

| 1.  | Title of the course   | Principles of Reinforcement Learning           |
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| 2.  | Course number   | CS603L   |
| 3.  | Structure of credits  | 3-0-0-3  |
| 4.  | Offered to  | PG   |
| 5.  | New course/modification to  | Modification To CS6021/17                      |
| 6.  | To be offered by  | Department of Computer Science and Engineering |
| 7.  | To take effect from   | July 2022                                      |
| 8.  | Prerequisite  | СоТ  |
| 9.  | <b>Course Objective(s):</b> To study the mathematical concepts towards understanding various elements of reinforcement learning and associated stochastic decision processes. To gain intuition behind the computational aspects of reinforcement learning algorithms.  |  |
| 10. | <b>Course Content:</b> Review of probability, random variables and their convergence, stochastic processes, expectation, conditional expectation, sufficient statistics, martingales; Introduction to real analysis; Sequences of functions: MCT and BCT; Markov chains: mixing times and HMM; Bergman loss functions; Dynamical system: fixed point and stability theorems; Monotonicity and contraction properties of dynamic programming, submodularity, stochastic and monotone likelihood ratio dominance; Stochastic approximation; Sample complexity; Concentration of measure: Tail probabilities, Markov inequality, Chebyshev inequality, Cramer-Chernoff method, Azuma-Hoeffding's inequality, Cauchy-Schwarz inequality; Occam's razor bound; VC dimension. |  |
| 11. | <ul> <li>Textbook(s):</li> <li>1. Agarwal A, Jian N,Kakade S M and Sun W, <i>Reinforcement Learning: Theory and Algorithms</i>, 1st Edition, Research Monograph Draft (2020).</li> <li>2. Krishnamurthy V, <i>Partially Observed Markov Decision Processes: From Filtering to Controlled Sensing</i>, 1st Edition, Cambridge (2016).</li> </ul>   |  |
| 12. | <ul> <li>Reference(s):</li> <li>1. Bertsekas D P, <i>Dynamic Programming and Optimal Control</i>, <i>Vol II</i>, 4th Edition, Athena Scientific (2012).</li> <li>2. Borkar V S, <i>Stochastic Approximation: A Dynamical Systems Viewpoint</i>, 1st Edition, Hindustan Book Agency (2009).</li> <li>3. Hajek B, <i>Random Processes for Engineers</i>, 1st Edition, Cambridge (2015).</li> <li>4. Shwartz S S and David S B, <i>Understanding Machine Learning: From Theory to Algorithms</i>, 1st Edition, Cambridge (2015).</li> </ul>  |  |