

# 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

Construction enterprises, characterized by project-based production, capital intensity, and geographical dispersion (ref), recently face three aspects of pressures. First, traditional operational challenges have intensified, including escalating material costs and acute skilled labor shortages (ref). Second, digital transformation imperatives require fundamental shifts from conventional project management to integrated digital ecosystems, demanding substantial investments in BIM, AI, and IoT technologies across fragmented project networks (ref). Third, sustainability mandates pressure enterprises to adopt sustainable operations while maintaining strict project timelines and budget constraints (ref). These challenges exhibit compounding effects: labor shortages acceler-

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10 ate automation needs (ref), sustainability requirements demand both technological upgrades and  
11 workforce retraining, while digital transformation costs strain already tight project margins. Col-  
12 lectively, these compounding pressures are not merely cyclical downturns but signify a structural  
13 transformation of the construction industry. They create a "new normal" characterized by per-  
14 sistent volatility, systemic uncertainty, and heightened complexity, fundamentally threatening the  
15 long-term viability of enterprises reliant on conventional business models.

16 In this context of structural transformation, developing sustainable competitive advantages be-  
17 comes critical for construction enterprises' survival and growth. Traditional competitive advan-  
18 tages, often based on temporary market opportunities or replicable efficiencies, are insufficient to  
19 withstand such persistent and systemic pressures. Sustainable competitive advantage (SCA), in  
20 contrast, denotes an enterprise's capacity to sustain superior performance relative to competitors  
21 through the deployment of unique resources and capabilities that exhibit temporal persistence and  
22 inimitability (ref). Characterized by deep embeddedness in organizational capabilities, path de-  
23 pendence, and resistance to imitation (ref), SCA has become the core strategic resource enabling  
24 construction enterprises to navigate structural industry transformations (e.g., green and digital tran-  
25 sitions) and cyclical market shocks (e.g., financial crises or pandemics) (ref). Consequently, con-  
26 struction enterprises must construct sustainable competitive advantages; failure to do so will expose  
27 them to severe risk of marginalization amid intensifying market competition.

28 Existing Architecture, Engineering, and Construction (AEC) literature has examined construc-  
29 tion enterprises' competitive strategies from perspectives including political relationships, social  
30 responsibility, Enterprise Resource Planning (ERP), digital transformation, and blockchain (ref).  
31 However, existing research exhibits two significant limitations.

32 First, the majority of AEC literature focuses predominantly on traditional competitive advan-  
33 tages rather than sustainable competitive advantages, yet these concepts differ fundamentally. Tra-  
34 ditional competitive advantages typically stem from external market opportunities(ref) or tempo-  
35 rary resource allocations (ref). This exhibits replicability and temporal constraints that render them  
36 susceptible to displacement through competitor imitation or technological catch-up. They primarily  
37 emphasize static resource allocation efficiency. Rather, sustainable competitive advantages origi-  
38 nate from enterprises' internal unique resource bundles and capability systems, possessing VRIN  
39 characteristics—valuable, rare, inimitable, and non-substitutable—enabling enterprises to gener-

40 ate superior profits over extended periods while resisting competitive pressures(ref). Moreover,  
41 sustainable competitive advantages emphasize the construction and evolution of dynamic capabil-  
42 ities—enterprises' capacity to sense environmental changes, seize market opportunities, and re-  
43 configure resources. This fundamental distinction means that insights from traditional competitive  
44 advantage research may inadequately guide construction enterprises seeking long-term resilience  
45 and performance.

46 Second, previous studies have predominantly analyzed the impact of individual factors on com-  
47 petitive advantage. This literature fails to capture the complex, systemic nature of success in the  
48 construction industry. Competitive advantage in this sector rarely stems from a single "silver  
49 bullet"; rather, it emerges from the synergistic interplay among an enterprise's strategy, its envi-  
50 ronment, and its internal resources (ref). This reality exposes the limitations of traditional linear  
51 models and necessitates a theoretical lens that can explain how different combinations of conditions  
52 lead to superior performance. The configurational school of strategic management provides such  
53 a lens, positing that an enterprise's advantage derives from the holistic fit of these multidimen-  
54 sional elements (ref). By examining these combinations, or configurations, it becomes possible  
55 to understand why enterprises with similar resources in the same environment can exhibit vastly  
56 different outcomes. Therefore, this study adopts a configurational perspective to ask its first key  
57 question: From the multi-dimensional perspective of strategy, environment, and resources, what  
58 are the effective configurations for building a sustainable competitive advantage in the construction  
59 industry?

60 While sustainable competitive advantage determines construction enterprises' market position  
61 and profitability, organizational resilience ensures their survival and stability in turbulent environ-  
62 ments. Organizational resilience refers to an enterprise's potential ability to anticipate, avoid, and  
63 respond to environmental shocks (especially crisis events) (ref), with its core characteristic be-  
64 ing the stability and reliability of maintaining competitive advantage and business performance  
65 under crisis impacts. Unlike competitive advantage which focuses on exceptional operational per-  
66 formance, organizational resilience focuses on stable operational performance. Existing research  
67 demonstrates that organizationally resilient construction enterprises can better absorb shocks, main-  
68 tain operational continuity, and recover more quickly from disruptions. For example, during the  
69 COVID-19 pandemic, construction enterprises with strong organizational resilience maintained

70 project continuity through flexible resource reallocation and adaptive management practices, while  
71 less resilient competitors faced project suspensions and financial distress (ref). This divergence  
72 in crisis responses underscores the critical importance of organizational resilience in sustaining  
73 competitive advantage over time.

74 Existing research largely focuses separately on the singular dimensions of either organiza-  
75 tional resilience or competitive advantage. For instance, some scholars concentrated on functional  
76 resilience rather than systemic resilience (ref); other scholars provided organizational resilience  
77 frameworks (ref). However, existing research largely focuses on either organizational resilience or  
78 competitive advantage as separate dimensions. The crucial interplay between them, particularly  
79 how resilience contributes to the sustainability of an advantage, remains underexplored. A truly  
80 sustainable advantage must be a resilient one; otherwise, it is merely a temporary high-performance  
81 state vulnerable to the next crisis.

82 The Resource-Based View (RBV) provides a theoretical foundation for understanding this re-  
83 lationship, positing that enterprises' sustained competitive advantage stems primarily from inter-  
84 nal rather than external factors. From this perspective, organizational resilience capability can be  
85 viewed as a strategic internal resource that enhances sustainable competitive advantage. Organiza-  
86 tional resilience enables enterprises to withstand pressure, continuously innovate, and rapidly adapt  
87 to changes, thereby protecting and reinforcing their competitive advantages during environmental  
88 turbulence. In the context of construction enterprises facing increasingly complex and volatile  
89 market environments characterized by supply chain disruptions, regulatory changes, and economic  
90 fluctuations, the interplay between organizational resilience and competitive advantage becomes  
91 particularly critical. Solely pursuing high competitive advantage without resilience leaves enter-  
92 prises vulnerable to crisis-induced performance collapse, while focusing exclusively on resilience  
93 without competitive advantage results in stable mediocrity. Construction enterprises must there-  
94 fore develop strategic configurations that simultaneously cultivate both organizational resilience  
95 and sustainable competitive advantage—enabling them to not only excel in stable periods but also  
96 maintain stability during crisis events and capitalize on post-crisis opportunities. Therefore, this  
97 study presents its second research question: How can enterprises simultaneously maintain organi-  
98 zational resilience while achieving high competitive advantage?

99 To address these two research questions, this study employs Time-Series Qualitative Com-

100 parative Analysis (TSQCA) (ref). This configurational method was selected because traditional  
101 regression analysis, which focuses on the net effects of isolated variables, is ill-suited to explain  
102 how competitive advantage and resilience emerge from the complex, synergistic interplay of an  
103 enterprise's strategy, environment, and resources. TSQCA, in contrast, is designed to identify the  
104 multiple, distinct combinations of conditions that produce an outcome. By embracing the prin-  
105 ciples of conjunctural causation (outcomes stem from combinations of factors) and equifinality  
106 (multiple pathways lead to success), this approach provides a powerful tool for uncovering the dif-  
107 ferent "recipes for success." In this study, TSQCA is used to determine which configurations of  
108 strategic, environmental, and organizational conditions are sufficient for achieving high sustainable  
109 competitive advantage and, subsequently, which of those configurations also ensure organizational  
110 resilience.

111 On this basis, this paper collected 2,150 data points from 118 listed construction enterprises in  
112 China from 2014 to 2023. It analyzes nine conditions across three major dimensions—strategic ori-  
113 entation (diversification strategy, differentiation strategy, cost leadership strategy), environmental  
114 characteristics (environmental dynamism, environmental richness), and organizational resources  
115 (cost stickiness, organizational size, social responsibility, digital transformation)—to determine  
116 which strategic configurations help construction enterprises gain sustainable competitive advan-  
117 tages and which of these can also help enterprises obtain organizational resilience in response to  
118 external uncertainties.

## 119 **2. Literature Review**

### 120 *2.1. Game Theory Applications in Engineering*

121 Game theory has been extensively applied to engineering problems since Nash's groundbreak-  
122 ing work (?). Recent developments have extended these concepts to complex multi-agent scenarios  
123 (?).

## 124 **3. Model Building**

### 125 *3.1. The Baseline Two-Player Evolutionary Game Model*

126 We first establish a baseline evolutionary game model to analyze the strategic interactions be-  
127 tween two horizontal governments during a disaster emergency. The model focuses on the decision-

128 making process regarding cooperation on resource sharing, which includes both relief supplies and  
129 critical information. It assumes that the governments are boundedly rational and dynamically ad-  
130 just their strategies based on the payoffs from previous interactions. This baseline game model does  
131 not include a higher-level (vertical) government. The baseline model is built upon the following  
132 key assumptions:

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

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

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

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

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

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

149 **Assumption 5 (Local Government's Strategic Considerations).** When choosing to cooperate  
150 with the Neighboring Government, the LG can obtain additional relief supplies through regional  
151 coordination. When both LG and NG actively cooperate, both governments incur a cooperation  
152 cost  $H$ , and the LG gains public credibility  $G_L$  for its collaborative efforts. According to the In-  
153 terim Measures for the Management of Central Emergency and Disaster Relief Material Reserves  
154 (ref), following the principle of "user pays," the LG bears the transportation cost for the shared  
155 supplies. In this simplified model, we assume the transportation cost is proportional to the quantity  
156 of supplies transferred, expressed as  $T = k(X_L - Q_L)$ , where  $k$  represents the per-unit transporta-

tion 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:

| 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<br>(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                              |

187

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

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

188

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

189

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

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

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

190 
$$= y(1 - y)(G_N - H_N + x((\alpha - \alpha_L)P_N + (m - W)(X_L - Q_L) - H + H_N)) \quad (5)$$

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

194 **4. Computational Experiments**

195 *4.1. Experimental Setup*

196 We implemented our algorithm in MATLAB and conducted experiments with the parameters  
 197 shown in Table 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      |

198 *4.2. Performance Metrics*

199 We evaluate our approach using the following metrics:

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

203 **5. Results and Discussion**

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

208 **6. Conclusion and Future Work**

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

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

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

216 **References**