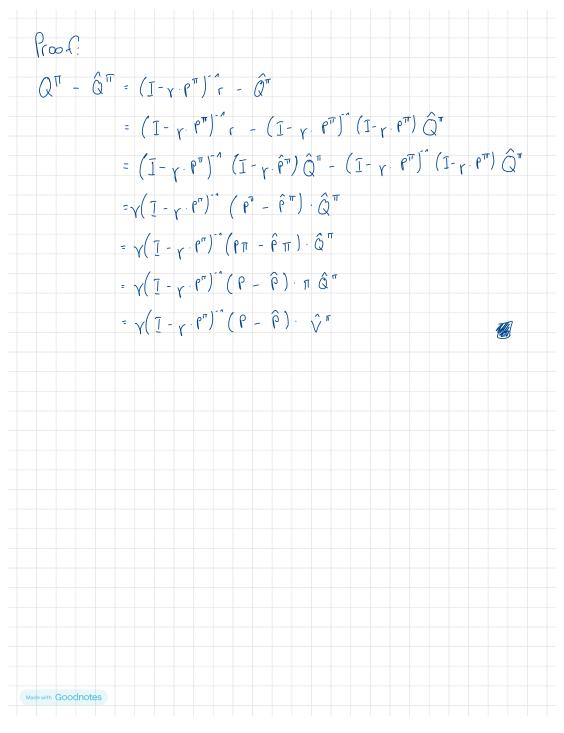
[Mustafa Suman] Assignment 4

Due November 03, 11:59 pm

2 Analysis

2.1 Alternative Simulation Lemma



2.2 Statements

- 1. True.
- 2. False.
- 3. True.
- 4. False.

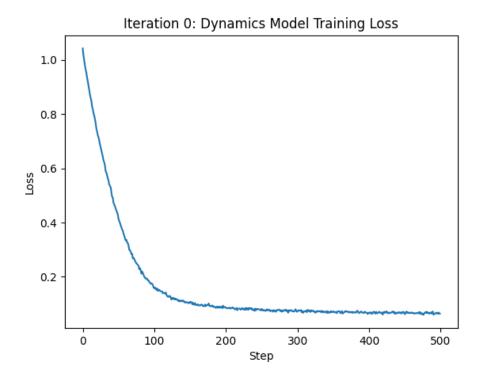


Figure 1: halfcheetah_0_iter; default settings

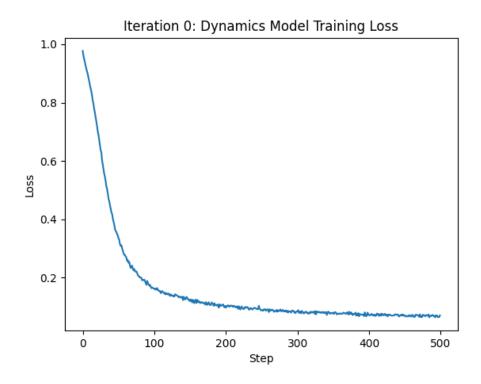


Figure 2: halfcheetah_0_iter; 3 layers

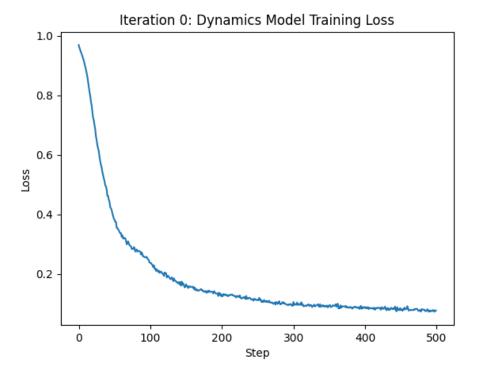


Figure 3: halfcheetah $_0$ _iter; 5 layers

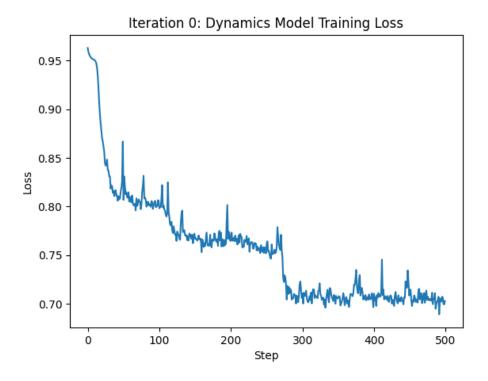


Figure 4: halfcheetah_0_iter; 20 layers

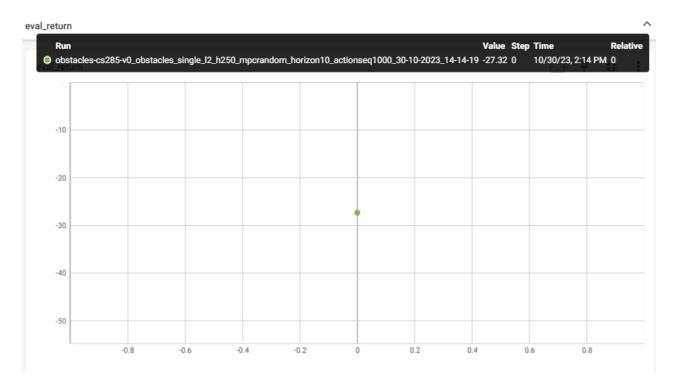


Figure 5: obstacles_1_iter.yaml eval is above minus seventy

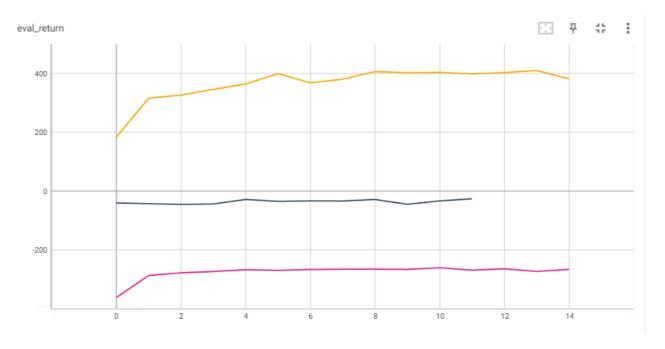


Figure 6: yellow = cheetah, black = obstacles, orange = reacher

Problem 4

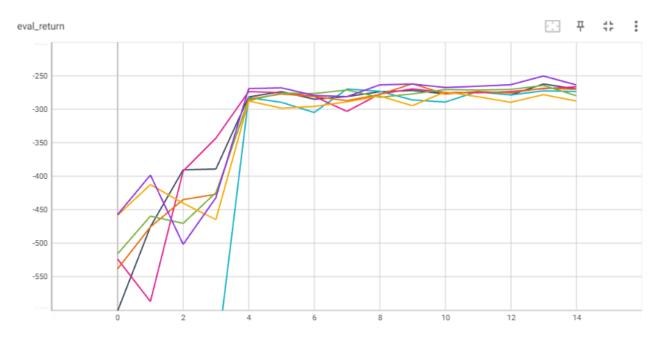


Figure 7: dark blue = default, blue = ensemble size d(ecreased), pink = ensemble size i(increased), orange = horizon i, purple = Horizon d, green = # ac_sequences d, orange (light) = # ac_sequences i

First of all, we can observe that after iteration 4, all of the approaches seem to perform pretty similar. Before that, there is some discrepancy.

- Ensemble size: The default performs very similar compared to the increased one. The decreased one is performing far worse at the beginning. Overall, it has the lowest starting and steepest learning phase.
- Horizon: No big impact. The decreased as well as the increased version perform worse than the default at the beginning. Might be due to randomness though.
- #Action sequences: No big impact/variation.

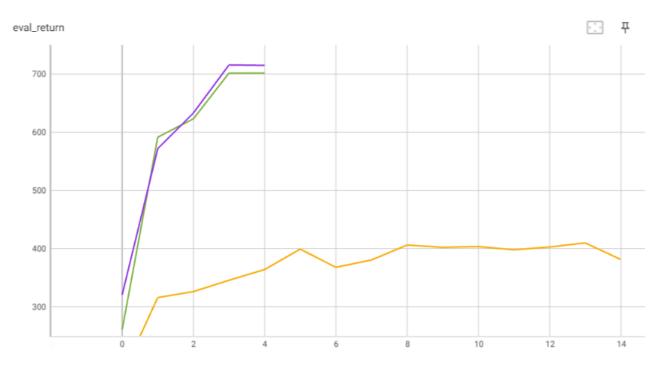


Figure 8: purple = n_iter 2, green = n_iter 4, orange = random shooting

The CEM strategy performs far better than the random shooting approach. The number of iterations does not seem to have a significant impact.

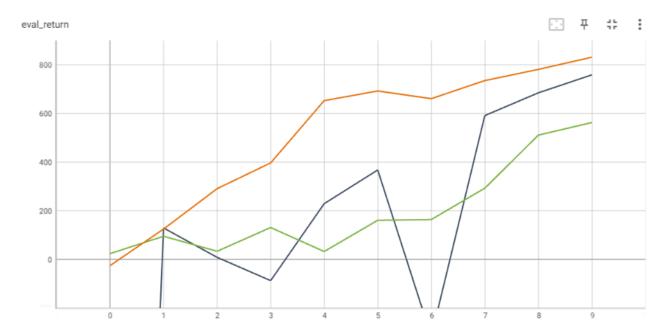


Figure 9: blue = model-free, green = Dyna-like, orange = full MBPO

We can see that the model-free and Dyna-like method perform quite similar, but the model-free method is way more volatile. The full MBPO approach clearly outperforms the other two, reaching higher rewards, doing so quicker and even in a more stable process!