## ECE183DA (Winter 2022)

Design of Robotic Systems I

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Problem set 9 | More learning, gaussian processes

## Key takeaways

After this lecture, you should understand:

- How different assumptions / constraints on problem statements and datasets result in the different forms of learning problems, and how to identify a suitable approach for a given problem;
- How datasets are used to generate planning algorithms using imitation learning, and the challenges associated with doing so; and
- How Gaussian Processes (GPs) can be used more generally to apply datasets to arbitrary computational problems.

## Assignment

- 9(a). Why is it generally difficult to use an inverse reinforcement learning based approach by generating a data set of "inexpert" trajectories of what **not** to do? When might this approach be successful anyway?
- 9(b). Describe some characteristics of an appropriate kernel function  $k_{ij} = k(x_i, x_j)$  to use when approximating a value function V(s) using a Gaussian Process regression on a 2D gridworld system with chess bishop system dynamics. That is, what can we say about  $k_{ij}$  for various relationships between  $x_i$  and  $x_j$ ?
- 9(c). Would you be willing to let us use your correct responses as (anonymized) examples for the class?