RL基础

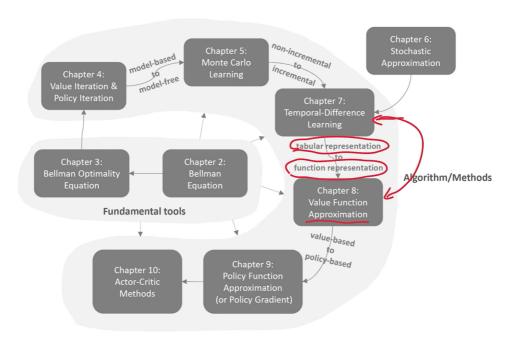
【强化学习的数学原理】课程:从零开始到透彻理解(完结) 哔哩哔哩 bilibili

8-值函数近似

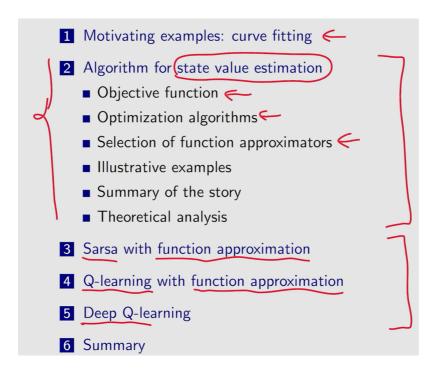
⊘ Note

第一次引入神经网络;

mindmap



主要内容



引言

- 表格型方法: 表格是指action-value的二维的表格表示;
- 缺点: state space或者action space较大、甚至连续时,表格型难以存储、难以泛化;

• 例子

- ullet Suppose there are one-dimensional states $s_1,\ldots,s_{|\mathcal{S}|}.$
- Their state values are $v_{\pi}(s_1), \ldots, v_{\pi}(s_{|\mathcal{S}|})$, where π is a given policy.
- Suppose |S| is very large and we hope to use a simple curve to approximate these dots to save storage.

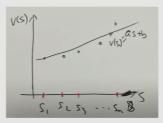


Figure: An illustration of function approximation of samples.

- 一维state数量非常多,**使用一个v函数来近似表示各个state的value** $(\hat{v}(s,w)\approx v_{\pi}(s))$; 如此,不用存储各个state对应的精确的value,而只需要存储曲线的参数(如线性拟合);
 - 节省大量的存储;
 - 增强了泛化性;访问一个状态,会改变函数参数,其他state-value也会变化;
- 对于非线性的曲线而言,本质上对于参数w而言仍然是线性变化,对于变量s而言需要先设计一个kernel function进行变换;

$$\underbrace{\hat{v}(s,w)}_{} = as^2 + bs + c = \underbrace{[s^2,s,1]}_{\phi^T(s)} \underbrace{\begin{bmatrix} a \\ b \\ c \end{bmatrix}}_{w} = \underbrace{\phi^T(s)w}.$$

• 也可以用神经网络做非线性的拟合;

state-value 的近似