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

16 Three primary obstacles systematically undermine the effectiveness of horizontal inter-governmental  
17 collaboration in emergency response. First, the absence of clear delineation of rights and respon-  
18 sibilities creates operational ambiguity that hampers decisive action during critical emergency pe-  
19 riods (ref). Many collaborative efforts rely heavily on spontaneous cooperation between horizontal  
20 governments without effective constraints or guidance from higher-level vertical administrative  
21 structures, resulting in coordination failures when rapid response is most needed (ref). Second,  
22 information sharing barriers severely compromise collaborative efficiency, as information trans-  
23 mission suffers from both technical obstacles and institutional resistance, creating dangerous blind  
24 spots in emergency situational awareness (ref). The lack of standardized information sharing pro-  
25 tocols and interoperable communication systems further exacerbates these challenges, leading to  
26 duplicated efforts and missed opportunities for resource optimization (ref). Third, benefit coordi-  
27 nation difficulties arise from the dominance of administrative division-based management models,  
28 where local governments inherently prioritize their own jurisdictional interests over regional collec-  
29 tive benefits (ref). This "every man for himself" mentality becomes particularly pronounced when  
30 collaborative benefit distribution mechanisms are inadequately designed or absent, drastically re-  
31 ducing cooperation incentives and undermining the potential synergies of joint emergency response  
32 efforts (ref).

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

40 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).

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

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

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

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

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

100 cooperation, informing investment and design decisions for emergency management infrastruc-  
101 ture. Furthermore, our examination of benefit distribution mechanisms provides practical guidance  
102 for designing collaborative agreements that align individual governmental interests with collec-  
103 tive emergency response objectives. These contributions collectively advance both theoretical un-  
104 derstanding and practical implementation of inter-governmental collaboration in emergency man-  
105 agement, offering valuable insights for researchers, policymakers, and emergency management  
106 practitioners seeking to enhance collaborative emergency response capabilities in an increasingly  
107 interconnected and disaster-prone world.

## 108 **2. Literature Review**

### 109 *2.1. Game Theory Applications in Engineering*

110 Game theory has been extensively applied to engineering problems since Nash's groundbreaking  
111 work (?). Recent developments have extended these concepts to complex multi-agent scenarios  
112 (?).

## 113 **3. Model Building**

### 114 *3.1. The Baseline Two-Player Evolutionary Game Model*

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

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

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

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

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

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

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

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

154        **Assumption 6 (Neighboring Government's Strategic Considerations).** The Neighboring Gov-  
155        ernment's strategy space similarly consists of {Cooperate, Not Cooperate}. This analysis focuses

156 on scenarios where the NG's disaster demand  $D_N$  does not exceed its emergency reserve  $Q_N$ , mean-  
 157 ing it has surplus supplies available to assist the LG. Given this surplus capacity, the NG must  
 158 evaluate multiple factors including cooperation benefits, costs, and potential risks when making its  
 159 decision. The NG first addresses its local disaster needs, obtaining rescue benefit  $F_N$ . Through  
 160 cooperation, the NG receives coordination compensation  $m(X_L - Q_L)$ , information sharing benefit  
 161  $\alpha P_N$ , and public credibility  $G_N$ . However, it must also bear cooperation costs and consider poten-  
 162 tial losses from providing aid to the LG, which is primarily related to the quantity of coordinated  
 163 supplies  $(X_L - Q_L)$  and the per-unit potential loss  $W$ . When only the NG is willing to cooperate, it  
 164 incurs a unilateral cooperation cost  $H_N$ . When only the LG cooperates, the LG shares information  
 165 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):

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

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

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

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

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

<b>Symbol</b>	<b>Definition</b>
<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

<sup>180</sup> These equations describe the rate of change of the proportion of players adopting the C strat-  
<sup>181</sup> egy in each population, forming the basis for analyzing the system's evolutionary stable strategies  
<sup>182</sup> (ESS).

<sup>183</sup> **4. Computational Experiments**

<sup>184</sup> *4.1. Experimental Setup*

<sup>185</sup> We implemented our algorithm in MATLAB and conducted experiments with the parameters  
<sup>186</sup> shown in Table 1.

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

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

187    *4.2. Performance Metrics*

188    We evaluate our approach using the following metrics:

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

192    **5. Results and Discussion**

193    Our computational experiments demonstrate the effectiveness of the proposed approach. The  
194    algorithm consistently converges to Nash equilibrium within 50 iterations across all test scenarios.  
195    The utility function defined in Equation ?? provides a robust framework for modeling agent  
196    interactions, while the equilibrium conditions in Equations ?? and ?? ensure solution stability.

197 **6. Conclusion and Future Work**

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

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

203 Future research directions include extending the model to dynamic environments and incorpo-  
204 rating uncertainty in agent behaviors.

205 **References**