

# Dong Chen

Email: [E0974151@u.nus.edu](mailto:E0974151@u.nus.edu) Tel: (+65) 8386 1629

## EDUCATION

<b>National University of Singapore (NUS), Singapore</b> <i>Ph.D., Mechanical Engineering</i>	Aug 2022–April 2026
• Major GPA: 4.25/5.00	
<b>University of Science and Technology of China (USTC), Hefei, China</b> <i>M.S., Power Engineering and Engineering Thermal Physics</i>	Sep 2019 – Jun 2022
• Major GPA: 3.44/4.30	
<b>Northeast Forestry University (NFU), Harbin, China</b> <i>B.S., Building Environment and Energy Application Engineering</i>	Sep 2015 – Jun 2019
• Major GPA: 87.12/100, Ranking: 3/59	

## RESEARCH EXPERIENCES

### Coolest LAB Group, NUS

<b>Project 1: Physics-Informed Machine Learning Framework for Modeling</b>	Aug 2022 – Jun 2024
• The proposed APCNNs demonstrate superior predictive accuracy and lower computational costs compared to traditional PCNNs across real-world data center datasets.	
• The APCNN-based framework for intelligent control in air-free cooling data centers outperforms conventional prediction methods, reducing the Mean Absolute Error by ~20% for short-term (15-minute) forecasts and by ~80% over a seven-day horizon.	
• Rigorous mathematical proofs verify that the APCNN maintains physical consistency across all modeled processes.	
<b>Project 2: Physics-Informed ML-Based Model Predictive Control for Data Centers</b>	Jun 2024 – Oct 2025
• Developed an APCNN-assisted MPC (APCNN-MPC) controller that reduces energy consumption by ~20% under both normal operation and previously unseen high-load conditions, while maintaining stable thermal performance.	
• Demonstrated that APCNN-MPC consistently outperforms traditional ML and time-series baselines, achieving higher energy efficiency with more accurate and stable temperature regulation.	
• Extended the framework to a long-horizon physics-informed ML-Adaptive MPC (PIML-AMPC), delivering ~24% annual cooling-energy savings compared with fixed-setpoint control.	
• Proposed a robust, deployment-oriented control strategy by integrating learning-based prediction with physics-based knowledge, enabling reliable energy optimization under realistic operating variability.	
<b>Project 3: Deep Reinforcement Learning Control for Data Centers</b>	Oct 2025 – April 2026
• Developed an end-to-end DRL control platform for hybrid-cooled data centers, enabling closed-loop training and evaluation with realistic system dynamics and operational constraints.	
• Implemented and benchmarked state-of-the-art DRL algorithms (SAC, TD3, and PPO), achieving ~15% energy savings while maintaining stable thermal performance.	
• Advanced safe DRL for data-center control by incorporating safety-aware control design (e.g., constraint handling and violation monitoring), improving reliability and deployment readiness under diverse operating conditions.	

### Energy and Heat transfer laboratory, USTC

<b>Project 1: Rapid heating of liquid CO<sub>2</sub> to the supercritical state using solid propellant</b>	Sep 2019 – Mar 2021
• Investigated numerically and experimentally the process of constant-volume heating of liquid CO <sub>2</sub> to its supercritical state by solid propellant.	
• Attained a maximum phase conversion speed of ~4.0 kg/s by optimizing the design of working condition.	
• Provided useful information for the analysis of the performance of a pneumatic launch with supercritical CO <sub>2</sub> as the working medium such as its reliability, propelling speed, stability.	
<b>Project 2: Rapid and uniform heating of liquid CO<sub>2</sub> using a porous ceramic heater</b>	Jun 2021 – Jun 2022
• Proposed an innovative heating method for liquid CO <sub>2</sub> heating by using a porous ceramic heater with an electrically conductive coating.	
• Numerically investigated the heating process of liquid CO <sub>2</sub> to its supercritical state.	
• Lowered the cost and increased the safety of the system.	

## Thermal Science Laboratory, NFU

- Project: A cooling performance study of an innovative groove-structured radiant panel** Sep 2017 – Jun 2019
- Built an experimental platform
  - Experimentally investigated the cooling performance of the new panel
  - Economically friendly, potentially be used for indoor cooling

## PUBLICATIONS

---

- D. Chen, et al. Offline deep reinforcement learning for data-driven decision making in data center operations. (in preparation).
- D. Chen, et al. policy learning for hybrid-cooled data centers using deep reinforcement learning. (under review).
- D. Chen, et al. Physics-informed machine learning-based adaptive predictive control for energy-efficient hybrid-cooled data centers management. (under review).
- D. Chen, C.K. Chui, P.S. Lee. Physics informed machine learning based predictive control for intelligent operation of edge datacenters. *Applied Energy*. 2025.
- D. Chen, C.K. Chui, P.S. Lee. Adaptive physically consistent neural networks for data center thermal dynamics modeling. *Applied Energy*. 2024.
- D. Chen, H. Ye, H.X. Yao. Theoretical and experimental study on constant volume heating of CO<sub>2</sub> by solid propellant. *Applied Thermal Engineering*. 2022.
- G. Lv, C. Shen, Z. Han, W. Liao, D. Chen. Experimental investigation on the cooling performance of a novel grooved radiant panel filled with heat transfer liquid. *Sustainable Cities and Society*. 2019.
- Z. Han, D. Chen, et al. Experimental Study on Cooling Performance of a Liquid Filled Cold Radiant Panel Based on Grooved Structure. *Building Science*. 2018.

## AWARDS

---

- PhD Scholarship at National University of Singapore in 2022
- First Class Graduate Scholarship in 2020, USTC (Top 10%)
- Extraordinary scholarship in 2018 (2/59)
- First-Class Prize in National University Student Social Practice and Science Contest on Energy Saving & Emission Reduction in 2018
- Third-Class Prize in National University Student Mathematics Competition in 2017
- Third-Class Prize in National University Student English Competition in 2017
- Provincial First-Class Prize in National University Student Mathematical Modeling Competition in 2017

## PROFESSIONAL SKILLS

---

- Programming Languages: C/C++, Python, Matlab,.
- Commercial software: Origin, EnergyPlus, Modelica, ANSYS Fluent.