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# Emergency response strategy and simulation analysis considering inter-government coordination and information sharing

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## Abstract

This paper investigates the application of game-theoretic approaches to engineering management problems, focusing on multi-agent optimization in complex systems. We develop a mathematical framework that combines Nash equilibrium concepts with optimization theory to model decision-making processes in distributed engineering environments. Our computational experiments demonstrate significant improvements in system efficiency and resource allocation. The proposed methodology provides both theoretical foundations and practical implementation strategies for modern engineering management challenges.

**Keywords:** Game theory, Multi-agent systems, Engineering management, Optimization, Nash equilibrium

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## 1. Introduction

Rapid urbanization and technological advancement have significantly expanded disaster impacts, creating unprecedented challenges for governmental emergency management (ref). Modern urban systems' interconnected nature demands more sophisticated emergency response mechanisms (ref). Inter-regional governmental collaboration has emerged as a critical solution for enhancing resource efficiency and response capabilities (ref). This represents a shift from hierarchical to flexible, network-based emergency management systems (ref).

At the policy level, the importance of inter-governmental collaboration has been explicitly recognized in national emergency planning frameworks. For instance, China's 14th Five-Year National

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Emergency System Plan explicitly mandates the establishment of robust regional collaborative response mechanisms (ref). Similar policy initiatives across various countries emphasize the need for horizontal governmental cooperation to address the increasingly trans-boundary nature of disaster impacts (ref). Despite this policy emphasis and theoretical recognition, the practical implementation of such collaborative mechanisms faces substantial challenges that significantly impede their effectiveness in real-world emergency scenarios.

Three primary obstacles systematically undermine the effectiveness of horizontal inter-governmental collaboration in emergency response. First, the absence of clear delineation of rights and responsibilities creates operational ambiguity that hampers decisive action during critical emergency periods (ref). Many collaborative efforts rely heavily on spontaneous cooperation between horizontal governments without effective constraints or guidance from higher-level vertical administrative structures, resulting in coordination failures when rapid response is most needed (ref). Second, information sharing barriers severely compromise collaborative efficiency, as information transmission suffers from both technical obstacles and institutional resistance, creating dangerous blind spots in emergency situational awareness (ref). The lack of standardized information sharing protocols and interoperable communication systems further exacerbates these challenges, leading to duplicated efforts and missed opportunities for resource optimization (ref). Third, benefit coordination difficulties arise from the dominance of administrative division-based management models, where local governments inherently prioritize their own jurisdictional interests over regional collective benefits (ref). This "every man for himself" mentality becomes particularly pronounced when collaborative benefit distribution mechanisms are inadequately designed or absent, drastically reducing cooperation incentives and undermining the potential synergies of joint emergency response efforts (ref).

The existing body of literature has extensively explored the theoretical foundations and practical implications of inter-governmental collaboration in emergency management from various perspectives. Scholars have investigated the fundamental necessity and influencing factors of governmental collaboration through both theoretical frameworks and empirical case studies (ref). For instance, research on humanitarian organizations has examined inventory cooperation mechanisms and resource sharing strategies that could inform governmental collaboration models (ref). Additionally, studies adopting macro-level perspectives have analyzed the game-theoretic relationships between

central and local governments, providing insights into the strategic interactions that shape collaborative behaviors (ref). Evolutionary game theory has emerged as a particularly valuable analytical tool for modeling multi-agent coordination in emergency management contexts, offering dynamic perspectives on how cooperation patterns evolve over time under different institutional and environmental conditions (ref).

However, significant research gaps persist despite these valuable contributions to the field. First, there is a notable absence of rigorous modeling and analysis regarding information sharing platforms as specific solutions to collaboration challenges (ref). While information sharing barriers are widely recognized as critical obstacles to effective collaboration, few studies have employed mathematical models to quantitatively analyze how information sharing platforms might influence collaborative strategy evolution and emergency response outcomes (ref). Second, existing analytical approaches remain predominantly macro-level, focusing on aggregate benefits and losses without adequately capturing the micro-level practical factors that shape actual collaborative behaviors (ref). Critical operational details such as specific material coordination quantities, transportation costs, and the nonlinear characteristics of rescue benefits—which this paper models using S-shaped functions—have received insufficient attention in current research (ref). Third, the literature has largely overlooked the crucial role of benefit allocation and distribution mechanisms in horizontal governmental collaboration (ref). The design and implementation of specific benefit distribution schemes between horizontal governments, which are essential for overcoming local protectionism and sustaining long-term collaborative relationships, remain understudied despite their fundamental importance to collaborative success (ref).

These research gaps become particularly problematic when considering the practical implementation of emergency collaboration systems. The lack of quantitative models for information sharing platforms prevents policymakers from understanding the potential returns on investment in such infrastructure or optimizing their design for maximum collaborative benefit (ref). Similarly, the absence of micro-level analysis limits our understanding of how specific operational factors influence collaboration decisions, making it difficult to identify targeted interventions that could enhance cooperation likelihood (ref). Furthermore, without adequate attention to benefit distribution mechanisms, even well-intentioned collaborative initiatives may fail due to perceived inequities or misaligned incentives among participating governments (ref).

To address these critical gaps, this research pursues three primary objectives that collectively advance our understanding of horizontal governmental collaboration in emergency response. First, we construct a comprehensive evolutionary game model that incorporates both internal and external micro-level factors affecting resource sharing, including both material supplies and information exchange, in horizontal governmental emergency collaboration (ref). This model explicitly captures the complex interdependencies between disaster characteristics, regional positioning, cooperation efficiency, rescue benefits, and benefit coordination mechanisms that shape collaborative decisions in real-world emergency scenarios. Second, we conduct a comparative analysis of strategy evolution paths under two distinct scenarios: one with a vertical government-established information sharing platform and one without such infrastructure (ref). This comparison enables quantitative assessment of how information platforms influence the emergence and stability of cooperative equilibria, providing concrete evidence for the value of such investments. Third, we aim to provide theoretical foundations and policy recommendations for constructing effective inter-governmental collaboration mechanisms that can overcome the identified barriers to cooperation (ref).

The research questions guiding this investigation focus on understanding the micro-level determinants and macro-level interventions that shape collaborative behaviors in emergency response contexts. Specifically, we seek to identify the key micro-level factors—including disaster characteristics, regional location, cooperation efficiency, rescue benefits, and benefit coordination schemes—that influence horizontal local governments' strategic choices between cooperation and non-cooperation in disaster emergency response (ref). Additionally, we examine how information sharing platforms established by vertical governments can alter information efficiency and introduce incentive mechanisms to guide and influence the strategic evolution paths of horizontal governments toward more cooperative outcomes (ref). These questions are addressed through rigorous mathematical modeling and systematic analysis that bridges theoretical insights with practical implementation considerations.

This paper makes several significant contributions to the emergency management literature and practice. By developing a detailed evolutionary game model that captures previously overlooked micro-level factors, we provide a more nuanced understanding of the conditions under which horizontal governmental cooperation emerges and persists in emergency contexts. Our quantitative analysis of information sharing platforms offers concrete evidence for their value in promoting

cooperation, informing investment and design decisions for emergency management infrastructure. Furthermore, our examination of benefit distribution mechanisms provides practical guidance for designing collaborative agreements that align individual governmental interests with collective emergency response objectives. These contributions collectively advance both theoretical understanding and practical implementation of inter-governmental collaboration in emergency management, offering valuable insights for researchers, policymakers, and emergency management practitioners seeking to enhance collaborative emergency response capabilities in an increasingly interconnected and disaster-prone world.

## 2. Literature Review

### 2.1. Game Theory Applications in Engineering

Game theory has been extensively applied to engineering problems since Nash’s groundbreaking work (?). Recent developments have extended these concepts to complex multi-agent scenarios (?).

## 3. Model Building

### 3.1. The Baseline Two-Player Evolutionary Game Model

We first establish a baseline evolutionary game model to analyze the strategic interactions between two horizontal governments during a disaster emergency. The model focuses on the decision-making process regarding cooperation on resource sharing, which includes both relief supplies and critical information. It assumes that the governments are boundedly rational and dynamically adjust their strategies based on the payoffs from previous interactions. This baseline game model does not include a higher-level (vertical) government. The baseline model is built upon the following key assumptions:

**Assumption 1.** The game involves two players, i.e., two governments at the same administrative level. The first player is the Local Government (LG), which represents the government whose jurisdiction is primarily affected by the disaster and is in need of assistance. The second player is the Neighboring Government (NG), which represents the government of an adjacent region that possesses surplus resources and can offer aid.

**Assumption 2.** Each player has a strategy set of {Cooperate (C), Not Cooperate (NC)}. Let  $x$  be the probability that the LG chooses C, and  $(1 - x)$  be the probability it chooses NC, where  $x \in [0, 1]$ . Similarly, let  $y$  be the probability that the NG chooses C, and  $(1 - y)$  be the probability it chooses NC, where  $y \in [0, 1]$ .

**Assumption 3.** The players are not perfectly rational; instead, they learn and adapt their strategies over time based on the relative success of past choices.

**Assumption 4.** The rescue benefit derived from relief supplies follows an “S”-shaped function, which realistically captures the marginal utility of resources, from scarcity to abundance. The function is defined as:

$$F(\theta) = \frac{c}{1 + e^{-a\theta+b}} \quad (1)$$

where  $\theta = X/D$  represents the material satisfaction rate (the ratio of allocated supplies  $X$  to demand  $D$ ), and  $a, b, c$  are benefit coefficients.

**Assumption 5** (Local Government’s Strategic Considerations). When choosing to cooperate with the Neighboring Government, the LG can obtain additional relief supplies through regional coordination. When both LG and NG actively cooperate, both governments incur a cooperation cost  $H$ , and the LG gains public credibility  $G_L$  for its collaborative efforts. According to the Interim Measures for the Management of Central Emergency and Disaster Relief Material Reserves (ref), following the principle of “user pays,” the LG bears the transportation cost for the shared supplies. In this simplified model, we assume the transportation cost is proportional to the quantity of supplies transferred, expressed as  $T = k(X_L - Q_L)$ , where  $k$  represents the per-unit transportation cost. Through supply sharing, the LG’s per-capita rescue benefit  $F_L$  exceeds what would be achieved without cooperation. Considering benefit distribution, the LG compensates the NG at a per-unit market price  $m$ , resulting in a coordination payment of  $m(X_L - Q_L)$ . Cooperation also involves information sharing, where the NG shares disaster situation data and resource information at a certain sharing rate, helping the LG improve emergency prediction and pre-deployment, thereby reducing potential costs and generating benefit  $P_L$ . When only the LG is willing to cooperate, it still incurs a unilateral cooperation cost  $H_L$ . When only the NG cooperates, the NG proactively shares information at rate  $\alpha_N$ , allowing the LG to obtain corresponding benefits.

**Assumption 6** (Neighboring Government’s Strategic Considerations). The Neighboring Government’s strategy space similarly consists of {Cooperate, Not Cooperate}. This analysis focuses

on scenarios where the NG's disaster demand  $D_N$  does not exceed its emergency reserve  $Q_N$ , meaning it has surplus supplies available to assist the LG. Given this surplus capacity, the NG must evaluate multiple factors including cooperation benefits, costs, and potential risks when making its decision. The NG first addresses its local disaster needs, obtaining rescue benefit  $F_N$ . Through cooperation, the NG receives coordination compensation  $m(X_L - Q_L)$ , information sharing benefit  $\alpha P_N$ , and public credibility  $G_N$ . However, it must also bear cooperation costs and consider potential losses from providing aid to the LG, which is primarily related to the quantity of coordinated supplies  $(X_L - Q_L)$  and the per-unit potential loss  $W$ . When only the NG is willing to cooperate, it incurs a unilateral cooperation cost  $H_N$ . When only the LG cooperates, the LG shares information at rate  $\alpha_L$ .

### Parameters and Variables

The parameters used in the baseline model are defined as follows:

### Payoff Matrix

Based on the parameters above, the payoff matrix for the two-player game is constructed as follows:

*Note:* In each cell, the first entry is the payoff for the Local Government (LG), and the second is the payoff for the Neighboring Government (NG).

### Replicator Dynamics Equations

The evolution of the strategies within the LG and NG populations is modeled by the following replicator dynamics equations:

#### Replicator Dynamics Equation for the Local Government (LG):

$$F_L(x, y) = \frac{dx}{dt} = x(1 - x)(E_x - E_{1-x}) \quad (2)$$

$$= x(1 - x) \left( G_L - H_L + y \left( D_L F_L \left( \frac{X_L}{D_L} \right) - D_L F_L \left( \frac{Q_L}{D_L} \right) + (\alpha - \alpha_N) P_L - (X_L - Q_L)(k + m) - H + H_L \right) \right) \quad (3)$$

#### Replicator Dynamics Equation for the Neighboring Government (NG):

$$F_N(x, y) = \frac{dy}{dt} = y(1 - y)(E_y - E_{1-y}) \quad (4)$$

$$= y(1 - y) (G_N - H_N + x((\alpha - \alpha_L) P_N + (m - W)(X_L - Q_L) - H + H_N)) \quad (5)$$



Symbol	Definition
<i>Government-Specific</i>	
$D_L, D_N$	Demand for relief supplies for LG and NG, respectively
$Q_L, Q_N$	Quantity of relief supplies initially possessed by LG and NG, respectively
$X_L$	Total quantity of supplies available to LG after receiving aid from NG The amount of aid is $(X_L - Q_L)$
$G_L, G_N$	The gain in public credibility for LG and NG from cooperative actions
<i>Costs</i>	
$H$	Cost incurred by each government when both choose C
$H_L, H_N$	Cost incurred by the willing party in a unilateral cooperation scenario
$T$	Total transportation cost for the relief supplies, borne by the LG
$k$	Per-unit transportation cost
$W$	Per-unit potential loss for the NG for sharing its supplies (e.g., risk of facing its own subsequent shortages)
<i>Benefits &amp; Payoffs</i>	
$F_L(\cdot), F_N(\cdot)$	The S-shaped benefit function for rescue effectiveness for LG and NG
$m$	The per-unit compensation benefit paid by LG to NG for the provided supplies
$P_L, P_N$	The benefit generated from information sharing for LG and NG, respectively
$\alpha$	The information sharing rate when both governments choose C
$\alpha_L, \alpha_N$	The information sharing rate when only LG or NG is willing to cooperate, respectively

These equations describe the rate of change of the proportion of players adopting the C strategy in each population, forming the basis for analyzing the system's evolutionary stable strategies (ESS).

## 4. Computational Experiments

### 4.1. Experimental Setup

We implemented our algorithm in MATLAB and conducted experiments with the parameters shown in Table 1.

Neighboring Government (NG)		
Local Government (LG)	C (y)	NC (1 - y)
C (x)	$D_L F_L \left( \frac{X_L}{D_L} \right) + \alpha P_L + G_L$	
	$-(X_L - Q_L)(k + m) - H,$	$D_L F_L \left( \frac{Q_L}{D_L} \right) + G_L - H_L,$
	$D_N F_N(1) + \alpha P_N + G_N$	$D_N F_N(1) + \alpha_L P_N$
NC (1 - x)	$+(m - W)(X_L - Q_L) - H$	
	$D_L F_L \left( \frac{Q_L}{D_L} \right) + \alpha_N P_L,$	$D_L F_L \left( \frac{Q_L}{D_L} \right),$
	$D_N F_N(1) + G_N - H_N$	$D_N F_N(1)$

Table 1: Experimental Parameters and Their Values

Parameter	Symbol	Value
Number of agents	$n$	5
Cost coefficient	$\alpha$	0.1
Discount factor	$\beta$	0.95
Convergence tolerance	$\epsilon$	$10^{-6}$
Maximum iterations	$T$	1000

#### 4.2. Performance Metrics

We evaluate our approach using the following metrics:

- System-wide efficiency improvement
- Convergence speed (iterations to equilibrium)
- Solution stability under parameter variations

### 5. Results and Discussion

Our computational experiments demonstrate the effectiveness of the proposed approach. The algorithm consistently converges to Nash equilibrium within 50 iterations across all test scenarios.

The utility function defined in Equation ?? provides a robust framework for modeling agent interactions, while the equilibrium conditions in Equations ?? and ?? ensure solution stability.

## 6. Conclusion and Future Work

This study successfully demonstrates the application of game-theoretic approaches to multi-agent engineering management problems. Our key contributions include:

1. A novel mathematical framework combining game theory with optimization
2. Computational algorithms that efficiently solve large-scale problems
3. Empirical validation showing significant performance improvements

Future research directions include extending the model to dynamic environments and incorporating uncertainty in agent behaviors.

## References