

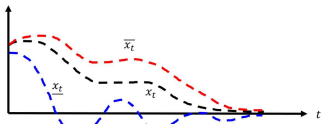


<p>Machine Decision Intelligence and Learning Research Group (MDILRG)</p> 	<p>■ <b>Contact information</b></p> <p>Email : <a href="mailto:donghwan@kaist.ac.kr">donghwan@kaist.ac.kr</a> Lab.: N1 314</p> 
<p>■ <b>Current state of the Lab. (in 2025 Spring Semester)</b></p> <p>Postdoctoral Fellows : 1      PhD Students: 7      Master's Student: 2</p>	
<p>■ <b>Research Areas</b></p> <p>Development of advanced reinforcement learning algorithms, theory and applications, such as robots and self-driving cars, control theory and applications, machine learning algorithms, interplay among control, reinforcement learning, and optimization, optimization algorithms and theories.</p>	
<p><math>\underbrace{\frac{d}{dt}\underline{x}_t = f(\underline{x}_t)}_{\text{Lower comparison system}} \leq \frac{d}{dt}x_t = A_{\sigma(x_t)}x_t + b_{\sigma(x_t)} \leq \underbrace{\frac{d}{dt}\bar{x}_t = \bar{f}(\bar{x}_t)}_{\text{Upper comparison system}}</math></p> <p><math>\downarrow \frac{1}{8}</math>      <math>\downarrow \frac{1}{8}</math>      <math>\downarrow \frac{1}{8}</math></p> <p><math>\underline{x}_t \rightarrow 0 \leq x_t \rightarrow 0 \leq \bar{x}_t \rightarrow 0</math></p>  <p>Comparison principle</p>	<p>Discrete-time stochastic domain</p> <p>GTD2 ( Sutton, R. S. (2009) )      Backstepping TD</p> <p><math>\eta \rightarrow 0</math></p> <p>Continuous-time deterministic domain</p> <p>Control design problem      Control design problem</p> <p><math>\frac{d}{dt}\lambda_t = -C\lambda_t + Ax_t</math>      <math>\frac{d}{dt}\lambda_t = (-C + \eta A)\lambda_t + Ax_t</math> <math>\frac{d}{dt}x_t = A^T\lambda_t</math>      <math>\frac{d}{dt}x_t = u_t</math> Select <math>u_t</math> using backstepping method <math>u_t = (A^T + \eta^*A - \eta C)\lambda_t - \eta Ax_t</math></p> <p><math>\eta \rightarrow 0</math></p>
<p>■ <b>Recommended courses &amp; Career after graduation</b></p> <p>Recommended courses: control system engineering, linear system, nonlinear system, optimal control, machine learning, reinforcement learning, probability theory, real analysis, measure theory</p> <p>Career after graduation: national labs, start up, industry, silicon valley, academia</p>	<p>■ <b>Introduction to other activities besides research</b></p> <p>Domestic/International Conferences</p> <p>Lab Seminar / Group Study</p>
<p>■ <b>Introduction to the Lab.</b></p> <p>Our research covers theory and application of control, machine learning, reinforcement learning, and interplay among them.</p>	
<p>■ <b>Recent research achievements ('23~'25)</b></p> <p>Narim Jeong, Donghwan Lee*, "Finite-time analysis of minimax Q-learning," RLC2025</p> <p>Hyeann Lee, Donghwan Lee*, "Suppressing overestimation in Q-learning through adversarial behaviors," Annual Allerton Conference on Communication, Control, and Computing, 2024</p> <p>Narim Jeong, Donghwan Lee*, "Unified finite-time error analysis of soft Q-learning," Neurocomputing, 2025</p> <p>Han-Dong Lim, Donghwan Lee*, "A primal-dual perspective for distributed TD-learning," UCAI2025</p> <p>Han-Dong Lim, Donghwan Lee*, "Regularized Q-learning," NeurIPS2024</p> <p>Han-Dong Lim, Donghwan Lee, "Backstepping temporal-difference learning " ICLR2023</p> <p>Donghwan Lee, Jianghai Hu, and Niao He, "A discrete-time switching system analysis of Q-learning," SICON, 2023</p>	