
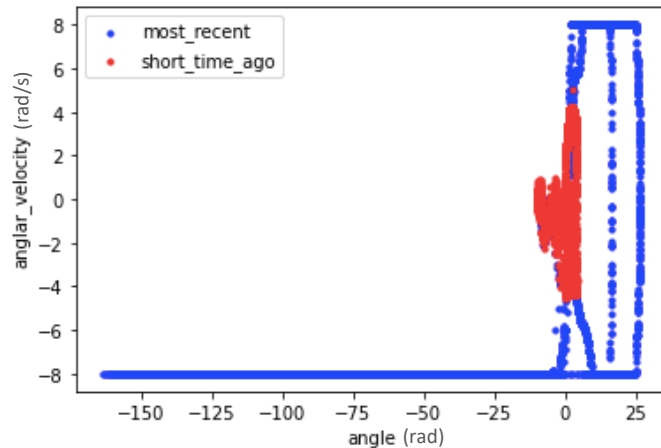


- Report on last week
  - It was possible to apply RL to event-triggered control
  - It will be applicable to self-triggered control as well
- (At least with same steps,) it was not conducted well
  - Algorithm did not converge
- Conjecture:  
**The control performance changes rapidly with little parameter change**
- If the conjecture is valid, research the condition and consider solution
  -  Check the validity (conclusion: NO)

# Weekly Report

M2 Ibuki Takeuchi

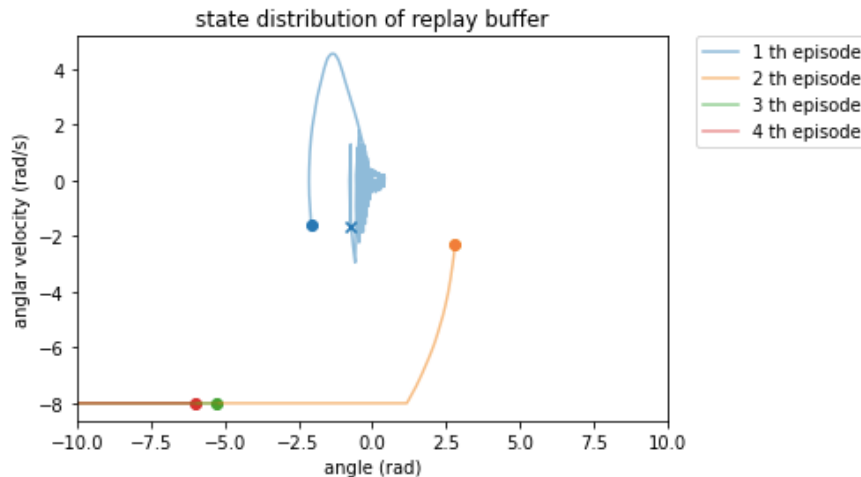
- Recently experienced data



- Although parameter changes slightly, control path changes drastically(?)

- This scatter plot does not show control path

- Depict control path for each episode(new starting point)



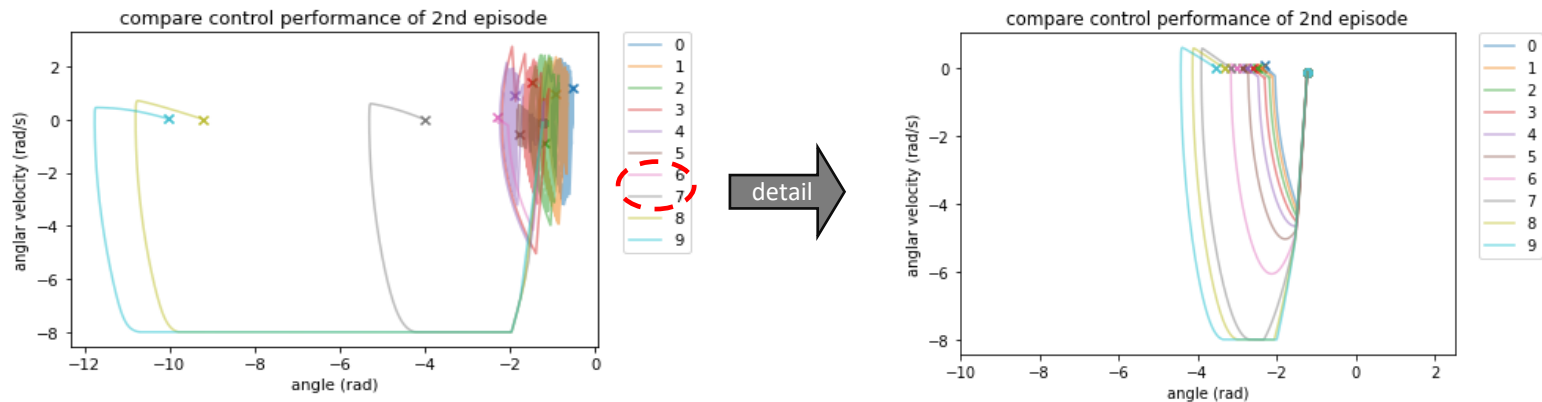
- Assume
  - Policy parameters are updated in all steps
  - Initial policy stabilize all initial states

- It is not possible to see why policy became bad with this picture
  - (P1) Control path suddenly changed in **second** episode (conjecture)
  - (P2) Overfit to **blue** path
  - (P3) Other reason

# Weekly Report

M2 Ibuki Takeuchi

- How the policy changes in 2<sup>nd</sup> episode
  - Pick up policy in equally step interval  
ex.) step 100, 200, 300, ...



(Attention: policy are not updated in each path)

- Control path does not change suddenly
  - The conjecture is not valid

- Next step: Reconsider the cause for policy deterioration
  - policy gradient approximation

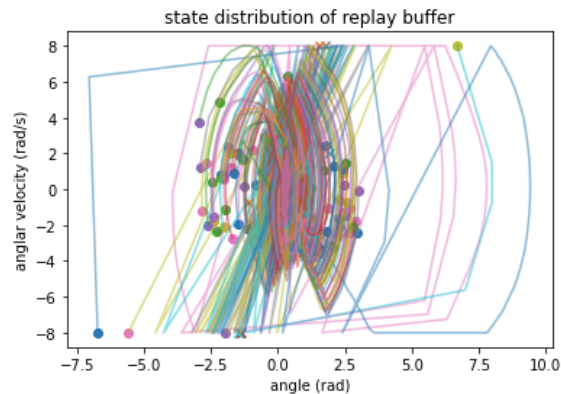
$$g = \frac{1}{N} \sum_{s \in E} [\nabla_{\theta} \pi(s|\theta) \nabla_a Q(s, a|\omega)|_{a=\pi(s|\theta)}]$$
$$\approx \mathbb{E}_{s \sim \rho^{\pi_{\theta}}} [\nabla_{\theta} \pi_{\theta}(s) \nabla_a Q^{\theta}(s, a)|_{a=\pi(s|\theta)}]$$

- This assumes experienced states is well approximates  $\rho^{\pi_{\theta}}$
  - In other words, improve policy by **prioritizing only experienced states**
  - Lack of the number of episodes makes policy overfitting
- For data efficiency, by enlarging minimum interval time, increase the number of control paths in replay buffer ( $N$  steps experienced data)

# Weekly Report

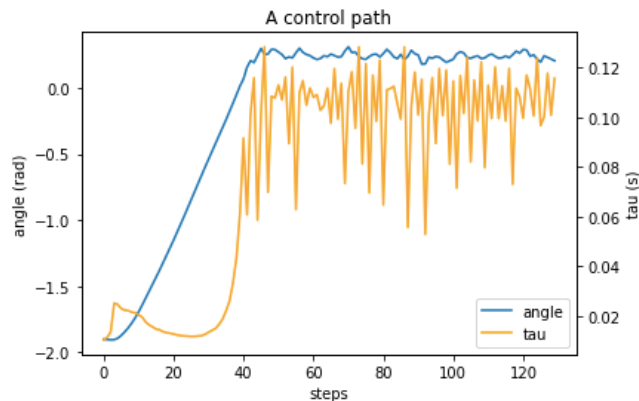
M2 Ibuki Takeuchi

- By changing learning configuration as last slide, policy may improved
  - Distribution of replay buffer



- There is no divergent paths

- Learned policy



There is no guarantee that this policy is the best policy ...

- Wide interval around origin and frequent otherwise
- Stabilize the system

# Weekly Report

M2 Ibuki Takeuchi

- This week
  - Reconsider and discuss the theme
  - Ideas for stable learning
    - Configuration of optimizer (learning rate hyperparameter etc..)
    - Safe learning