What Can We Learn from Harry Potter? An Exploratory Study of Visual Representation Learning from Atypical Videos

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pretraining

(w/o label)









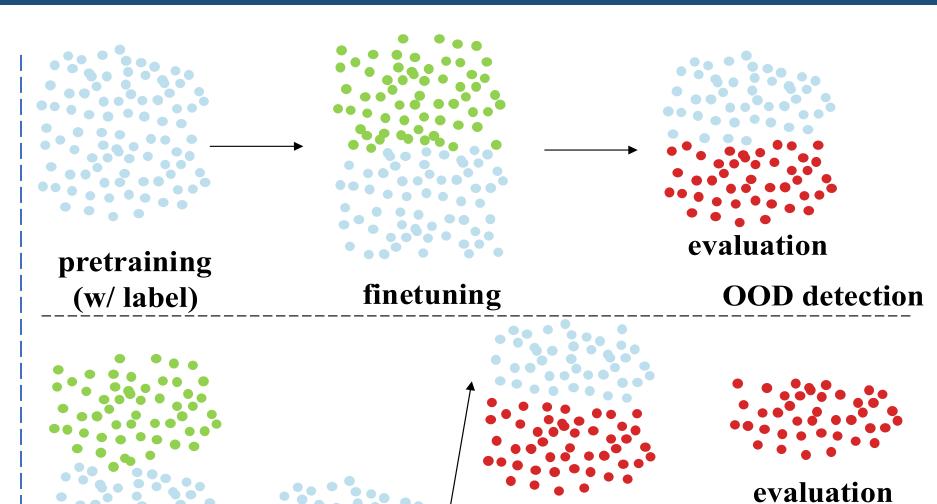
NCD

ZSAR

known data

unknown data

auxiliary data



finetuning







- ☐ Humans generalize well from *uncommon* experiences — can models do the same?
- ☐ Existing studies focus on *typical closed-set* videos.
- ☐ Atypical/unusual videos remain under-explored for open-world learning.
- ☐ Investigate whether *atypical videos* improve open world tasks (OOD, NCD, ZSAR).

Atypical datasets

joint training

evaluation



Table 1. Statistical details of the proposed <i>atypical</i> video dataset							
Category	# Videos	Avg duration	Data Source				
Sci-fi	1,119	4.00s	YouTube film trailers				
Animation	1,058	4.04s	YouTube animated trailers				
Unintentional	2,835	9.77s	Oops Dataset [7]				
Abnormal	1,103	7.53s	Ped2 [24], CUHK Avenue [23], UCF-Crime [35]				
Surreal	1,024	3.18s	Surreal dataset [37]				
Theatre	679	4.81s	YouTube stage performances				
atypical	7,818	5.70s	Diverse sources				

Table 2. Data splits for open world learning							
Task	Known	Unknown	Auxiliary				
OOD detection	UCF101	HMDB51, MiT-v2, [atypical]-surreal, [atypical]-theatre	Gaussian Noise, Diving48, Kinetics400, atypical				
NCD /ZSAR	UCF_80, HMDB	_40 UCF_21, HMDB_11	UCF101 / HMDB51, Kinetics400, atypical				

Experiments

unintentional

[atypical] - theatre

[atypical] - surreal

□ OOD detection

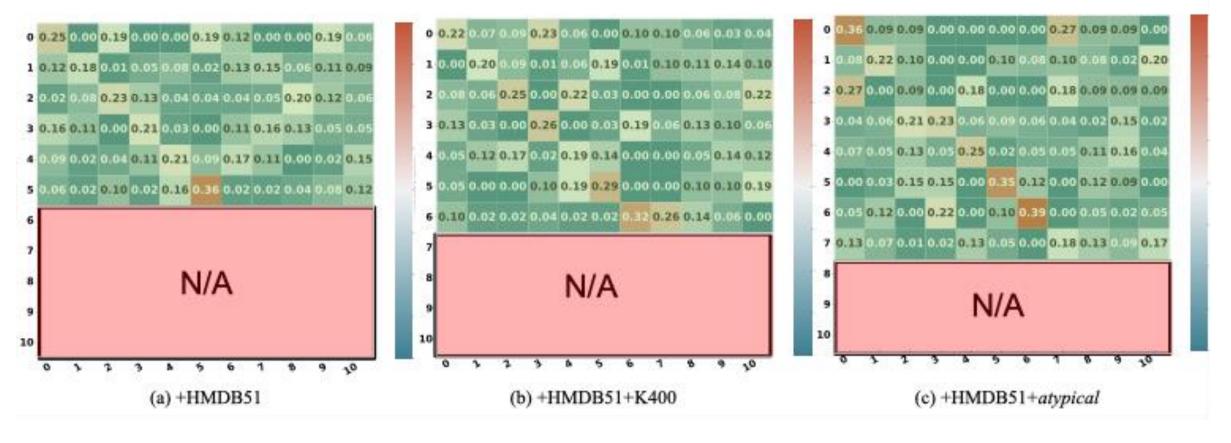
[atypical] - animation

HMDB51

• Gaussian Noise • HMDB51 • UCF101 • Kinetics400 • MiT-v2 ▲ atypical

Dataset	ataset Metric		+Gaussian	+Diving48	+K400	+Atypical
HMDB51	FPR95↓ AUROC↑ AUPR↑	77.08 63.85 22.54	81.11 63.36 23.03	81.14 64.84 24.04	75.52 66.84 25.13	73.07 69.43 27.07
MiT-v2	FPR95↓ AUROC↑ AUPR↑	77.73 64.94 23.76	77.51 65.14 24.12	80.87 65.46 27.24	67.72 72.53 30.86	66.62 74.01 32.59
[atypical]-surreal	FPR95↓ AUROC↑ AUPR↑	62.22 75.50 33.42	67.37 73.32 32.26	73.14 69.13 25.71	53.66 79.78 39.13	49.99 82.03 43.19
[atypical]-theatre	FPR95↓ AUROC↑ AUPR↑	79.75 61.80 20.34	80.71 63.95 21.97	82.66 60.05 19.04	66.41 71.29 27.17	61.91 74.62 34.57
MEAN	FPR95↓ AUROC↑ AUPR↑	74.20 66.52 25.02	76.68 66.44 25.35	79.45 64.87 24.01	65.83 72.61 30.62	62.90 75.02 34.36

I Novel category discovery

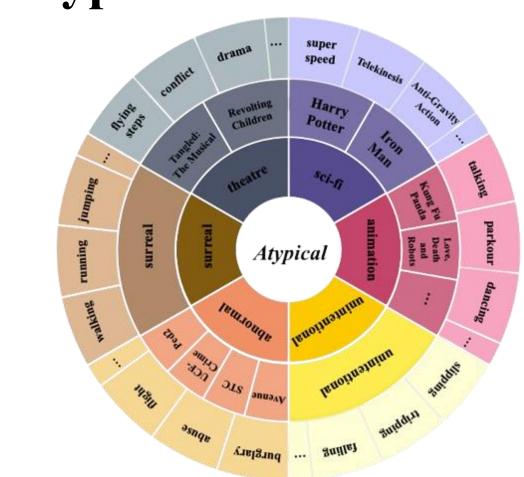


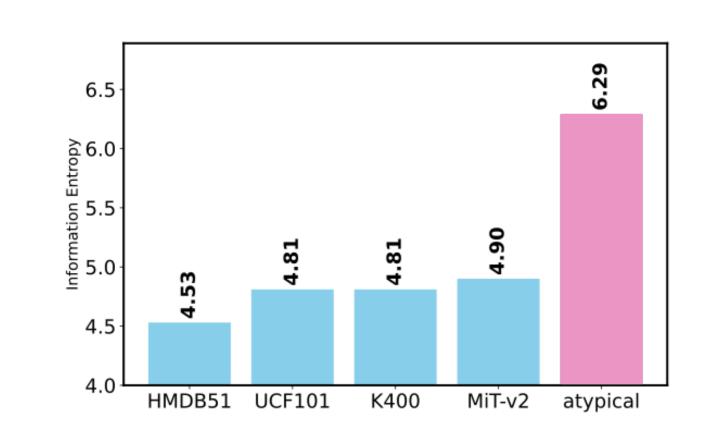
☐ Zero-shot action recognition

Metric	Baseline	+HMDB51	+K400	+atypical	+HMDB51+K400	+HMDB51+atypical
Top-1 ↑	57.40	57.02	58.23	58.30	61.31	62.96
Top-5 ↑	94.97	96.62	96.84	97.45	96.62	96.39

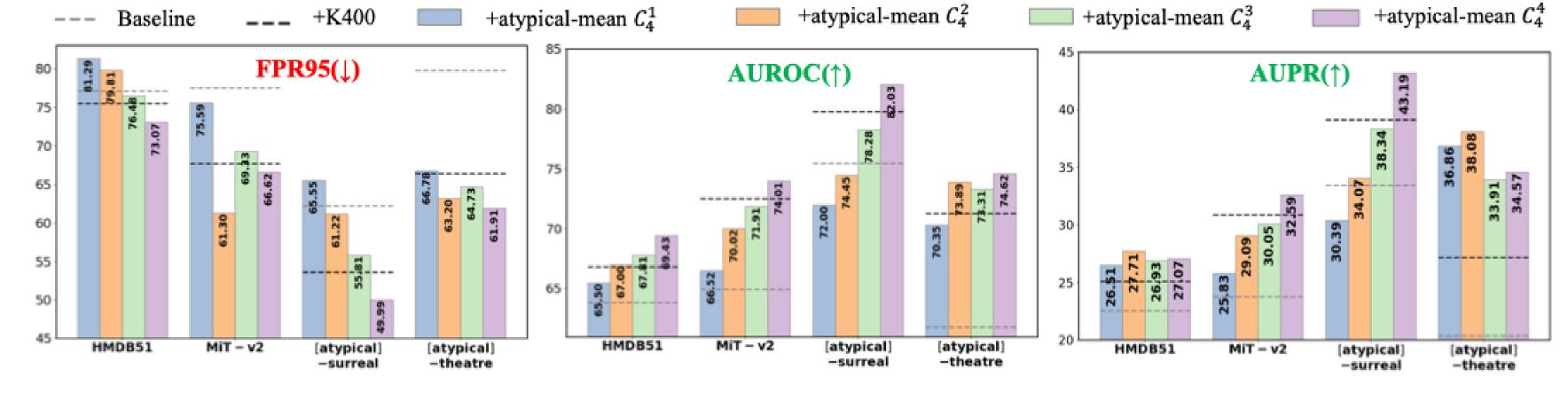
What can we learn from Harry Potter?

☐ Atypical dataset exhibits semantic and distributional diversity





☐ Semantic diversity helps open-world learning



☐ Atypical data coverage helps open-world learning

Test Sets	Metric	+abn_sci	+abn_uni	+abn_ani	+ani_sci	+ani_uni	+sci_uni	Mean ± Std Dev
HMDB51 [25]	FPR95↓	81.86	<u>78.36</u>	80.72	82.06	79.06	76.77	79.81 ± 2.10
	AUROC ↑	65.50	65.77	66.63	66.03	<u>68.09</u>	69.97	67.00 ± 1.72
	AUPR ↑	<u>29.38</u>	23.39	28.40	30.49	25.93	28.67	27.71 ± 2.60
MiT v2	FPR95↓	79.98	68.26	78.49	79.70	61.40	64.83	72.11 ± 8.27
MiT-v2 [32]	AUROC ↑	63.89	73.46	67.23	65.35	75.30	<u>74.87</u>	70.02 ± 5.11
	AUPR ↑	24.36	31.82	26.91	25.36	<u>32.94</u>	33.16	29.09 ± 4.00
[atypical]	FPR95↓	75.99	48.32	73.31	74.11	47.32	48.29	61.22 ± 14.54
-surreal	AUROC ↑	64.54	83.28	67.74	65.95	<u>82.94</u>	82.22	74.45 ± 9.23
	AUPR ↑	22.26	44.81	25.18	23.31	<u>44.69</u>	44.14	34.07 ± 11.52
[atypical] -theatre	FPR95↓	75.96	81.27	57.40	41.13	58.72	64.70	63.20 ± 14.37
	AUROC ↑	69.89	56.26	<u>82.45</u>	91.18	73.34	70.19	73.89 ± 11.94
	AUPR ↑	31.75	16.73	<u>52.84</u>	73.82	28.64	26.72	38.42 ± 21.02