

THESIS PLAN: "SCALABLE CONTINUOUS RL WITH KULLBACK-LEIBLER POLICY CHAIN"
UNDERGRADUATE FINAL YEAR PROJECT
CHRIS DOYLE

DEADLINES:

- 25th January – Introduction/Background
- 25th March – [Entire Thesis] & Demo to be completed
- 25th March – Demo Presentation
- 12th April – Final Deadline for Thesis

BY 25th JANUARY:

ABSTRACT (OUTLINE PROBLEM & SOLUTION):

- ☒ Define lifelong learning.
- ☒ Define lifelong reinforcement learning.
- ☒ Existing issues being solved (scalability, catastrophic forgetting).
- ☐ Introduce idea of Kullback-Leibler Policy Chain (i.e. 2 parts, KL & RL).

INTRODUCTION

- ☒ Reinforcement Learning -> Lifelong Learning.
- ☒ Define catastrophic forgetting.
- ☐ What the worst case scenario etc. is.

BACKGROUND

- ☐ Define Reinforcement Learning.
- ☐ Define Policy Gradient Methods.
- ☐ Define Kullback-Leibler Divergence & how it is being used in this project.
- Compare policies in adjacent figures to show effect

BY 25th MARCH

THEORY OF APPROACH

- ☐ Explain the concept of Kullback-Leibler policy chain in a mathematical sense.
- ☐ Derive & explain conceptualised cost function.

PRACTICAL EXPERIMENT

- ☐ Create GridWorld environment & ability to change reward etc.
- ☐ Create RL agent to learn GW.
- ☐ Add functionality to record and turn to GIF.
- ☐ Add functionality to change the reward.
- ☐ Add KLPC functionality.

RESULTS

- ☐ Output of grid with snapshot of arrows whose length reflects the prob of that action.
- ☐ DEMO: GIF of standard RL agent vs GIF of KLPC agent

CONCLUSIONS

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How generalizable is this approach to other environments?

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Compare to other approaches.

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Shortcomings of this approach.

- Outside boundary smaller % of grid as grid grows
- Technique works best when goals of tasks align similarly
- Worst case scenario with infinite tasks: uniform inside, perfect outside