

# Application of DeepSeek in Intelligent Optoelectronic Countermeasure

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**Abstract:** With the rapid advancement of war to intelligence, electro-optical countermeasure faces challenges such as high fidelity of situational awareness and high intelligence of game decision-making. Starting from the technical architecture of DeepSeek, this paper systematically discusses its application prospects and challenges in the field of intelligent photoelectric countermeasures. DeepSeek can effectively integrate multi-modal perception, hierarchical reinforcement learning, hardware collaborative optimization and other technologies, realize the 'perception-decision-confrontation' closed-loop feedback from sensor perception, system decision-making to confrontation behavior, and expand the development trend of deep learning in the electro-optical countermeasure system, in order to eliminate the shortcomings of the current electro-optical countermeasure system, such as long-standing but undetected system response lag and strategy rigidity. Based on the research needs of data-driven, algorithm robustness and computing resource constraints, the new difficulties that DeepSeek will encounter in the application of intelligent photoelectric countermeasure are further discussed and solutions are proposed, such as multi-agent game, brain-like photoelectric cooperation and other technical methods.

**Keywords:** DeepSeek; Intelligent photoelectric countermeasure; Application.

## 1. Introduction

At present, with the development of information warfare, the technology based on electromagnetic confrontation and photoelectric confrontation has become one of the focuses of arms competition among countries around the world. With the advantages of high resolution and strong anti-electromagnetic damage ability, the photoelectric system has been widely used in enemy target reconnaissance, target guidance, electromagnetic attack and other fields. However, under the background of information warfare, the diversification and complexity of the application scenarios of weapon equipment systems are highly related to the battlefield environment faced by photoelectric equipment systems, which puts forward high requirements for the detection resolution and anti-interference ability of photoelectric countermeasure systems. In this case, the traditional photoelectric stealth materials and anti-stealth interference technologies are more and more difficult to adapt, and the photoelectric system identification algorithms and tactical strategies using these technologies cannot be stable and reliable. At the same time, the modern battlefield environment requires the response speed of the photoelectric system to reach the millisecond level. The intelligent processing method based on static rules is no longer suitable for the intelligent countermeasure system. The photoelectric equipment system urgently needs the support of artificial intelligence technology. The artificial intelligence platform technology represented by DeepSeek products can effectively solve the 'static problem', complete independent decision-making and dynamic response under the support of self-learning, self-evolution, algorithm matching and other capabilities, and promote the upgrading and development of the photoelectric intelligent countermeasure system.

## 2. DeepSeek's technical architecture and core capabilities

The underlying technical architecture support determines

the performance level of the intelligent photoelectric countermeasure system. By integrating deep learning, multi-modal fusion, dynamic game theory and other technical architectures, DeepSeek can establish an intelligent decision-making system for complex battlefield scenarios, and realize the autonomy and intelligence level of electro-optical countermeasure system around 'perception-decision-confrontation'. This chapter analyzes the technical empowerment of DeepSeek to photoelectric countermeasure from two aspects of technical principle and core function.

### 2.1. Technical principles: hierarchical architecture and multi-modal collaboration

The technical architecture of DeepSeek adopts the hybrid mode of 'data-driven + knowledge embedding', which is divided into three modules: perception layer, decision layer and execution layer.

(1) Perception layer : multi-source heterogeneous data fusion

When the photoelectric countermeasure involves different types and heterogeneous data of various sensors such as infrared, visible light, lidar, etc., the Cross-Modal Alignment Network (CMAN) can be designed through DeepSeek. The correlation characteristics between spectrum, space and time are obtained through the self-attention mechanism, and the sensor noise and battlefield environment interference are eliminated or reduced through confrontation training. In the target recognition task, CMAN can fuse the radiation features of infrared thermal imaging with the texture features of visible light images to generate characterizations with strong robustness to occlusion and interference, thereby improving the accuracy of target recognition and detection under severe weather conditions.

(2) Decision-making layer : dynamic game strategy generation

The strategy layer adopts the DRL system. According to the hierarchical Markov Game (HMGM) model, the

photoelectric countermeasure strategy of both sides is modeled. The deep Q network is used to evaluate the effectiveness of different interference strategies, and the interference strategy is deduced by Monte Carlo Trees Teaching (MCTS). For example, in the process of jamming waveform design, according to the analysis law of frequency switching of local radar on the battlefield, multiple frequency points that may be switched by enemy radar are calculated in real time, and the jamming waveform is planned to interfere with multiple frequency points, so that the enemy receiver appears ' frequency mixing ' phenomenon. The online learning mechanism of the decision-making layer can make the strategy library feedback the combat data in time and evolve continuously to adapt to the confrontation environment.

(3) Execution layer: hardware-algorithm collaborative optimization

Aiming at the millisecond delay requirement of photoelectric countermeasure, DeepSeek can deploy the deep learning model on the FPGA + GPU acceleration platform based on the heterogeneous computing architecture. Through operator-level hardware optimization ( such as pipelined parallel strategy of convolution computing ) and model optimization ( such as channel pruning and quantitative perception training ), the system-level inference delay is further reduced to less than 5ms, which is an order of magnitude improvement compared with the general computing platform. The execution layer is deeply integrated with the hardware interfaces of photoelectric devices such as laser jammers and infrared decoys, supporting closed-loop adjustment of parameters such as interference energy and operating distance, and truly completing the seamless connection of ' decision-action '.

## 2.2. Core competence: the triple breakthrough of battlefield intelligence

(1) DeepSeek 's technical architecture gives it three core capabilities, which can solve the key bottlenecks of traditional electro-optical countermeasure systems.

High real-time performance: from second-level response to millisecond game

The traditional photoelectric system uses the method of artificial rule base and off-line calculation to make decision processing. The response time is usually in seconds, which is difficult to meet the needs of high dynamic battlefield. DeepSeek adopts the edge-cloud collaborative computing architecture, and deploys the lightweight model on the front-end photoelectric platform to realize front-end data processing. The global strategy is optimized and trained through a distributed reinforcement learning cluster. For example, in anti-UAV combat, the trajectory and laser interference angle prediction of such targets can be completed in about 20 ms.

(2) Strong robustness: stable confrontation in complex environment

The working environment of photoelectric detection is complex, such as strong electromagnetic interference in battlefield, bad weather, camouflage deception technology and so on, which often reduce the performance of photoelectric system. DeepSeek improves the robustness of the model by using adversarial training and domain adaptation techniques. On the one hand, noise samples are added in the training process by injecting Gaussian noise and motion blur, so that the model can learn invariant features in

adversarial training. On the other hand, the meta-learning framework is used to quickly adapt to unknown combat environments. In the dense environment of electronic warfare, the infrared target tracking of DeepSeek can maintain a low error discrimination rate, and it is expected to achieve an error reduction of more than 50 % compared with the traditional algorithm.

(3) Self-adaptability: from static rules to dynamic evolution

The original rule pre-training is only applicable to common threats, and the online incremental learning of the DeepSeek model can find subtle changes from the abnormal agents found under the newly discovered threat attack. The online incremental learning can fine-tune the parameters of the model according to the feedback information. If the system finds that the other party uses a new stealth coating, the recognition confidence of the other party 's stealth target is reduced. DeepSeek can automatically fine-tune the original model and re-fit the feature space under a small amount of data information. At the same time, the simulation environment of DeepSeek multi-agent can interact with the virtual reality of the agent through thousands of simulation attack scenarios, so that the agent can update the decision boundary from the mastered experience. When encountering enemy interference in unknown frequency bands, DeepSeek can achieve adaptive decision-making and reconstruct its own communication.

## 2.3. Typical application scenarios

In order to verify the effectiveness of the technical architecture, DeepSeek can be tested in multiple electro-optical countermeasure scenarios:

Scene 1 : In the shipborne electro-optical countermeasure system, DeepSeek identifies the sea-grazing anti-ship missile through multi-spectral fusion, controls the laser directional jammer to blind, and tests the increase of interception success rate.

Scenario 2: In the plateau mountain environment, the system uses transfer learning to quickly adapt to the sensor performance offset caused by low oxygen and low temperature, and test the target tracking accuracy offset.

## 2.4. Summary of technical value

The full-stack technical architecture of ' multi-modal perception-dynamic game decision-hardware collaborative execution ' is DeepSeek to break through the ' dead holes ' of traditional photoelectric countermeasures, such as slow response, environmental vulnerability and rigid strategy. It obtains reinforcement capabilities from three modules: multi-source and multi-dimensional information perception, complex dynamic game decision-making and hardware execution coordination, and establishes an enabling closed-loop from data to decision-making and from software to hardware outside the algorithm, which will become a powerful starting point for realizing intelligent photoelectric countermeasures from theory to practice.

## 3. Three application directions of DeepSeek in intelligent photoelectric countermeasure

The underlying architecture of DeepSeek constructs the technical basis for supporting intelligent photoelectric countermeasures. The practical application value of DeepSeek also needs to be fully demonstrated in three typical

application scenarios : ' intelligent target recognition and tracking ', ' photoelectric interference strategy generation and optimization ', and ' cooperative / networking self-decision '. The following three typical applications of DeepSeek and the new mode of realizing intelligent photoelectric countermeasure are studied and discussed respectively.

### **3.1. Target intelligent recognition and tracking: from ' passive detection ' to ' active perception '.**

In the past, the photoelectric system used the known feature library to identify the target, which is easily affected by environmental interference and camouflage means. DeepSeek can significantly improve the effectiveness of target perception in complex environment by means of multi-modal feature fusion and adversarial learning technology.

#### **(1) Multi-spectral data fusion and feature enhancement**

The multi-sensor information fusion module uses multi-source perceptual information and multi-source fusion Cross-Modal Attention Network (CMAN) to align and complement features of infrared, visible light, lidar and other data sources. For example, under the condition of dense fog weather occlusion, the visibility of the visible light sensor is reduced and the target shape is blurred, while the infrared sensor can normally obtain the target radiation heat information. CMAN dynamically assigns and fuses the weights according to the different characteristics of the two data to generate an anti-interference fusion feature map.

#### **(2) Adversarial sample defense and robustness enhancement**

For adversarial sample attacks against adversaries ( such as adding spoofing recognition algorithms ), DeepSeek can respond by adversarial training and feature decoupling. Adding adversarial samples to the training sample set requires the model to decouple the essential characteristics of the target and the characteristics of noise interference, which can cope with typical attack methods such as gradient mask and FGSM (fast gradient sign method).

#### **(3) Dynamic trajectory prediction and adaptive tracking**

Based on Spatiotemporal Graph Convolutional Network (ST-GCN), DeepSeek models the spatio-temporal relationship of the target trajectory to achieve accurate prediction of the trajectory. For example, in the application of intercepting supersonic cruise missiles, the trajectory deviation within 0.5 s is quickly predicted by on-line detecting the acceleration, attitude angle change and atmospheric disturbance of the missile, and the laser gun is guided to complete the dynamic correction.

### **3.2. The generation and optimization of photoelectric interference strategy: from ' fixed mode ' to ' intelligent game '.**

The traditional interference strategy based on the artificial experience library can only be ' passive defense ' for the dynamic changing confrontation environment, while DeepSeek can realize the autonomous generation and dynamic update of the interference strategy by virtue of deep learning combined with the game theory model.

#### **(1) Adversarial interference waveform design**

DeepSeek can use the combination of generative adversarial network (GAN) and deep Q network (DQN) as the interference signal generation mechanism of intelligent deception enemy photoelectric system. For example, when

dealing with pulse Doppler radar, comb jamming covering multiple frequency bands can be generated according to the pulse repetition frequency ( PRF ) and coding law of the other radar, so that the enemy radar processor is overloaded.

#### **(2) Adaptive resource allocation and cooperative jamming**

In the case of limited interference resources (including transmit power, frequency bandwidth, etc.), DeepSeek 's multi-agent reinforcement learning (MARL) framework can be used to optimize resource allocation. For example, in the ship-borne photoelectric jamming, the system can respond and allocate the priority of the laser jammer and the infrared decoy according to the threat level : for the high-threat anti-ship missile, the directional blind jamming is implemented, and the low-threat anti-ship missile releases the decoy to deceive.

#### **(3) Online learning and counter-strategy evolution**

The online incremental learning module of DeepSeek can optimize the anti-jamming means in real time for the enemy 's new jamming technology. For example, an enemy system adopts frequency hopping communication means to anti-jamming. DeepSeek can use online deep learning to iterate and optimize, change the center frequency point of jamming in real time, and realize convergence optimization within 30 s.

### **3.3. Collaborative engagement and system autonomous decision-making: from ' single point of confrontation ' to ' system linkage '.**

Optoelectronic equipment often faces cross-platform cooperative combat scenarios in the information battlefield. DeepSeek can realize the intelligent scheduling of global combat resources through multi-agent distributed decision-making structure and multi-agent cooperative algorithm, which can effectively improve the efficiency of optoelectronic equipment in information warfare.

#### **(1) Cooperative detection of multi-platform photoelectric system**

In order to realize the sharing of target features by heterogeneous platforms such as satellites, drones, and ground radars, while ensuring that local data is not leaked, DeepSeek can use the federated learning framework to build a federated learning model. In the low-altitude combat scenario, multiple drone groups cooperate to identify low-altitude aircraft, so that heterogeneous platforms share low-altitude target features, which can expand the observation field of view and improve the recognition response efficiency.

#### **(2) Dynamic battlefield situation deduction and decision-making**

DeepSeek can use Monte Carlo tree search MCTS based on game theory to conduct strategic game between adversaries and adversaries, and finally find the optimal solution for users. In the field of air defense and anti-missile, the most effective interception scheme is given by deducing the possible attack direction of enemy missiles and combining the deployment combination of interceptors, electronic countermeasures equipment and baits.

#### **(3) Autonomous mission planning and fault-tolerant control**

HTN (Hierarchical Task Network) can be used for multi-task decomposition and multi-task scheduling management. For example, after the photoelectric device located at a node is destroyed, the system quickly reconstructs the communication link and re-schedules the task to other nodes.

### 3.4. Summary of application value

In summary, the application of DeepSeek in the above scenarios marks the transformation of electro-optical countermeasures from 'rule-based passive combat' to 'data-based game active'. It not only improves the recognition accuracy and optimizes the utilization rate of jamming resources in technical indicators, but also includes the 'OODA' cycle that reshapes the new mode of 'observation, judgment and action' of the campaign, which provides a technical way to promote the development of iterative technology in intelligent warfare.

## 4. Challenges and future trends

DeepSeek's technological breakthrough in enabling intelligent photoelectric countermeasure represents a new level of application of deep learning combined with military, but it also faces many challenges at present. This section analyzes the factors that may hinder the development of DeepSeek in intelligent optoelectronic countermeasure applications and the future breakthrough path from the aspects of technical challenges, ethical issues and industrial issues.

### 4.1. Current technical bottleneck : multi-dimensional constraints from algorithm to hardware

#### (1) Data dependence and small sample learning problem

Although DeepSeek can improve the recognition ability of the target by means of multi-modal fusion, it needs to be trained on the basis of large enough labeled data. In actual combat scenarios, there will be some unpredictable situations, such as the enemy's new metamaterial stealth coating or temporary special meteorological conditions (such as the interference of plasma clouds), which will cause the data distribution in the original training model to shift, which will easily lead to model failure. In order to cope with this situation, DeepSeek can try to introduce a meta-learning framework and use limited data to quickly rebuild a feature library, but the general applicability of the model and the accuracy of individual applications are still unknown.

#### (2) Computing power dependence and hardware adaptation challenges

Although DeepSeek (3% of the same type of OpenAI model) with low inference cost makes the deployment of the algorithm no longer difficult, its inference decision model (such as hierarchical Markov game) is still quite complex, and the requirements for high-performance computing resources still exist. At present, the peak computing power of domestic AI chips (such as Cambrian MLU370) is only 65% of that of NVIDIA A100, and the lack of software ecological support makes it difficult to meet the needs of the system's wide deployment in equipment such as individual photoelectric countermeasure equipment. The problem that the response rate of the photoelectric device (such as laser frequency tuning) is not synchronized with the algorithm decision rhythm remains to be solved. There will be a 'decision-execution' link delay in individual scenarios.

#### (3) Algorithm robustness defects in complex battlefield environment

In the face of extreme scenarios such as strong electromagnetic environment and high-speed moving large clusters, DeepSeek's anti-attack response ability still has shortcomings. The ZeroDay attack defense ability faced by

the algorithm itself is also a key area to be improved. The robustness enhancement technology based on causal reasoning needs to be strengthened.

### 4.2. Future technology direction : cross-domain integration and paradigm innovation

#### (1) Deep combination of multi-agent reinforcement learning and game theory

DeepSeek can consider adding hierarchical multi-agent reinforcement learning (HMA-RL) technology to meet the needs of multi-level cross-platform coordinated confrontation. For example, satellites, drones and ground radars form a distributed multi-agent system, and the optimal interference resource allocation scheme is obtained by Nash equilibrium solution, so as to effectively improve spectrum efficiency and reduce energy consumption in electronic warfare systems.

#### (2) Neuromorphic integration of brain-like computing and optoelectronic hardware

Based on the spatio-temporal coding information of spiking neural network (SNN), DeepSeek can be combined with optoelectronic spiking synaptic devices, and the pulse communication information between spiking neurons of synapses is simulated by photon pulses, so as to realize the photoelectric countermeasure judgment delay below microsecond level.

#### (3) Edge computing and lightweight breakthrough of end-to-side AI

Aiming at the computing power limitation of front-line devices, DeepSeek-R1 inference small model compresses the model volume to less than 200 MB through dynamic network pruning and 8-bit fixed-point quantization, which can realize real-time decision-making on mobile phone-level chips. This technology can promote the rapid popularization of end-to-side equipment such as AI glasses and individual tactical helmets. It is expected that the market scale of end-to-side photoelectric countermeasure equipment will reach more than billions of yuan.

### 4.3. Application prospect: from tactical weapons to strategic system upgrade

#### (1) Popularization of industrial intelligent countermeasure system

In addition to military applications, DeepSeek can also be developed in the fields of military and civilian security and industrial detection to achieve 'peacetime and wartime integration'. For example, in the industrial control system, the infrared and vibration data are analyzed to effectively improve the accuracy of the failure rate prediction of the equipment.

#### (2) Intelligent coordination of global combat system

As the core AI engine, DeepSeek realizes the connection of each 'point' in the 'sky-space-ground' global data link. The U.S. military has begun to use similar applications in the 'Joint Global Command and Control' JADC2 test in 2025. The intelligent target assignment of the photoelectric detector from the 'satellite-UAV-ground station' multi-platform accelerates the threat response time from 10.8 seconds to 2.8 seconds. This will be a major innovation in the rules of information warfare.

#### (3) Ethical regulation and global technology co-opetition

The rapid development and application of generative artificial intelligence platforms in the military and civilian fields in recent years have caused global concern about the

loss of control of AI weapons. On the one hand, open source ecological construction (partial disclosure of model parameters) promotes the rapid development and diffusion of technology, on the other hand, it also enhances the possibility of reverse engineering by hostile forces. The role of innovation incentives and security constraints will become the core factor in determining whether to implement them at the technical level.

## 5. Conclusion

The future development of DeepSeek in the field of intelligent photoelectric countermeasure is not only facing the technical upper limit of data, computing power and robustness, but also facing the development opportunities brought by technological breakthroughs such as multi-agent collaboration, brain-like hardware and edge computing. At the same time, with the empowerment of generative artificial intelligence such as DeepSeek, it not only brings about the improvement of equipment performance, but also reorganizes the global combat chain of ' perception-decision-confrontation '. With the improvement of AI ethics and technical standards, the new generation of military AI key technology platform and subversive technology innovation driven by DeepSeek and other generative artificial intelligence will be accelerated under the integration of algorithm and photoelectricity, which also indicates that the integration of algorithm and photoelectricity in the era of intelligent war is no longer a choice, but a must-answer.

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