



## **AI in Space Exploration Autonomous Navigation for Rovers Planetary Habitat Planning & AI for Satellite Energy Efficiency.**

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### **ABSTRACT :**

Artificial Intelligence (AI) has become a transformative force in space exploration, addressing critical challenges and enhancing mission capabilities. This paper reviews AI applications in planetary missions, focusing on autonomous navigation, data processing, scientific discovery, predictive analytics, and mission planning. AI technologies, particularly machine learning and deep learning, have revolutionized decision-making by enabling autonomous operations.

Intelligent autonomous navigation systems that are AI-enabled may be considered equal to great revolution in space exploration and should be given momentum because these systems are able to operate independently and also do what is expected of them at a particular time. This paper presents the design philosophy which is to construct such systems based on the integration of several sensors, the utilization of modern machine learning techniques, and the employment of strong path planning and hazard detection mechanism.

the utilization of artificial intelligence (AI) in the realm of space exploration, highlighting its significant role in advancing our understanding of the cosmos. As space exploration ventures continue to push the boundaries of human knowledge, AI has emerged as a powerful tool to augment and enhance scientific endeavors. This abstract explores how AI is employed in autonomous navigation systems, data analysis techniques, decision-making processes, and mission planning strategies.

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**Keywords:** Artificial Intelligence, space exploration, satellite energy, autonomous navigation.

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### **1. Introduction**

Artificial Intelligence is revolutionizing space exploration by enabling smarter, faster, and more efficient decision-making beyond Earth. One major application is in autonomous navigation for rovers, where AI allows planetary rovers to detect obstacles, choose optimal routes, and conduct scientific experiments with minimal human intervention—essential for missions on Mars and the Moon where communication delays are significant.

AI also plays a crucial role in planetary habitat planning, helping scientists design sustainable living environments by analyzing terrain, resources, and environmental risks to support long-term human presence on other planets. Additionally, in the domain of satellite efficiency, AI algorithms optimize orbital paths, manage communication networks, predict maintenance needs, and process large volumes of space data, thereby enhancing the performance and lifespan of satellites.

Together, these applications demonstrate how AI is becoming an indispensable tool for advancing humanity's ability to explore, inhabit, and utilize space effectively.

Artificial Intelligence (AI) is transforming the way humanity approaches space exploration, offering solutions to challenges that would otherwise be difficult or impossible to manage. With vast distances, communication delays, and unpredictable environments, AI provides the ability to analyze data, make decisions, and adapt to new conditions in real-time. This makes it a key technology in shaping the future of missions to the Moon, Mars, and beyond.

One of the most promising applications of AI lies in autonomous navigation for planetary rovers. Traditionally, rovers on Mars or the Moon rely on commands sent from Earth, which can take several minutes to arrive due to signal delays. AI-powered navigation systems allow rovers to independently identify obstacles, map terrains, and select efficient paths, reducing dependence on Earth-based control. This not only increases the efficiency of exploration but also enables rovers to conduct scientific experiments and adapt quickly to unexpected conditions on alien landscapes.

AI is also becoming vital in planetary habitat planning, where it assists in designing and managing sustainable living environments for astronauts. By analyzing terrain data, resource availability, and environmental risks, AI can suggest suitable sites for habitats and optimize the use of local resources,

such as water or solar energy. It can further simulate human life-support systems, predict potential hazards, and develop adaptive strategies to ensure safety and sustainability in hostile extraterrestrial conditions—paving the way for long-term human presence on other planets.

In addition, AI is enhancing satellite efficiency and space operations. Satellites generate enormous amounts of data that must be processed quickly and accurately. AI systems help in data compression, image recognition, and anomaly detection, making satellite operations more reliable. They also optimize orbital paths, improve communication networks, and predict maintenance needs, thereby extending the functional lifespan of satellites. This efficiency not only reduces mission costs but also improves the quality of space-based services such as Earth observation, weather forecasting, and deep-space communication.

Overall, AI is becoming an indispensable partner in space exploration. By enabling autonomy, improving efficiency, and supporting sustainable planning, it bridges the gap between human limitations and the immense possibilities of the cosmos. As technology advances, AI will play an even greater role in exploring uncharted worlds and preparing humanity for life beyond Earth.

One of the major reason of AI been used in Self Driving is that it can get the most data in the shortest time and then make decisions without human interference. The most valuable and indispensable capability in this highly dynamic and harsh environment is swift and accurate vice-response, as even in few seconds of delay can lead to mission failure. The AI can perform sensor data analysis, detect any dangerous spots and guide spacecraft trajectory on-the-go, in order to guarantee safe, speedy travel.

In addition to bettering the performance and decision- making speed of human beings in Space missions, AI technology, which is applied to autonomous navigation systems, reduces its operational cost. Through technologically-enabled system solutions which do not require extensive human support and supervision overall, it is possible to decrease the need for control backend infrastructure and staff which are usually based on the ground[8]. This becomes particularly prominent or useful for long-term missions which would not be possible or too expensive to sustain the entire human involvement through out.

### ***1.1. AUTONOMOUS NAVIGATION AND CONTROL***

AI plays a critical role in navigation and control systems for planetary exploration vehicles. Rovers operating in harsh extraterrestrial environments must make autonomous decisions due to communication delays between Earth and deep-space missions. AI- powered systems allow spacecraft to function independently, enhancing mission efficiency.

A prime example is NASA's Perseverance rover, which utilizes AI algorithms to traverse Martian terrain autonomously. By processing sensor data, including high-resolution imagery and environmental parameters, AI helps the rover identify hazards such as rocks, steep slopes, and craters. This capability enables the rover to plan optimal routes while minimizing energy consumption and avoiding obstacles. Real-time autonomous navigation ensures mission continuity, even in communication blackouts, significantly improving mission success rates and expanding exploration capabilities.

### ***1.2. DATA PROCESSING AND ANALYSIS***

Planetary missions generate vast amounts of data that require efficient processing. Traditional methods rely on transmitting data back to Earth for analysis, which is limited by bandwidth and transmission delays. AI technologies, such as Machine Learning (ML) and Compressed Sensing (CS), are now embedded in onboard processing systems to overcome these limitations.

ML models are trained on extensive datasets to detect patterns and anomalies in real-time. CS techniques enhance data compression, allowing spacecraft to transmit only the most relevant information while preserving critical details like geological features and potential biosignatures. These advancements improve mission efficiency, enabling faster scientific discoveries and reducing the burden on communication systems.

### ***1.3. AI IN MISSION PLANNING & OPTIMIZATION***

AI plays a crucial role in optimizing space missions by analyzing large datasets to enhance mission design and execution. AI-driven algorithms assess factors such as launch windows, orbital trajectories, and landing sites, improving mission efficiency.

For instance, AI uses historical mission data and space weather forecasts to recommend optimal launch timings and flight paths, minimizing risks. Additionally, AI enhances spacecraft design by simulating performance under various conditions, allowing engineers to refine configurations for maximum efficiency. Real-time AI-driven optimization also ensures adaptive decision-making during missions. AI continuously monitors fuel, power, and other resources, adjusting mission parameters as needed to optimize operations. Moreover, AI-based decision- support systems assist mission planners in evaluating multiple strategies, improving overall mission success rates.

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## **2. Related Works**

### ***2.1 AUTONOMOUS NAVIGATION FOR ROVERS***

Rovers like Sojourner, Spirit, Opportunity, Curiosity, and Perseverance rely on autonomy to mitigate Earth–Mars communication delays. Notably, Perseverance employs stereo vision and visual odometry to navigate independently, increasing daily traversal distances, despite hardware limitations and radiation challenges.

### **2.1.1 ROVER AUTONOMY AND SCIENCE AUTONOMY**

NASA's Mars rovers have progressively increased autonomy. The AutoNav system on Perseverance builds on stereo vision and onboard mapping to plan traverses at higher speeds with fewer ground interventions. AEGIS (Autonomous Exploration for Gathering Increased Science) enables onboard target selection for instruments (e.g. ChemCam, SuperCam), letting rovers identify and interrogate scientifically valuable rocks without waiting for Earth commands. These systems have been documented across mission notes and academic venues.

### **2.1.2 AI FOR HABITAT PLANNING**

Studies on lunar base design and site selection increasingly integrate data-driven approaches: combining illumination, temperature cycles, terrain hazards, and proximity to volatiles to recommend candidate outpost regions and power/storage strategies. Although the field is emerging, recent work demonstrates how optimization and machine learning can co-design sites and systems that exploit local resources.

### **2.1.3 AI IN SATELLITE OPERATIONS**

ESA's OPS-SAT provides a flying laboratory to flight-test AI for onboard image processing, autonomous mission operations, and anomaly detection. Recent datasets and competitions based on OPS-SAT encourage reproducible benchmarking. Parallel literature in predictive maintenance outlines AI techniques—from anomaly detection in telemetry to component health forecasting—now being adapted to spacecraft.

## **2.2 AI FOR AUTONOMOUS ROVER NAVIGATION**

### **2.2.1 PERCEPTION**

Modern rover stacks use stereo vision, terrain classification, and hazard detection to create traversability maps. Machine learning models (e.g., convolutional nets) distinguish benign regolith from slip-prone sand or sharp rocks. The NOAH-H system exemplifies ML-based terrain classification that can be adapted for Mars analogs, improving route planning.

### **2.2.2 PLANNING AND CONTROL**

Auto Nav on Perseverance fuses geometry, slip risk, and occlusions to plan safe paths at higher average velocities than predecessors. Paired with advanced traction control and continuous replanning, this reduces the number of ground-in-the-loop driving sols and increases science ops per unit time.

### **2.2.3 SCIENCE AUTONOMY**

AEGIS enables onboard target detection, ranking, and retargeting for instruments—e.g., selecting outcrops for laser spectroscopy based on visual salience or geologic criteria. This closes the loop between navigation and science: the rover not only reaches a site but immediately conducts context-aware measurements, boosting campaign efficiency under tight energy and communication budgets.

## **2.3 AI FOR PLANETARY HABITAT PLANNING**

### **2.3.1 SITE SELECTION UNDER MULTI-CRITERIA CONSTRAINTS**

Selecting lunar or Martian habitat sites requires balancing illumination (for solar power), thermal stability, line-of-sight communications, slope/hazard risks, and ISRU potential (e.g., water ice). Optimization and learning methods can perform multi-objective ranking of candidate tiles from orbital datasets (e.g., LRO, HiRISE). Studies show designs that exploit local resources and environmental cycles to reduce imported mass and risk.

### **2.3.2 SYSTEMS CO-DESIGN AND DIGITAL TWINS**

AI-enabled digital twins can co-optimize power systems, thermal loops, radiation shielding, and life support with habitat geometry. Surrogate models accelerate scenario testing across dust storms, eclipse periods, and equipment faults, informing redundancy levels and storage sizing.

### **2.3.3 OPERATIONS AND SAFETY**

Online learning can adapt environmental controls to occupancy and external conditions, while anomaly detection protects critical subsystems (ECLSS, power electronics). Mission-level planners can allocate crew time, schedule ISRU runs, and manage consumables based on predictive models derived from telemetry and simulation.

## **2.4 AI FOR SATELLITE EFFICIENCY**

### **2.4.1 ONBOARD PROCESSING AND EDGE AI**

OPS-SAT demonstrations show that neural networks can compress, classify, and prioritize imagery in orbit, downlinking only high-value products and enabling faster response to dynamic events (e.g., fires, storms). Onboard autonomy can also optimize pointing, tasking, and inter-satellite routing under bandwidth constraints.

#### 2.4.2 ANOMALY DETECTION AND HEALTH MONITORING

AI models trained on telemetry streams detect outliers in attitude control, power, and thermal subsystems. The newly published OPS-SAT anomaly benchmark provides real satellite telemetry to compare methods—a crucial step toward standardizing evaluations. Predictive maintenance reviews Highlight Model Classes (Autoencoders) And The Importance Of Explainability And Verification For Operational Adoption.

#### 2.4.3 MISSION AUTONOMY AND DYNAMIC TARGETING

Emerging systems enable satellites to decide what to observe and when, skipping cloud-obscured scenes and opportunistically retasking for transient events. Demonstrations with compact EO platforms underline how onboard AI can cut latency from minutes to seconds, increasing yield per orbit.

#### 2.4.4 SCHEDULING AND GROUND SEGMENT OPTIMIZATION

AI can optimize downlink schedules against weather, pass geometry, and network congestion—an area being pursued by commercial and agency teams alike. Early in-orbit experiments (and ground pilot projects) suggest improved throughput and reduced operator workload when AI is introduced into the mission planning loop.

### 2.5 RELATED WORKS

The navigation systems should have the capacity of making autonomous decisions when necessary based on the stated mission objectives. Procedures and algorithms for making decisions, determine priority, channel resources and success in dynamic situations. They approach uncertainties using problems modeling techniques like Markov decision processes (MDPs) and decision trees.

Besides adaptability and learning capability which are the essential part of AI-based systems it is also necessary. Machine learning models are updated at each data collection stage and performance of the models improves over time by the addition of new data acquired. The online learning techniques employed in this system enable it to update its models and strategies which are precisely aimed at coping well with real time input and are thereby effective in dynamic and inconsistent conditions.

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## Applications Of AI in Space Exploration

1. Communication between the spacecraft and the Earth has been a major issue, as it takes an enormous amount of time for a message to be delivered. Some space missions have even had weeks-long delays as a result of this phenomenon. However, with artificial intelligence-based technologies, such as machine-learning robots, users no longer need to worry about communication because the device itself chooses its next course of action using its software. Also, while researching places where communication is not possible and therefore proper coordination is lost, artificial intelligence helps robots act independently and avoid obstacles. Rovers like the Mars Exploration Rover use sensors that detect rocks, craters, and other environmental hazards, and their specially built AI determines the best path by analyzing the received data. This helps the rover avoid any collisions and function without suffering any technical harm. Other vision-based detection systems use their software to find interesting and unique rocks to sample, and then the users access the images of the samples that the device captures.

2. To navigate and prevent accidents with other satellites in orbit, major corporations are using AI-based technology. For instance, SpaceX uses algorithms that analyze satellite location and velocity information to find potentially dangerous maneuvers. The satellite's internal computer uses the data to change the speed and direction of the satellite to prevent collisions with outside objects like comets. Using rule-based systems that combine expert knowledge about space navigation and collision avoidance is an alternative strategy. These systems are capable of being programmed with rules that specify how the satellite ought to react to various objects and circumstances, such as slowing down or altering course. AI can also be used to monitor the condition of the satellite and its parts, like its solar arrays and thrusters, and to decide how to improve performance and avoid malfunctions. Even if in-built software machine learning robots get damaged and fail to complete their space missions, the consequence will not be fatal, as human safety is more vital than potential threats to AI-based robots.

3. By offering more precise and effective ways to analyze data from space missions, artificial intelligence (AI) can aid in data analysis for space exploration. In order to spot anomalies that can be signs of possible discoveries or dangers, machine learning algorithms can help recognize patterns in data from satellites, probes, and other space research gear. For example, an AI system could be trained to recognize patterns in the data that indicate the presence of a new object or phenomenon, such as a comet or asteroid. The system could then alert human operators or make decisions about how to respond to the discovery. Similarly, AI can be used to recognize patterns in satellite data by analyzing large amounts of data and identifying correlations and trends that may not be immediately apparent to human operators. This can include identifying changes in weather patterns, tracking the movement of ocean currents, or detecting changes in vegetation cover on the Earth's surface. To recognize patterns in data, AI systems use a variety of techniques, including machine learning algorithms, neural networks, and deep learning. These systems are trained on large datasets of historical data and can learn to recognize patterns and make predictions based on that data. Through the following chart, comparisons of the accuracy of data analysis between physicians and AI are provided.

Also, it is impossible for humans to travel and examine the particular sites due to extreme conditions that are deadly for human life. Several planets' surrounding radiation shields prevented mankind from traveling there. But, with the use of AI-based technologies, people no longer worry about such circumstances and may use machine-learning robots to investigate the place. Robots can effectively operate in harsh space environments by utilizing machine learning algorithms that enable them to adjust to changing conditions, such as temperature or atmospheric pressure. For instance, if a robot detects a change in temperature or atmospheric pressure, it can predict how these changes will impact its performance and modify its behavior accordingly. This could involve altering its speed, changing its trajectory, or improving its sensors to better detect changes in the environment. By continuously analyzing data and adapting to changing conditions, robots can function efficiently in challenging space environments and accomplish their objectives successfully. This adaptation ensures that the robot can continue to function optimally despite the challenging conditions.

Environment	Physical capabilities	AI Robot capabilities
Extreme temperatures	-150°C to +120°C	-270°C to +400°C
Radiation exposure	1,200 millisieverts	Higher than humans
No atmosphere	10 minutes	Limitless
Low gravity	Dangerous	Operates successfully

**TABLE I Comparisons between the physical prowess of humans and robots in adverse space environments.**

## 4. Challenges & Future Roles

### 4.1 RECENT ADVANCES

The application of AI to space exploration has progressed considerably over the last few decades. The application of machine learning methods in space operations has been one of the most important developments in recent years. A subset of AI known as machine-learning algorithms enables computers to learn from data and make decisions based on that data. In space operations, machine learning algorithms have been used to evaluate data, forecast results, and make mission-related choices. The Mars 2020 expedition was one of the most notable recent instances of artificial intelligence being used in space exploration. An AI-based autonomous guidance system that enabled the Mars 2020 mission to choose its course and avoid obstructions was installed on board. A machine-learning method was used by the automated navigation system to examine data from the spacecraft's cameras and make mission-related choices. The use of deep learning algorithms is another new development in the application of AI to space travel. Artificial neural networks are used by deep learning algorithms, a subset of machine learning algorithms, to learn from data. Deep learning algorithms have been used to evaluate data, make forecasts, and find anomalies in space operations.

### 4.2 CHALLENGES

Making AI systems dependable in the harsh conditions of space is one of the largest challenges facing AI use in space flight, despite the fact that AI offers several advantages over humans when it comes to resistance to harsh environments. Electrical components on spacecraft and rovers can be harmed or interfered with by extreme heat, radiation, and other hazards. To do this, AI systems must be built to be robust and resilient, with redundant components and fail- safe techniques. For instance, the Mars Exploration Rovers were built with several processors and sensors to ensure that they could continue to operate even if one or more of their components failed.

Utilizing machine learning algorithms that can adjust to shifting circumstances is another strategy for ensuring reliability. Machine learning techniques are used, for instance, by the Mars Reconnaissance Orbiter to recognize and categorize characteristics on the Martian landscape. The programs can adjust to alterations in the lighting and atmosphere, gradually increasing their precision.

Keeping the data gathered by spacecraft and rovers private and secure is a major obstacle to using AI in space research. These devices can collect extremely confidential data, including details on the makeup of planets, the existence of water, and other important scientific information. AI systems must be built with strong security features, such as encryption and identification methods, to meet this challenge. Data must be transmitted only to approved employees on Earth and kept securely. Utilizing on-board processing and analysis is another strategy for guaranteeing data security and anonymity. Sensitive data can be kept private and only pertinent information relayed back to Earth by analyzing data on board satellites and rovers. This strategy can also lessen data transfer delays, enabling quicker decision-making.

Although AI systems have the potential to be extremely effective and strong, they are not perfect. The requirement for human supervision is one of the most important obstacles to using AI in space travel. To make sure AI systems are functioning properly and making the right choices, human oversight is required. For instance, during the Mars Exploration Rover expedition, human controllers on Earth kept an eye on the rovers' progress and decided where to go and what to do based on the information the rovers had collected. Human oversight is also necessary to ensure that AI systems are not making decisions that could put humans or other spacecraft at risk. For example, in the event of a collision course with another spacecraft or object, human operators must have the ability to override the AI system and take control of the spacecraft or rover. Training and inference are the two stages that AI approaches for onboard processing must go through. The AI model first goes through the training process. To anticipate the satellite

configuration for the system conditions, the goal is to identify the ideal model parameters. Then, based on the input data, the trained model is obtained and utilized to suggest system parameters. In some circumstances, training expenses can be higher than anticipated, which makes it more difficult to manufacture the technology on a budget. The trends of estimated expenditure spent on training on machine learning systems can be viewed through this graph.

### 4.3 FUTURE ROLES OF AI IN SPACE EXPLORATION

Deep space travel is one of the most intriguing potential uses of AI in space exploration. Deep space travel presents significant difficulties, including the need for independent systems and the immense distances involved. AI can assist us in overcoming these difficulties. AI, for instance, can assist in the development of autonomous spacecraft that can traverse the vastness of space and decide what to do and where to go based on the information they collect. In order to make sense of the complicated and constantly shifting environments of deep space, AI can also assist us in the analysis of the enormous amounts of data collected by deep space projects. AI can also contribute to the creation of new technologies that will enable deep space travel. AI, for instance, can assist us in research, and we have made enormous strides toward independent AI systems. AI will be the persistent competitor for all manned missions and robotics beyond low-Earth orbit. In the near future, an enormous number of machine-learning robots will be trained to conduct space surveys and research in space, replacing humans in environments dangerous to them. While AI-based robots can conduct more accurate investigations and gather more reliable data, they do not require as much funding as astronauts.

## 5. Conclusion

Artificial Intelligence is redefining the future of space exploration by providing the autonomy and efficiency needed to overcome communication delays, resource constraints, and unpredictable extraterrestrial environments. In autonomous rover navigation, AI allows planetary rovers to safely traverse complex terrains, make independent decisions, and even conduct scientific analysis without constant human input—maximizing mission efficiency and scientific yield. In planetary habitat planning, AI assists in selecting safe and resource-rich sites, optimizing life-support systems, and enabling sustainable human settlement on the Moon and Mars through intelligent use of local resources and adaptive management. Meanwhile, in satellite efficiency, AI enhances operations by performing onboard data processing, anomaly detection, predictive maintenance, and dynamic scheduling, which extend mission lifespans and improve the quality of space-based services.

Together, these applications highlight AI's critical role as a partner in humanity's quest to explore and inhabit space. While challenges remain in terms of verification, trust, and robustness, AI-driven autonomy continues to accelerate our ability to explore new worlds, safeguard human missions, and harness space technologies more effectively than ever before.

AI has moved from concept to capability in spaceflight. On Mars, autonomous navigation and science autonomy now measurably increase traverse rates and scientific return. For future habitats, AI can integrate site selection with system co-design to reduce risk and logistics burden. In orbit, AI turns satellites into responsive agents through onboard processing, anomaly detection, and predictive maintenance. Real-world deployments (Auto Nav, AEGIS) and in-orbit labs (OPS-SAT) provide a clear path to scalable, trustworthy autonomy—provided the community invests in rigorous V&V, shared datasets.

The use of AI in space exploration has given a new dimension in space missions with higher efficiency and increased rate of success. Recent innovations in the field of AI have led to the new ways of optimizing autonomous navigation, implementing real-time data processing, and performing predictions that make the mission planning and efficient management of the resources much more efficient when it comes to dealing with the unforeseen problems. These advancements suggest that future AI technologies, particularly quantum and neuromorphic computing, will continue to transform space exploration. Nonetheless, to leverage AI to the maximum, it is critical to overcome all the issues relevant to the system's durability, interaction with human personnel, and cooperation with other systems and devices. Thus, the further directions should include the enhancement of the AI systems' resilience, forming the set of universally accepted norms, and the estimation of the potential ethically-loaded risks and outcomes for the AI-effective space mission. Current development in AI shows that there is more progress to be made, and such development will be of great value as it opens up new frontiers towards the exploration of the universe.

The introduction of AI-driven navigation systems is a decisive step up in the manner of space exploration experimentation. This type of system focus on the seamless flow of integrating sophistication of sensors technic, advanced machine learning algorithms, and real-time decision-making approach for the successful operation of spacecraft under such exorbitant space conditions. Unmanned aircraft technologies provide many advantages such as flight conformity, enhanced safety of missions and greater mission efficiency which as a whole brings down the overall operational costing. Throughput, stringent simulation, and testing protocols, the components remain dependable and resilient, responding intellectually to the unknown and unpredictable, and shifting in dynamics when the settings vary. We will continue inching further towards challenging of space exploration, and AI-enabled autonomous navigation systems will stay crucial element in enabling missions at deep space, hence, broadening our understanding of the universe and our place in it. Space missions of the last century had been a joint effort between different space agencies. There had also been interactions between research institutions and the private sector. These cutting-edge technologies will be the drivers behind the next stage of space exploration, a stage that will be full of discoveries and multitude of opportunities.

## 6. REFERENCE

1. Ahamad, A. Iqbal, M. A. Alotaibi, H. Malik, F. P. García Márquez, and A. Afthanorhan, "Packed U Cell seven level and five level inverter topologies for renewable energy applications," *Int. J. Math. Eng. Manag. Sci.*, vol. 9, no. 4, pp. 881–901, 2024, doi: 10.33889/IJMEMS.2024.9.4.046.

2. S. Malik et. al., "Ambiguous Fuzzy Einstein Ordered Averaging Operator: Application to the classification of Power Generation methods," in ICACITE, IEEE, May 2024, pp. 1853–1857. doi: 10.1109/ICACITE60783.2024.10617360.
3. S. Rathee, A. Mittal, N. Kumar and A. Singh, "Assessment of Two 20Kw PV Solar Energy Generation Plants in Homogeneous Environments," 2023 PEEIC, Greater Noida, India, 2023, pp. 449-455, doi: 10.1109/PEEIC59336.2023.10451686.  
A. Hardas, "Moving Object Detection for Video Surveillance using Different Color Spaces," vol. 118, no. 13, pp. 39–43, 2015.
4. V. R. Saxena, P. Singh, and A. Tiwari, "Role of artificial intelligence based applications used for space technologies," *Journal of Applied Science and Education (JASE)*, vol. 2, no. 1, pp. 1–8, 2022.
5. S. Kumar and R. Tomar, "The role of artificial intelligence in space exploration," in 2018 International conference on communication, computing and internet of things (IC3IoT). IEEE, 2018, pp. 499–503.
6. J. Maki, D. Gruel, C. McKinney, M. Ravine, M. Morales, D. Lee,
7. R. Willson, D. Copley-Woods, M. Valvo, T. Goodsall et al., "The mars 2020 engineering cameras and microphone on the perseverance rover: A next-generation imaging system for mars exploration," *Space Science Reviews*, vol. 216, pp. 1–48, 2020.
8. P. Tompkins, A. Stentz, and D. Wettergreen, "Global path planning for mars rover exploration," in 2004 IEEE Aerospace Conference Proceed- ings (IEEE Cat. No. 04TH8720), vol. 2. IEEE, 2004, pp. 801–815.
9. M. W. Maimone, P. C. Leger, and J. J. Biesiadecki, "Overview of the mars exploration rovers' autonomous mobility and vision capabilities," in IEEE international conference on robotics and automation (ICRA) space robotics workshop, 2007.
10. J. Kua, S. W. Loke, C. Arora, N. Fernando, and C. Ranaweera, "Internet of things in space: A review of opportunities and challenges from satellite-aided computing to digitally-enhanced space living," *Sensors*, vol. 21, no. 23, p. 8117, 2021.
11. D. Olawade, O. Z. Wada, A. O. Ige, and B. Egbewole, "Artificial Intelligence in Environmental Monitoring: Advancements, Challenges, and Future Directions," University of East London, Hamad bin Khalifa University, University of Ibadan, Virginia Tech, 2023.
12. Efficient AI Models for Extreme Edge Environments: A Comprehensive Review for Space, Underwater, and Disaster Zones
13. Russo, A., & Lax, G. (2022). Using Artificial Intelligence for Space Challenges: A Survey. *Appl. Sci.* 12(10), 5106;
14. <https://doi.org/10.3390/app12105106>
15. Oche, P.A., Ewa, G.A., & Ibekwe, N. (2021).
16. Applications and Challenges of Artificial Intelligence in Space Missions. *IEEE Access*, PP, 1-1.  
<https://doi.org/10.1109/ACCESS.2021.3132500>
17. Burroughes, G., & Gao, Y. (2016). Ontology-Based Self-Reconfiguring Guidance, Navigation, and Control for Planetary Rovers. *J. Aerosp. Inf. Syst.*, 13, 316-328. <https://doi.org/10.2514/1.1010378>
18. Azkarate, M., Gerdes, L., Joudrier, L., & P'erez-del- Pulgar, C.J. (2019). A GNC Architecture for Planetary Rovers with Autonomous Navigation. 2020 IEEE International Conference on Robotics and Automation (ICRA), 3003-3009.