

Reinforcement Learning

Training Gym's LunarLander with different scenarios and models



Reinforcement Learning

FEUP-M.EIC009-2022/2023-2S

Group B

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Proposal

We want to explore:

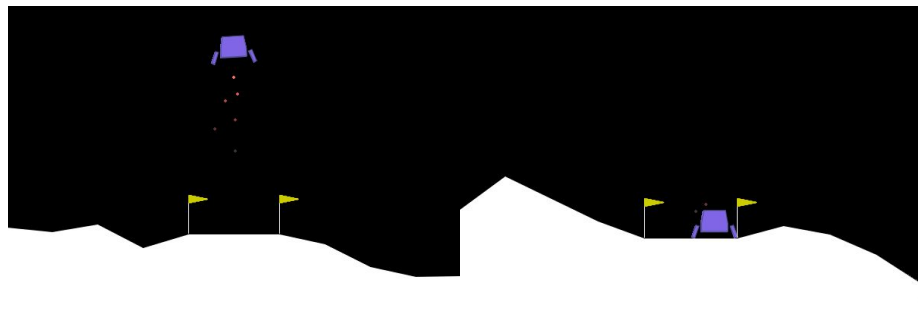
- Stable baselines models and capacities.
- Gym API and environments.
- Reinforcement Learning.

Approach:

- Choose a base environment.
- Test different models.
- Test different environments.

Environment - Lunar Lander V2

- Action space \rightarrow Box(-1,1,(2,))
 - Two floats between -1 and 1
 - Main engine(fires if value > 0.5)
 - Side engines (left fires if value > 0.5, right fires if value < 0.5)
- Observation space \rightarrow (8,)
 - Lander coordinates, and velocity
 - Lander angle and angular velocity
 - If the legs contact the ground or not
- State:
 - Initial: Center of screen, with a random velocity
 - During: observation
 - Done:
 - Lander crashes
 - Lander disappears
 - Lander at rest



- Rewards
 - Negative:
 - Firing engines
 - Moving away from landing pad
 - Crashing
 - Positive
 - Moving in the direction of the landing pad
 - Finishing at rest

Algorithms applied

- A2C - Advantage Actor-Critic
- DDPG - Deep Deterministic Policy Gradient
- PPO - Proximal Policy Optimization
- SAC - Soft Actor-Critic

Training

- Per model per environment
- 15 episodes of 10k iterations
- 150k iterations

Environments used

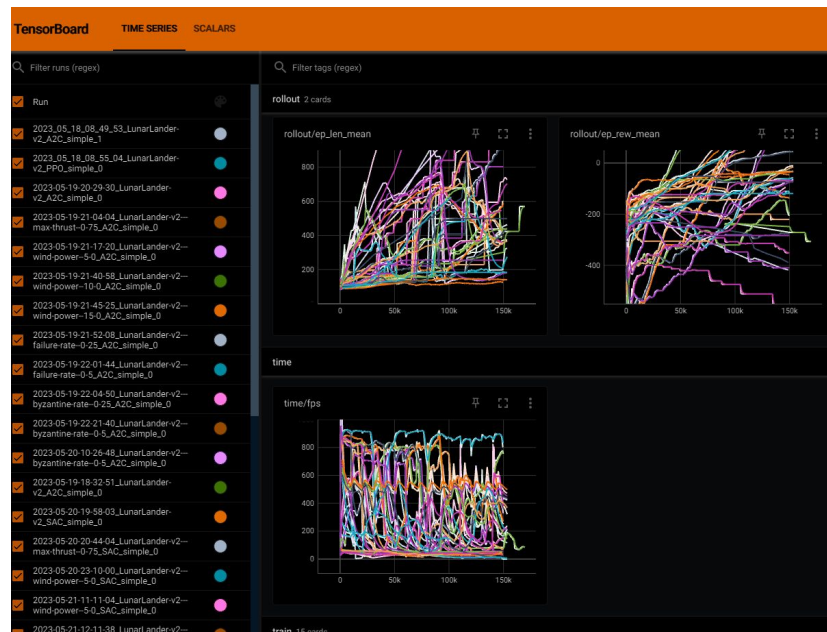
- Max thrust = 75%
- Wind power = 5
- Wind power = 10
- Wind power = 15
- Failure Rate = 5%
- Failure Rate = 10%
- Failure Rate = 25%
- Failure Rate = 50%
- Byzantine Rate = 25%
- Byzantine Rate = 50%

Visualization during and after training

- Models have a verbose option:
 - Too many prints cluttering the terminal.

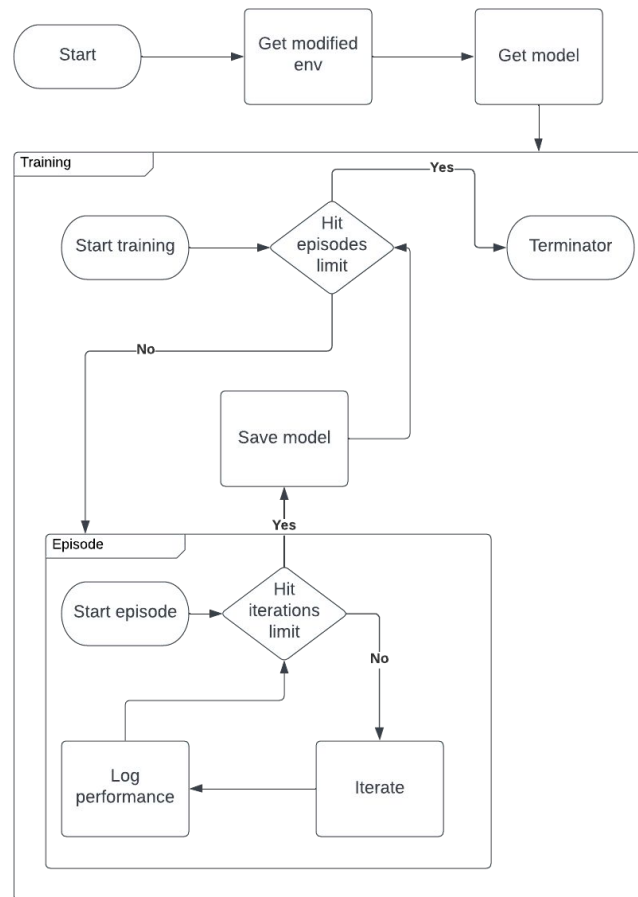
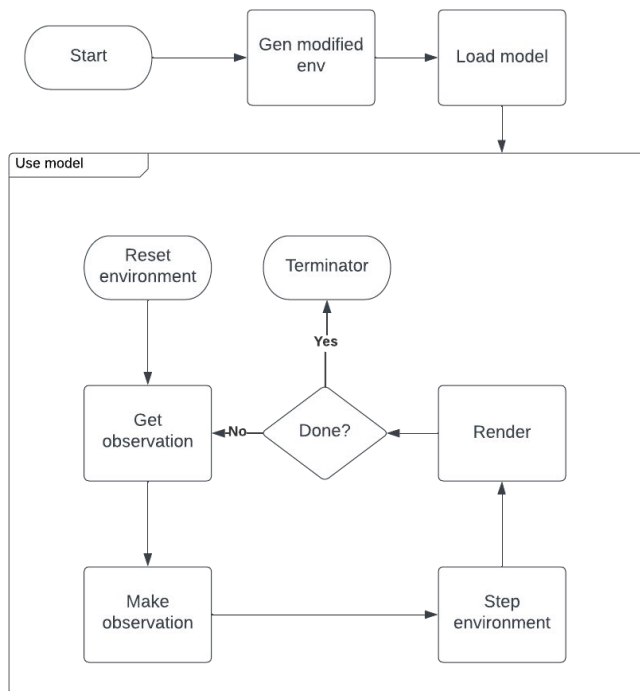
Implemented our own mechanisms:

- Training by episode:
 - Variable number of episodes and iterations per episode.
 - Episode conclusion is displayed.
- Tensorboard:
 - Real time logging.
 - After the fact visualization.
- Using models:
 - Validating learning.

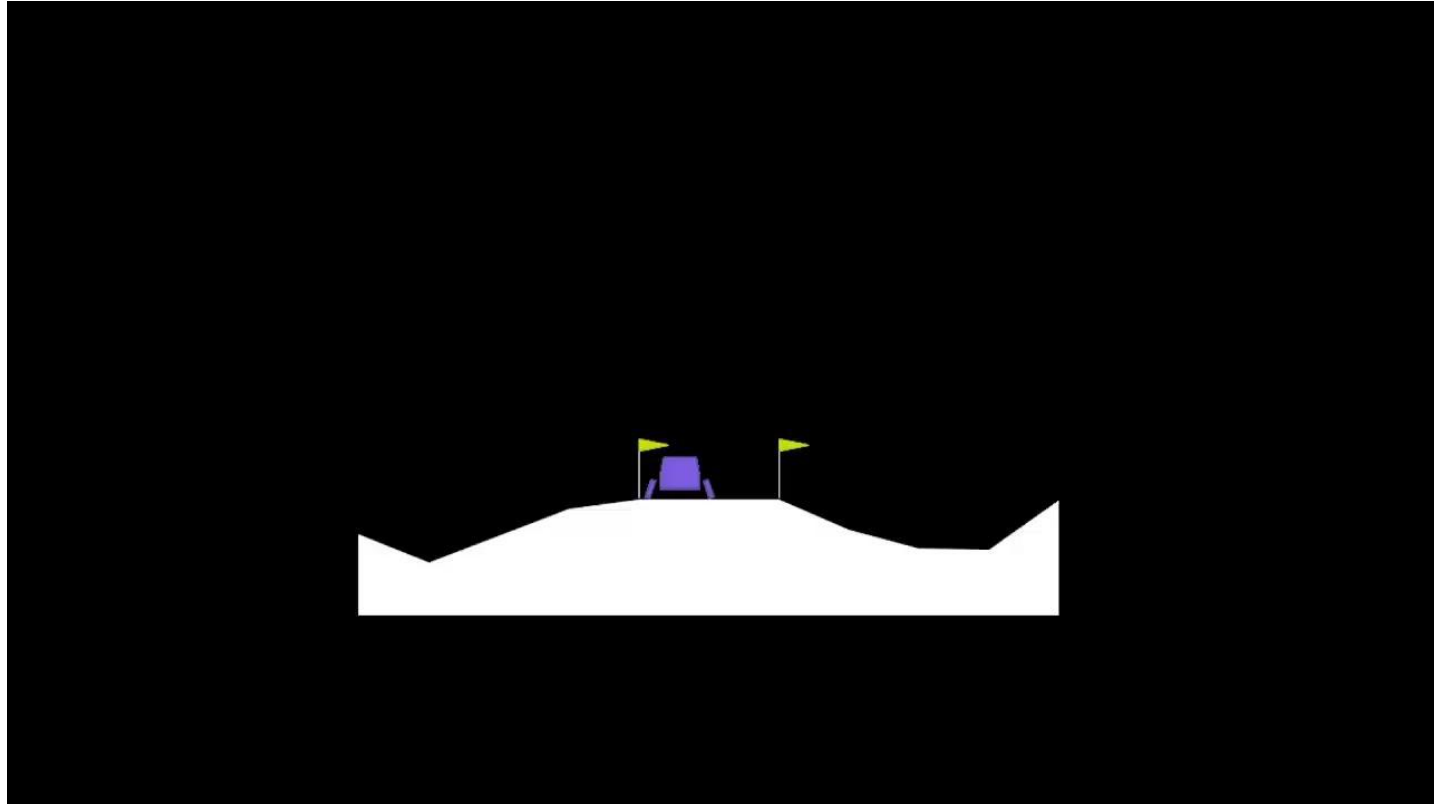


```
Starting training DDPG on LunarLander-v2---max-thrust--0-75
C:\Users\henri\AppData\Local\Programs\Python\Python310\lib\site-packages\stable_baselines3\common\saf
55_LunarLander-v2---max-thrust--0-75_DDPG_simple' does not exist. Will create it.
warnings.warn(f"Path '{path.parent}' does not exist. Will create it.")
Finished 1/15 episodes for DDPG in 2023-05-23-10-18-55_LunarLander-v2---max-thrust--0-75_DDPG_simple
Finished 2/15 episodes for DDPG in 2023-05-23-10-18-55_LunarLander-v2---max-thrust--0-75_DDPG_simple
Finished 3/15 episodes for DDPG in 2023-05-23-10-18-55_LunarLander-v2---max-thrust--0-75_DDPG_simple
Finished 4/15 episodes for DDPG in 2023-05-23-10-18-55_LunarLander-v2---max-thrust--0-75_DDPG_simple
Stopping training due to interrupt
Finished 5/15 episodes for DDPG in 2023-05-23-10-18-55_LunarLander-v2---max-thrust--0-75_DDPG_simple
Finished training routine
```

Pipeline



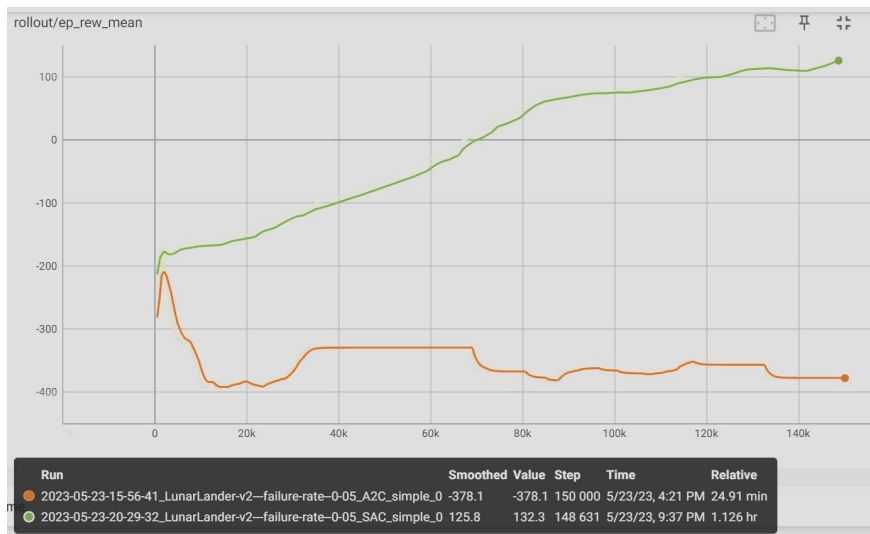
Demo



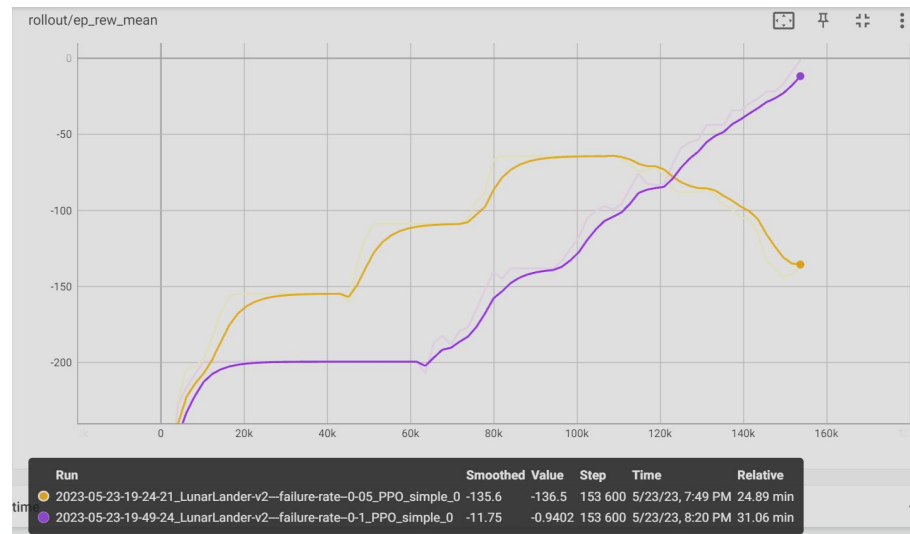
SAC with wind power 15.0

Analysis

- Comparing some models considering a failure rate in the environment:



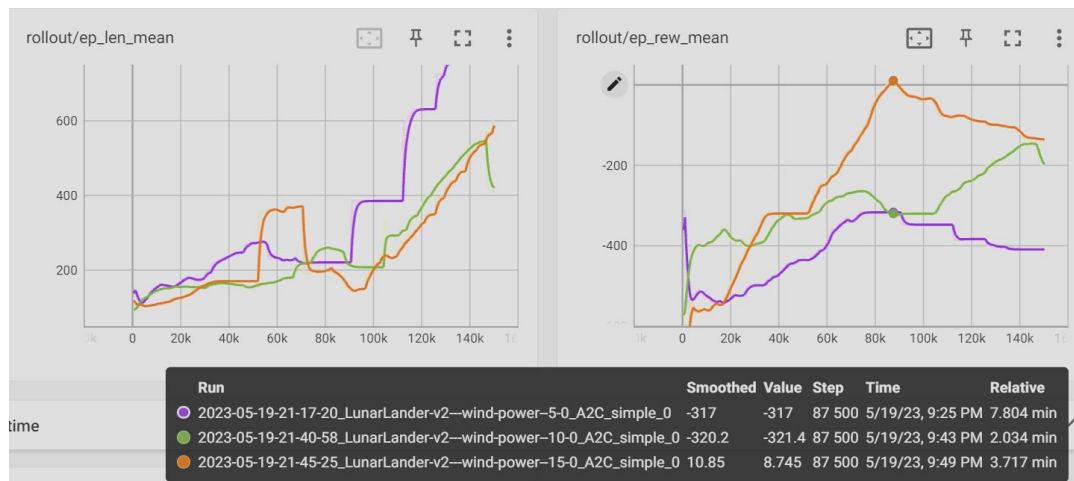
Orange: A2C Failure rate 0.05
Green: SAC Failure rate 0.05



Orange: PPO Failure rate 0.05
Purple: PPO Failure rate 0.1

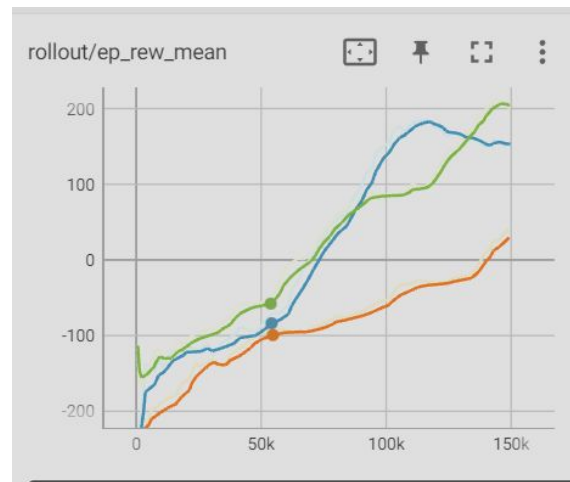
Analysis

- Comparing episode length



Purple: A2C Wind power 5.0
Green: A2C Wind power 10.0
Orange: A2C Wind power 15.0

- Comparing models



3 version of SAC

Conclusion and future work

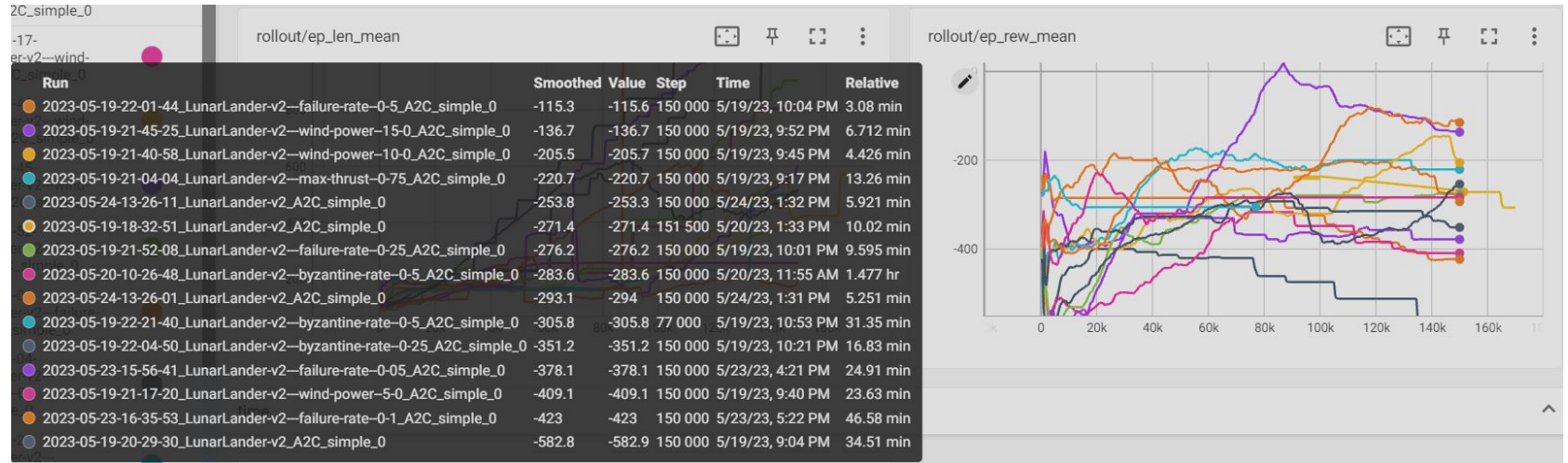
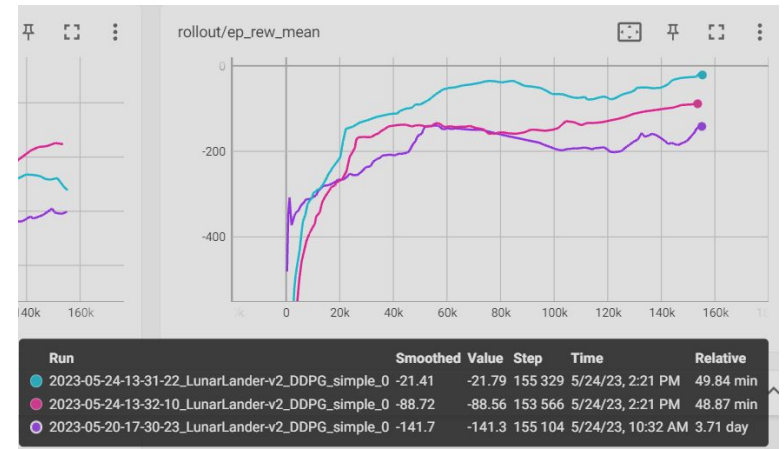
- There are libraries that make it incredibly easy to experiment with Reinforcement Learning:
 - It is even possible to get decent to good results on simple tasks.
- An adequate methodology can lead to better results and faster development.
- Training models can take a long time.
- Not all models are created equal:
 - There is enough randomization that the same conditions may lead to slightly different results.
- Future work:
 - Obtain positive result for every environment - is that even possible?
 - Find ways to speed up training or prevent local optima - adjust reward?

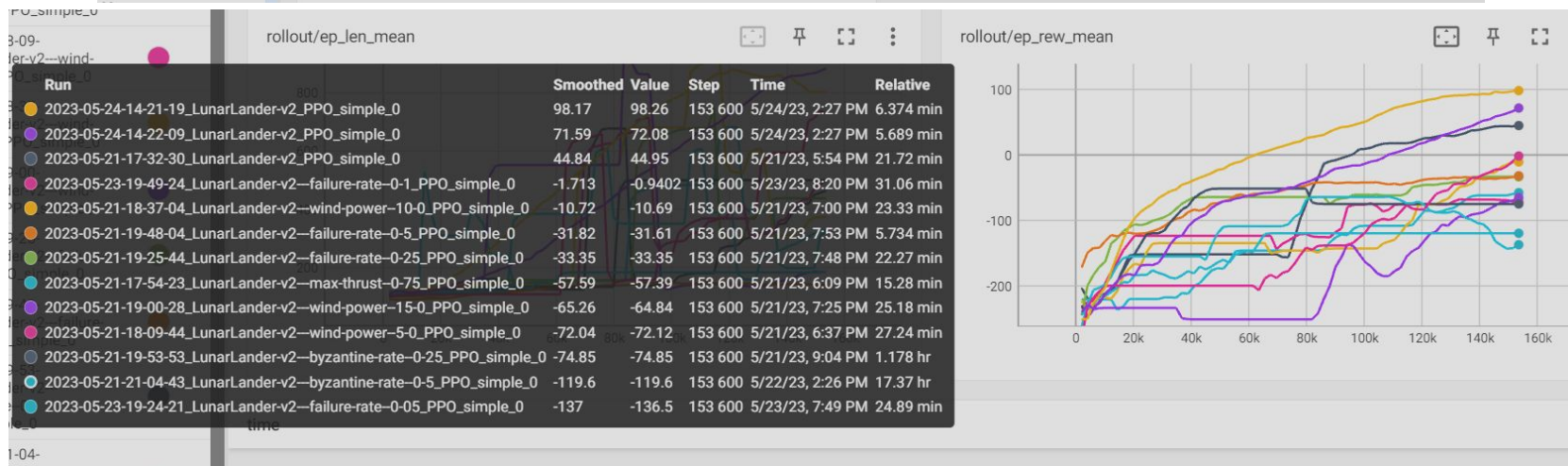
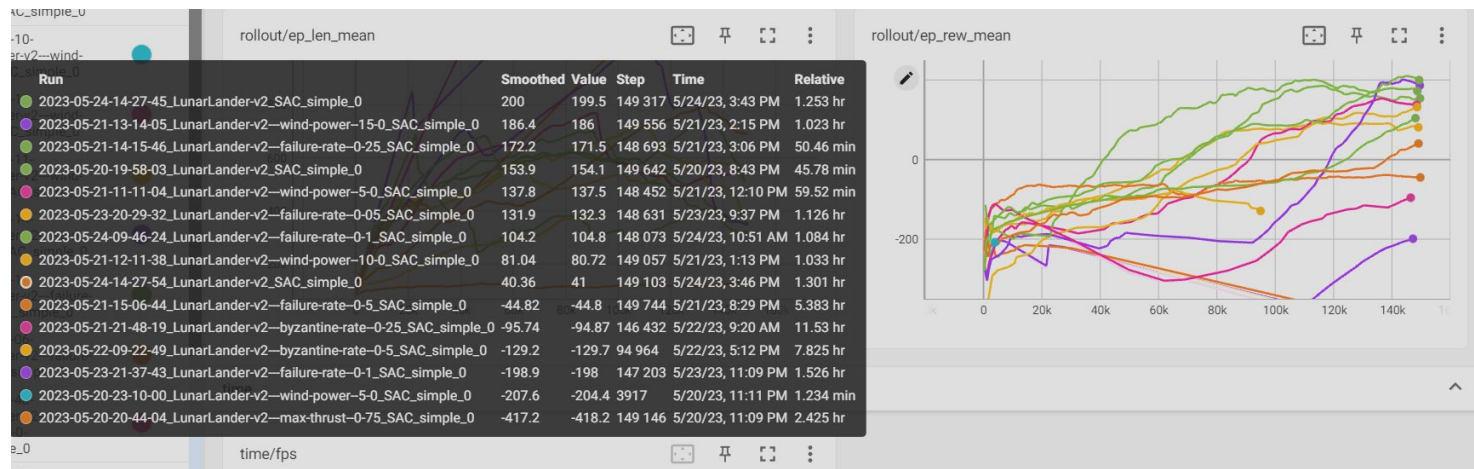
Annexes

Note: The README.md in the code has instructions, other details on running, and the available features.

Other results

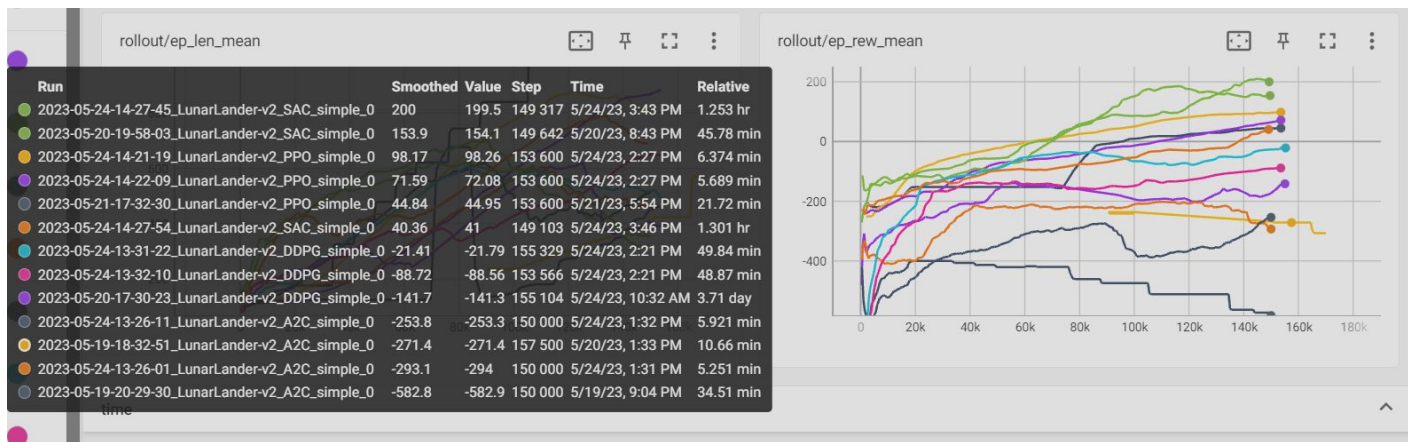
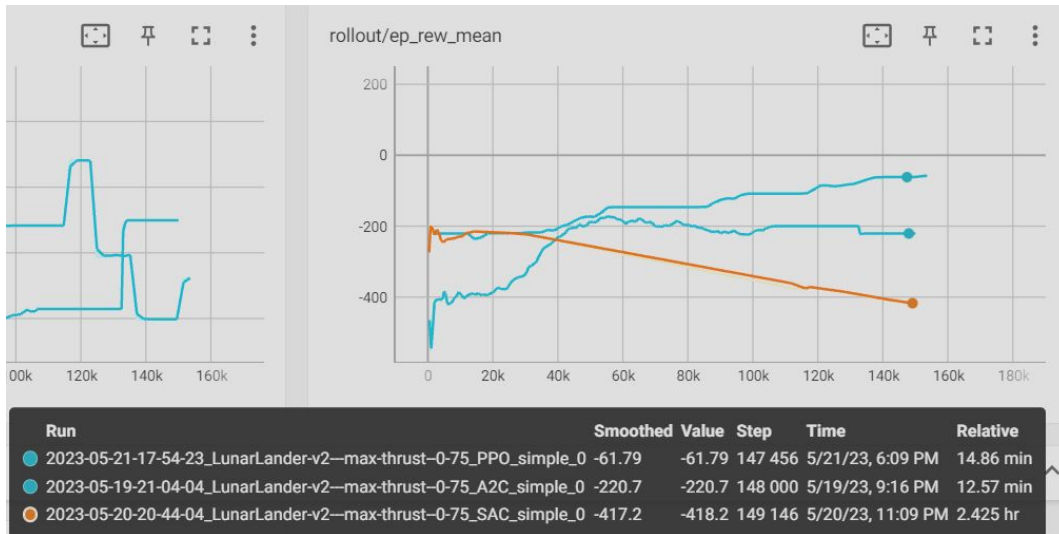
- Reward graphs per model

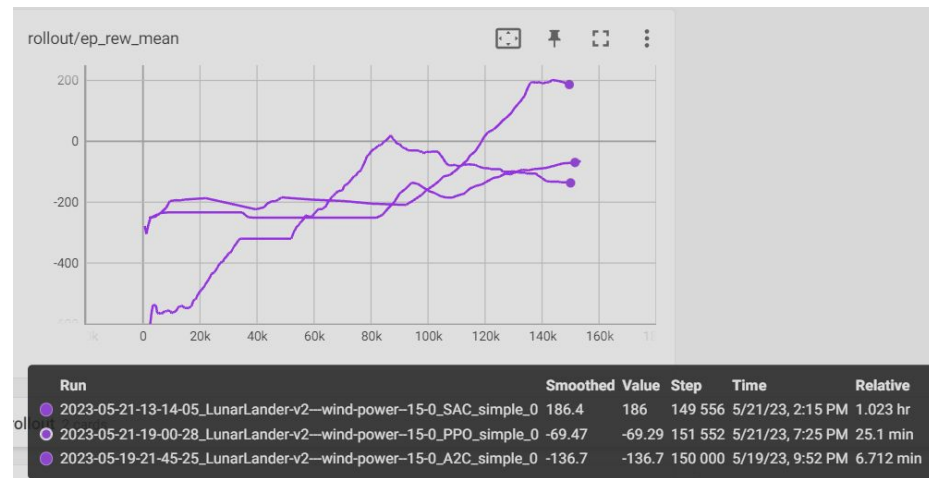
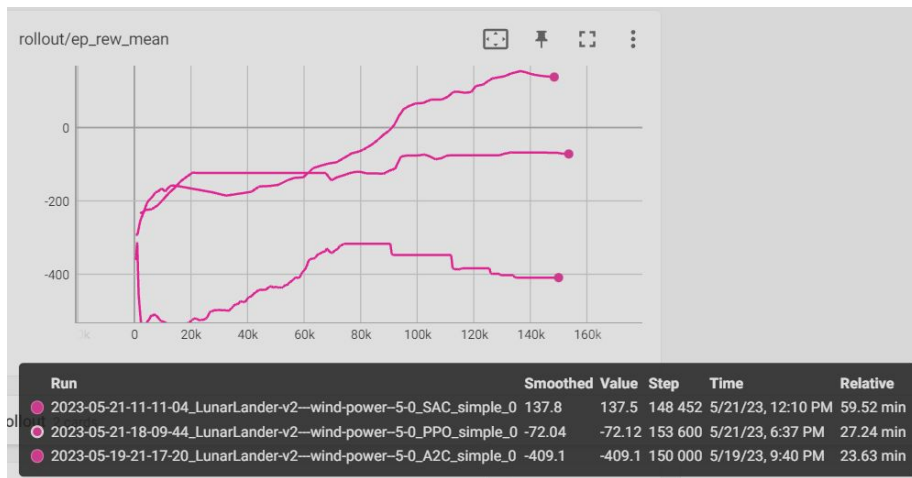


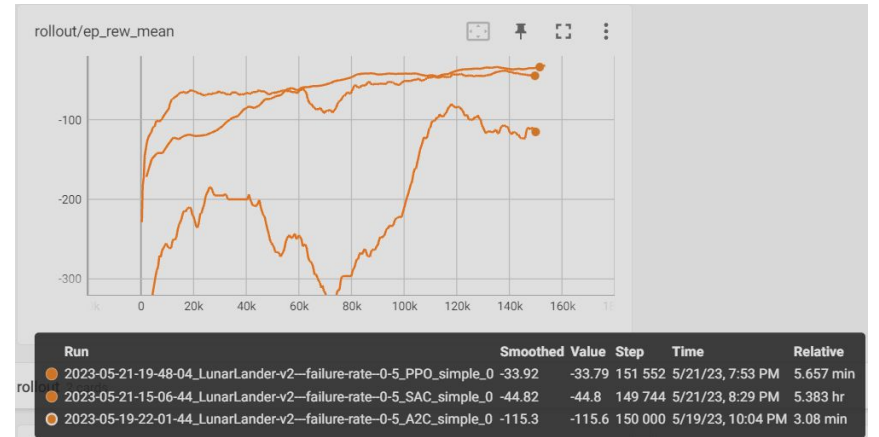
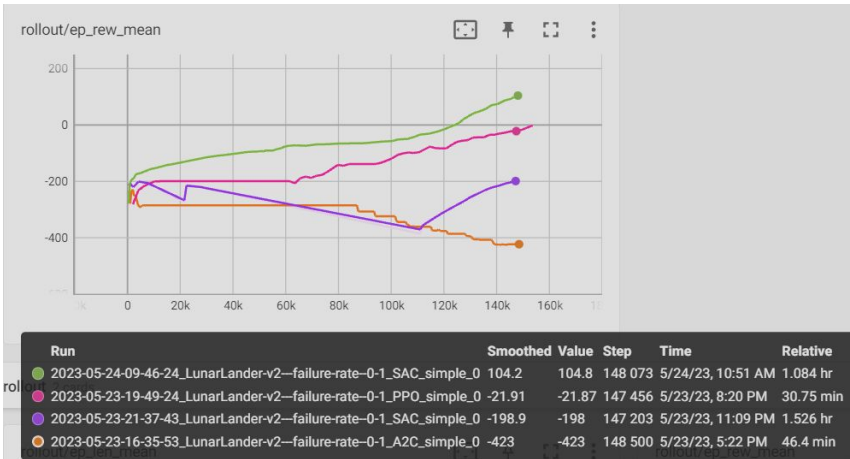
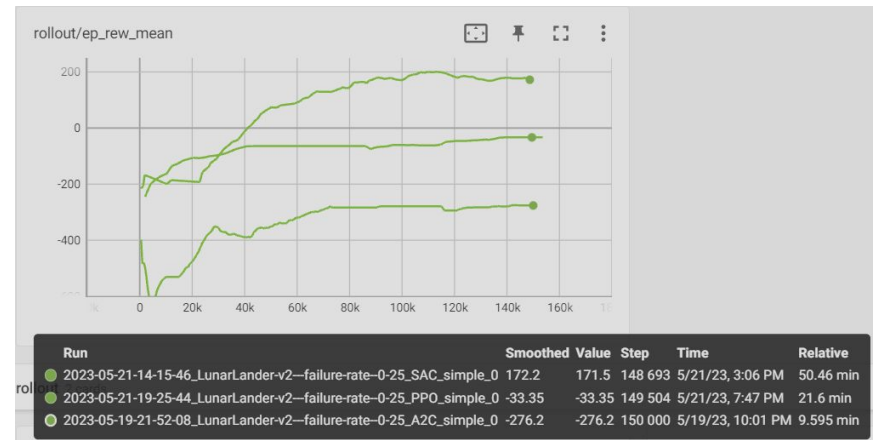
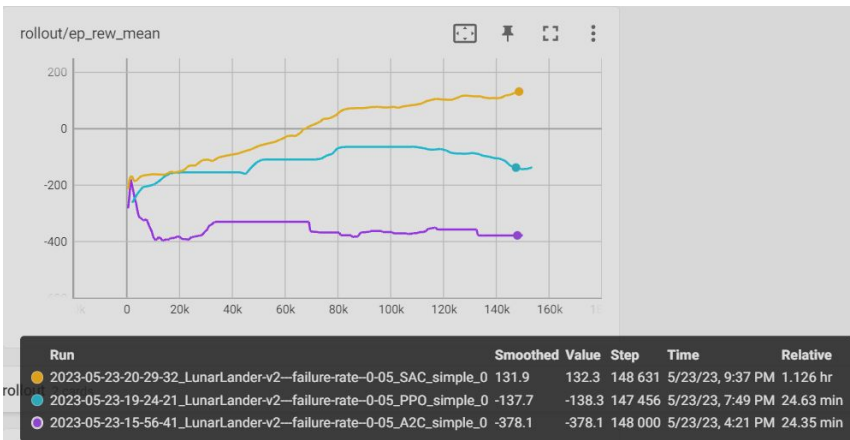


Other results

- Reward graphs per environment







Detailed data - By model

A2C								
	env	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
0	LunarLander-v2	170000	-236.452652	-306.650452	-306.650452	570.500000	233.830002	570.500000
1	LunarLander-v2	150000	-400.023376	-754.365417	-582.866028	1449.329956	97.800003	1449.329956
5	LunarLander-v2	150000	-202.001465	-410.694122	-293.969849	612.530029	109.500000	450.079987
6	LunarLander-v2	150000	-253.250946	-636.849670	-253.250946	622.510010	106.864868	518.020020
33	LunarLander-v2---byzantine-rate--0-25	150000	-292.394989	-484.809967	-351.236298	1125.810059	98.500000	1125.810059
36	LunarLander-v2---byzantine-rate--0-5	77000	-234.007477	-310.266235	-305.794556	159.727280	114.750000	159.727280
37	LunarLander-v2---byzantine-rate--0-5	150000	-227.212433	-436.405945	-283.610931	255.330002	93.800003	255.330002
40	LunarLander-v2---failure-rate--0-05	150000	-173.486984	-397.261200	-378.094635	1158.979980	124.333336	1158.979980
45	LunarLander-v2---failure-rate--0-1	150000	-228.266678	-425.797180	-422.983215	2429.694336	113.000000	1922.657837
27	LunarLander-v2---failure-rate--0-25	150000	-275.416351	-626.930847	-276.242920	903.770020	89.080002	903.770020
30	LunarLander-v2---failure-rate--0-5	150000	-79.922859	-407.965912	-115.642410	239.330002	84.727272	187.190002
13	LunarLander-v2---max-thrust--0-75	150000	-172.319672	-551.877502	-220.667114	498.779999	108.250000	498.779999
21	LunarLander-v2---wind-power--10-0	150000	-143.933395	-572.376404	-205.673065	547.859985	93.599998	417.170013
24	LunarLander-v2---wind-power--15-0	150000	19.821482	-668.412476	-136.707581	601.859985	101.367348	601.859985
17	LunarLander-v2---wind-power--5-0	150000	-308.635834	-563.784119	-409.120087	804.000000	105.878784	804.000000

DDPG								
	env	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
2	LunarLander-v2	155104	-139.497116	-480.652466	-141.257126	623.659973	82.500	398.799988
7	LunarLander-v2	155329	-16.730101	-833.337769	-21.794130	608.609985	88.000	477.000000
8	LunarLander-v2	153566	-88.562225	-853.943726	-88.562225	652.830017	78.000	647.059998
41	LunarLander-v2---failure-rate--0-05	1792	-272.985596	-450.430786	-371.169525	115.833336	75.250	112.000000
42	LunarLander-v2---failure-rate--0-05	157548	-173.591629	-951.069641	-220.480194	758.400024	102.750	370.459991
15	LunarLander-v2---max-thrust--0-75	79358	-151.214157	-344.607300	-172.113037	342.880005	79.625	338.839996

PPO								
	env	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
4	LunarLander-v2	153600	44.948971	-220.450470	44.948971	677.580017	107.052635	464.160004
9	LunarLander-v2	153600	98.261688	-250.998184	98.261688	881.880005	105.526314	802.080017
10	LunarLander-v2	153600	72.078133	-240.324020	72.078133	884.369995	105.111115	884.369995
34	LunarLander-v2---byzantine-rate--0-25	153600	-51.564663	-248.864975	-74.847038	499.429993	99.500000	499.429993
38	LunarLander-v2---byzantine-rate--0-5	153600	-119.621590	-244.223297	-119.621590	183.449997	108.555557	183.449997
43	LunarLander-v2---failure-rate--0-05	153600	-63.530300	-262.285828	-136.515945	612.950012	108.244682	327.890015
46	LunarLander-v2---failure-rate--0-1	153600	-0.940206	-281.396973	-0.940206	659.739990	103.923080	502.200012
29	LunarLander-v2---failure-rate--0-25	153600	-33.346172	-244.939301	-33.346172	758.989990	99.016129	438.200012
32	LunarLander-v2---failure-rate--0-5	153600	-31.614159	-171.878326	-31.614159	149.880005	87.742859	140.419998
16	LunarLander-v2---max-thrust--0-75	153600	-57.385582	-237.672867	-57.385582	682.820007	114.333336	325.070007
23	LunarLander-v2---wind-power--10-0	153600	-10.691111	-237.685669	-10.691111	919.140015	102.947365	373.309998
26	LunarLander-v2---wind-power--15-0	153600	-64.839897	-251.497894	-64.839897	1089.727295	112.694443	183.860001
20	LunarLander-v2---wind-power--5-0	153600	-67.407654	-234.716034	-72.116722	741.359985	103.789474	727.190002

SAC								
	env	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
3	LunarLander-v2	149642	185.473572	-271.136230	154.138931	702.640015	94.00	349.549988
11	LunarLander-v2	149317	210.785843	-167.376083	199.461029	621.570007	161.75	319.290009
12	LunarLander-v2	149103	40.995003	-246.109467	40.995003	791.419983	86.50	689.359985
35	LunarLander-v2---byzantine-rate--0-25	146432	-94.873909	-307.104462	-94.873909	1558.366699	133.00	1413.920044
39	LunarLander-v2---byzantine-rate--0-5	94964	-76.314713	-185.968613	-129.672638	388.869995	104.50	388.869995
44	LunarLander-v2---failure-rate--0-05	148631	132.274673	-213.371399	132.274673	593.929993	121.50	505.429993
47	LunarLander-v2---failure-rate--0-1	147203	-198.012543	-385.652435	-198.012543	1682.474976	115.00	1176.147705
48	LunarLander-v2---failure-rate--0-1	148073	104.778748	-279.042908	104.778748	738.700012	101.50	508.899994
28	LunarLander-v2---failure-rate--0-25	148693	200.833145	-214.601364	171.485413	714.593750	90.00	386.649994
31	LunarLander-v2---failure-rate--0-5	149744	-38.510250	-228.608490	-44.803474	128.559998	83.25	126.000000
14	LunarLander-v2---max-thrust--0-75	149146	-192.448166	-418.174591	-418.174591	1477.229980	92.75	1477.229980
22	LunarLander-v2---wind-power--10-0	149057	93.922272	-419.526001	80.723900	734.010010	106.50	394.929993
25	LunarLander-v2---wind-power--15-0	149556	202.200546	-306.832642	185.989029	994.109985	143.25	347.970001
18	LunarLander-v2---wind-power--5-0	3917	-204.409363	-289.090729	-204.409363	195.850006	105.00	195.850006
19	LunarLander-v2---wind-power--5-0	148452	154.536194	-293.235901	137.542816	769.119995	122.50	565.590027

Detailed data - By environment

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
0	A2C	170000	-236.452652	-306.650452	-306.650452	570.500000	233.830002	570.500000
1	A2C	150000	-400.023376	-754.365417	-582.866028	1449.329956	97.800003	1449.329956
2	DDPG	155104	-139.497116	-480.652466	-141.257126	623.659973	82.500000	398.799988
3	SAC	149642	185.473572	-271.136230	154.138931	702.640015	94.000000	349.549988
4	PPO	153600	44.948971	-220.450470	44.948971	677.580017	107.052635	464.160004
5	A2C	150000	-202.001465	-410.694122	-293.969849	612.530029	109.500000	450.079987
6	A2C	150000	-253.250946	-636.849670	-253.250946	622.510010	106.864868	518.020020
7	DDPG	155329	-16.730101	-833.337769	-21.794130	608.609985	88.000000	477.000000
8	DDPG	153566	-88.562225	-853.943726	-88.562225	652.830017	78.000000	647.059998
9	PPO	153600	98.261688	-250.998184	98.261688	881.880005	105.526314	802.080017
10	PPO	153600	72.078133	-240.324020	72.078133	884.369995	105.111115	884.369995
11	SAC	149317	210.785843	-167.376083	199.461029	621.570007	161.750000	319.290009
12	SAC	149103	40.995003	-246.109467	40.995003	791.419983	86.500000	689.359985

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
33	A2C	150000	-292.394989	-484.809967	-351.236298	1125.810059	98.5	1125.810059
34	PPO	153600	-51.564663	-248.864975	-74.847038	499.429993	99.5	499.429993
35	SAC	146432	-94.873909	-307.104462	-94.873909	1558.366699	133.0	1413.920044

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
36	A2C	77000	-234.007477	-310.266235	-305.794556	159.727280	114.750000	159.727280
37	A2C	150000	-227.212433	-436.405945	-283.610931	255.330002	93.800003	255.330002
38	PPO	153600	-119.621590	-244.223297	-119.621590	183.449997	108.555557	183.449997
39	SAC	94964	-76.314713	-185.968613	-129.672638	388.869995	104.500000	388.869995

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
40	A2C	150000	-173.486984	-397.261200	-378.094635	1158.979980	124.333336	1158.979980
41	DDPG	1792	-272.985596	-450.430786	-371.169525	115.833336	75.250000	112.000000
42	DDPG	157548	-173.591629	-951.069641	-220.480194	758.400024	102.750000	370.459991
43	PPO	153600	-63.530300	-262.285828	-136.515945	612.950012	108.244682	327.890015
44	SAC	148631	132.274673	-213.371399	132.274673	593.929993	121.500000	505.429993

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
45	A2C	150000	-228.266678	-425.797180	-422.983215	2429.694336	113.000000	1922.657837
46	PPO	153600	-0.940206	-281.396973	-0.940206	659.739990	103.92308	502.200012
47	SAC	147203	-198.012543	-385.652435	-198.012543	1682.474976	115.000000	1176.147705
48	SAC	148073	104.778748	-279.042908	104.778748	738.700012	101.500000	508.899994

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
27	A2C	150000	-275.416351	-626.930847	-276.242920	903.77002	89.080002	903.770020
28	SAC	148693	200.833145	-214.601364	171.485413	714.59375	90.000000	386.649994
29	PPO	153600	-33.346172	-244.939301	-33.346172	758.98999	99.016129	438.200012

	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
30	A2C	150000	-79.922859	-407.965912	-115.642410	239.330002	84.727272	187.190002
31	SAC	149744	-38.510250	-228.608490	-44.803474	128.559998	83.250000	126.000000
32	PPO	153600	-31.614159	-171.878326	-31.614159	149.880005	87.742859	140.419998

Detailed data - By environment

LunarLander-v2---max-thrust--0-75								
	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
13	A2C	150000	-172.319672	-551.877502	-220.667114	498.779999	108.250000	498.779999
14	SAC	149146	-192.448166	-418.174591	-418.174591	1477.229980	92.750000	1477.229980
15	DDPG	79358	-151.214157	-344.607300	-172.113037	342.880005	79.625000	338.839996
16	PPO	153600	-57.385582	-237.672867	-57.385582	682.820007	114.333336	325.070007

LunarLander-v2---wind-power--5-0								
	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
17	A2C	150000	-308.635834	-563.784119	-409.120087	804.000000	105.878784	804.000000
18	SAC	3917	-204.409363	-289.090729	-204.409363	195.850006	105.000000	195.850006
19	SAC	148452	154.536194	-293.235901	137.542816	769.119995	122.500000	565.590027
20	PPO	153600	-67.407654	-234.716034	-72.116722	741.359985	103.789474	727.190002

LunarLander-v2---wind-power--10-0								
	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
21	A2C	150000	-143.933395	-572.376404	-205.673065	547.859985	93.599998	417.170013
22	SAC	149057	93.922272	-419.526001	80.723900	734.010010	106.500000	394.929993
23	PPO	153600	-10.691111	-237.685669	-10.691111	919.140015	102.947365	373.309998

LunarLander-v2--wind-power--15-0								
	model	length	reward_max	reward_min	reward_last	len_max	len_min	len_last
24	A2C	150000	19.821482	-668.412476	-136.707581	601.859985	101.367348	601.859985
25	SAC	149556	202.200546	-306.832642	185.989029	994.109985	143.250000	347.970001
26	PPO	153600	-64.839897	-251.497894	-64.839897	1089.727295	112.694443	183.860001

Running all models

```
Starting training A2C on LunarLander-v2
C:\Users\henri\AppData\Local\Programs\Python\Python310\lib\site-packages\stable_baselines3\common\save_util.py:278: UserWarning: Path 'models\2023-05-24-18-13-56_LunarLander-v2_A2C_simple' does not exist. Will create it.
  warnings.warn(f"Path '{path.parent}' does not exist. Will create it.")
Finished 1/5 episodes for A2C in 0:00:04 in 2023-05-24-18-13-56_LunarLander-v2_A2C_simple
Finished 2/5 episodes for A2C in 0:00:01 in 2023-05-24-18-13-56_LunarLander-v2_A2C_simple
Finished 3/5 episodes for A2C in 0:00:01 in 2023-05-24-18-13-56_LunarLander-v2_A2C_simple
Finished 4/5 episodes for A2C in 0:00:01 in 2023-05-24-18-13-56_LunarLander-v2_A2C_simple
Finished 5/5 episodes for A2C in 0:00:01 in 2023-05-24-18-13-56_LunarLander-v2_A2C_simple
Finished training routine
Finished training A2C on LunarLander-v2
Starting training DDPG on LunarLander-v2
C:\Users\henri\AppData\Local\Programs\Python\Python310\lib\site-packages\stable_baselines3\common\save_util.py:278: UserWarning: Path 'models\2023-05-24-18-14-06_LunarLander-v2_DDPG_simple' does not exist. Will create it.
  warnings.warn(f"Path '{path.parent}' does not exist. Will create it.")
Finished 1/5 episodes for DDPG in 0:00:12 in 2023-05-24-18-14-06_LunarLander-v2_DDPG_simple
Finished 2/5 episodes for DDPG in 0:00:14 in 2023-05-24-18-14-06_LunarLander-v2_DDPG_simple
Finished 3/5 episodes for DDPG in 0:00:14 in 2023-05-24-18-14-06_LunarLander-v2_DDPG_simple
Finished 4/5 episodes for DDPG in 0:00:43 in 2023-05-24-18-14-06_LunarLander-v2_DDPG_simple
Finished 5/5 episodes for DDPG in 0:00:20 in 2023-05-24-18-14-06_LunarLander-v2_DDPG_simple
Finished training routine
Finished training DDPG on LunarLander-v2
Starting training PPO on LunarLander-v2
C:\Users\henri\AppData\Local\Programs\Python\Python310\lib\site-packages\stable_baselines3\common\save_util.py:278: UserWarning: Path 'models\2023-05-24-18-15-51_LunarLander-v2_PPO_simple' does not exist. Will create it.
  warnings.warn(f"Path '{path.parent}' does not exist. Will create it.")
Finished 1/5 episodes for PPO in 0:00:03 in 2023-05-24-18-15-51_LunarLander-v2_PPO_simple
Finished 2/5 episodes for PPO in 0:00:03 in 2023-05-24-18-15-51_LunarLander-v2_PPO_simple
Finished 3/5 episodes for PPO in 0:00:03 in 2023-05-24-18-15-51_LunarLander-v2_PPO_simple
Finished 4/5 episodes for PPO in 0:00:03 in 2023-05-24-18-15-51_LunarLander-v2_PPO_simple
Finished 5/5 episodes for PPO in 0:00:03 in 2023-05-24-18-15-51_LunarLander-v2_PPO_simple
Finished training routine
Finished training PPO on LunarLander-v2
Starting training SAC on LunarLander-v2
C:\Users\henri\AppData\Local\Programs\Python\Python310\lib\site-packages\stable_baselines3\common\save_util.py:278: UserWarning: Path 'models\2023-05-24-18-16-08_LunarLander-v2_SAC_simple' does not exist. Will create it.
  warnings.warn(f"Path '{path.parent}' does not exist. Will create it.")
Finished 1/5 episodes for SAC in 0:00:21 in 2023-05-24-18-16-08_LunarLander-v2_SAC_simple
Finished 2/5 episodes for SAC in 0:00:29 in 2023-05-24-18-16-08_LunarLander-v2_SAC_simple
Finished 3/5 episodes for SAC in 0:00:35 in 2023-05-24-18-16-08_LunarLander-v2_SAC_simple
Finished 4/5 episodes for SAC in 0:00:34 in 2023-05-24-18-16-08_LunarLander-v2_SAC_simple
Finished 5/5 episodes for SAC in 0:00:39 in 2023-05-24-18-16-08_LunarLander-v2_SAC_simple
Finished training routine
Finished training SAC on LunarLander-v2
Finished training all models
```