

```
!pip install swig
```

```
Collecting swig
  Downloading swig-4.4.0-py3-none-manylinux_2_12_x86_64.manylinux2010_x86_64.whl.metadata (3.5 kB)
  Downloading swig-4.4.0-py3-none-manylinux_2_12_x86_64.manylinux2010_x86_64.whl (1.9 MB)
    1.9/1.9 MB 81.8 MB/s eta 0:00:00
```

```
Installing collected packages: swig
Successfully installed swig-4.4.0
```

```
!pip install stable_baselines3 gymnasium[box2d]
```

```
Collecting stable_baselines3
  Downloading stable_baselines3-2.7.0-py3-none-any.whl.metadata (4.8 kB)
Requirement already satisfied: gymnasium[box2d] in /usr/local/lib/python3.12/dist-packages (1.2.2)
Requirement already satisfied: numpy<3.0,>=1.20 in /usr/local/lib/python3.12/dist-packages (from stable_baselines3) (2.0.2)
Requirement already satisfied: torch<3.0,>=2.3 in /usr/local/lib/python3.12/dist-packages (from stable_baselines3) (2.9.0+cu126)
Requirement already satisfied: cloudpickle in /usr/local/lib/python3.12/dist-packages (from stable_baselines3) (3.1.2)
Requirement already satisfied: pandas in /usr/local/lib/python3.12/dist-packages (from stable_baselines3) (2.2.2)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.12/dist-packages (from stable_baselines3) (3.10.0)
Requirement already satisfied: typing-extensions>=4.3.0 in /usr/local/lib/python3.12/dist-packages (from gymnasium[box2d]) (4.15.0)
Requirement already satisfied: farama-notifications>=0.0.1 in /usr/local/lib/python3.12/dist-packages (from gymnasium[box2d]) (0.0.4)
Collecting box2d-py==2.3.5 (from gymnasium[box2d])
  Downloading box2d-py-2.3.5.tar.gz (374 kB)
    374.4/374.4 kB 29.1 MB/s eta 0:00:00
Preparing metadata (setup.py) ... done
Requirement already satisfied: pygame>=2.1.3 in /usr/local/lib/python3.12/dist-packages (from gymnasium[box2d]) (2.6.1)
Requirement already satisfied: swig==4.* in /usr/local/lib/python3.12/dist-packages (from gymnasium[box2d]) (4.4.0)
Requirement already satisfied: filelock in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (3.20.0)
Requirement already satisfied: setuptools in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (75.2.0)
Requirement already satisfied: sympy>=1.13.3 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (1.13.3)
Requirement already satisfied: networkx>=2.5.1 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (3.5)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (3.1.6)
Requirement already satisfied: fsspec>=0.8.5 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (2025.3.0)
Requirement already satisfied: nvidia-cuda-nvrtc-cu12==12.6.77 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.77)
Requirement already satisfied: nvidia-cuda-runtime-cu12==12.6.77 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.77)
Requirement already satisfied: nvidia-cuda-cupti-cu12==12.6.80 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.80)
Requirement already satisfied: nvidia-cudnn-cu12==9.10.2.21 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (9.10.2.21)
Requirement already satisfied: nvidia-cublas-cu12==12.6.4.1 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.4.1)
Requirement already satisfied: nvidia-cufft-cu12==11.3.0.4 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (11.3.0.4)
Requirement already satisfied: nvidia-curand-cu12==10.3.7.77 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (10.3.7.77)
Requirement already satisfied: nvidia-cusolver-cu12==11.7.1.2 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (11.7.1.2)
Requirement already satisfied: nvidia-cusparse-cu12==12.5.4.2 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.5.4.2)
Requirement already satisfied: nvidia-cusparselt-cu12==0.7.1 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (0.7.1)
Requirement already satisfied: nvidia-nccl-cu12==2.27.5 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (2.27.5)
Requirement already satisfied: nvidia-nvhsmem-cu12==3.3.20 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (3.3.20)
Requirement already satisfied: nvidia-nvtx-cu12==12.6.77 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.77)
Requirement already satisfied: nvidia-nvjitlink-cu12==12.6.85 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (12.6.85)
Requirement already satisfied: nvidia-cufile-cu12==1.11.1.6 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (1.11.1.6)
Requirement already satisfied: triton==3.5.0 in /usr/local/lib/python3.12/dist-packages (from torch<3.0,>=2.3->stable_baselines3) (3.5.0)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (1.3.3)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (4.60.1)
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (1.4.9)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (25.0)
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (11.3.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (3.2.5)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.12/dist-packages (from matplotlib->stable_baselines3) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas->stable_baselines3) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas->stable_baselines3) (2025.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil>=2.7->matplotlib->stable_baselines3) (1.17.0)
```

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Requirement already satisfied: mpmath<1.4,>=1.1.0 in /usr/local/lib/python3.12/dist-packages (from sympy>=1.13.3->torch<3.0,>=2.3->stable_baselines3) (1.3.0)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.12/dist-packages (from jinja2->torch<3.0,>=2.3->stable_baselines3) (3.0.3)
Downloading stable_baselines3-2.7.0-py3-none-any.whl (187 kB)
    187.2/187.2 kB 18.3 MB/s eta 0:00:00
Building wheels for collected packages: box2d-py
  Building wheel for box2d-py (setup.py) ... done
    Created wheel for box2d-py: filename=box2d_py-2.3.5-cp312-cp312-linux_x86_64.whl size=2398998 sha256=959307cfcc226c6163272eb02fed9697ca18371b77f2a39c0b4fe2bf
    Stored in directory: /root/.cache/pip/wheels/22/00/60/771d0bcd07f7dc7761a9500f22d065e1060568a78dcde2210
```

```
import os, platform, warnings
warnings.filterwarnings("ignore", category=DeprecationWarning)
warnings.filterwarnings("ignore", message="This system does not have CUDA")
warnings.filterwarnings("ignore", message="pkg_resources is deprecated")

import numpy as np
import torch
import matplotlib.pyplot as plt

import gymnasium as gym
from stable_baselines3 import SAC
from stable_baselines3.common.atari_wrappers import WarpFrame # 84x84 grayscale
from stable_baselines3.common.env_util import make_vec_env
from stable_baselines3.common.vec_env import (
    VecFrameStack, VecTransposeImage, VecNormalize, VecVideoRecorder
)
from stable_baselines3.common.callbacks import EvalCallback
from stable_baselines3.common.evaluation import evaluate_policy
from stable_baselines3.common.torch_layers import NatureCNN
from importlib.metadata import version

print(f"Python: {platform.python_version()}")
print(f"Torch: {version('torch')} | CUDA avail: {torch.cuda.is_available()} | torch.cuda: {torch.version.cuda}")
print(f"Gymnasium: {version('gymnasium')}")
print(f"SB3: {version('stable_baselines3')}")
print(f"Numpy: {version('numpy')} Matplotlib: {version('matplotlib')}"
```

```
Python: 3.12.12
Torch: 2.9.0+cu126 | CUDA avail: True | torch.cuda: 12.6
Gymnasium: 1.2.2
SB3: 2.7.0
Numpy: 2.0.2 Matplotlib: 3.10.0
Gym has been unmaintained since 2022 and does not support NumPy 2.0 amongst other critical functionality.
Please upgrade to Gymnasium, the maintained drop-in replacement of Gym, or contact the authors of your software and request that they upgrade.
See the migration guide at https://gymnasium.farama.org/introduction/migration\_guide/ for additional information.
/usr/local/lib/python3.12/dist-packages/jupyter_client/session.py:203: DeprecationWarning: datetime.datetime.utcnow() is deprecated and scheduled for removal in
return datetime.utcnow().replace(tzinfo=utc)
```

```
# -----
# Config (faster run)
# -----
ENV_ID = "CarRacing-v3"
LOG_DIR = f"./logs/{ENV_ID}"
VIDEO_DIR = "./videos"
os.makedirs(LOG_DIR, exist_ok=True)
os.makedirs(VIDEO_DIR, exist_ok=True)

TOTAL_STEPS = 500_000      # ↓ from 1M
```

```

EVAL_FREQ = 20_000          # evaluate a bit less often
EVAL_EPISODES = 5           # fewer episodes during training to save time
FINAL_EVAL_EPISODES = 20    # do a fuller eval at the end for your report
SEED = 42

USE_GRAY = True
WRAPPER_CLASS = WarpFrame if USE_GRAY else None

def make_training_env(seed=SEED):
    env = make_vec_env(ENV_ID, n_envs=1, seed=seed, wrapper_class=WRAPPER_CLASS)
    env = VecFrameStack(env, n_stack=4)
    env = VecTransposeImage(env)
    # Reward normalization helps SAC stability on pixels
    env = VecNormalize(env, norm_obs=False, norm_reward=True, clip_reward=10.0)
    return env

def make_eval_env(seed=SEED+1):
    env = make_vec_env(ENV_ID, n_envs=1, seed=seed, wrapper_class=WRAPPER_CLASS)
    env = VecFrameStack(env, n_stack=4)
    env = VecTransposeImage(env)
    env = VecNormalize(env, norm_obs=False, norm_reward=True, clip_reward=10.0, training=False)
    return env

train_env = make_training_env(SEED)
eval_env = make_eval_env(SEED + 1)

policy_kwargs = dict(
    features_extractor_class=NatureCNN,
    features_extractor_kwargs=dict(features_dim=512), # stronger CNN head
    net_arch=[256, 256],
)

eval_callback = EvalCallback(
    eval_env,
    best_model_save_path=LOG_DIR,
    log_path=LOG_DIR,
    eval_freq=EVAL_FREQ,
    n_eval_episodes=EVAL_EPISODES,
    deterministic=True,
    render=False,
)

```

```

/usr/local/lib/python3.12/dist-packages/pygame/pkgdata.py:25: DeprecationWarning: pkg_resources is deprecated as an API. See https://setuptools.pypa.io/en/latest/
  from pkg_resources import resource_stream, resource_exists
/usr/local/lib/python3.12/dist-packages/pkg_resources/_init_.py:3154: DeprecationWarning: Deprecated call to `pkg_resources.declare_namespace('google')`.
Implementing implicit namespace packages (as specified in PEP 420) is preferred to `pkg_resources.declare_namespace`. See https://setuptools.pypa.io/en/latest/declare\_namespace\(pkg\)
/usr/local/lib/python3.12/dist-packages/pkg_resources/_init_.py:3154: DeprecationWarning: Deprecated call to `pkg_resources.declare_namespace('google.cloud')`.
Implementing implicit namespace packages (as specified in PEP 420) is preferred to `pkg_resources.declare_namespace`. See https://setuptools.pypa.io/en/latest/declare\_namespace\(pkg\)
/usr/local/lib/python3.12/dist-packages/pkg_resources/_init_.py:3154: DeprecationWarning: Deprecated call to `pkg_resources.declare_namespace('sphinxcontrib')`.
Implementing implicit namespace packages (as specified in PEP 420) is preferred to `pkg_resources.declare_namespace`. See https://setuptools.pypa.io/en/latest/declare\_namespace\(pkg\)
/usr/local/lib/python3.12/dist-packages/jupyter_client/session.py:203: DeprecationWarning: datetime.datetime.utcnow() is deprecated and scheduled for removal in
  return datetime.utcnow().replace(tzinfo=utc)

```

```
# -----
# SAC hyperparams (slightly lighter/faster)
# -----
model = SAC(
    "CnnPolicy",
    train_env,
    learning_rate=1e-4,
    buffer_size=500_000,           # ↓ from 1,000,000 to reduce RAM/time
    batch_size=256,
    learning_starts=5_000,         # ↓ warmup so learning begins earlier
    gamma=0.99,
    tau=0.005,
    ent_coef="auto",
    target_entropy="auto",
    train_freq=(1, "step"),
    gradient_steps=1,
    policy_kwargs=policy_kwargs,
    verbose=1,
    tensorboard_log=LOG_DIR,
    seed=SEED,
)
```

```
Using cuda device
/usr/local/lib/python3.12/dist-packages/jupyter_client/session.py:203: DeprecationWarning: datetime.datetime.utcnow() is deprecated and scheduled for removal in
return datetime.utcnow().replace(tzinfo=utc)
```

```
# -----
# Train
# -----
model.learn(total_timesteps=TOTAL_STEPS, callback=eval_callback, progress_bar=True)

# Save final model and normalization stats
final_model_path = os.path.join(LOG_DIR, "sac_car_racing_final_500k")
model.save(final_model_path)
train_env.save(os.path.join(LOG_DIR, "vecnormalize.pkl"))
```



```
Logging to ./logs/CarRacing-v3/SAC_1
100% _____ 500,000/500,000 [ 7:26:35 < 0:00:00 , ? it/s ]
/usr/local/lib/python3.12/dist-packages/ipywidgets/widgets/widget_output.py:111: DeprecationWarning:
Kernel._parent_header is deprecated in ipykernel 6. Use .get_parent()
  if ip and hasattr(ip, 'kernel') and hasattr(ip.kernel, '_parent_header'):

rollout/
  ep_len_mean    1e+03
  ep_rew_mean   -31.6
time/
  episodes      4
  fps           94
  time_elapsed  42
  total_timesteps 4000
-----
rollout/
  ep_len_mean    1e+03
  ep_rew_mean   -35
time/
  episodes      8
  fps           38
  time_elapsed  206
  total_timesteps 8000
train/
  actor_loss    -23.3
  critic_loss   0.0994
  ent_coef      0.741
  ent_coef_loss -1.51
  learning_rate 0.0001
  n_updates     2999
-----
rollout/
  ep_len_mean    1e+03
  ep_rew_mean   -34.5
time/
  episodes      12
  fps           29
  time_elapsed  406
  total_timesteps 12000
train/
  actor_loss    -40.1
  critic_loss   0.107
  ent_coef      0.497
  ent_coef_loss -3.53
  learning_rate 0.0001
  n_updates     6999
-----
rollout/
  ep_len_mean    1e+03
  ep_rew_mean   -33.6
time/
  episodes      16
  fps           26
  time_elapsed  606
  total_timesteps 16000
train/
  actor_loss    -47.3
  critic_loss   0.0478
  ent_coef      0.333
  ent_coef_loss -5.57
  learning_rate 0.0001
```

n_updates	10999
<hr/>	
Eval num_timesteps	=20000, episode_reward=-18.57 +/- 16.33
Episode length	: 1000.00 +/- 0.00
eval/	
mean_ep_length	1e+03
mean_reward	-18.6
time/	
total_timesteps	20000
train/	
actor_loss	-48
critic_loss	0.0523
ent_coef	0.223
ent_coef_loss	-7.57
learning_rate	0.0001
n_updates	14999

New best mean reward!

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.9
<hr/>	
time/	
episodes	20
fps	23
time_elapsed	867
total_timesteps	20000

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.5
<hr/>	
time/	
episodes	24
fps	22
time_elapsed	1070
total_timesteps	24000
train/	
actor_loss	-45.5
critic_loss	0.0299
ent_coef	0.15
ent_coef_loss	-9.62
learning_rate	0.0001
n_updates	18999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.9
<hr/>	
time/	
episodes	28
fps	21
time_elapsed	1273
total_timesteps	28000
train/	
actor_loss	-41.2
critic_loss	0.0263
ent_coef	0.1
ent_coef_loss	-11.6
learning_rate	0.0001
n_updates	22999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.9
time/	
episodes	32
fps	21
time_elapsed	1476
total_timesteps	32000
train/	
actor_loss	-36.4
critic_loss	0.059
ent_coef	0.0672
ent_coef_loss	-13.7
learning_rate	0.0001
n_updates	26999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-33.2
time/	
episodes	36
fps	21
time_elapsed	1678
total_timesteps	36000
train/	
actor_loss	-31.4
critic_loss	0.0397
ent_coef	0.0451
ent_coef_loss	-15.7
learning_rate	0.0001
n_updates	30999

Eval num_timesteps=40000, episode_reward=-30.38 +/- 5.07
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-30.4
time/	
total_timesteps	40000
train/	
actor_loss	-26.8
critic_loss	0.0473
ent_coef	0.0302
ent_coef_loss	-17.7
learning_rate	0.0001
n_updates	34999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.7
time/	
episodes	40
fps	20
time_elapsed	1938
total_timesteps	40000

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.7
time/	
episodes	41

cpu_usage	44
fps	20
time_elapsed	2139
total_timesteps	44000
train/	
actor_loss	-22.5
critic_loss	0.0511
ent_coef	0.0203
ent_coef_loss	-19.6
learning_rate	0.0001
n_updates	38999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.6
time/	
episodes	48
fps	20
time_elapsed	2341
total_timesteps	48000
train/	
actor_loss	-18.6
critic_loss	0.032
ent_coef	0.0136
ent_coef_loss	-21.6
learning_rate	0.0001
n_updates	42999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.8
time/	
episodes	52
fps	20
time_elapsed	2547
total_timesteps	52000
train/	
actor_loss	-15.3
critic_loss	0.029
ent_coef	0.0091
ent_coef_loss	-23.5
learning_rate	0.0001
n_updates	46999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-32.7
time/	
episodes	56
fps	20
time_elapsed	2750
total_timesteps	56000
train/	
actor_loss	-12.3
critic_loss	0.0281
ent_coef	0.0061
ent_coef_loss	-25.5
learning_rate	0.0001
n_updates	50999

Eval num_timesteps=60000, episode_reward=-23.61 +/- 14.95
 Episode length: 1000.00 +/- 0.00

Episode Length: 1000.00 / 1000		
eval/		
mean_ep_length	1e+03	
mean_reward	-23.6	
time/		
total_timesteps	60000	
train/		
actor_loss	-9.97	
critic_loss	0.0286	
ent_coef	0.00409	
ent_coef_loss	-27.7	
learning_rate	0.0001	
n_updates	54999	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-32	
time/		
episodes	60	
fps	19	
time_elapsed	3012	
total_timesteps	60000	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-31.6	
time/		
episodes	64	
fps	19	
time_elapsed	3213	
total_timesteps	64000	
train/		
actor_loss	-7.94	
critic_loss	0.0579	
ent_coef	0.00275	
ent_coef_loss	-28.4	
learning_rate	0.0001	
n_updates	58999	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-31.4	
time/		
episodes	68	
fps	19	
time_elapsed	3414	
total_timesteps	68000	
train/		
actor_loss	-6.15	
critic_loss	0.0237	
ent_coef	0.00188	
ent_coef_loss	-19.7	
learning_rate	0.0001	
n_updates	62999	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-33.9	
time/		
episodes	72	

fps	19
time_elapsed	3616
total_timesteps	72000
<hr/>	
train/	
actor_loss	-5.09
critic_loss	0.0184
ent_coef	0.00138
ent_coef_loss	-5.69
learning_rate	0.0001
n_updates	66999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-36.6
<hr/>	
time/	
episodes	76
fps	19
time_elapsed	3822
total_timesteps	76000
<hr/>	
train/	
actor_loss	-4.06
critic_loss	0.0177
ent_coef	0.00109
ent_coef_loss	-5.21
learning_rate	0.0001
n_updates	70999

Eval num_timesteps=80000, episode_reward=-79.25 +/- 0.77
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-79.3
<hr/>	
time/	
total_timesteps	80000
<hr/>	
train/	
actor_loss	-3.54
critic_loss	0.0215
ent_coef	0.000899
ent_coef_loss	-1.23
learning_rate	0.0001
n_updates	74999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-38.6
<hr/>	
time/	
episodes	80
fps	19
time_elapsed	4088
total_timesteps	80000

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-40.5
<hr/>	
time/	
episodes	84
fps	19
time_elapsed	4293
total_timesteps	84000
<hr/>	
train/	

actor_loss	-2.75
critic_loss	0.0146
ent_coef	0.000594
ent_coef_loss	-21.9
learning_rate	0.0001
n_updates	78999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-41.4
time/	
episodes	88
fps	19
time_elapsed	4496
total_timesteps	88000
train/	
actor_loss	-2.23
critic_loss	0.0152
ent_coef	0.000393
ent_coef_loss	-17.8
learning_rate	0.0001
n_updates	82999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-40.9
time/	
episodes	92
fps	19
time_elapsed	4700
total_timesteps	92000
train/	
actor_loss	-1.87
critic_loss	0.0114
ent_coef	0.000269
ent_coef_loss	-12.5
learning_rate	0.0001
n_updates	86999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-40.2
time/	
episodes	96
fps	19
time_elapsed	4904
total_timesteps	96000
train/	
actor_loss	-1.47
critic_loss	0.0153
ent_coef	0.000187
ent_coef_loss	-8.67
learning_rate	0.0001
n_updates	90999

rollout/	
ep_len_mean	997
ep_rew_mean	-42.4
time/	
episodes	100

fps	19
time_elapsed	5092
total_timesteps	99713
train/	
actor_loss	-1.05
critic_loss	0.00879
ent_coef	0.000154
ent_coef_loss	-4.21
learning_rate	0.0001
n_updates	94712

Eval num_timesteps=100000, episode_reward=-78.17 +/- 3.04
Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-78.2
time/	
total_timesteps	100000
train/	
actor_loss	-1.01
critic_loss	0.0106
ent_coef	0.000154
ent_coef_loss	4.87
learning_rate	0.0001
n_updates	94999

rollout/	
ep_len_mean	997
ep_rew_mean	-42.9
time/	
episodes	104
fps	19
time_elapsed	5358
total_timesteps	103713
train/	
actor_loss	-0.848
critic_loss	0.00764
ent_coef	0.000205
ent_coef_loss	10.5
learning_rate	0.0001
n_updates	98712

rollout/	
ep_len_mean	997
ep_rew_mean	-41.2
time/	
episodes	108
fps	19
time_elapsed	5562
total_timesteps	107713
train/	
actor_loss	-0.571
critic_loss	0.00884
ent_coef	0.000307
ent_coef_loss	9
learning_rate	0.0001
n_updates	102712

rollout/	
ep_len_mean	997

ep_rew_mean	-36.9
time/	
episodes	112
fps	19
time_elapsed	5769
total_timesteps	111713
train/	
actor_loss	-0.523
critic_loss	0.00763
ent_coef	0.000425
ent_coef_loss	3.87
learning_rate	0.0001
n_updates	106712

rollout/	
ep_len_mean	997
ep_rew_mean	-23.9
time/	
episodes	116
fps	19
time_elapsed	5975
total_timesteps	115713
train/	
actor_loss	-0.817
critic_loss	0.00343
ent_coef	0.00059
ent_coef_loss	5.63
learning_rate	0.0001
n_updates	110712

rollout/	
ep_len_mean	997
ep_rew_mean	-7.96
time/	
episodes	120
fps	19
time_elapsed	6181
total_timesteps	119713
train/	
actor_loss	-0.714
critic_loss	0.00354
ent_coef	0.000821
ent_coef_loss	0.815
learning_rate	0.0001
n_updates	114712

Eval num_timesteps=120000, episode_reward=524.31 +/- 302.41
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	524
time/	
total_timesteps	120000
train/	
actor_loss	-0.885
critic_loss	0.00501
ent_coef	0.000836
ent_coef_loss	2.47
learning_rate	0.0001
n_updates	114999

New best mean reward!

new best mean reward:		
rollout/		
ep_len_mean	995	
ep_rew_mean	18.4	
time/		
episodes	124	
fps	19	
time_elapsed	6436	
total_timesteps	123460	
train/		
actor_loss	-0.997	
critic_loss	0.00685	
ent_coef	0.00103	
ent_coef_loss	-3.11	
learning_rate	0.0001	
n_updates	118459	
<hr/>		
rollout/		
ep_len_mean	995	
ep_rew_mean	39.5	
time/		
episodes	128	
fps	19	
time_elapsed	6641	
total_timesteps	127460	
train/		
actor_loss	-1.06	
critic_loss	0.00446	
ent_coef	0.00106	
ent_coef_loss	0.716	
learning_rate	0.0001	
n_updates	122459	
<hr/>		
rollout/		
ep_len_mean	995	
ep_rew_mean	66	
time/		
episodes	132	
fps	19	
time_elapsed	6847	
total_timesteps	131460	
train/		
actor_loss	-1.11	
critic_loss	0.00618	
ent_coef	0.00101	
ent_coef_loss	-1.65	
learning_rate	0.0001	
n_updates	126459	
<hr/>		
rollout/		
ep_len_mean	995	
ep_rew_mean	86.6	
time/		
episodes	136	
fps	19	
time_elapsed	7051	
total_timesteps	135460	
train/		
actor_loss	-1.11	
critic_loss	0.0035	
ent_coef	0.00103	

ent_coef	0.0001
ent_coef_loss	1.82
learning_rate	0.0001
n_updates	130459

rollout/	
ep_len_mean	991
ep_rew_mean	111
time/	
episodes	140
fps	19
time_elapsed	7239
total_timesteps	139135
train/	
actor_loss	-1.44
critic_loss	0.00666
ent_coef	0.00108
ent_coef_loss	-2.39
learning_rate	0.0001
n_updates	134134

Eval num_timesteps=140000, episode_reward=723.94 +/- 143.86
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	724
time/	
total_timesteps	140000
train/	
actor_loss	-1.38
critic_loss	0.00522
ent_coef	0.00105
ent_coef_loss	-0.594
learning_rate	0.0001
n_updates	134999

New best mean reward!

rollout/	
ep_len_mean	991
ep_rew_mean	133
time/	
episodes	144
fps	19
time_elapsed	7511
total_timesteps	143135
train/	
actor_loss	-1.36
critic_loss	0.00461
ent_coef	0.00118
ent_coef_loss	4.96
learning_rate	0.0001
n_updates	138134

rollout/	
ep_len_mean	991
ep_rew_mean	157
time/	
episodes	148
fps	19
time_elapsed	7715
total_timesteps	147135

train/	
actor_loss	-1.39
critic_loss	0.00464
ent_coef	0.00118
ent_coef_loss	-1.83
learning_rate	0.0001
n_updates	142134

rollout/	
ep_len_mean	991
ep_rew_mean	155
time/	
episodes	152
fps	19
time_elapsed	7916
total_timesteps	151135
train/	
actor_loss	-1.56
critic_loss	0.00558
ent_coef	0.00118
ent_coef_loss	1.44
learning_rate	0.0001
n_updates	146134

rollout/	
ep_len_mean	991
ep_rew_mean	158
time/	
episodes	156
fps	19
time_elapsed	8119
total_timesteps	155135
train/	
actor_loss	-1.42
critic_loss	0.00452
ent_coef	0.00121
ent_coef_loss	-0.137
learning_rate	0.0001
n_updates	150134

rollout/	
ep_len_mean	991
ep_rew_mean	164
time/	
episodes	160
fps	19
time_elapsed	8321
total_timesteps	159135
train/	
actor_loss	-1.51
critic_loss	0.00802
ent_coef	0.00138
ent_coef_loss	0.493
learning_rate	0.0001
n_updates	154134

Eval num_timesteps=160000, episode_reward=296.36 +/- 331.31
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03

mean_reward	296
time/total_timesteps	160000
train/actor_loss	-1.66
train/critic_loss	0.00581
train/ent_coef	0.00139
train/ent_coef_loss	-1.45
train/learning_rate	0.0001
train/n_updates	154999

rollout/ep_len_mean	991
rollout/ep_rew_mean	170
time/episodes	164
time/fps	18
time/time_elapsed	8586
time/total_timesteps	163135
train/actor_loss	-1.71
train/critic_loss	0.00602
train/ent_coef	0.00137
train/ent_coef_loss	-0.711
train/learning_rate	0.0001
train/n_updates	158134

rollout/ep_len_mean	991
rollout/ep_rew_mean	179
time/episodes	168
time/fps	19
time/time_elapsed	8788
time/total_timesteps	167135
train/actor_loss	-1.68
train/critic_loss	0.00392
train/ent_coef	0.00126
train/ent_coef_loss	-2.15
train/learning_rate	0.0001
train/n_updates	162134

rollout/ep_len_mean	991
rollout/ep_rew_mean	185
time/episodes	172
time/fps	19
time/time_elapsed	8990
time/total_timesteps	171135
train/actor_loss	-1.67
train/critic_loss	0.00357
train/ent_coef	0.00112
train/ent_coef_loss	0.252
train/learning_rate	0.0001
train/n_updates	166134

rollout/	...
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ep_len_mean	991
ep_rew_mean	190
time/	
episodes	176
fps	19
time_elapsed	9195
total_timesteps	175135
train/	
actor_loss	-1.64
critic_loss	0.0082
ent_coef	0.000987
ent_coef_loss	-2.36
learning_rate	0.0001
n_updates	170134

rollout/	
ep_len_mean	991
ep_rew_mean	190
time/	
episodes	180
fps	19
time_elapsed	9399
total_timesteps	179135
train/	
actor_loss	-1.57
critic_loss	0.00375
ent_coef	0.000918
ent_coef_loss	1.24
learning_rate	0.0001
n_updates	174134

Eval num_timesteps=180000, episode_reward=-85.45 +/- 6.25
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-85.5
time/	
total_timesteps	180000
train/	
actor_loss	-1.57
critic_loss	0.00359
ent_coef	0.000938
ent_coef_loss	2.82
learning_rate	0.0001
n_updates	174999

rollout/	
ep_len_mean	991
ep_rew_mean	190
time/	
episodes	184
fps	18
time_elapsed	9661
total_timesteps	183135
train/	
actor_loss	-1.44
critic_loss	0.00354
ent_coef	0.00103
ent_coef_loss	-1.88
learning_rate	0.0001
n_updates	178134

rollout/	
ep_len_mean	991
ep_rew_mean	207
time/	
episodes	188
fps	18
time_elapsed	9865
total_timesteps	187135
train/	
actor_loss	-1.48
critic_loss	0.00452
ent_coef	0.000923
ent_coef_loss	-0.992
learning_rate	0.0001
n_updates	182134

rollout/	
ep_len_mean	991
ep_rew_mean	233
time/	
episodes	192
fps	18
time_elapsed	10069
total_timesteps	191135
train/	
actor_loss	-1.57
critic_loss	0.00411
ent_coef	0.000886
ent_coef_loss	1.74
learning_rate	0.0001
n_updates	186134

rollout/	
ep_len_mean	991
ep_rew_mean	261
time/	
episodes	196
fps	18
time_elapsed	10273
total_timesteps	195135
train/	
actor_loss	-1.53
critic_loss	0.0056
ent_coef	0.000933
ent_coef_loss	2.89
learning_rate	0.0001
n_updates	190134

rollout/	
ep_len_mean	994
ep_rew_mean	285
time/	
episodes	200
fps	19
time_elapsed	10478
total_timesteps	199135
train/	
actor_loss	-1.49
critic_loss	0.00406
ent_coef	0.000933

ent_coef	0.000000
ent_coef_loss	2.23
learning_rate	0.0001
n_updates	194134

Eval num_timesteps=200000, episode_reward=-11.20 +/- 151.92
Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-11.2
time/	
total_timesteps	200000
train/	
actor_loss	-1.52
critic_loss	0.00888
ent_coef	0.000974
ent_coef_loss	2.47
learning_rate	0.0001
n_updates	194999

rollout/	
ep_len_mean	994
ep_rew_mean	292
time/	
episodes	204
fps	18
time_elapsed	10745
total_timesteps	203135
train/	
actor_loss	-1.62
critic_loss	0.00438
ent_coef	0.000963
ent_coef_loss	-1.91
learning_rate	0.0001
n_updates	198134

rollout/	
ep_len_mean	994
ep_rew_mean	293
time/	
episodes	208
fps	18
time_elapsed	10949
total_timesteps	207135
train/	
actor_loss	-1.45
critic_loss	0.00763
ent_coef	0.00107
ent_coef_loss	-4.41
learning_rate	0.0001
n_updates	202134

rollout/	
ep_len_mean	994
ep_rew_mean	286
time/	
episodes	212
fps	18
time_elapsed	11152
total_timesteps	211135
train/	

.....	
actor_loss	-1.99
critic_loss	0.0185
ent_coef	0.00115
ent_coef_loss	2.75
learning_rate	0.0001
n_updates	206134

rollout/	
ep_len_mean	994
ep_rew_mean	280
time/	
episodes	216
fps	18
time_elapsed	11355
total_timesteps	215135
train/	
actor_loss	-1.87
critic_loss	0.0143
ent_coef	0.00123
ent_coef_loss	2.39
learning_rate	0.0001
n_updates	210134

rollout/	
ep_len_mean	994
ep_rew_mean	262
time/	
episodes	220
fps	18
time_elapsed	11558
total_timesteps	219135
train/	
actor_loss	-2.16
critic_loss	0.00917
ent_coef	0.00128
ent_coef_loss	0.064
learning_rate	0.0001
n_updates	214134

Eval num_timesteps=220000, episode_reward=-88.99 +/- 2.04
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-89
time/	
total_timesteps	220000
train/	
actor_loss	-2.02
critic_loss	0.0111
ent_coef	0.00127
ent_coef_loss	-1.2
learning_rate	0.0001
n_updates	214999

rollout/	
ep_len_mean	997
ep_rew_mean	233
time/	
episodes	224
fps	18

time_elapsed	11822
total_timesteps	223135
train/	
actor_loss	-2.23
critic_loss	0.0242
ent_coef	0.00117
ent_coef_loss	0.81
learning_rate	0.0001
n_updates	218134

rollout/	
ep_len_mean	997
ep_rew_mean	210
time/	
episodes	228
fps	18
time_elapsed	12022
total_timesteps	227135
train/	
actor_loss	-1.96
critic_loss	0.0152
ent_coef	0.00112
ent_coef_loss	-0.293
learning_rate	0.0001
n_updates	222134

rollout/	
ep_len_mean	997
ep_rew_mean	182
time/	
episodes	232
fps	18
time_elapsed	12226
total_timesteps	231135
train/	
actor_loss	-2.4
critic_loss	0.033
ent_coef	0.00114
ent_coef_loss	0.47
learning_rate	0.0001
n_updates	226134

rollout/	
ep_len_mean	997
ep_rew_mean	159
time/	
episodes	236
fps	18
time_elapsed	12428
total_timesteps	235135
train/	
actor_loss	-2.49
critic_loss	0.0396
ent_coef	0.00121
ent_coef_loss	-0.673
learning_rate	0.0001
n_updates	230134

rollout/	
ep_len_mean	1e+03

ep_rew_mean	134
time/	
episodes	240
fps	18
time_elapsed	12631
total_timesteps	239135
train/	
actor_loss	-2.75
critic_loss	0.0658
ent_coef	0.00137
ent_coef_loss	-1.23
learning_rate	0.0001
n_updates	234134

Eval num_timesteps=240000, episode_reward=-77.94 +/- 17.44
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-77.9
time/	
total_timesteps	240000
train/	
actor_loss	-2.2
critic_loss	0.033
ent_coef	0.00139
ent_coef_loss	-0.617
learning_rate	0.0001
n_updates	234999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	111
time/	
episodes	244
fps	18
time_elapsed	12895
total_timesteps	243135
train/	
actor_loss	-2.94
critic_loss	0.0379
ent_coef	0.00145
ent_coef_loss	0.616
learning_rate	0.0001
n_updates	238134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	84.7
time/	
episodes	248
fps	18
time_elapsed	13098
total_timesteps	247135
train/	
actor_loss	-2.41
critic_loss	0.0642
ent_coef	0.00167
ent_coef_loss	-3.13
learning_rate	0.0001
n_updates	242134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	85.3
time/	
episodes	252
fps	18
time_elapsed	13301
total_timesteps	251135
train/	
actor_loss	-2.93
critic_loss	0.0948
ent_coef	0.00206
ent_coef_loss	2.89
learning_rate	0.0001
n_updates	246134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	80.9
time/	
episodes	256
fps	18
time_elapsed	13504
total_timesteps	255135
train/	
actor_loss	-3.02
critic_loss	0.0886
ent_coef	0.0026
ent_coef_loss	-0.217
learning_rate	0.0001
n_updates	250134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	74.7
time/	
episodes	260
fps	18
time_elapsed	13707
total_timesteps	259135
train/	
actor_loss	-3.55
critic_loss	0.115
ent_coef	0.00289
ent_coef_loss	0.117
learning_rate	0.0001
n_updates	254134

Eval num_timesteps=260000, episode_reward=-53.58 +/- 2.52
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-53.6
time/	
total_timesteps	260000
train/	
actor_loss	-4.2
critic_loss	0.493
ent_coef	0.00302
ent_coef_loss	1.16
learning_rate	0.0001

learning_rate	0.0001
n_updates	254999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	66.9
time/	
episodes	264
fps	18
time_elapsed	13973
total_timesteps	263135
train/	
actor_loss	-4.05
critic_loss	0.0581
ent_coef	0.00348
ent_coef_loss	0.947
learning_rate	0.0001
n_updates	258134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	57.3
time/	
episodes	268
fps	18
time_elapsed	14177
total_timesteps	267135
train/	
actor_loss	-4.72
critic_loss	0.0577
ent_coef	0.00345
ent_coef_loss	-1.24
learning_rate	0.0001
n_updates	262134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	51.3
time/	
episodes	272
fps	18
time_elapsed	14378
total_timesteps	271135
train/	
actor_loss	-4.42
critic_loss	0.0845
ent_coef	0.0033
ent_coef_loss	1
learning_rate	0.0001
n_updates	266134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	46.8
time/	
episodes	276
fps	18
time_elapsed	14582
total_timesteps	275135
train/	
actor_loss	4.71

actor_loss	-4.11
critic_loss	0.341
ent_coef	0.00302
ent_coef_loss	0.352
learning_rate	0.0001
n_updates	270134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	47.6
time/	
episodes	280
fps	18
time_elapsed	14785
total_timesteps	279135
train/	
actor_loss	-5.81
critic_loss	0.105
ent_coef	0.00309
ent_coef_loss	0.34
learning_rate	0.0001
n_updates	274134

Eval num_timesteps=280000, episode_reward=-60.06 +/- 10.65
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-60.1
time/	
total_timesteps	280000
train/	
actor_loss	-5.43
critic_loss	0.63
ent_coef	0.00303
ent_coef_loss	-0.336
learning_rate	0.0001
n_updates	274999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	48
time/	
episodes	284
fps	18
time_elapsed	15051
total_timesteps	283135
train/	
actor_loss	-4.95
critic_loss	0.15
ent_coef	0.00321
ent_coef_loss	-0.114
learning_rate	0.0001
n_updates	278134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	29.9
time/	
episodes	288
fps	18
time_elapsed	15254

	total_timesteps
train/	
actor_loss	-5.79
critic_loss	0.302
ent_coef	0.00326
ent_coef_loss	-1.66
learning_rate	0.0001
n_updates	282134

	rollout/
	ep_len_mean
	1e+03
	ep_rew_mean
	2.95
time/	
episodes	292
fps	18
time_elapsed	15457
total_timesteps	291135
train/	
actor_loss	-5.39
critic_loss	0.156
ent_coef	0.00324
ent_coef_loss	0.902
learning_rate	0.0001
n_updates	286134

	rollout/
	ep_len_mean
	1e+03
	ep_rew_mean
	-25.9
time/	
episodes	296
fps	18
time_elapsed	15658
total_timesteps	295135
train/	
actor_loss	-5.48
critic_loss	0.308
ent_coef	0.00342
ent_coef_loss	1.86
learning_rate	0.0001
n_updates	290134

	rollout/
	ep_len_mean
	1e+03
	ep_rew_mean
	-49.2
time/	
episodes	300
fps	18
time_elapsed	15861
total_timesteps	299135
train/	
actor_loss	-6.06
critic_loss	0.115
ent_coef	0.00372
ent_coef_loss	-0.585
learning_rate	0.0001
n_updates	294134

Eval num_timesteps=300000, episode_reward=-73.17 +/- 11.71
 Episode length: 1000.00 +/- 0.00

	eval/

mean_ep_length	1e+03
mean_reward	-73.2
time/	
total_timesteps	300000
train/	
actor_loss	-5.34
critic_loss	0.34
ent_coef	0.00378
ent_coef_loss	-0.41
learning_rate	0.0001
n_updates	294999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-57.9
time/	
episodes	304
fps	18
time_elapsed	16131
total_timesteps	303135
train/	
actor_loss	-6.56
critic_loss	0.23
ent_coef	0.00387
ent_coef_loss	1.09
learning_rate	0.0001
n_updates	298134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-60.5
time/	
episodes	308
fps	18
time_elapsed	16334
total_timesteps	307135
train/	
actor_loss	-5.9
critic_loss	0.279
ent_coef	0.00413
ent_coef_loss	0.17
learning_rate	0.0001
n_updates	302134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-59.6
time/	
episodes	312
fps	18
time_elapsed	16537
total_timesteps	311135
train/	
actor_loss	-5.95
critic_loss	0.384
ent_coef	0.00457
ent_coef_loss	-0.825
learning_rate	0.0001
n_updates	306134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-67.8
time/	
episodes	316
fps	18
time_elapsed	16740
total_timesteps	315135
train/	
actor_loss	-6.62
critic_loss	0.309
ent_coef	0.00524
ent_coef_loss	1.54
learning_rate	0.0001
n_updates	310134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-66.9
time/	
episodes	320
fps	18
time_elapsed	16944
total_timesteps	319135
train/	
actor_loss	-7.4
critic_loss	0.682
ent_coef	0.00594
ent_coef_loss	1.52
learning_rate	0.0001
n_updates	314134

Eval num_timesteps=320000, episode_reward=-48.35 +/- 28.16
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-48.4
time/	
total_timesteps	320000
train/	
actor_loss	-6.83
critic_loss	0.406
ent_coef	0.00615
ent_coef_loss	-0.399
learning_rate	0.0001
n_updates	314999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-64.8
time/	
episodes	324
fps	18
time_elapsed	17211
total_timesteps	323135
train/	
actor_loss	-7.87
critic_loss	0.589
ent_coef	0.00677
ent_coef_loss	0.725
learning_rate	0.0001

	n_updates	318134
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-62.8	
time/		
episodes	328	
fps	18	
time_elapsed	17413	
total_timesteps	327135	
train/		
actor_loss	-7.9	
critic_loss	0.385	
ent_coef	0.00713	
ent_coef_loss	-0.718	
learning_rate	0.0001	
n_updates	322134	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-60.1	
time/		
episodes	332	
fps	18	
time_elapsed	17613	
total_timesteps	331135	
train/		
actor_loss	-9.73	
critic_loss	0.387	
ent_coef	0.00691	
ent_coef_loss	-1.62	
learning_rate	0.0001	
n_updates	326134	
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rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-58.2	
time/		
episodes	336	
fps	18	
time_elapsed	17816	
total_timesteps	335135	
train/		
actor_loss	-8.52	
critic_loss	0.426	
ent_coef	0.00607	
ent_coef_loss	0.907	
learning_rate	0.0001	
n_updates	330134	
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rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-55.6	
time/		
episodes	340	
fps	18	
time_elapsed	18018	
total_timesteps	339135	
train/		
actor_loss	-7.57	

critic_loss	0.355
ent_coef	0.00637
ent_coef_loss	-1.84
learning_rate	0.0001
n_updates	334134

Eval num_timesteps=340000, episode_reward=-59.16 +/- 8.49
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-59.2
time/	
total_timesteps	340000
train/	
actor_loss	-8.04
critic_loss	0.463
ent_coef	0.00622
ent_coef_loss	0.527
learning_rate	0.0001
n_updates	334999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-52
time/	
episodes	344
fps	18
time_elapsed	18285
total_timesteps	343135
train/	
actor_loss	-9.57
critic_loss	0.629
ent_coef	0.00557
ent_coef_loss	-0.654
learning_rate	0.0001
n_updates	338134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-48.7
time/	
episodes	348
fps	18
time_elapsed	18488
total_timesteps	347135
train/	
actor_loss	-8.45
critic_loss	0.542
ent_coef	0.00537
ent_coef_loss	-1.55
learning_rate	0.0001
n_updates	342134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-45.4
time/	
episodes	352
fps	18
time_elapsed	18690

	total_timesteps	501155
train/		
actor_loss	-11	
critic_loss	0.866	
ent_coef	0.00598	
ent_coef_loss	-0.71	
learning_rate	0.0001	
n_updates	346134	

	rollout/	
	ep_len_mean	1e+03
	ep_rew_mean	-46
time/		
	episodes	356
	fps	18
	time_elapsed	18893
	total_timesteps	355135
train/		
	actor_loss	-8.39
	critic_loss	0.492
	ent_coef	0.0058
	ent_coef_loss	-0.649
	learning_rate	0.0001
	n_updates	350134

	rollout/	
	ep_len_mean	1e+03
	ep_rew_mean	-46.5
time/		
	episodes	360
	fps	18
	time_elapsed	19098
	total_timesteps	359135
train/		
	actor_loss	-9.98
	critic_loss	0.846
	ent_coef	0.00601
	ent_coef_loss	-0.16
	learning_rate	0.0001
	n_updates	354134

Eval num_timesteps=360000, episode_reward=-56.11 +/- 15.32
 Episode length: 1000.00 +/- 0.00

	eval/	
	mean_ep_length	1e+03
	mean_reward	-56.1
time/		
	total_timesteps	360000
train/		
	actor_loss	-12.1
	critic_loss	4.39
	ent_coef	0.00589
	ent_coef_loss	-2.49
	learning_rate	0.0001
	n_updates	354999

	rollout/	
	ep_len_mean	1e+03
	ep_rew_mean	-46.9
time/		
	episodes	364

fps	18
time_elapsed	19363
total_timesteps	363135
train/	
actor_loss	-9.24
critic_loss	1.48
ent_coef	0.006
ent_coef_loss	-0.277
learning_rate	0.0001
n_updates	358134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-46.7
time/	
episodes	368
fps	18
time_elapsed	19565
total_timesteps	367135
train/	
actor_loss	-9.34
critic_loss	2.01
ent_coef	0.0063
ent_coef_loss	0.286
learning_rate	0.0001
n_updates	362134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-45.8
time/	
episodes	372
fps	18
time_elapsed	19767
total_timesteps	371135
train/	
actor_loss	-8.9
critic_loss	1.31
ent_coef	0.00719
ent_coef_loss	-0.749
learning_rate	0.0001
n_updates	366134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-45.4
time/	
episodes	376
fps	18
time_elapsed	19970
total_timesteps	375135
train/	
actor_loss	-12
critic_loss	14.4
ent_coef	0.00828
ent_coef_loss	2.33
learning_rate	0.0001
n_updates	370134

rollout/	

.....	
ep_len_mean	1e+03
ep_rew_mean	-45.5
time/	
episodes	380
fps	18
time_elapsed	20172
total_timesteps	379135
train/	
actor_loss	-11.3
critic_loss	4.01
ent_coef	0.00893
ent_coef_loss	1.5
learning_rate	0.0001
n_updates	374134

Eval num_timesteps=380000, episode_reward=-78.72 +/- 12.76
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-78.7
time/	
total_timesteps	380000
train/	
actor_loss	-9.7
critic_loss	3.74
ent_coef	0.00913
ent_coef_loss	1.16
learning_rate	0.0001
n_updates	374999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-45.2
time/	
episodes	384
fps	18
time_elapsed	20436
total_timesteps	383135
train/	
actor_loss	-13
critic_loss	5.46
ent_coef	0.0102
ent_coef_loss	0.19
learning_rate	0.0001
n_updates	378134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-45.4
time/	
episodes	388
fps	18
time_elapsed	20639
total_timesteps	387135
train/	
actor_loss	-21.5
critic_loss	15.8
ent_coef	0.0131
ent_coef_loss	-0.185
learning_rate	0.0001
n_updates	382134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-46.5
time/	
episodes	392
fps	18
time_elapsed	20840
total_timesteps	391135
train/	
actor_loss	-22.4
critic_loss	15.2
ent_coef	0.0163
ent_coef_loss	0.68
learning_rate	0.0001
n_updates	386134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-47.5
time/	
episodes	396
fps	18
time_elapsed	21048
total_timesteps	395135
train/	
actor_loss	-36.3
critic_loss	38.6
ent_coef	0.021
ent_coef_loss	1.78
learning_rate	0.0001
n_updates	390134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-47.8
time/	
episodes	400
fps	18
time_elapsed	21255
total_timesteps	399135
train/	
actor_loss	-26
critic_loss	29.2
ent_coef	0.024
ent_coef_loss	1.55
learning_rate	0.0001
n_updates	394134

Eval num_timesteps=400000, episode_reward=-81.05 +/- 4.42
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-81.1
time/	
total_timesteps	400000
train/	
actor_loss	-18.2
critic_loss	174
ent_coef	0.0258

ent_coef_loss	0.987
learning_rate	0.0001
n_updates	394999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-48.1
time/	
episodes	404
fps	18
time_elapsed	21524
total_timesteps	403135
train/	
actor_loss	-44.6
critic_loss	133
ent_coef	0.0306
ent_coef_loss	1
learning_rate	0.0001
n_updates	398134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-49
time/	
episodes	408
fps	18
time_elapsed	21729
total_timesteps	407135
train/	
actor_loss	-49.7
critic_loss	168
ent_coef	0.0353
ent_coef_loss	0.691
learning_rate	0.0001
n_updates	402134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-49.4
time/	
episodes	412
fps	18
time_elapsed	21934
total_timesteps	411135
train/	
actor_loss	-60.8
critic_loss	129
ent_coef	0.0407
ent_coef_loss	0.00503
learning_rate	0.0001
n_updates	406134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-49.5
time/	
episodes	416
fps	18
time_elapsed	22138
total_timesteps	415135

train/	
actor_loss	-47.2
critic_loss	67.1
ent_coef	0.0402
ent_coef_loss	-1.49
learning_rate	0.0001
n_updates	410134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-50.1
time/	
episodes	420
fps	18
time_elapsed	22342
total_timesteps	419135
train/	
actor_loss	-56
critic_loss	168
ent_coef	0.0448
ent_coef_loss	0.622
learning_rate	0.0001
n_updates	414134

Eval num_timesteps=420000, episode_reward=-73.35 +/- 8.27

eval/	
mean_ep_length	1e+03
mean_reward	-73.4
time/	
total_timesteps	420000
train/	
actor_loss	-50.5
critic_loss	122
ent_coef	0.0452
ent_coef_loss	-0.676
learning_rate	0.0001
n_updates	414999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-52
time/	
episodes	424
fps	18
time_elapsed	22613
total_timesteps	423135
train/	
actor_loss	-56.2
critic_loss	1.52e+03
ent_coef	0.0484
ent_coef_loss	-0.323
learning_rate	0.0001
n_updates	418134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-53.9
time/	
episodes	428
fps	18

time_elapsed	22819
total_timesteps	427135
train/	
actor_loss	-67.8
critic_loss	72
ent_coef	0.0465
ent_coef_loss	-0.389
learning_rate	0.0001
n_updates	422134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-56.2
time/	
episodes	432
fps	18
time_elapsed	23027
total_timesteps	431135
train/	
actor_loss	-70.6
critic_loss	305
ent_coef	0.0517
ent_coef_loss	-0.589
learning_rate	0.0001
n_updates	426134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-57.2
time/	
episodes	436
fps	18
time_elapsed	23232
total_timesteps	435135
train/	
actor_loss	-86.1
critic_loss	619
ent_coef	0.05
ent_coef_loss	-0.805
learning_rate	0.0001
n_updates	430134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-61.2
time/	
episodes	440
fps	18
time_elapsed	23437
total_timesteps	439135
train/	
actor_loss	-124
critic_loss	420
ent_coef	0.0556
ent_coef_loss	0.813
learning_rate	0.0001
n_updates	434134

Eval num_timesteps=440000, episode_reward=-80.17 +/- 10.75
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-80.2
time/	
total_timesteps	440000
train/	
actor_loss	-97
critic_loss	167
ent_coef	0.0592
ent_coef_loss	-0.212
learning_rate	0.0001
n_updates	434999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-63.4
time/	
episodes	444
fps	18
time_elapsed	23706
total_timesteps	443135
train/	
actor_loss	-104
critic_loss	600
ent_coef	0.069
ent_coef_loss	0.213
learning_rate	0.0001
n_updates	438134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-66.3
time/	
episodes	448
fps	18
time_elapsed	23910
total_timesteps	447135
train/	
actor_loss	-114
critic_loss	329
ent_coef	0.076
ent_coef_loss	0.0772
learning_rate	0.0001
n_updates	442134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-69.5
time/	
episodes	452
fps	18
time_elapsed	24114
total_timesteps	451135
train/	
actor_loss	-203
critic_loss	1.18e+03
ent_coef	0.0882
ent_coef_loss	0.381
learning_rate	0.0001
n_updates	446134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-69.5
<hr/>	
time/	
episodes	456
fps	18
time_elapsed	24317
total_timesteps	455135
<hr/>	
train/	
actor_loss	-217
critic_loss	636
ent_coef	0.109
ent_coef_loss	-0.314
learning_rate	0.0001
n_updates	450134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-70.1
<hr/>	
time/	
episodes	460
fps	18
time_elapsed	24520
total_timesteps	459135
<hr/>	
train/	
actor_loss	-168
critic_loss	1.43e+03
ent_coef	0.117
ent_coef_loss	-0.317
learning_rate	0.0001
n_updates	454134

Eval num_timesteps=460000, episode_reward=-82.05 +/- 12.66
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-82.1
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time/	
total_timesteps	460000
<hr/>	
train/	
actor_loss	-172
critic_loss	670
ent_coef	0.115
ent_coef_loss	-0.368
learning_rate	0.0001
n_updates	454999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-70.3
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time/	
episodes	464
fps	18
time_elapsed	24789
total_timesteps	463135
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train/	
actor_loss	-195
critic_loss	1.2e+03
ent_coef	0.108
ent_coef_loss	0.384
learning rate	0.0001

	n_updates	458134
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rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-71.9	
time/		
episodes	468	
fps	18	
time_elapsed	24994	
total_timesteps	467135	
train/		
actor_loss	-373	
critic_loss	1.38e+04	
ent_coef	0.118	
ent_coef_loss	0.207	
learning_rate	0.0001	
n_updates	462134	
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rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-73.2	
time/		
episodes	472	
fps	18	
time_elapsed	25199	
total_timesteps	471135	
train/		
actor_loss	-226	
critic_loss	1.6e+03	
ent_coef	0.115	
ent_coef_loss	0.197	
learning_rate	0.0001	
n_updates	466134	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-73.6	
time/		
episodes	476	
fps	18	
time_elapsed	25406	
total_timesteps	475135	
train/		
actor_loss	-268	
critic_loss	2.18e+03	
ent_coef	0.122	
ent_coef_loss	0.126	
learning_rate	0.0001	
n_updates	470134	
<hr/>		
rollout/		
ep_len_mean	1e+03	
ep_rew_mean	-74.4	
time/		
episodes	480	
fps	18	
time_elapsed	25608	
total_timesteps	479135	
train/		
actor loss	-238	

critic_loss	1.52e+03
ent_coef	0.135
ent_coef_loss	0.0812
learning_rate	0.0001
n_updates	474134

Eval num_timesteps=480000, episode_reward=-90.92 +/- 1.49
 Episode length: 1000.00 +/- 0.00

eval/	
mean_ep_length	1e+03
mean_reward	-90.9
time/	
total_timesteps	480000
train/	
actor_loss	-280
critic_loss	1.09e+03
ent_coef	0.135
ent_coef_loss	-0.236
learning_rate	0.0001
n_updates	474999

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-75.5
time/	
episodes	484
fps	18
time_elapsed	25873
total_timesteps	483135
train/	
actor_loss	-195
critic_loss	759
ent_coef	0.133
ent_coef_loss	0.277
learning_rate	0.0001
n_updates	478134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-75.7
time/	
episodes	488
fps	18
time_elapsed	26077
total_timesteps	487135
train/	
actor_loss	-286
critic_loss	2.39e+03
ent_coef	0.136
ent_coef_loss	-0.0232
learning_rate	0.0001
n_updates	482134

rollout/	
ep_len_mean	1e+03
ep_rew_mean	-75.5
time/	
episodes	492
fps	18
time_elapsed	26280

```
| total timesteps | 491135 |
# -----
# Helpers to eval with true env rewards
# -----
def load_eval_env_with_stats():
    env = make_eval_env(SEED + 100)
    env = VecNormalize.load(os.path.join(LOG_DIR, "vecnormalize.pkl"), env)
    env.training = False
    env.norm_reward = False # report true env rewards
    return env

# -----
# Evaluate FINAL model
# -----
eval_env_final = load_eval_env_with_stats()
final_model = SAC.load(final_model_path, env=eval_env_final)
mean_r, std_r = evaluate_policy(final_model, eval_env_final, n_eval_episodes=FINAL_EVAL_EPISODES, deterministic=True)
print(f"[FINAL @500k] Mean reward: {mean_r:.2f} +/- {std_r:.2f}")
eval_env_final.close()
```

[FINAL @500k] Mean reward: 1344.95 +/- 10.34

```
# -----
# Evaluate BEST model
# -----
best_model_path = os.path.join(LOG_DIR, "best_model.zip")
if os.path.exists(best_model_path):
    eval_env_best = load_eval_env_with_stats()
    best_model = SAC.load(best_model_path, env=eval_env_best)
    mean_r_b, std_r_b = evaluate_policy(best_model, eval_env_best, n_eval_episodes=FINAL_EVAL_EPISODES, deterministic=True)
    print(f"[BEST @500k] Mean reward: {mean_r_b:.2f} +/- {std_r_b:.2f}")
    eval_env_best.close()
else:
    print("No best_model.zip found (EvalCallback may not have improved over initial policy.)")
```

[BEST @500k] Mean reward: 682.02 +/- 268.29
Eval num_timesteps=500000, episode_reward=-81.52 +/- 7.12

```
# -----
# Record short video (best if available else final)
# -----
def record_video(model_path, out_prefix="best_model"):
    env = load_eval_env_with_stats()
    env = VecVideoRecorder(
        env,
        VIDEO_DIR,
        record_video_trigger=lambda step: step == 0,
        video_length=1000, # CarRacing horizon ~1000 steps
        name_prefix=f"{out_prefix}_car_racing_sac_500k",
    )
    obs = env.reset()
    # Load model once outside the loop
    policy = SAC.load(model_path, env=env)
    for _ in range(1000):
        action, _ = policy.predict(obs, deterministic=True)
```

```

obs, rewards, dones, infos = env.step(action)
if dones[0]:
    break
env.close()

if os.path.exists(best_model_path):
    record_video(best_model_path, out_prefix="best_model")
else:
    record_video(final_model_path, out_prefix="final_model")

print(f"Video(s) saved to: {VIDEO_DIR}")

Saving video to /content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4
/usr/local/lib/python3.12/dist-packages/moviepy/config_defaults.py:47: SyntaxWarning: invalid escape sequence '\P'
    IMAGEMAGICK_BINARY = r"C:\Program Files\ImageMagick-6.8.8-Q16\magick.exe"
Moviepy - Building video /content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4.
Moviepy - Writing video /content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4

/usr/local/lib/python3.12/dist-packages/jupyter_client/session.py:203: DeprecationWarning: datetime.datetime.utcnow() is deprecated and scheduled for removal in
    return datetime.utcnow().replace(tzinfo=utc)
Moviepy - Done !
Moviepy - video ready /content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4
Video(s) saved to: ./videos

```

```

# See it
import glob
glob.glob("/content/videos/*.mp4")

# Download to your computer
from google.colab import files
files.download("/content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4")

```

```

from google.colab import drive
drive.mount('/content/drive')

import os, shutil
DEST = "/content/drive/MyDrive/CarRacingVideos"
os.makedirs(DEST, exist_ok=True)
shutil.copy("/content/videos/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4",
            f"{DEST}/best_model_car_racing_sac_500k-step-0-to-step-1000.mp4")
print("Copied to:", DEST)

```

Mounted at /content/drive
Copied to: /content/drive/MyDrive/CarRacingVideos

```

# -----
# Plot evaluation curve (from EvalCallback)
# -----
eval_npz = os.path.join(LOG_DIR, "evaluations.npz")
if os.path.exists(eval_npz):
    data = np.load(eval_npz)
    timesteps = data["timesteps"]
    results = data["results"]

```

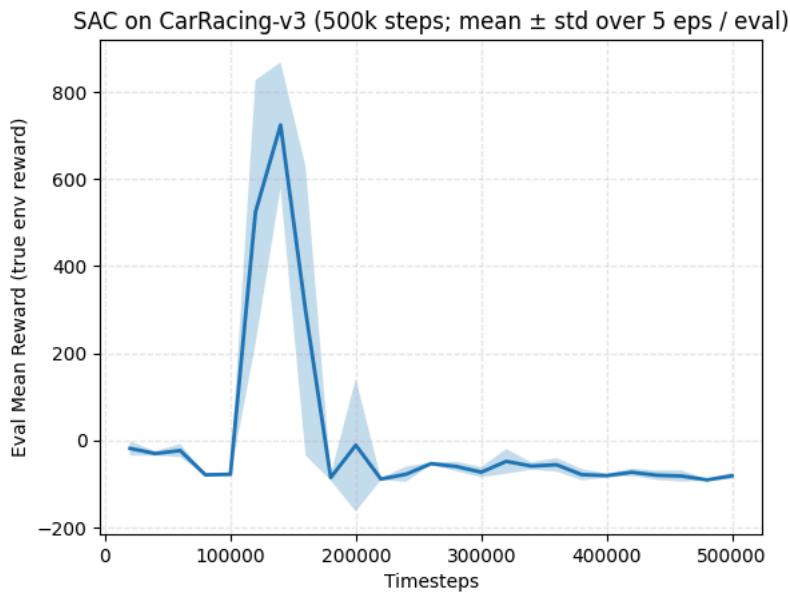
```

mean_results = results.mean(axis=1)
std_results = results.std(axis=1)

plt.figure()
plt.plot(timesteps, mean_results, linewidth=2)
plt.fill_between(timesteps, mean_results - std_results, mean_results + std_results, alpha=0.25)
plt.xlabel("Timesteps")
plt.ylabel("Eval Mean Reward (true env reward)")
plt.title(f"SAC on {ENV_ID} (500k steps; mean ± std over {EVAL_EPISODES} eps / eval)")
plt.grid(True, linestyle="--", alpha=0.3)
plt.show()

else:
    print("evaluations.npz not found; was EvalCallback attached?")

```



```

import numpy as np
import matplotlib.pyplot as plt
import os

LOG_DIR = "./logs/CarRacing-v3" # change if yours differs
npz_path = os.path.join(LOG_DIR, "evaluations.npz")
data = np.load(npz_path)

# results shape: (num_evals, episodes_per_eval)
results = data["results"]

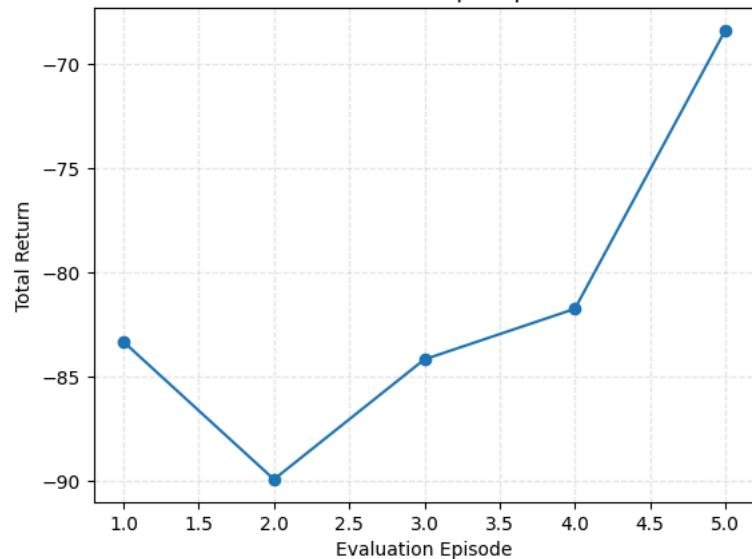
# Option 1: last evaluation only
last_returns = results[-1] # shape (episodes_per_eval,)
plt.figure()
plt.plot(range(1, len(last_returns)+1), last_returns, marker="o")
plt.xlabel("Evaluation Episode")
plt.ylabel("Total Return")
plt.title("Last Eval: Return per Episode")

```

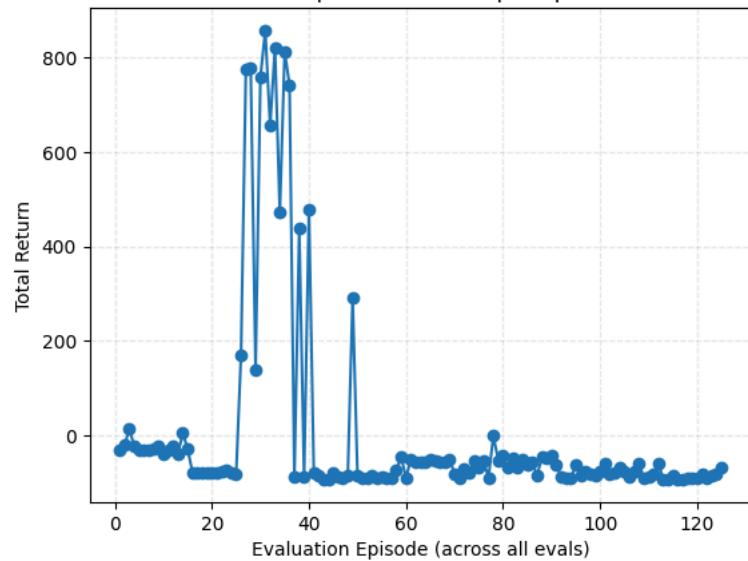
```
plt.grid(True, linestyle="--", alpha=0.3)
plt.show()

# Option 2: all evaluations flattened into a single sequence
flat_returns = results.reshape(-1)
plt.figure()
plt.plot(range(1, len(flat_returns)+1), flat_returns, marker="o")
plt.xlabel("Evaluation Episode (across all evals)")
plt.ylabel("Total Return")
plt.title("All Eval Episodes: Return per Episode")
plt.grid(True, linestyle="--", alpha=0.3)
plt.show()
```

Last Eval: Return per Episode



All Eval Episodes: Return per Episode



```
# Rebuild eval env with saved VecNormalize stats, then disable reward norm
eval_env = load_eval_env_with_stats()

# Load whichever model you want to plot
model_to_plot = SAC.load(best_model_path, env=eval_env) # or final_model_path

# Get per-episode returns
episode_returns, episode_lengths = evaluate_policy()
```

```
model_to_plot, eval_env,
n_eval_episodes=20,
deterministic=True,
return_episode_rewards=True
)
```