

Applications of Artificial Intelligence for Physics Simulations and Data Analysis

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DOI: <https://doi.org/10.51584/IJRIAS.2026.11070150>

Received: 02 July 2026; Accepted: 08 July 2026; Published: 14 August 2026

ABSTRACT

Artificial Intelligence is maybe among the most revolutionary technological inventions of the 21st century, which has changed research in almost all sciences. In physics, artificial intelligence has immensely helped scientific research by facilitating data analysis, automated simulations, improving experimental design, and making some breakthroughs computationally impossible before AI. Physics experiments create huge volumes of data using particle accelerators, telescopes, satellites, and laboratory equipment. Classical computing methods usually take much time for computations and human effort while methods based on artificial intelligence, such as machine learning, deep learning, neural networks, reinforcement learning, and computer vision, are very fast and accurate.

Currently, artificial intelligence is an essential element in quantum physics, astrophysics, condensed matter physics, nuclear physics, and materials science research. AI-powered simulations help physicists model interactions between atoms, calculate material properties, detect gravitational waves, find new planets outside our Solar System, and optimize particle collision experiments. Also, simulations powered by artificial intelligence reduce costs, risks, and speed up scientific discoveries by predicting experiment results without expensive physical experiments.

Though there have been many developments, some issues such as data quality, computing costs, transparency of algorithms, ethical considerations, and the need for interdisciplinary expertise persist. This paper investigates the role of Artificial Intelligence in the study of physics and simulations, reviews various uses of AI in the field of physics, addresses current problems, and identifies opportunities for the future. The conclusion from this paper is that AI will revolutionize the future of physics through faster discoveries, improved simulation capabilities, and other innovations in science.

Keywords: Artificial Intelligence in Physics, Machine Learning for Physics, Physics Research, Deep Learning Physics Simulation, Simulations, Quantum Computing, Scientific Computing, Data Analysis, Physics-informed Neural Networks.

INTRODUCTION

Physics is the core science that studies physical principles underlying matter, energy, space, and time. For centuries, developments in physics have depended significantly on observations, experiments, mathematical models, and computer simulations. As the scientific apparatus has evolved, the volume of data gathered from experiments has risen dramatically. Modern scientific installations like particle colliders, astronomical observatories, fusion reactors, and space missions produce a petabyte of data annually, rendering it impractical to analyze manually.

Artificial Intelligence (AI) has appeared as a disruptive technology, which helps solve these problems.

Artificial Intelligence defined as a computer technology, which performs tasks usually requiring human intelligence such as learning, reasoning, decision-making, problem solving, and pattern recognition. In physics, AI allows researchers to work with big data sets, find patterns, automate processes, optimize experiment conditions, and simulate physical systems in a very fast and precise way.

The involvement of AI in physics has revolutionized both theoretical and experimental sciences. Machine learning algorithms can forecast material characteristics without running costly laboratory tests, while deep learning technologies help classify celestial bodies based on telescope data. Furthermore, AI allows physicists to detect rare events occurring in high energy particle collisions, improve climate and weather simulations.

Physics simulation has greatly benefited from the development of AI. Physics simulation usually requires solving very complicated equations, which need a lot of computation power. With the help of surrogate models created using AI, computational time has been significantly reduced, while accuracy of predictions has remained high.

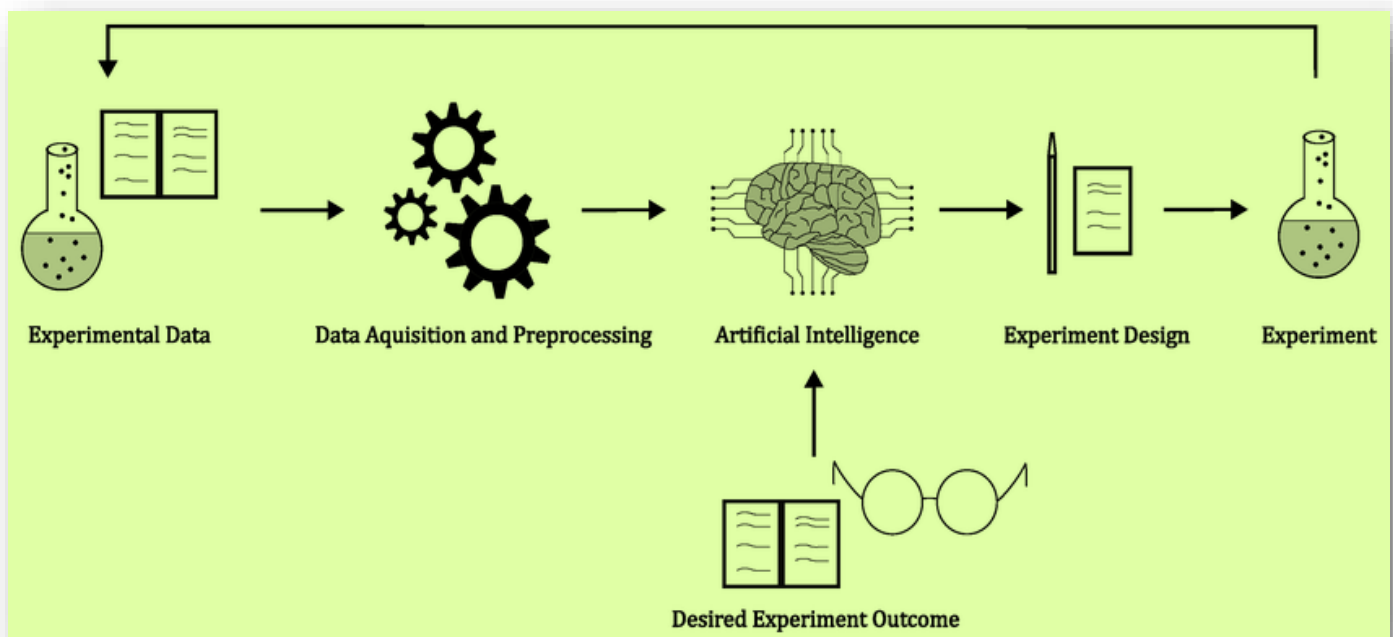
Because of the interaction between AI and physics, a new interdisciplinary research field has emerged that uses computer science, mathematics, and physics for solving some of the hardest scientific problems faced by humans.

OBJECTIVES OF THE STUDY

The main objectives of this research include:

- To comprehend the concept of Artificial Intelligence.
- To explore the use of AI in modern physics research.
- To investigate the use of AI in scientific simulations.
- To determine the benefits of using AI for physics research.
- To evaluate the challenges faced in implementing AI in physics.
- To identify the future prospects of AI in scientific research.

SCHEMATIC DIAGRAM:



Schematic illustration of the AI-driven experimental design workflow. Available experimental data used to train AI. The scientist defines an experimental objective, and AI suggests experimental designs to the scientist accordingly or performs them autonomously.

RESEARCH METHODOLOGY

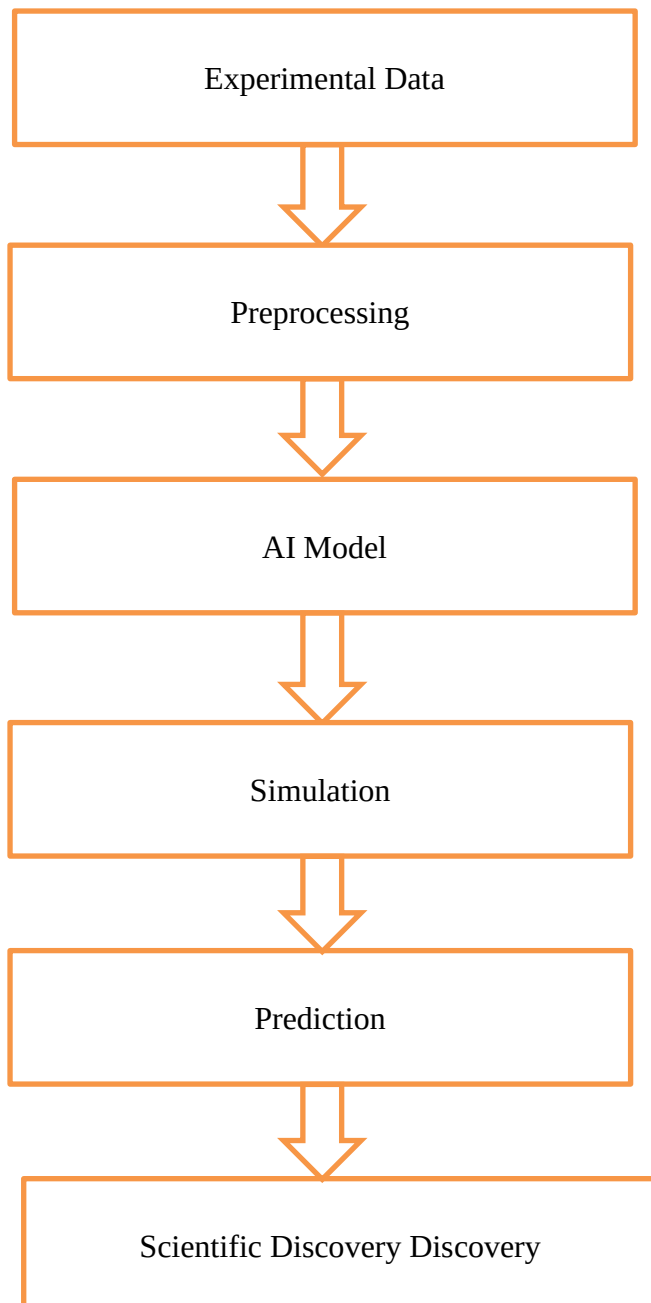
This research paper has been prepared using the secondary research methodology.

Sources of information included:

- a) Peer-reviewed research journals
- b) Scientific publications
- c) IEEE explore Digital Library
- d) Springer Nature
- e) Elsevier Science Direct
- f) Google Scholar
- g) Government and research organization reports
- h) Books on Artificial Intelligence and Computational Physics

The data has been analyzed using the descriptive and qualitative analysis method.

FLOW CHART:



LITERATURE REVIEW

A number of studies explored the incorporation of Artificial Intelligence in scientific research.

1. In Engineering Applications of Artificial Intelligence Advances in Physics-Informed Neural Networks for Solving Complex Partial Differential Equations and Their Engineering Applications (2025)
2. Schmeing, L., & Pioch, F. (2026), Advancements in Physics-Informed Neural Networks for Solving Maxwell's Equations: A Systematic Literature Review
3. Machine Learning and the Physical Sciences – Giuseppe Carleo et al. (2019)
4. Machine Learning for Molecular and Materials Science – Keith T. Butler et al. (2018)
5. Machine Learning for Molecular Simulation – Frank Noé et al. (2020)
6. Machine Learning at the Energy and Intensity Frontiers of Particle Physics – Alexander Radovic et al. (2018)
7. As per Carleo et al. (2019), machine learning had emerged as an indispensable approach for tackling complex quantum many-body problems to carry out precise simulations.
8. Butler et al. (2018) found that the usage of machine learning in AI was helpful in accelerating the process of discovery of advanced materials by predicting the properties of materials without performing costly laboratory tests.
9. Schawinski et al. (2017) discussed how deep learning was revolutionizing the field of astrophysics by facilitating galaxy classification, object identification and efficient image processing from telescopes.
10. According to Radovic et al. (2018), deep learning approaches were widely adopted in high-energy particle physics to detect events, particles and anomalies at major scientific facilities.
11. Current studies show that AI is now being used increasingly in climate modeling, fusion research, quantum computing, gravitational waves detection, and computational fluid dynamics.

From the above literature review, it is clear that AI is playing a very significant role in enhancing the process of research in several areas of physics.

Artificial Intelligence is a branch of computer science that focuses on developing intelligent systems capable of learning from data, recognizing patterns, making predictions, and solving complex problems with minimal human intervention.

THE MAJOR COMPONENTS OF AI INCLUDE:

Machine Learning (ML)

With Machine Learning, computers are able to learn from data through the absence of programming. In other words, ML algorithms develop themselves through pattern detection in datasets.

1. Supervised Learning
2. Unsupervised Learning
3. Reinforcement Learning

Deep Learning

Deep Learning is a Machine Learning technique that uses multilayered Artificial Neural Networks.

Applications are:

1. Image recognition
2. Speech recognition
3. Scientific simulations
4. Pattern recognition
5. Particle classification

Artificial Neural Networks (ANN)

Artificial Neural Networks emulate the behavior of human brain using information exchange between neurons.

Uses of Artificial Neural Networks are:

1. Predictive modeling
2. Optimization
3. Complicated simulations
4. Quantum states prediction

Computer Vision

Computer Vision helps computers understand and process visual information.

In physics, applications of Computer Vision are:

1. Astronomical image analysis
2. Microscopic material analysis
3. Particle tracking
4. Satellite images processing

Reinforcement Learning

By training an intelligent system using rewards and penalties, Reinforcement Learning makes an intelligent system learn what to do and not to do.

Uses are:

1. Robotics
2. Experiments optimization
3. Autonomous lab systems
4. Quantum control

ARTIFICIAL INTELLIGENCE IN PHYSICS RESEARCH

Physics researchers have embraced artificial intelligence in the way they conduct research. Rather than relying only on theoretical equations and conventional simulations, AI is now employed in analyzing large data sets, building predictive models, and carrying out other computational procedures.

SOME APPLICATIONS ARE:

Data Analysis

Large amounts of scientific data collected during modern-day experiments.

1. Some examples are:
2. Accelerator experiment data
3. Observations from telescopes
4. Data gathered by satellites
5. Experiments involving plasma physics

Such data sets processed quickly by AI algorithms, which uncover patterns that cannot be detected through conventional statistics.

Pattern Recognition

Machine learning is applied in pattern recognition in scientific data sets.

Some of its applications are:

1. Particle recognition
2. Detection of gravitational waves
3. Recognizing galaxies
4. Identifying atomic structures

Experimental Automation

AI can control scientific instruments by:

1. Carrying out monitoring
2. Changing parameters
3. Collecting observations
4. Optimization of measurements

Predictive Modeling

Without having to conduct costly physical experiments repeatedly, AI makes predictions about the results of such experiments using data collected earlier.

The examples include the following:

1. Materials property prediction
2. Plasma properties
3. Molecular interaction
4. Qualitative state prediction

ARTIFICIAL INTELLIGENCE FOR PHYSICS SIMULATIONS

Physics simulation is one way to make researchers study those systems that cannot be created in real life or are too risky.

Artificial Intelligence improves physics simulations by:

1. Reducing simulation time
2. Increasing simulation accuracy
3. Optimizing numerical models
4. Predicting physics behavior
5. Learning from past simulations

AI-powered simulations are currently widely used in:

1. Quantum mechanics
2. Fluid dynamics
3. Nuclear power plants
4. Space missions
5. Climatology
6. Fusion energy studies

In contrast to regular simulations that solve differential equations repeatedly, AI simulations are capable of learning from their past and making predictions very fast.

APPLICATIONS OF ARTIFICIAL INTELLIGENCE IN VARIOUS BRANCHES OF PHYSICS

Artificial Intelligence has revolutionized virtually every branch of physics by providing fast computations, correct predictions and automations of difficult scientific research processes. The following paragraphs will elaborate on the various applications of AI.

AI IN QUANTUM PHYSICS

Quantum physics is concerned with the behavior of matter and energy at an atomic and sub-atomic level. Quantum mechanics calculations are very difficult due to the large number of particles involved in the process. AI makes such calculations easier through optimization and predictive modeling.

Applications

Benefits

- | | |
|--|--|
| 1. Prediction of quantum states and wave function. | 1. Fast quantum simulations. |
| 2. Design of quantum algorithms for quantum computers. | 2. Low computational complexity. |
| 3. Optimization of quantum circuits. | 3. Higher accuracy in quantum experiments. |
| 4. Error correction in quantum computing. | 4. Better development of quantum technology. |
| 5. Simulation of molecular interactions. | |

AI IN PARTICLE PHYSICS

Particle physics is the branch of physics that studies the fundamental particles of matter and the fundamental forces governing their interactions. Experiments conducted at large particle accelerators, such as the Large Hadron Collider (LHC), produce enormous volumes of complex data from particle collisions. Artificial Intelligence (AI) and machine learning techniques are widely used to analyze these datasets, identify particles, classify collision events, detect rare phenomena, reduce background noise, and accelerate the discovery of new particles and physical processes.

Applications

Benefits

- | | |
|---|---|
| 1. Particle identification. | 1. Faster processing of experimental data. |
| 2. Event classification during collisions. | 2. Enhancement in the discovery of new particles. |
| 3. Detection of rare interactions between particles. | 3. Increased experimental accuracy. |
| 4. Signal noise reduction in detectors. | 4. Decreased manual analysis. |
| 5. Real-time monitoring of experiments at accelerator facilities. | |

AI IN ASTROPHYSICS

Modern astronomical facilities produce vast amounts of data through the use of telescopes, satellites, and space probes. AI helps astronomers analyze such observations more efficiently.

Applications

Benefits

- | | |
|--------------------------------------|--|
| 1. Classification of galaxies. | 1. Automated image processing. |
| 2. Detection of exoplanets. | 2. Faster astronomical discoveries. |
| 3. Identification of black holes. | 3. Increased accuracy of observations. |
| 4. Detection of gravitational waves. | 4. Efficient work with large datasets. |
| 5. Prediction of supernovas. | |
| 6. Space image enhancement. | |

AI IN MATERIALS SCIENCE

Traditional search for new materials usually takes several years of laboratory research. AI allows to accelerate the process by predicting properties of new materials based on existing databases.

Applications

1. Discovery of superconductors.
2. Optimization of battery materials.
3. Semiconductor development.
4. Crystal structures prediction.
5. Nanomaterials development.

Benefits

1. Cost savings on experiments.
2. Innovation acceleration.
3. Better material performance.
4. Sustainable material development.

AI IN NUCLEAR PHYSICS

Nuclear physics involves highly complex systems where safety and accuracy are critical.

Applications

1. Nuclear reactors monitoring.
2. Radiation measurements.
3. Simulations of nuclear fusion.
4. Optimization of reactors' safety.
5. Predictive maintenance.

Benefits

1. Safety enhancement.
2. Less accident risks.
3. Energy efficiency improvement.
4. Improved performance of the reactor.

APPLICATIONS OF AI IN PLASMA PHYSICS AND FUSION ENERGY

Fusion energy considered one of the most promising clean and sustainable energy sources for the future because it has the potential to generate enormous amounts of energy with minimal greenhouse gas emissions and long-lived radioactive waste. In fusion reactors, maintaining stable, high-temperature plasma is one of the greatest scientific and engineering challenges. Artificial Intelligence (AI) plays a vital role by monitoring plasma behavior in real time, predicting plasma instabilities, optimizing magnetic confinement, controlling reactor parameters, detecting equipment faults, and improving the overall safety and efficiency of fusion experiments. AI-driven control systems help accelerate the development of reliable and commercially viable fusion energy technologies.

Applications

- 1.Forecasting of the plasma stability.
- 2.Optimization of magnetic confinement.
- 3.Control systems for reactors.
- 4.Real-time monitoring.
- 5.Detection of malfunctions.

Benefits

1. Safety and efficiency improvement of the reactor.
2. Safety and efficiency of the plasma management.
3. Reduced expenses for running the reactor.
4. Faster development of commercial fusion energy.

Table 1:

Physics Domain	AI Technique	Outcome
Quantum Physics	PINNs	Faster PDE solutions
Particle Physics	Deep Learning	Event classification
Astrophysics	CNNs	Galaxy classification
Materials Science	ML	Property prediction
Nuclear Physics	Reinforcement Learning	Reactor optimization
Climate Physics	Foundation Models	Better forecasting

BENEFITS OF AI IN PHYSICS RESEARCH

There are many benefits provided by the artificial intelligence in the scientific research.

Faster Data Processing

The AI algorithm can process terabytes of data in just minutes while conventional processing takes several weeks.

Improved Accuracy

The machine learning models prevent human errors and improve the accuracy of predictions.

Reduction in Costs

Computer simulation replaces expensive laboratory experiments.

Automation

Artificial Intelligence automates scientific processes, which include image analysis, data classification and parameter optimization.

Informed Decisions

Research decisions can be made based on the results provided by AI.

Discovery of Hidden Patterns

AI discovers connections in large scientific data, which is hard for a human brain to recognize.

Timesaving Process

Scientific research projects, which once used to take years, can be done much quicker using AI.

CHALLENGES

Although Artificial Intelligence has great potential in physics research, it has some limitations as well.

Extensive Datasets

To train a successful AI model, large amounts of data are required. Bad quality data affects negatively the model performance.

High Energy Consumption

Highly advanced computational models need powerful computers with high-energy consumption for training.

The Problem of Explain ability

Most of AI algorithms work as a "black box" and decision-making process cannot be understood.

Bias in Data

There might be inaccuracy in scientific prediction because of bias in the training dataset.

Lack of Generalization

A model that can solve one scientific problem cannot solve other scientific problems related to another physical system.

Problems of ethics and security

It is important to consider data security in case of sensitive scientific data.

Requirement for Interdisciplinary Skills

Implementation of AI requires a combination of skills in physics and computer science.

Table 2:

AI Technique	Strengths	Limitations	Computational Cost	Best Physics Applications
Machine Learning	Fast prediction	Needs labeled data	Medium	Materials science
Deep Learning	High accuracy	Black-box	High	Particle physics
Reinforcement Learning	Adaptive control	Long training	Very High	Fusion reactors
PINNs	Physics-consistent	Difficult optimization	High	PDE simulations
GANs	Data generation	Training instability	High	Astronomy
Diffusion Models	High-quality generation	Slow inference	High	Scientific imaging
LLMs	Scientific assistance	Hallucination	Medium	Literature review
Foundation Models	Generalization	Very expensive	Very High	Multidisciplinary research

FUTURE SCOPE

Artificial Intelligence is going to become the vital part of future physics.

The future scope includes:

1. AI-based autonomous laboratories.
2. Completely automated scientific experiments.
3. Space exploration by means of AI.
4. Detection of new quantum materials.
5. Smart fusion reactor.
6. Climate change prediction.
7. Discovery of drugs with help of computational physics.
8. Merging of AI and quantum computing.
9. New smart scientific devices.
10. Digital twin of a physical system.

The increasing progress in computing technologies will lead to the faster scientific discovery.

DISCUSSION

Critically compares AI techniques rather than only describing them.

Discusses trade-offs between accuracy, interpretability, data requirements, and computational cost.

Explains where PINNs outperform traditional deep learning (e.g., physics-constrained simulations) and where LLMs and foundation models are best suited (e.g., scientific assistance and multi-domain knowledge).

Identifies open research challenges such as explain ability, uncertainty quantification, reproducibility, and energy-efficient AI.

CONCLUSION

The arrival of artificial intelligence technology has been proven to have a revolutionary effect on current research in physics and simulations. With the use of machine learning, deep learning, computer vision, and computational power, scientists can analyze huge sets of data, simulate complex physical phenomena, conduct automatic experiments, and expedite their research work.

The applications of AI technology include quantum mechanics, particle physics, astrophysics, material physics, plasma physics, and nuclear physics. Not only does AI help scientists conduct research faster and more effectively, but it also decreases the cost of computations and increases their precision. This makes it possible to resolve the problems, which were once impossible to compute.

Nevertheless, the issues of data reliability, computational expenses, algorithm interpretability, ethical questions, and interdisciplinary will stay relevant in this context. Overcoming them will involve cooperation between physicists, computer scientists, engineers, and policymakers.

AI technology expected to revolutionize science even more with time thanks to autonomous laboratories, smart simulations, quantum computations, and space exploration. The implementation of artificial intelligence technology in physics is a crucial step towards scientific progress and an important part of future research and technology.

REFERENCES

1. Karniadakis, G. E., et al. (2023). *Physics-informed machine learning*. Nature Reviews Physics.
2. Cuomo, S., et al. (2023). Scientific Machine Learning with PINNs.
3. Wang, S., et al. (2024). Recent Advances in Physics-Informed Neural Networks.
4. Bommasani, R., et al. (2023). *On the Opportunities and Risks of Foundation Models*.
5. Ren, Z., Zhou, S., Liu, D., & Liu, Q. (2025), Applied Sciences, 15(14), 8092.
6. Luo, K., Zhao, J., Wang, Y., et al. (2025), Artificial Intelligence Review
7. Meng, C., Griesemer, S., Cao, D., et al. (2025), Machine Learning for Computational Science and Engineering
8. Barman, K. G., et al. (2025), The European Physical Journal C
9. Suh, Y., Chandramowlishwaran, A., & Won, Y. (2024), npj Computational Materials
10. Hu, H., Qi, L., & Chao, X. (2024), Thin-Walled Structures
11. Alkhadhr, S., et al. (2024), Understanding Physics-Informed Neural Networks: Techniques, Applications, Trends, and Challenges
12. Schmeing, L., & Pioch, F. (2026), Electronics
13. Butler, K. T., Davies, D. W., Cartwright, H., Isayev, O., & Walsh, A. (2018). *Machine learning for molecular and materials science*. Nature, 559(7715), 547–555.
14. Carleo, G., et al. (2019). *Machine learning and the physical sciences*. Reviews of Modern Physics, 91(4), 045002.
15. Mehta, P., et al. (2019). A high-bias, low-variance introduction to Machine Learning for physicists. Physics Reports, 810, 1–124
16. Butler, K. T., et al. (2018). Machine learning for molecular and materials science. Nature, 559(7715), 547–555.
17. Radovic, A., et al. (2018). Machine learning at the energy and intensity frontiers of particle physics. Nature, 560(7716), 41–48.
18. Biamonte, J., et al. (2017). Quantum machine learning. Nature, 549(7671), 195–202 .
19. Noé, F., et al. (2020). Machine learning for molecular simulation. Annual Review of Physical Chemistry, 71, 361–390.
20. Zdeborová, L. (2017). Machine learning: New tool in the box. Nature Physics, 13(5), 420–421.
21. Radovic, A., et al. (2018). *Machine learning at the energy and intensity frontiers of particle physics*. Nature, 560, 41–48.

22. Schawinski, K., et al. (2017). *Generative adversarial networks recover features in astrophysical images*. Monthly Notices of the Royal Astronomical Society.
23. Goodfellow, I., Bengio, Y., & Courville, A. (2016). *Deep Learning*. MIT Press.
24. Bishop, C. M. (2006). *Pattern Recognition and Machine Learning*. Springer.
25. Russell, S., & Norvig, P. (2021). *Artificial Intelligence: A Modern Approach* (4th ed.). Pearson.
26. Nielsen, M. A., & Chuang, I. L. (2010). *Quantum Computation and Quantum Information*. Cambridge University Press.
27. LeCun, Y., Bengio, Y., & Hinton, G. (2015). *Deep Learning*. Nature, 521, 436–444.
28. Jordan, M. I., & Mitchell, T. M. (2015). *Machine Learning: Trends, Perspectives, and Prospects*. Science, 349(6245), 255–260.
29. Murphy, K. P. (2022). *Probabilistic Machine Learning*. MIT Press.
30. Bishop, C. M., & Nasrabadi, N. M. (2006). *Pattern Recognition and Machine Learning*. Springer.
31. Schmidhuber, J. (2015). *Deep Learning in Neural Networks: An Overview*. Neural Networks.
32. Silver, D., et al. (2017). *Mastering Chess and Shogi by Self-Play*. Nature.
33. National Academies of Sciences. (2018). *Data Science for Undergraduates: Opportunities and Options*. National Academies Press.