
Supplementary information

Mastering diverse control tasks through world models

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Atari learning curves

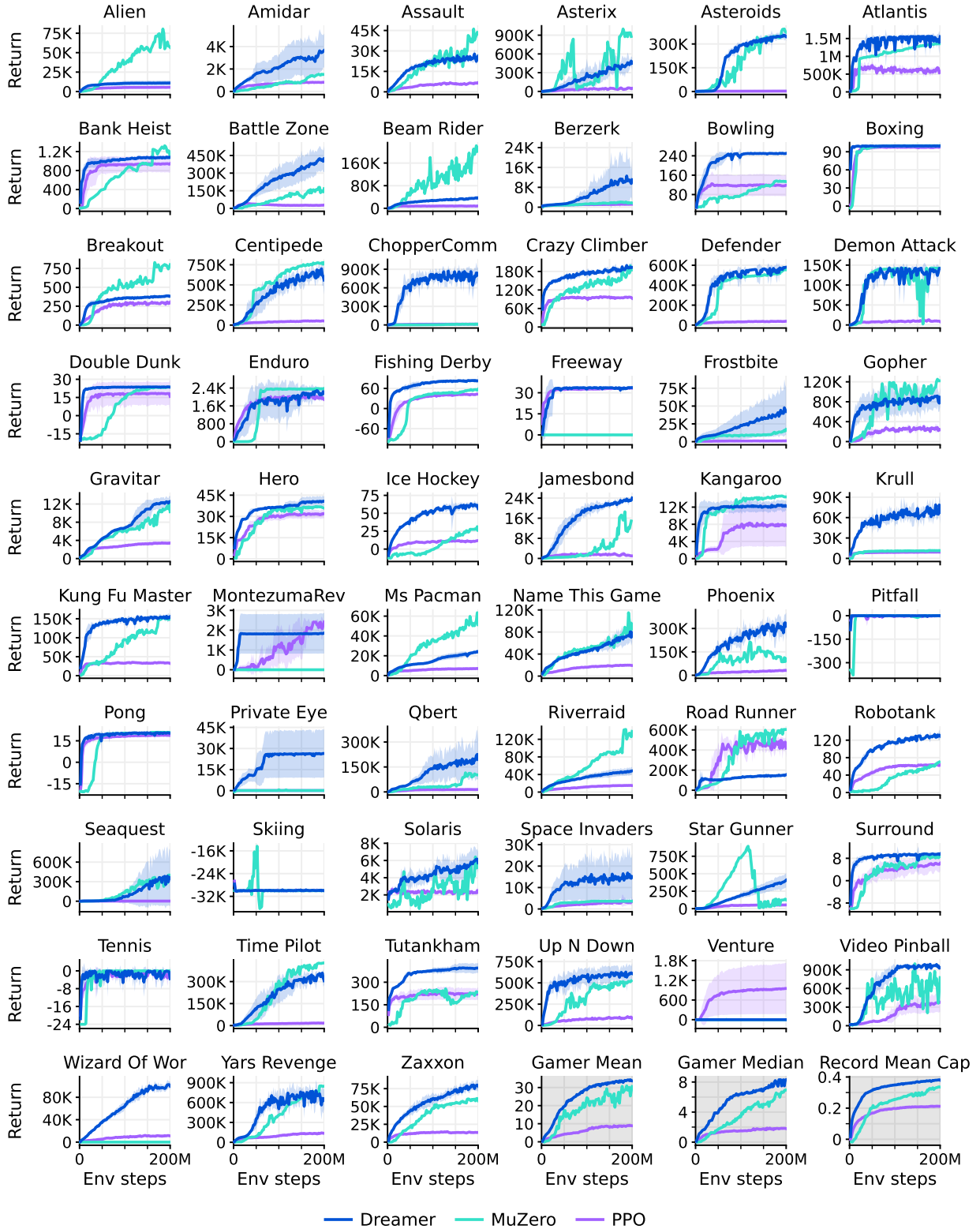


Figure 1: Atari learning curves.

Atari scores

Task	Random	Gamer	Record	PPO	MuZero	Dreamer
Environment steps	—	—	—	200M	200M	200M
Alien	228	7128	251916	5476	56835	10977
Amidar	6	1720	104159	817	1517	3612
Assault	222	742	8647	6673	42742	26010
Asterix	210	8503	1000000	47190	879375	441763
Asteroids	719	47389	10506650	2479	374146	348684
Atlantis	12850	29028	10604840	539721	1353617	1553222
Bank Heist	14	753	82058	946	1077	1083
Battle Zone	2360	37188	801000	27816	167412	419653
Beam Rider	364	16926	999999	7973	201154	37073
Berzerk	124	2630	1057940	1186	1698	10557
Bowling	23	161	300	118	133	250
Boxing	0	12	100	98	100	100
Breakout	2	30	864	299	799	384
Centipede	2091	12017	1301709	51833	774421	554553
Chopper Command	811	7388	999999	12667	8945	802698
Crazy Climber	10780	35829	219900	93176	184394	193204
Defender	2874	18689	6010500	38270	554492	579875
Demon Attack	152	1971	1556345	8229	142509	142109
Double Dunk	-19	-16	22	16	23	24
Enduro	0	860	9500	1887	2369	2166
Fishing Derby	-92	-39	71	43	58	82
Freeway	0	30	38	33	0	34
Frostbite	65	4335	454830	1123	17087	41888
Gopher	258	2412	355040	24792	122025	87600
Gravitar	173	3351	162850	3436	10301	12570
Hero	1027	30826	1000000	31967	36063	40677
Ice Hockey	-11	1	36	12	26	57
Jamesbond	29	303	45550	1019	14872	24010
Kangaroo	52	3035	1424600	7769	14380	12229
Krull	1598	2666	104100	9193	11476	69858
Kung Fu Master	258	22736	1000000	32335	148936	154893
Montezuma Revenge	0	4753	1219200	2368	0	1852
Ms Pacman	307	6952	290090	7041	51310	24079
Name This Game	2292	8049	25220	19441	85331	77809
Phoenix	761	7243	4014440	31412	105593	316606
Pitfall	-229	6464	114000	-2	0	0
Pong	-21	15	21	19	21	20
Private Eye	25	69571	101800	73	100	26432
Qbert	164	13455	2400000	14554	102129	201084
Riverraid	1338	17118	1000000	14860	137983	48080
Road Runner	12	7845	2038100	423995	604083	150402
Robotank	2	12	76	63	70	132
Seaquest	68	42055	999999	1927	399764	356584
Skiing	-17098	-4337	-3272	-29926	-30000	-29965
Solaris	1236	12327	111420	2368	5860	5851
Space Invaders	148	1669	621535	3489	3639	15005
Star Gunner	664	10250	77400	53439	127417	408961
Surround	-10	6	6	6	9	9
Tennis	-24	-8	21	-1	0	-3
Time Pilot	3568	5229	65300	17250	427209	314947
Tutankham	11	168	5384	225	235	395
Up N Down	533	11693	82840	83743	522962	614065
Venture	0	1188	38900	953	0	0
Video Pinball	16257	17668	89218328	382306	775304	940631
Wizard Of Wor	564	4756	395300	10910	0	99136
Yars Revenge	3093	54577	15000105	137164	846061	675774
Zaxxon	32	9173	83700	13599	58115	78443
Gamer median (%)	0	100	3716	180	693	830
Gamer mean (%)	0	100	123001	892	3054	3381
Record mean (%)	0	13	100	21	66	74
Record mean capped (%)	0	13	100	21	34	38

Table 1: Atari scores.

ProcGen learning curves

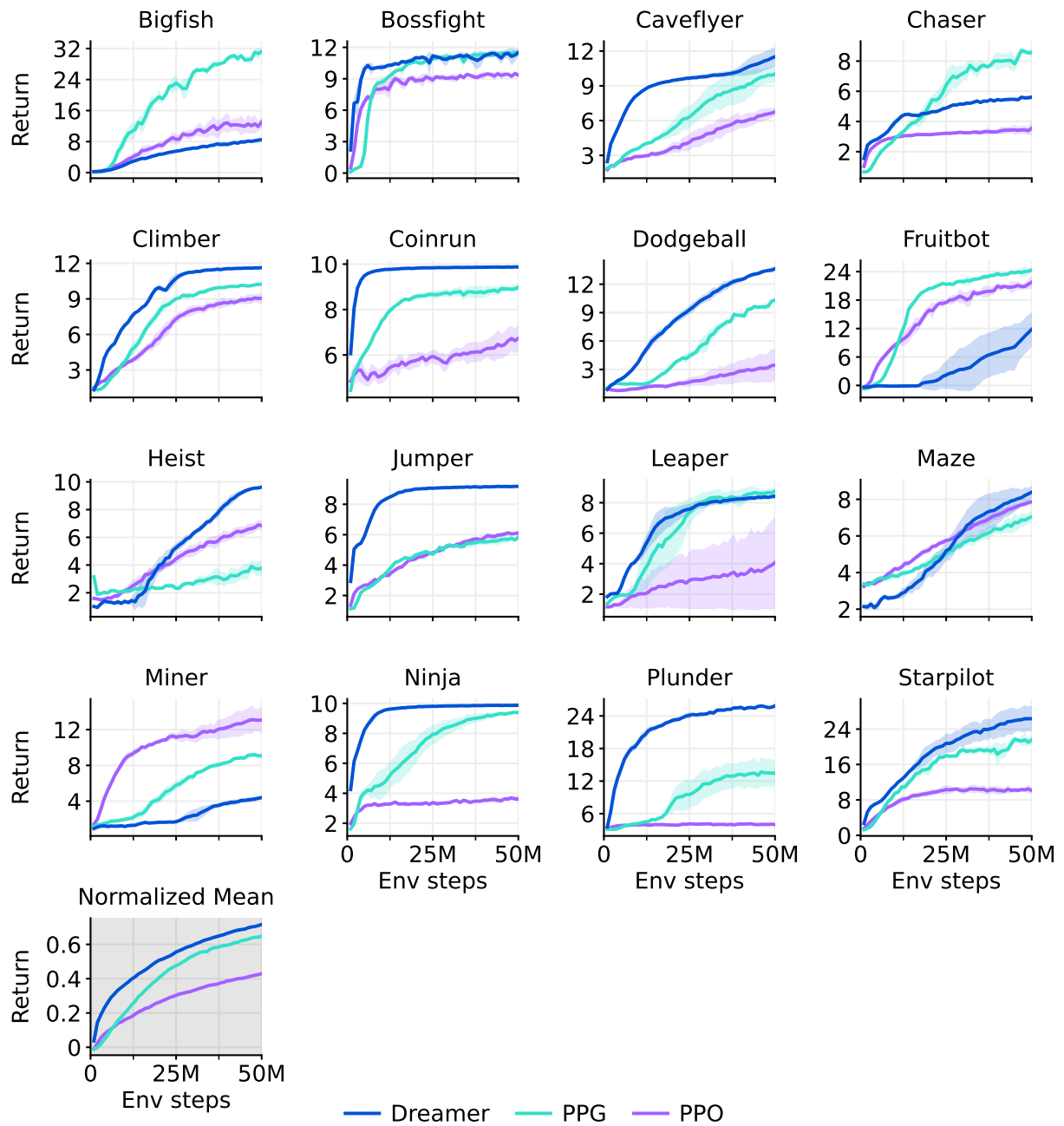


Figure 2: ProcGen learning curves.

ProcGen scores

Task	Original PPO	PPO	PPG	Dreamer
Environment steps	50M	50M	50M	50M
Bigfish	10.9	12.7	31.3	8.5
Bossfight	10.5	9.4	11.5	11.5
Caveflyer	6.0	6.7	10.0	11.5
Chaser	4.5	3.5	8.6	5.6
Climber	7.6	9.0	10.2	11.6
Coinrun	7.9	6.7	9.0	9.9
Dodgeball	4.8	3.4	10.3	13.6
Fruitbot	20.3	21.7	24.3	11.8
Heist	2.2	6.9	3.8	9.6
Jumper	5.1	6.1	5.8	9.2
Leaper	5.9	4.1	8.8	8.4
Maze	5.0	7.9	7.1	8.4
Miner	7.6	13.0	9.1	4.4
Ninja	6.2	3.6	9.4	9.9
Plunder	11.2	4.0	13.4	25.8
Starpilot	17.0	10.1	21.6	26.4
Normalized mean	41.2	42.8	64.9	71.6

Table 2: ProcGen scores. The PPO implementation we use under fixed hyperparameters throughout our paper performs on par or better than the original PPO, which its authors describe as highly tuned with near optimal hyperparameters in Cobbe et al. (2021).

DMLab learning curves

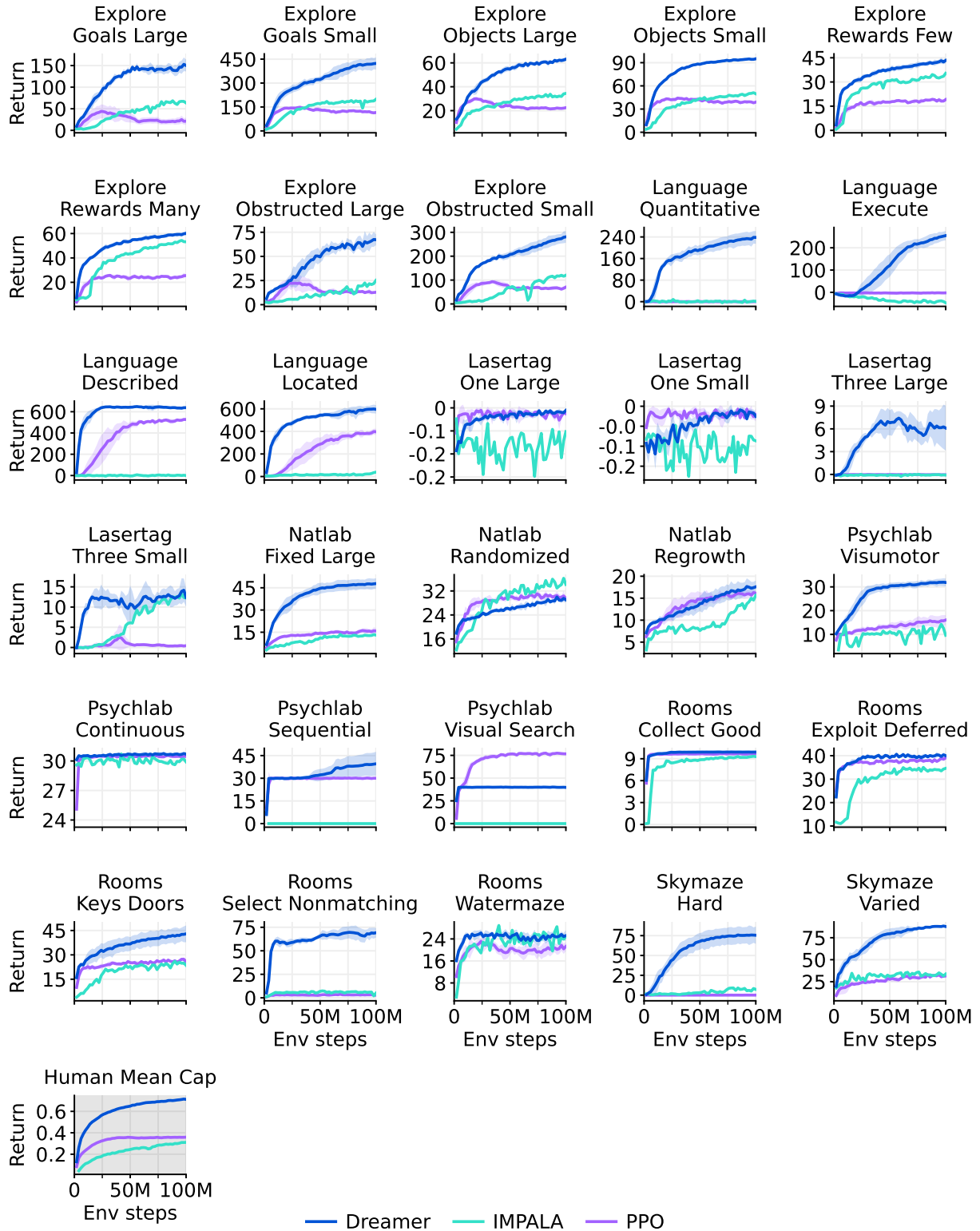


Figure 3: DMLab learning curves.

DMLab scores

Task	R2D2+	IMPALA	IMPALA	IMPALA	PPO	Dreamer
Environment steps	10B	10B	1B	100M	100M	100M
Explore Goal Locations Large	174.7	316.0	137.8	64.2	21.2	150.5
Explore Goal Locations Small	460.7	482.0	302.8	196.1	115.1	423.3
Explore Object Locations Large	60.6	91.0	55.1	34.3	22.5	62.9
Explore Object Locations Small	83.7	100.4	75.9	50.6	38.9	94.8
Explore Object Rewards Few	80.7	92.6	46.9	34.6	19.0	43.1
Explore Object Rewards Many	75.8	89.4	68.5	53.3	25.5	60.2
Explore Obstructed Goals Large	95.5	102.0	57.9	23.7	12.9	67.2
Explore Obstructed Goals Small	311.9	372.0	214.9	118.0	70.4	280.2
Language Answer Quantitative Question	344.4	362.0	304.7	1.0	0.3	238.5
Language Execute Random Task	497.4	465.4	140.8	-44.4	-2.5	254.7
Language Select Described Object	617.6	664.0	618.4	0.2	526.9	637.8
Language Select Located Object	772.8	731.4	413.0	38.2	397.5	595.9
Lasertag One Opponent Large	0.0	0.0	0.0	-0.1	0.0	0.0
Lasertag One Opponent Small	31.8	0.0	0.0	-0.1	0.0	0.0
Lasertag Three Opponents Large	28.6	32.2	10.4	-0.1	0.0	6.2
Lasertag Three Opponents Small	49.0	57.2	37.1	12.1	0.4	12.5
Natlab Fixed Large Map	60.6	63.4	53.8	13.0	15.8	47.7
Natlab Varying Map Randomized	42.4	47.0	40.5	35.3	29.6	29.4
Natlab Varying Map Regrowth	24.6	34.0	25.5	15.3	16.3	17.5
Psychlab Arbitrary Visuomotor Mapping	33.1	38.4	16.5	11.9	16.0	31.9
Psychlab Continuous Recognition	30.0	28.6	30.0	30.1	30.5	30.7
Psychlab Sequential Comparison	30.0	29.6	0.0	0.0	30.0	39.5
Psychlab Visual Search	79.9	80.0	0.0	0.0	76.6	39.9
Rooms Collect Good Objects Test	9.9	10.0	9.9	9.3	9.7	9.9
Rooms Exploit Deferred Effects Test	38.1	62.2	37.6	34.5	39.0	40.0
Rooms Keys Doors Puzzle	46.2	54.6	36.9	24.2	26.0	43.1
Rooms Select Nonmatching Object	63.6	39.0	63.2	4.0	2.7	69.0
Rooms Watermaze	49.0	47.0	50.1	23.6	21.2	25.2
Skymaze Irreversible Path Hard	76.0	80.0	46.4	7.7	0.0	75.8
Skymaze Irreversible Path Varied	76.0	100.0	69.8	32.7	31.3	87.9
Human mean capped (%)	85.4	85.1	66.3	31.0	35.9	71.3

Table 3: DMLab scores at 100M environment steps and at larger budgets. The IMPALA agent corresponds to “IMPALA (deep)” presented by Kapturowski et al. (2018) who made the learning curves available.

Atari100k learning curves

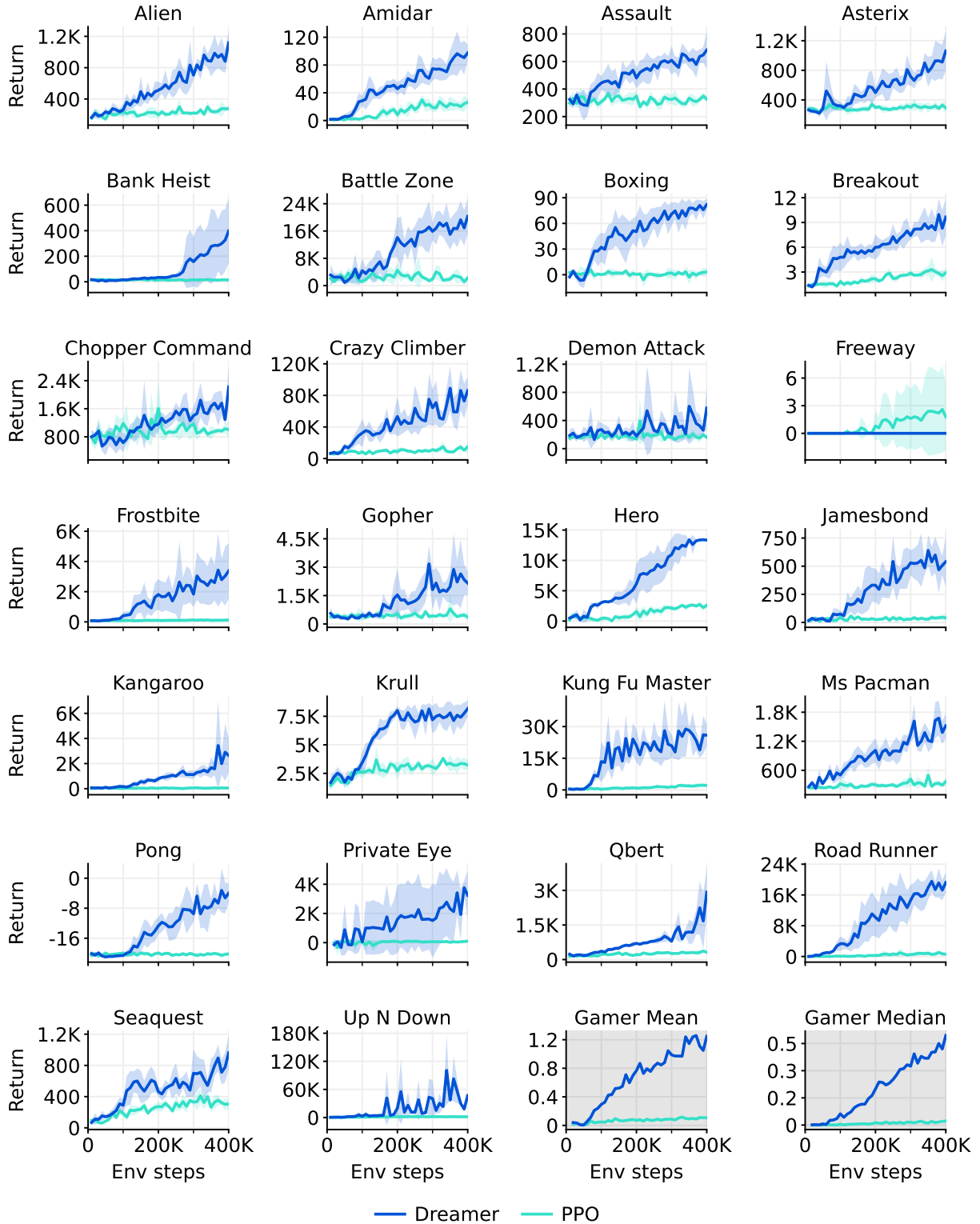


Figure 4: Atari100k learning curves.

Atari100k scores

Task	Random	Human	PPO	SimPLe	SPR	TWM	IRIS	Dreamer
Environment steps	—	—	400K	400K	400K	400K	400K	400K
Alien	228	7128	276	617	842	675	420	1118
Amidar	6	1720	26	74	180	122	143	97
Assault	222	742	327	527	566	683	1524	683
Asterix	210	8503	292	1128	962	1117	854	1062
Bank Heist	14	753	14	34	345	467	53	398
Battle Zone	2360	37188	2233	4031	14834	5068	13074	20300
Boxing	0	12	3	8	36	78	70	82
Breakout	2	30	3	16	20	20	84	10
Chopper Command	811	7388	1005	979	946	1697	1565	2222
Crazy Climber	10780	35829	14675	62584	36700	71820	59324	86225
Demon Attack	152	1971	160	208	518	350	2034	577
Freeway	0	30	2	17	19	24	31	0
Frostbite	65	4335	127	237	1171	1476	259	3377
Gopher	258	2412	368	597	661	1675	2236	2160
Hero	1027	30826	2596	2657	5859	7254	7037	13354
Jamesbond	29	303	41	100	366	362	463	540
Kangaroo	52	3035	55	51	3617	1240	838	2643
Krull	1598	2666	3222	2205	3682	6349	6616	8171
Kung Fu Master	258	22736	2090	14862	14783	24555	21760	25900
Ms Pacman	307	6952	366	1480	1318	1588	999	1521
Pong	−21	15	−20	13	−5	19	15	−4
Private Eye	25	69571	100	35	86	87	100	3238
Qbert	164	13455	317	1289	866	3331	746	2921
Road Runner	12	7845	602	5641	12213	9109	9615	19230
Seaquest	68	42055	305	683	558	774	661	962
Up N Down	533	11693	1502	3350	10859	15982	3546	46910
Gamer mean (%)	0	100	11	33	62	96	105	125
Gamer median (%)	0	100	2	13	40	51	29	49

Table 4: Atari100k scores at 400K environment steps, corresponding to 100k agent steps.

Proprio Control Suite learning curves

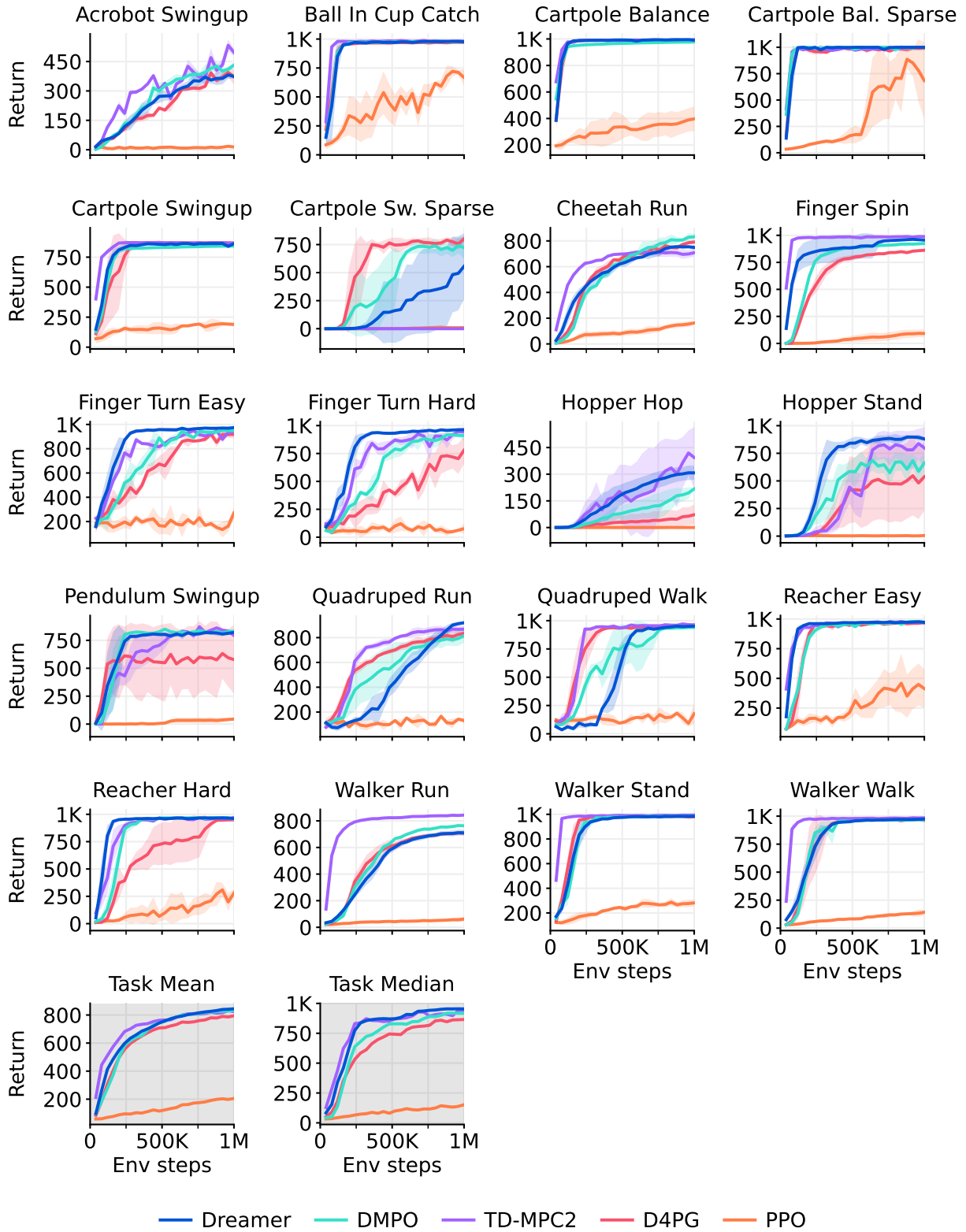


Figure 5: DeepMind Control Suite learning curves under proprioceptive inputs.

Proprio Control Suite scores

Task	PPO	D4PG	TD-MPC2	DMPO	Dreamer
Environment steps	1M	1M	1M	1M	1M
Acrobot Swingup	14	372	496	432	378
Ball In Cup Catch	668	972	982	979	978
Cartpole Balance	397	990	994	977	993
Cartpole Balance Sparse	684	990	1000	996	1000
Cartpole Swingup	189	869	864	842	852
Cartpole Swingup Sparse	9	801	0	722	525
Cheetah Run	163	791	708	832	752
Finger Spin	92	862	988	924	958
Finger Turn Easy	270	919	937	950	972
Finger Turn Hard	76	779	944	910	963
Hopper Hop	0	72	394	217	308
Hopper Stand	5	538	799	663	887
Pendulum Swingup	45	578	793	832	817
Quadruped Run	131	835	868	816	916
Quadruped Walk	178	953	963	940	951
Reacher Easy	414	970	967	975	974
Reacher Hard	280	950	971	956	963
Walker Run	62	707	843	763	711
Walker Stand	282	980	994	983	984
Walker Walk	141	963	985	967	973
Task mean	205	795	825	834	843
Task median	152	865	941	917	954

Table 5: DeepMind Control Suite scores under proprioceptive inputs.

Visual Control Suite learning curves

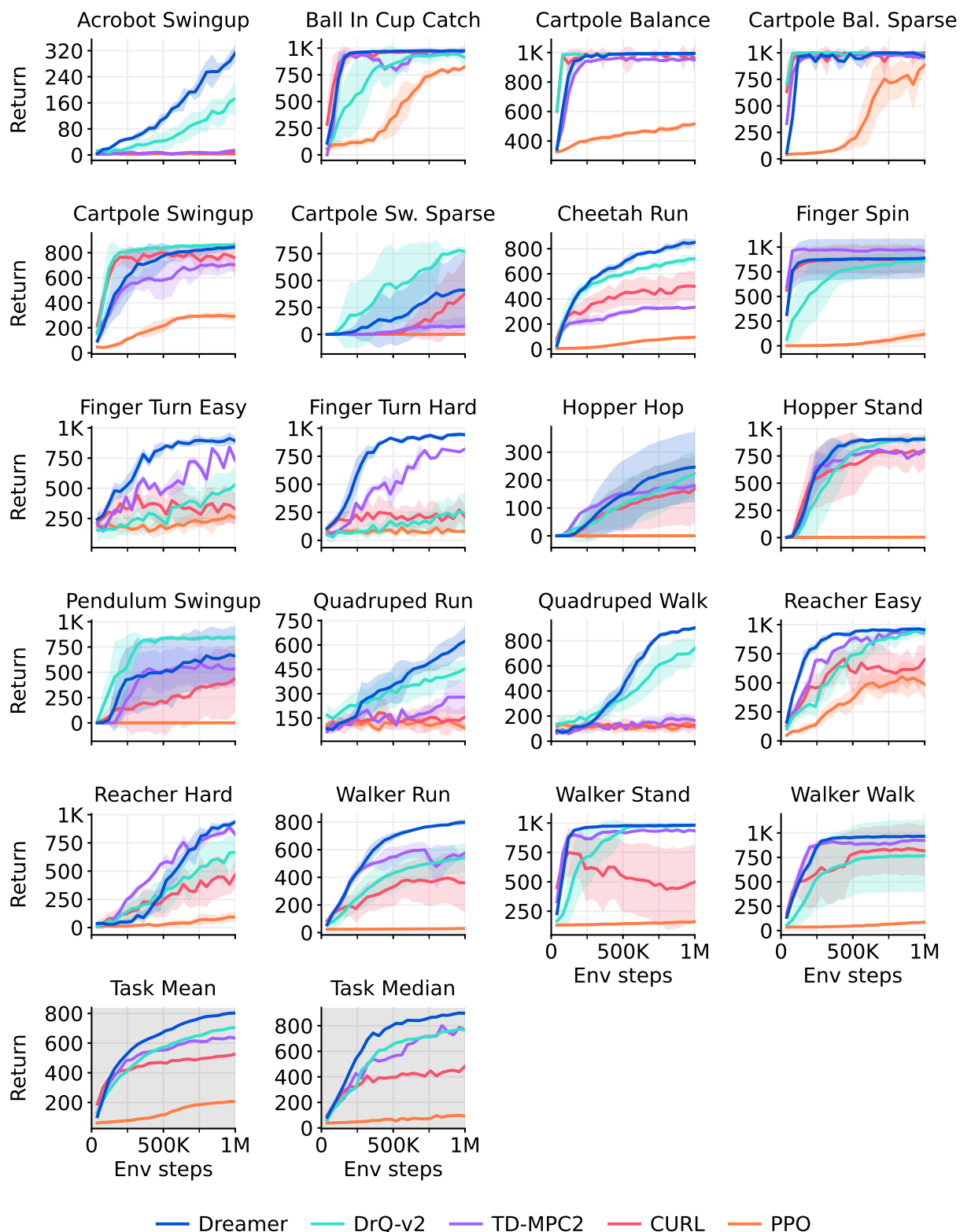


Figure 6: DeepMind Control Suite learning curves under visual inputs.

Visual Control Suite scores

Task	PPO	SAC	TD-MPC2	DrQ-v2	Dreamer
Environment steps	1M	1M	1M	1M	1M
Acrobot Swingup	3	4	14	166	300
Ball In Cup Catch	829	176	971	920	975
Cartpole Balance	516	937	949	992	994
Cartpole Balance Sparse	881	956	968	981	975
Cartpole Swingup	290	706	703	863	843
Cartpole Swingup Sparse	1	149	76	773	411
Cheetah Run	95	20	334	716	846
Finger Spin	118	291	961	861	886
Finger Turn Easy	253	200	730	525	903
Finger Turn Hard	79	94	814	247	944
Hopper Hop	0	0	180	221	246
Hopper Stand	4	5	805	912	904
Pendulum Swingup	1	592	534	845	668
Quadruped Run	88	54	279	450	614
Quadruped Walk	112	49	165	726	897
Reacher Easy	487	67	930	944	960
Reacher Hard	94	7	827	670	922
Walker Run	30	27	575	539	797
Walker Stand	161	143	934	979	982
Walker Walk	87	40	923	768	966
Task mean	206	226	634	705	802
Task median	94	81	767	770	900

Table 6: DeepMind Control Suite scores under visual inputs.

BSuite performance spectrum

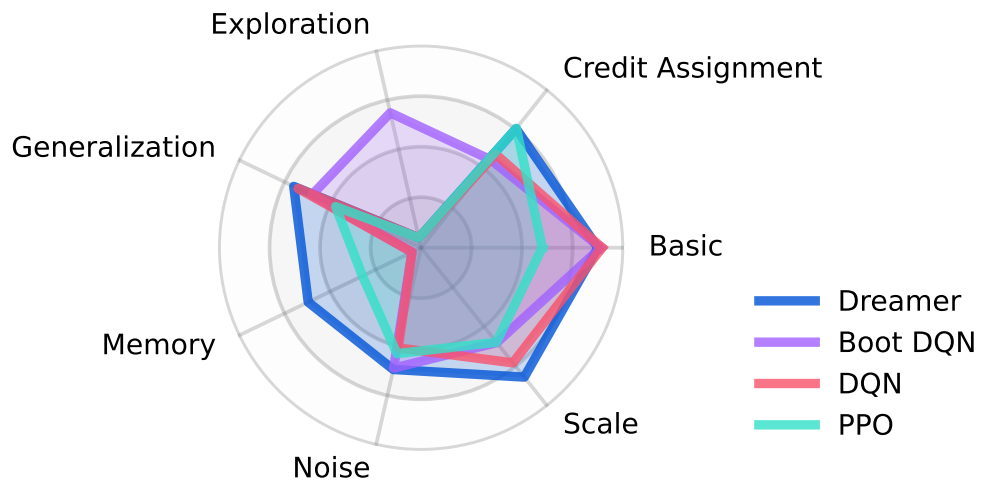


Figure 7: BSuite scores visualized by category. Dreamer exceeds previous methods in the categories scale and memory. The scale category measure robustness to reward scales.

BSuite scores

Task	Random	PPO	AC-RNN	DQN	Boot DQN	Dreamer
Bandit	0.00	0.38	1.00	0.93	0.98	0.96
Bandit Noise	0.00	0.61	0.63	0.71	0.80	0.75
Bandit Scale	0.00	0.39	0.60	0.74	0.83	0.78
Cartpole	0.04	0.84	0.40	0.85	0.69	0.93
Cartpole Noise	0.04	0.77	0.20	0.82	0.69	0.93
Cartpole Scale	0.04	0.83	0.12	0.72	0.65	0.92
Cartpole Swingup	0.00	0.00	0.00	0.00	0.15	0.03
Catch	0.00	0.91	0.87	0.92	0.99	0.96
Catch Noise	0.00	0.54	0.27	0.58	0.68	0.53
Catch Scale	0.00	0.90	0.17	0.85	0.65	0.94
Deep Sea	0.00	0.00	0.00	0.00	1.00	0.00
Deep Sea Stochastic	0.00	0.00	0.00	0.00	0.90	0.00
Discounting Chain	0.20	0.24	0.39	0.25	0.22	0.40
Memory Len	0.00	0.17	0.70	0.04	0.04	0.65
Memory Size	0.00	0.47	0.29	0.00	0.00	0.59
Mnist	0.05	0.77	0.56	0.85	0.85	0.61
Mnist Noise	0.05	0.41	0.22	0.38	0.34	0.34
Mnist Scale	0.05	0.76	0.09	0.49	0.31	0.55
Mountain Car	0.10	0.10	0.10	0.93	0.93	0.92
Mountain Car Noise	0.10	0.10	0.10	0.89	0.82	0.87
Mountain Car Scale	0.10	0.10	0.10	0.85	0.56	0.90
Umbrella Distract	0.00	1.00	0.09	0.30	0.26	0.74
Umbrella Length	0.00	0.87	0.43	0.39	0.39	0.78
Basic	0.04	0.60	0.58	0.90	0.89	0.88
Credit assignment	0.03	0.76	0.37	0.59	0.56	0.75
Exploration	0.00	0.00	0.00	0.00	0.68	0.01
Generalization	0.06	0.47	0.19	0.68	0.60	0.70
Memory	0.00	0.32	0.49	0.02	0.02	0.62
Noise	0.02	0.54	0.24	0.51	0.61	0.62
Scale	0.04	0.60	0.22	0.73	0.60	0.82
Task mean (%)	3	49	32	54	60	66
Category mean (%)	3	47	30	49	57	63

Table 7: BSuite scores for each task averaged over environment configurations, as well as aggregated performance by category and over all tasks.

Normalization ablations

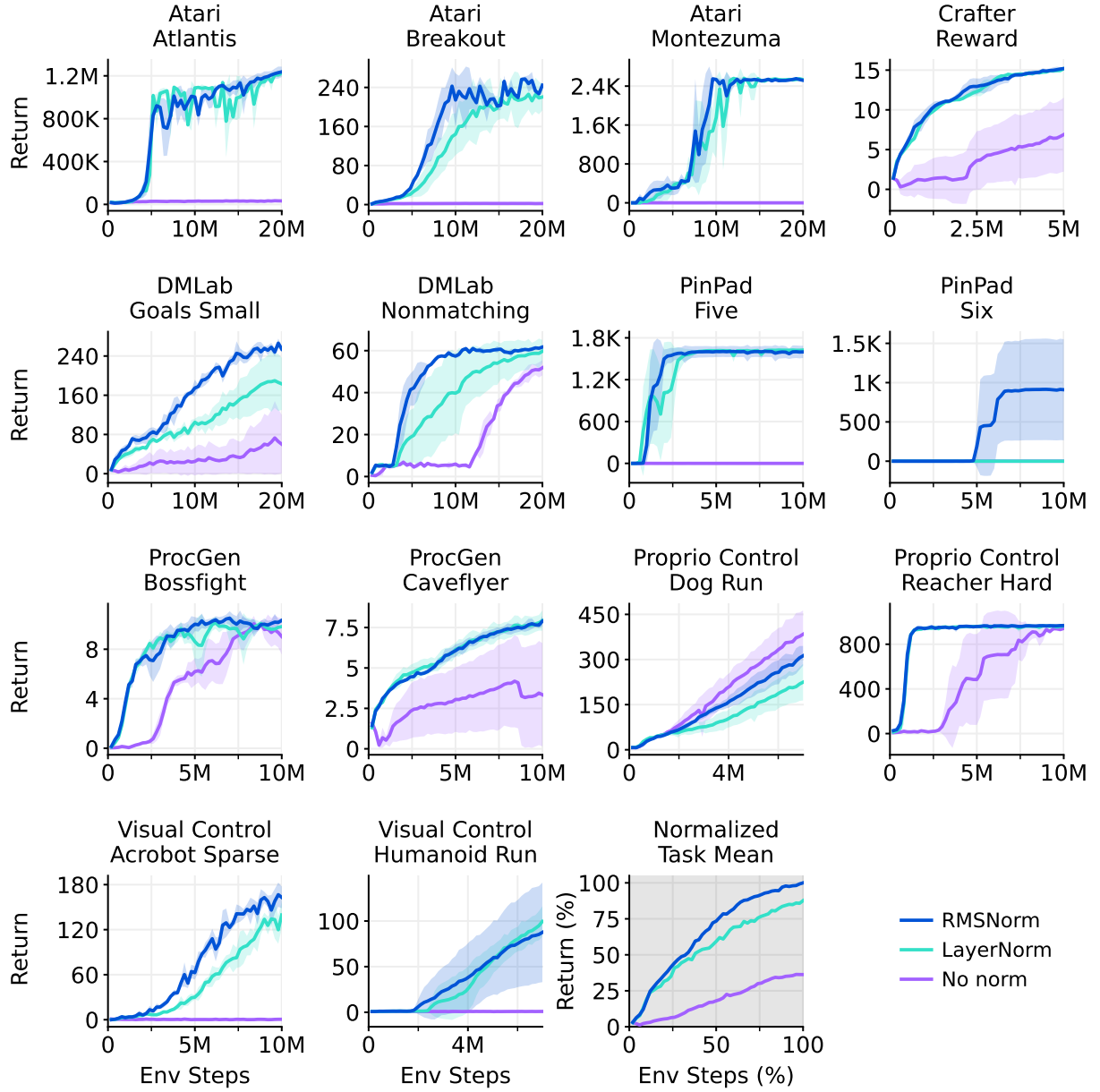


Figure 8: Individual learning curves for the normalization ablation experiment. Both LayerNorm and RMSNorm stabilize the learning dynamics of Dreamer. In practice, we opt for RMSNorm due to its simplicity.

Learning signal ablations

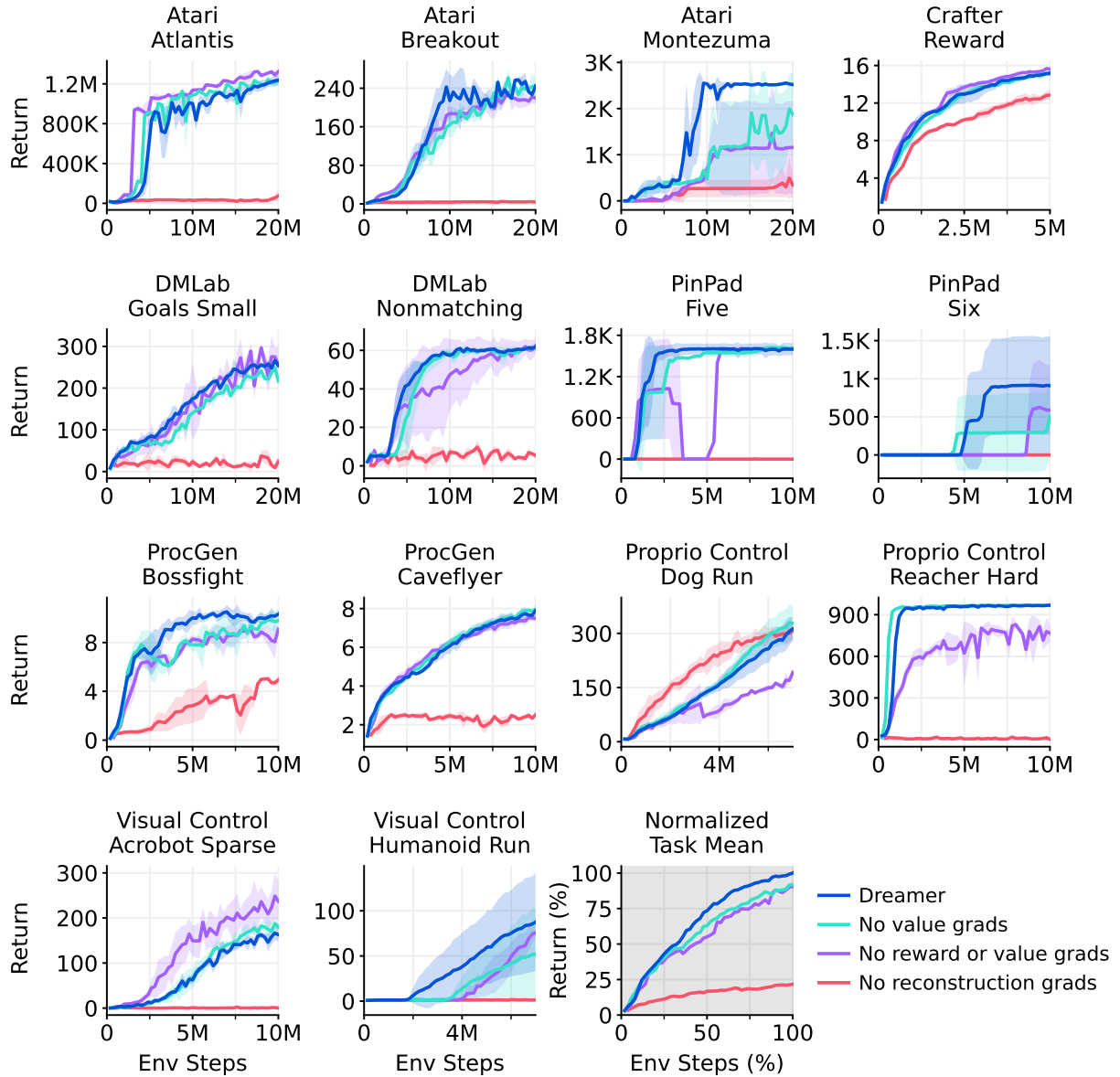


Figure 9: Individual learning curves for the learning signal ablation experiment. Dreamer relies predominantly on the undersupervised reconstruction objective of its world model and additional reward and value gradients further improve performance on a subset of tasks.

Robustness ablations

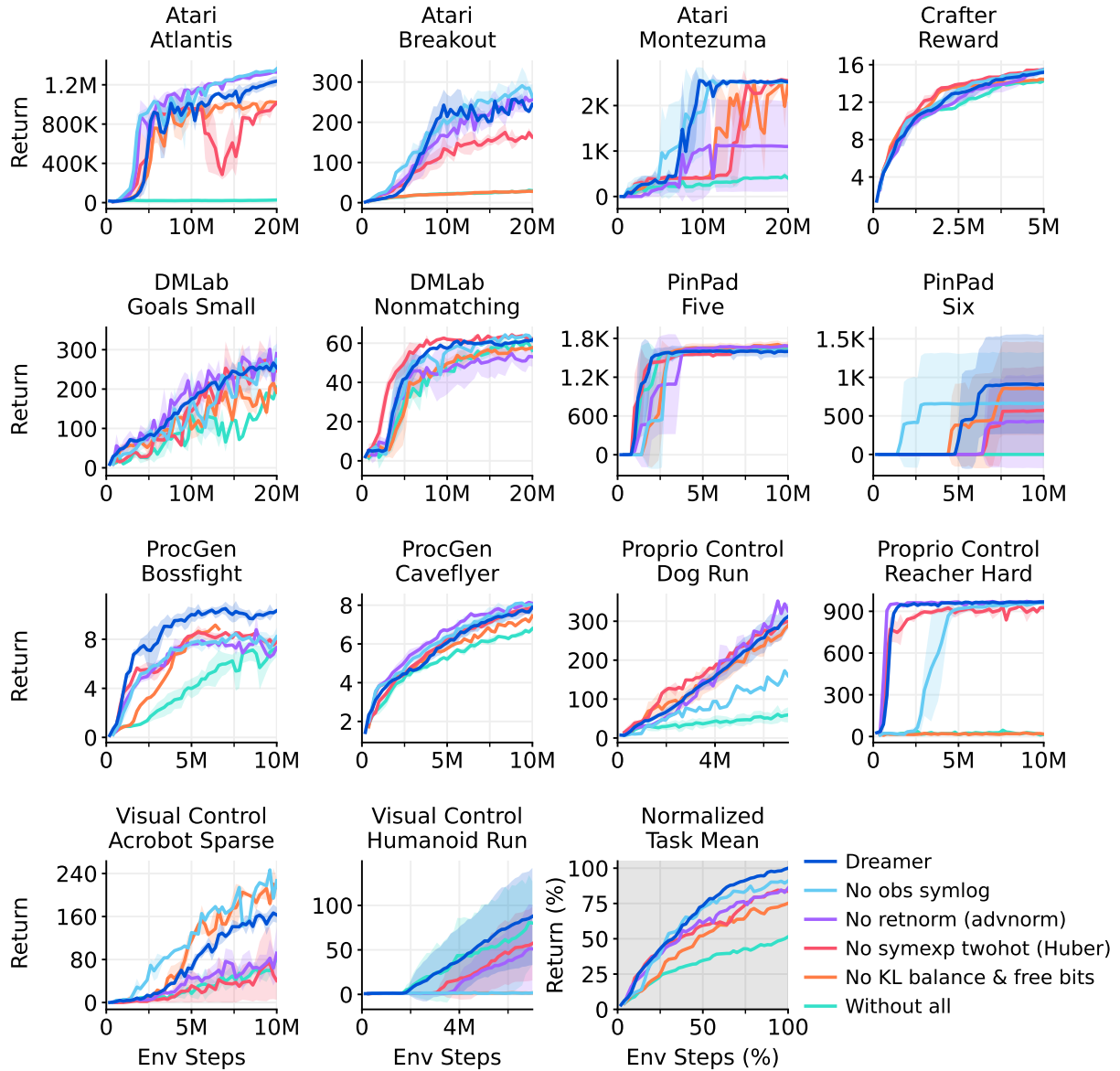


Figure 10: Individual learning curves for the robustness ablation experiment. All robustness techniques contribute to the overall performance of Dreamer, although each individual technique may only improve the performance on a subset of the tasks.