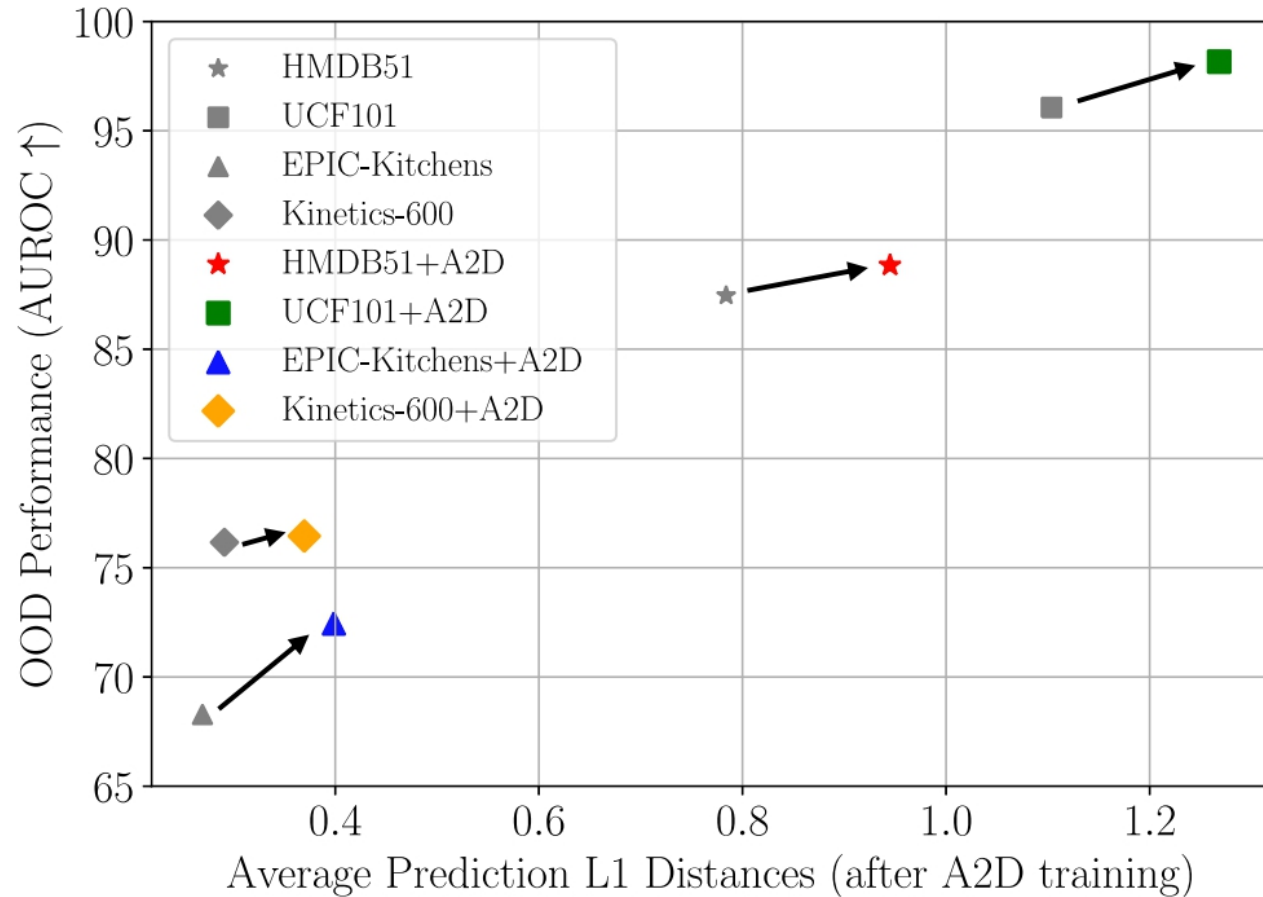
Modality Prediction Discrepancy



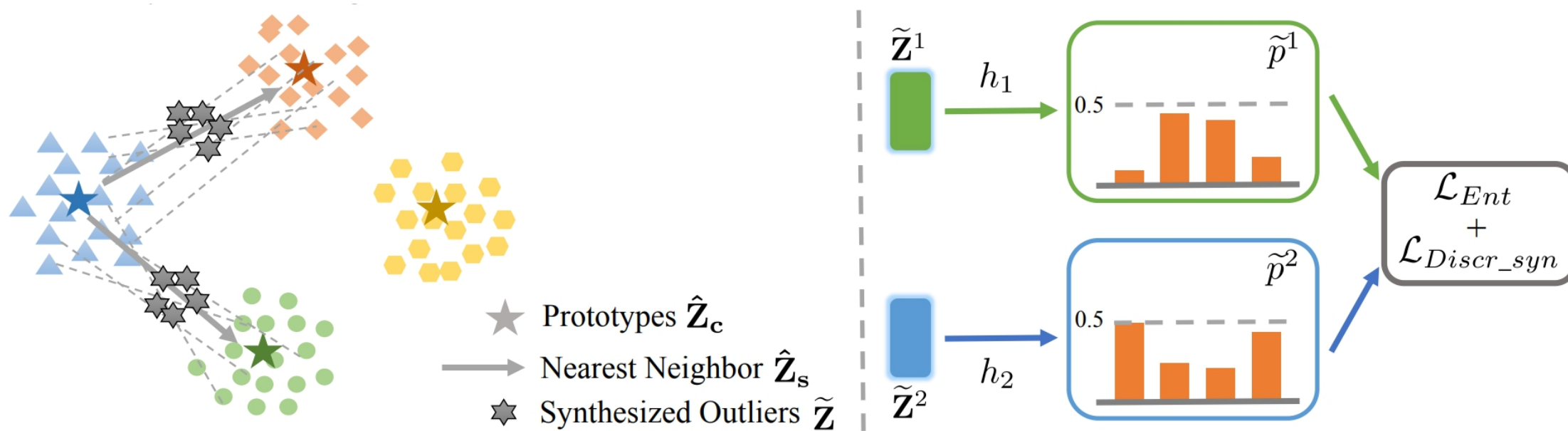
Agree-to-Disagree Algorithm



Agree-to-Disagree Algorithm



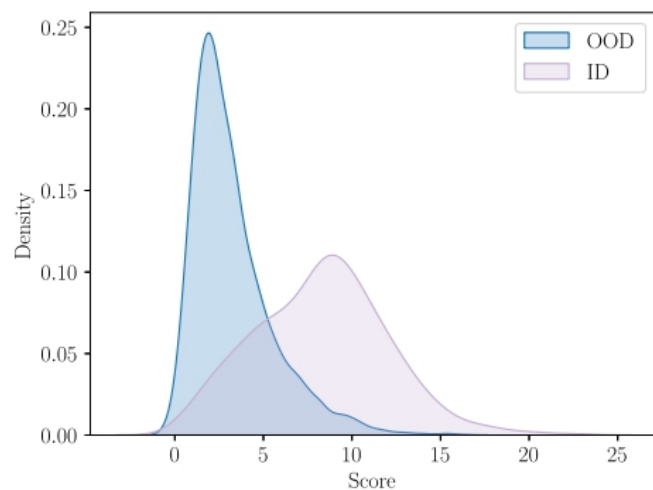
Nearest Neighbor Prototype-based Mixup



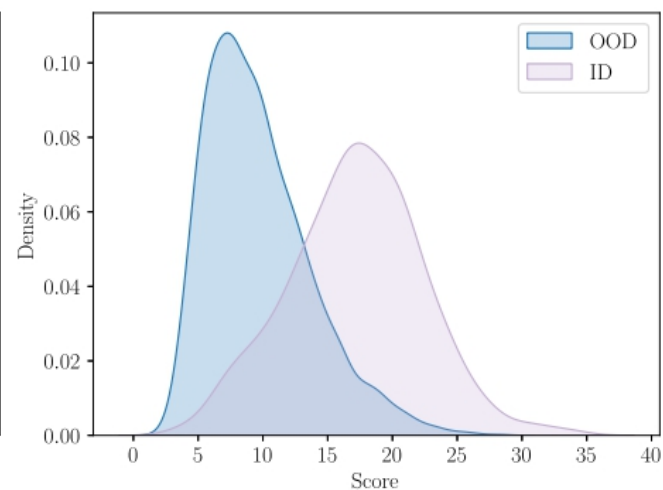
Multimodal OOD Detection with HMDB51 as ID

Methods	OOD Datasets										ID ACC \uparrow
	Kinetics-600		UCF101		EPIC-Kitchens		HAC		Average		
	FPR95 \downarrow	AUROC \uparrow	FPR95 \downarrow	AUROC \uparrow	FPR95 \downarrow	AUROC \uparrow	FPR95 \downarrow	AUROC \uparrow	FPR95 \downarrow	AUROC \uparrow	
Without A2D Training											
Energy	32.95	92.48	44.93	87.95	8.10	97.70	32.95	92.28	29.73	92.60	87.23
ASH	51.20	87.81	53.93	84.22	19.95	95.92	42.99	90.23	42.02	89.55	86.20
GEN	41.51	90.34	46.18	87.91	8.21	98.26	38.31	91.28	33.55	91.95	87.23
KNN	22.69	95.01	39.34	89.28	9.92	97.92	20.75	96.02	23.18	94.56	87.23
VIM	13.68	97.01	33.87	91.45	5.93	98.15	13.45	97.12	16.73	95.93	87.23
With A2D Training and NP-Mix Outlier Synthesis											
Energy++	24.52 $_{-8.43}$	93.96 $_{+1.48}$	36.49 $_{-8.44}$	89.67 $_{+1.72}$	6.96 $_{-1.14}$	97.53 $_{-0.17}$	22.92 $_{-10.14}$	94.41 $_{+2.13}$	22.72 $_{-7.01}$	93.89 $_{+1.29}$	86.89
ASH++	27.82 $_{-23.38}$	93.17 $_{+5.36}$	38.43 $_{-15.50}$	89.52 $_{+5.30}$	6.84 $_{-13.11}$	98.23 $_{+2.31}$	23.03 $_{-19.96}$	94.45 $_{+4.22}$	24.03 $_{-17.99}$	93.84 $_{+4.29}$	86.20
GEN++	25.66 $_{-15.85}$	93.50 $_{+3.16}$	37.40 $_{-8.78}$	91.19 $_{+3.28}$	5.25 $_{-2.96}$	98.98 $_{+0.72}$	24.63 $_{-13.68}$	94.28 $_{+3.00}$	23.24 $_{-10.31}$	94.49 $_{+2.54}$	86.89
KNN++	15.05 $_{-7.64}$	96.96 $_{+1.95}$	33.06 $_{-6.28}$	91.92 $_{+2.64}$	5.47 $_{-4.45}$	98.97 $_{+1.05}$	13.45 $_{-7.30}$	97.25 $_{+1.23}$	16.76 $_{-6.42}$	96.28 $_{+1.72}$	86.89
VIM++	9.24 $_{-4.44}$	98.04 $_{+1.03}$	26.45 $_{-7.42}$	92.34 $_{+0.89}$	5.36 $_{-0.57}$	98.09 $_{-0.06}$	6.04 $_{-7.41}$	98.56 $_{+1.44}$	11.77 $_{-4.96}$	96.76 $_{+0.83}$	86.89

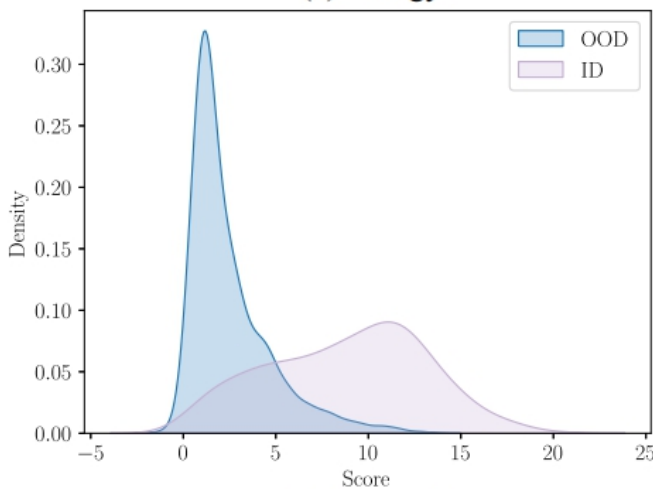
Score Distributions



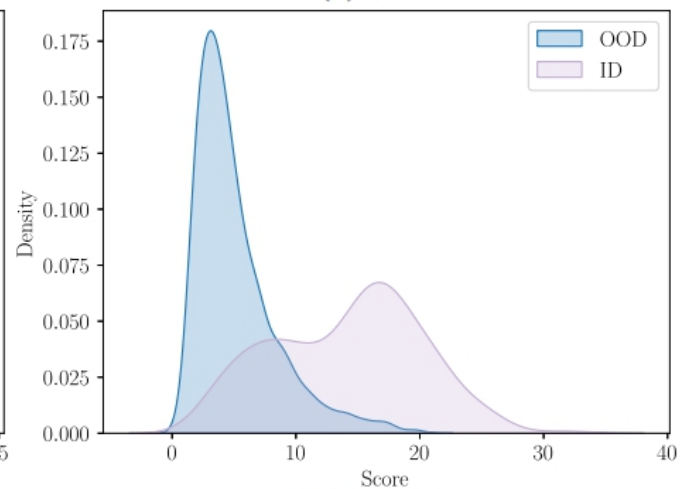
(a) Energy



(b) ASH



(e) Energy++



(f) ASH++

Conclusion

- The *first* benchmark for Multimodal OOD Detection
- A novel *A2D* training algorithm, inspired by the observation of the *Modality Prediction Discrepancy* phenomenon
- A new outlier synthesis algorithm *NP-Mix*
- Extensive evaluations



Code and Benchmark: <https://github.com/donghao51/MultiOOD>