

Introduction to Reinforcement Learning

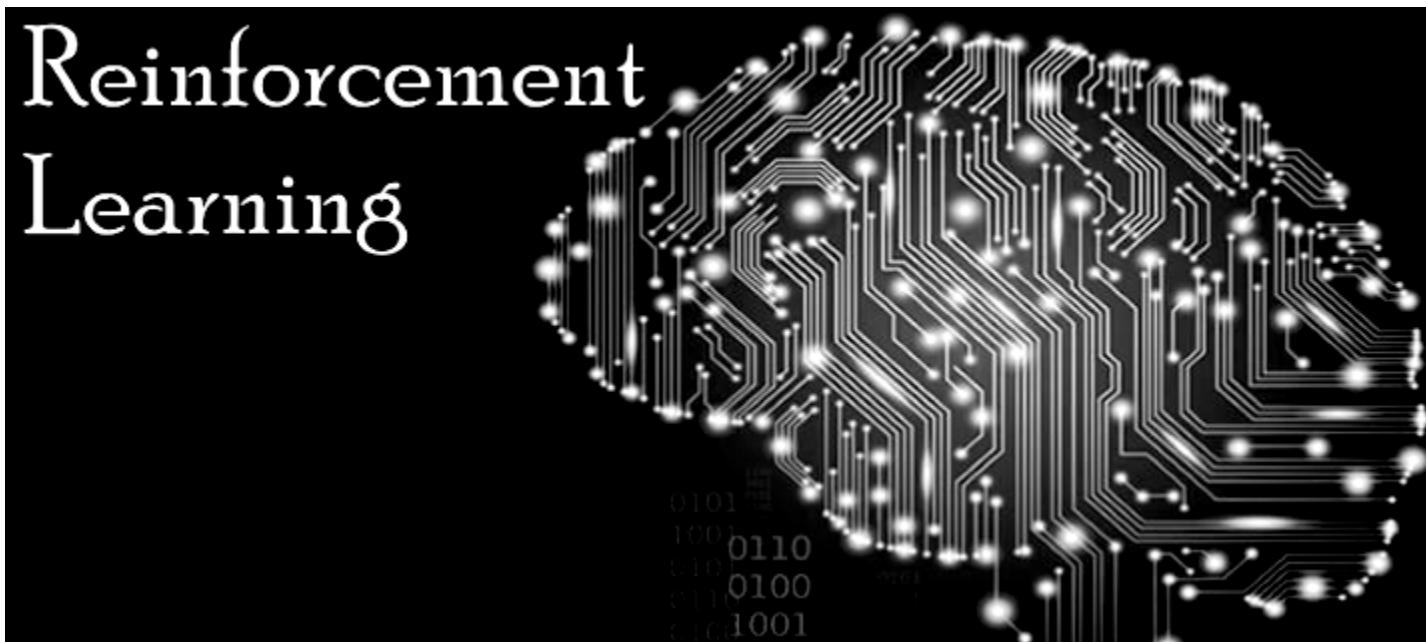
w/ some applications to Energy

LEC 1a: Course Organization

Instructors: Professor Scott Moura & Saehong Park

TA: Xinyi Zhao

Summer 2020



Why take this course?

*Obtain conceptual understanding of
reinforcement learning*

Why take this course? (details)

- Demystify reinforcement learning
- Obtain a solid foundation for the basic fundamental concepts
- Explore via an energy systems example
- It's only EIGHT days and 12 hours

This course is NOT about...

- Open source software
 - e.g. Google Deep Mind, OpenAI, TensorFlow, Matlab RL Toolbox, Pytorch, etc.
- Reducing RL to coding
- Deep RL tricks
- Surveying the most famous algorithms of 2020
 - We take a long time horizon view about the past, present, and future

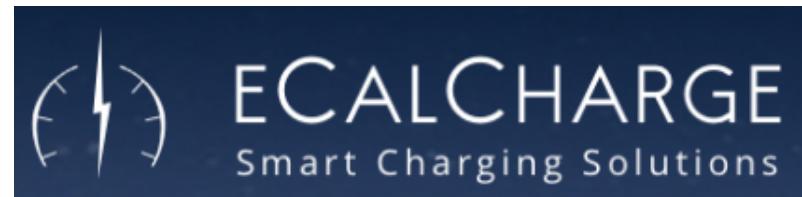
Prerequisite Previous Coursework

- Multivariable calculus
- Linear algebra
- Random variables & Probability
- Optimization
- Machine Learning
- Numerical computing

Very helpful (but not required) previous/concurrent coursework

- Dynamical Systems & Feedback Control
- Stochastic processes
- Statistical learning & Information Theory

Startups out of my courses



Selected Companies that recruit my students



Advanced Microgrid Solutions



TOTAL
COMMITTED TO BETTER ENERGY



Exclusive access to alumni above! Join the CE295 LinkedIn Group: <https://www.linkedin.com/groups/7068321>

Class Format

Live Lecture Dates:

China Time: 7, 8, 9, 10 (Tu – Fri), 14, 15, 16, 17 (Tu – Fri); 08:30 – 10:05

Berkeley Time: 6, 7, 8, 9 (M– Th), 13, 14, 15, 16 (M – Th); 17:30 – 19:05

Remote Teaching: Zoom (detailed login info coming soon)

Course Website: <https://scott-moura.github.io/rl/>

Discussions: WeChat & Slack



Professor Scott MOURA

smoura@berkeley.edu

Office Hours:

30min after class



Co-instructor: Saehong Park

sspark@berkeley.edu

Office Hours: 30min after class



Teaching Assistant:

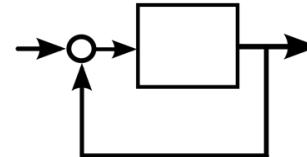
Xinyi Zhao

zxyyx48@163.com

About me

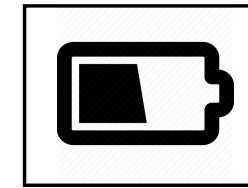


FUNDAMENTAL
RESEARCH

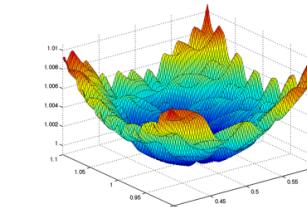


Dynamic Systems
& Control

APPLICATIONS



Battery Management
Systems (#BATT)



Optimization



Automated, Connected, &
Electric Vehicles (#ACES)



Data Science



Distributed Energy
Resources (#DER)

Professor Scott Moura

Postdoc – UC San Diego

PhD MechE – University of Michigan

MS MechE – University of Michigan

BS MechE – UC Berkeley

Selected Honors

- Carol D. Soc Distinguished Grad Student Mentoring Award
- NSF CAREER Awardee
- Hugo O. Schuck Best Paper Award
- Best Student Paper, as advisor (ACC, IFAC, DSAC)
- UC President's Postdoctoral Fellow
- NSF Graduate Research Fellow



Dr. Luis COUTO
(ULB/Oxford)

Laurel
DUNN

Zach
GIMA

Saehong
PARK

Dong
ZHANG

Zhe
ZHOU

Sangjae
BAE

Bertrand
TRAVACCA

Mathilde
BADOUAL



Soomin
WOO

Yiqi
ZHAO

Ioanna
KAVVADA

Aaron
KANDEL

Patrick
KEYANTUO

Dylan
KATO

Teng
ZENG

Jing YU
(TBSI)

Shirin
YOUSEFIZADEH
(Aalborg, DK)



Andrea POZZI
(Pavia, IT)

Upadhi
VIJAY

Pedro
ERRAZURIZ

Sihan
LIU

Yan
XIAO KESTELMAN

Jonathan
DAYARAMANI

Sonia
MARTIN

Vanessa
HERNANDEZ-CRUZ

German
PEREA



WeChat



RL for Energy Systems-2020



该二维码7天内(6月26日前)有效，重新进入将更新



Scott Moura

Shenzhen, Guangdong



Scan the QR code to add me on WeChat



Saehong Park (박세홍,...)



Scan the QR code to add me on WeChat

Slack



Click to Join!

https://join.slack.com/t/introtorlucbtbsi/shared_invite/zt-flhuxmhrmGkBUOlpxzI3d8l5F_yQ

Course Organization

Lectures

- Pre-recorded videos
- During lecture time, watch videos, pause, discuss

Notes

- Course notes available at
<https://scott-moura.github.io/rl/>

Assignments

- Two assignments
- Application to Offshore Wind Energy

Topic Outline

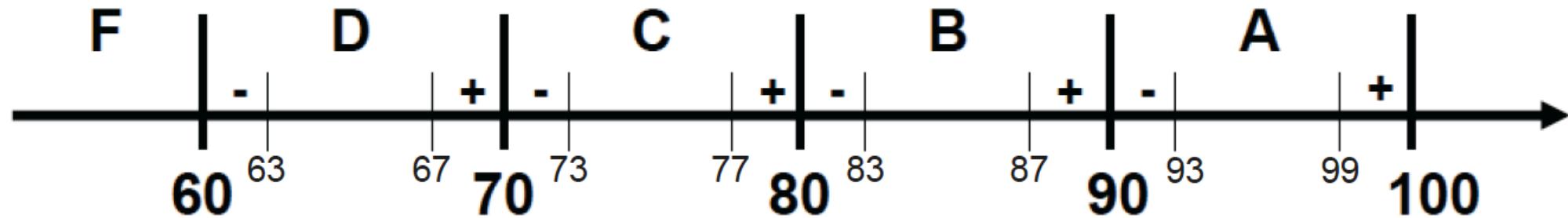
1. Optimal Control
2. Dynamic Programming
 1. Principal of Optimality & Value Functions
 - Case Study: Linear Quadratic Regulator (LQR)
3. Policy Iteration & Value Iteration
 - Case Study: LQR
4. Approximate Dynamic Programming (ADP)
 1. Temporal Difference (TD) Error
 2. Value Function Approximation
 - Case Study: LQR
3. Online RL with ADP
4. Actor-Critic Method
 - Case Study: Offshore Wind
5. Q-Learning
 1. Q-learning algorithm
 2. Advanced Q-learning algorithm, i.e., DQN
6. Policy Gradient
 1. Vanilla policy gradient (REINFORCE)
7. Actor-Critic using Policy Gradient
 1. Actor-Critic using Policy Gradient
 2. Advanced Actor-Critic algorithm, i.e., DDPG
8. RL for energy systems
 1. Case Study: Battery Fast-charging

Textbooks (none are required)

The following are recommended for additional background:

1. D. P. Bertsekas and J. Tsitsiklis, Neuro-Dynamic Programming, Athena Scientific, 1996
2. Powell, W. B. (2007). Approximate Dynamic Programming: Solving the curses of dimensionality (Vol. 703). John Wiley & Sons.
3. R. S. Sutton and A. G. Barto, Reinforcement Learning: An Introduction, 2017
4. Lewis, F., & Vrabie, D. (2009). Reinforcement learning and adaptive dynamic programming for feedback control. *IEEE Circuits and Systems Magazine*, 9(3), 32–50.
<http://doi.org/10.1109/MCAS.2009.933854>
5. Lewis, F. L., & Vrabie, D. (2009). Reinforcement learning and adaptive dynamic programming for feedback control. *IEEE Circuits and Systems Magazine*, 9(3), 32–50.
<http://doi.org/10.1109/MCAS.2009.933854>
6. J. Si, A. Barto, W. Powell D. Wunsch (2004). Handbook of Learning and Approximate Dynamic Programming.

Grading



Straight scale (we may curve up, but it has never been necessary)

Participation	50pts	Based on attendance and interaction. Award at instructor's discretion
Assignments	50pts	Two assignments, 25pts each
TOTAL	100pts	

How to Succeed (in Remote Learning Courses)

- Ask questions in class! MORE important in remote learning
 - LIVE discussion
- See instructors during office hours
- Be resourceful!
- Send us a WeChat/Slack.

- You are not alone!
- Work together! Lift each other! Succeed Together!
- You are responsible for making a better world for yourself & all