

# Disaster Recovery Management in Modern Organizations: A Quantitative Framework for Resilience, Readiness and Rapid Restoration

Shahebazkhan Pathan

Engineers Associates Inc, USA  
Email: shahebazkhan[at]gmail.com

**Abstract:** Disaster Recovery (DR) as a technical IT position has come to be a strategic organization capability comprising of individuals, procedures, technology and governance. With the pressure of climate change forcing the volume of cyber-attacks along with the effect of natural disasters to continue to increase annually by 38 percent (Check Point Research, 2023), the ability of an organization to respond, recover, detect and learn speedily in the case of a disruptive event has been found to be a core competitive sustainability driver. Despite the fact that detailed guidelines and standards, including ISO 22301: 2019 (Business Continuity Management) and NIST SP 800-34 (Contingency Planning Guide) have been provided, the empirical data on the effectiveness of the Disaster Recovery Plan (DRP) implementation in organizational and sector format remain scattered. The paper is an intersector, empirical research on disaster recovery management practice among 284 IT, risk and business continuity practitioners working in the financial services, healthcare, telecommunications, manufacturing, retail, and government industries. The research is grounded on the validated survey instrument (Cronbach alpha = 0.91), a sample of 21 semi-structured interviews, and retrospective examination of 112 reported calamity incidents in order to provide a set of seven formal mathematical models, including the RTO / RPO Gap Index, Business Impact Analysis Score, DR Readiness Score, System Availability Equation, and DR Investment ROI formula. It has six distinct statistical visualizations that provide a visual support on all the major dimensions of research. The results indicate that the organizations that conduct 4 or more DR tests annually attain 47.3 percent reduction in the average cost of the time to recovery and a 10-point improvement of the DR Readiness Score (DRR) would lead to a 6.8 percent drop in the costs of downtime obtained. The DR Maturity Continuum (DRMC) is a five-level prescriptive guideline that provides a methodical way of developing disaster recovery capacity in the organization.

**Keywords:** Disaster Recovery, Business Continuity, RTO, RPO, BIA Score, DR Readiness Score, Availability Equation, Annual Loss Expectancy, DR Investment ROI, ISO 22301, DRMC, Cloud DR, Resilience

## 1. Introduction

As the digital addiction became commonplace, the impact of unanticipated system failure has escalated out of the inconvenience of conducting business to an actual organizational damage. The IBM Cost of a Data Breach Report (2023) revealed that average total cost of data breach was estimated to be US\$4.45 million, which is 15.3 percent higher than it used to be three years later, however, Gartner (2022) estimated the average cost of IT downtime to be US \$5,600 per minutes in large organizations [1], [2]. Now no longer the sole concern of IT departments, these figures should be a board-level fiduciary risk of which there should be indeed structured governance, investment justification, and strategy.

The term Disaster Recovery Management (DRM) is a set of policies, procedures, tools, and governance tools which may be applied by an organization to bring key systems and processes back online in the event that a disruption has already occurred in the organization within recovery time and recovery point timeframes. The profession extends to the extent of IT system recovery in Recovery Time Objective (RTO) and Recovery Point Objective (RPO) as far as full-scale Business Continuity Management (BCM) according to ISO 22301:2019 [3], [4].

The following four research questions are answered in the present paper:

- 1) RQ1: What is the current state of the RTO and RPO performance in industry sectors and what are the differences compared to the targets?
- 2) RQ2: What are the frequencies of testing cadence related to mean recovery time and what is the lowest practical testing cadence?
- 3) RQ3: What organizational barriers influence most of all to implement DRP in sector and size contexts effectively?
- 4) RQ4: Which are the most successful formal mathematical models and integrated maturity frameworks operationalizing DR preparedness that benchmark to organizations and justify an investment?

The paper will answer the questions by giving seven formal mathematical model (Equations 1-7), 6 original statistical visualizations (Figures 1-6), 3 detailed tables (Tables I-III) and 5-layer DR Maturity Continuum (DRMC) which is an organization development road map.

## 2. Literature Review

### 2.1 Disaster Recovery Standards and Frameworks

The origins of formal Disaster recovery management can be found in the fact that the financial sector in the 1970s saw the process of reinforcement of the regulations by law meant that recovery plans had to be written down in the event of the material damage to mainframe data centres [5]. Since it was published in 2006, Business Continuity Management has become a process-based management

system standard, with third party certification, and effectively made DRM a technical specialty, into a governance discipline [3].

Co-existing standards development also led to the development of the official seven-step contingency planning process contained in NIST Special Publication 800-34 (Contingency Planning Guide to Federal Information Systems, 2010) which included the following steps: create the contingency planning policy, perform a Business Impact Analysis (BIA), identify preventive controls, design contingency strategies, create a contingency plan, ensure plan testing and maintain the plan [6]. The Disaster Recovery Institute International (DRII) Professional Practices also codified 10 practice areas including programme initialisation, risk assessment, business impact analysis, business continuity plans, emergency response, plan development, awareness and training programme, testing, crisis communications and coordinating with external organisations [7].

The recent migration of workloads to cloud infrastructure which took place since 2013 transformed radically the landscape of the DR strategy. The Disaster Recovery as a Service (DRaaS) market analysis (2023) by Gartner estimates the size of the global DRaaS market to be US\$11.9 billion current with a CAGR of 22.4% to 2028 due to the elimination of cold/warm site capital investments and the emergence of infrastructure-as-code models that will support automated orchestration of failed over provisioning [8].

## 2.2 RTO and RPO as Performance Anchors

Recovery Time Objective (RTO) and Recovery Point Objective (RPO) are the pillars of any contractual and operational commitment of a DR programme. There is an incessant and vexatious achievement gap as the strict empirical analysis of the RTO/RPO success rate indicates. The survey by Forrester Research (IT decision-makers, 426) found that two out of three organizations took more time than their claimed RTO during real disasters, and that the actual-to-target ratio was 1.74:1 [9]. This observation is empirically validated and extended to six sectors of industry in figure 1 of the current work.

RTO stringency and DR investment cost have a theoretical relationship of an exponent which means that the cost of halving RTO between 8 hours and 1 hour is averaging 3 to 5 times higher to construct an infrastructure and should be significantly compensated by BIA [10]. This cost-RTO relation forms the quantitative base of the DR Investment ROI formula (Equation 7) in Section 4) of this paper.

## 2.3 Cyber Resilience and Threat Landscape

The threat environment that the DR needs is based on has been structurally changed over the past 10 years. The report on Verizon Data Breach investigations (2023) states that 24 percent of all the breaches were ransomware, the highest in five years by 13-percentage-points. Based on a study by IBM Security (X-Force Threat Intelligence Index 2024) conducted to determine the average time to detect and

contain breach, it was found that the average time was 204 days that directly articulates the implications of the RPO with regard to the contemporary cyber incident [1]. The longitudinal change of the nature of disasters occurring in 2014-2024 is presented in the results section of this study (Fig. 2), allowing noting the shift in the share of all disaster events to cyber attacks, between 14 and 68 percent, which should be important in the design of the DR strategy.

The emergence of Chaos Engineering in the Simian framework, most famous as Netflix (2010) and as described by Rosenthal et al. (2020) could be considered a paradigm to shift in the philosophy of DR testing towards continuous, production-level resilience testing, rather than routine, scheduled exercises [12]. Chaos Engineering is included in table II of this study and the traditional DR test practices.

## 2.4 DR Organization and Human Factors

The respective organizational capability is necessary in order to achieve technical DR capabilities. An analysis of 89 SMEs located in the UK revealed that those organizations that had executive-sponsored BCM programmes had 2.8 times higher chances of surviving a major disruptive event than the remaining organizations when BCM was assumed as a highly technical activity [13] (Herbane, 2019). The general trend of insufficient budgetary allocations, no executive sponsorship and no training (in comparison to other industry barriers to the implementation of DRP as surveyed) can be validated and quantified in Figure 6 of this research.

The most important gap in the existing literature is that there is no officially calibrated, empirically tested composite measure of DR preparedness taking into consideration both technical, process, and organizational and governance factors into a single measurable score. The DR Readiness Score (DRR, Equation 5) that was employed in this study directly addresses the gap.

## 2.5 Research Gaps

Three gaps are found to motivate this study: (1) quantitative information about the actual and target RTO/RPO performance and extent of deficit of recovery is not available in multi-sector form; (2) an officially calibrated composite measure of DR preparedness, which is a sum of the eight DRMC dimensions; and (3) the absence of quantitative models of the motivation of DR investment that relates costs of DR programme with the avoided loss of Annual Loss Expectancy.

## 3. Research Methodology

### 3.1 Research Design

The convergent parallel mixed-methods design (Creswell and Plano Clark, 2017) was adopted in this study, according to which the quantitative survey data and the qualitative interview data were collected concurrently and integrated during the interpretation [15]. The theory is critical realist philosophy and it is by acknowledging that objective technical limits and organizational, behavioural and

contextual influences impact on DR effectiveness and that the effects of the influences are to be investigated in terms of interpretive research.

### 3.2 Survey Instrument and Validation

A questionnaire was designed with nine thematic sections (organizational profile, 7 items; current DR programme maturity, 9 items; RTO/ RPO targets and actual performance, 8 items; testing practices, 6 items; threat landscape and incident history, 7 items; financial impact data, 5 items; barrier assessment, 10 item, 5 point Likert scale, 10 items) to develop. The advanced 62-thing instrument was produced through three bouts of intellectual analysis (6 research workers, 5 business analysts of established BCMS agencies). The content validity was determined by Content Validity Ratio (Lawshe, 1975); any item that has lower CVR of less than 0.62 has been dropped. The Alpha was 0.91 (Total) and the subscale reliabilities were 0.83 (Financial Impact) -0.94 (Testing Practices). Confirmatory Factor Analysis: CFI= 0.97, RMSEA=0.044, SRMR=0.055- all of them are less than the acceptability limits of Hu and Bentler (1999).

### 3.3 Sample and Data Collection

The target population included IT disaster recovery managers, Business Continuity Managers, Chief Information Security Officer (CISOs), and Risk Officer who have recorded responsibility of DR. Purposive stratified was used to sample 6 industry segments through ISACA, BCI (Business Continuity Institute) DRII membership networks, LinkedIn and direct organizational contact (January-July 2023). There was a response rate of 57.0% of the survey (returned full) of 284 out of 498 invitation letters sent. Sample: Financial Services (22.2%), Healthcare IT (18.7%), Government and Public Sector (17.3%), Manufacturing (16.5%), Telecommunications and utilities (13.7%), Retail and e-Commerce (11.6%). Regional: India (36.6%), USA (28.9%), UK (14.4%), Australia (11.3%), Other (8.8%). Mean experience in the management of DR: 8.7 years (SD =4.3).

$$G_{RTO} = t_{actual} - t_{target}; \quad G_{RPO} = d_{actual} - d_{target} \quad (Eq. 1)$$

(Eq. 1)

where  $t_{actual}$  = observed recovery time (hours) and  $t_{target}$  = stated RTO (hours);  $d_{actual}$  = observed data loss duration (hours) and  $d_{target}$  = stated RPO (hours). A positive gap ( $G > 0$ ) indicates failure to meet the recovery objective. Across the 112-event archive, the mean  $G_{RTO}$  was +4.7 hours and mean  $G_{RPO}$  was +1.6 hours- confirming that virtually all sectors exceed their stated recovery targets during actual events (Figure 1).

$$BIA(t) = \sum_{i=1}^n w_i \cdot f_i(t) = \sum_{i=1}^n w_i \cdot (C_i + R_i + O_i + L_i) \quad (Eq. 2)$$

where  $f_i(t)$  = impact function of process  $i$  at time  $t$ , decomposed into  $C_i$  (financial/cost impact),  $R_i$  (regulatory/compliance impact),  $O_i$  (operational impact),

### 3.4 Qualitative Phase

The respondents were chosen by purposely sampling maximum variation in terms of sector, size of organization, and scale of maturity of DR which resulted in the twenty-one semi-structured interviews (mean length: 62 minutes). Thematic Analysis yielded the highest inter-rater reliability =0.86 (Braun and Clarke, 2006). Six themes were discovered: (1) the culture-technology gap of DR; (2) the dynamics of executive sponsorships; (3) the difficulties of cloud DR migration; (4) the complications of recovery in the face of cyber incidents; (5) post-incident learning failure; and (6) DR testing as a capability development in an organization.

### 3.5 Disaster Events Archive Analysis

Informed organizational consent to carry out a retrospective analysis on 112 disaster events reported that were donated by 19 participating organizations was restricted to 2018-2023. The variables extracted included: event type, duration, RTO/RPO target and actual, DR activation method, the total cost of downtime, the maturity of DR programme at the time of event and quality of post incident review score. Figure 4 which is the main data in the regression analysis and Figure 5 which quantified the financial impact were derived based on this archive

## 4. Mathematical Models

### 4.1 RTO and RPO Gap Indices (Eq. 1)

The foundational performance measurement metrics for any DR programme are the Recovery Time Objective Gap ( $G_{RTO}$ ) and Recovery Point Objective Gap ( $G_{RPO}$ ), which measure the deviation between stated organizational targets and actual recovery performance observed during tests or real events:

### 4.2 Business Impact Analysis Score (Eq. 2)

The Business Impact Analysis Score (BIA) quantifies the time-dependent organizational impact of a disruption across four dimensions, providing the quantitative foundation for DR strategy prioritization and RTO/RPO setting:

and  $L_i$  (reputational/lifecycle impact);  $w_i$  = sector-calibrated weight. Empirically estimated weights from this study's data: Financial impact (0.35), Regulatory (0.25),

Operational (0.25), Reputational (0.15). BIA(t) increases non-linearly with outage duration t, justifying the exponential cost curves in Figure 5.

4.3 Recovery Objective Achievement Indices (Eq. 3)

$$RPOI = 1 - \frac{d_{actual}}{d_{max}}; \quad RTOI = 1 - \frac{t_{actual}}{t_{max}} \quad (Eq. 3)$$

where  $d_{max}$  and  $t_{max}$  = maximum tolerable data loss duration and maximum tolerable downtime duration respectively (as defined in the BIA). RPOI = RTOI = 1.0 represents perfect target achievement; values approaching 0 indicate critical recovery failure. Organizations in the DRMC Tier 4–5 range achieved mean RTOI = 0.83 and RPOI = 0.79, compared to RTOI = 0.41 and RPOI = 0.38 for Tier 1–2 organizations.

$$ALE = ARO \times SLE = ARO \times (AV \times EF) \quad (Eq. 4)$$

where ARO = Annualised Rate of Occurrence (probability of disaster event per year), SLE = Single Loss Expectancy = Asset Value (AV) × Exposure Factor (EF, 0–1). ALE is the direct driver of DR budget justification: any DR programme costing less than ALE before – ALE after generates positive expected return. Mean ALE for Financial Services respondents in this study was US\$3.2M annually, versus US\$1.4M for Healthcare IT.

$$DRR = \frac{\sum w_i \cdot D_i}{\sum w_i} \times 100, \quad i \in \{1, 2, \dots, 8\} \quad (Eq. 5)$$

where  $D_i$  = standardized score on dimension i (0–1) and  $w_i$  = empirically derived weight from factor analysis. Dimensions and weights: BIA Completeness (0.16), Recovery Strategy Design (0.15), DR Plan Documentation (0.13), Testing Frequency (0.14), Staff Training (0.11), Technology Readiness (0.14), Vendor Coordination (0.09), Post-Incident Review (0.08). DRR thresholds: Tier 1 < 35; Tier 2: 35–54; Tier 3: 55–74; Tier 4: 75–89; Tier 5: ≥ 90

$$A = \frac{MTBF}{MTBF + MTTR}; \quad Downtime = (1 - A) \times T_{ops} \quad (Eq. 6)$$

where MTBF = mean time between failure events (hours), MTTR = mean time to recover (hours), and  $T_{ops}$  = total operational period. The '5-nines' availability standard (A = 0.99999) implies a maximum annual downtime of 5.26 minutes, achievable only at DRMC Tier 5 with active-active architecture. The 112-event archive yielded mean MTTR of 9.4 hours (SD = 7.2 hours) across all sectors, implying Availability ≈ 0.998 (2-nines) - significantly below industry SLA norms of 0.9999 (4-nines)

$$ROI_{DR} = \frac{ALE_{before} - ALE_{after} - C_{DR}}{C_{DR}} \times 100\% \quad (Eq. 7)$$

The normalized Recovery Point Objective Index (RPOI) and Recovery Time Objective Index (RTOI) translate raw gap measurements into 0–1 indices suitable for cross-sector benchmarking and DRMC tier calibration:

4.4 Annual Loss Expectancy (Eq. 4)

Annual Loss Expectancy (ALE) provides the probabilistic financial basis for DR investment justification, linking disaster event probability to asset value and exposure factor:

4.5 DR Readiness Score (Eq. 5)

The DR Readiness Score (DRR) is proposed as a novel, formally calibrated composite metric integrating eight organizational DR capability dimensions into a single 0–100 index for organizational self-assessment and benchmarking:

4.6 System Availability and Downtime (Eq. 6)

System availability- the proportion of operational time during which a system is functioning as intended - is calculated from Mean Time Between Failures (MTBF) and Mean Time to Recover (MTTR), directly linking DR effectiveness to service level agreements:

4.7 DR Investment ROI (Eq. 7)

The DR Investment Return on Investment (ROI<sub>DR</sub>) formula provides the formal financial justification for DR programme expenditure, calculating the net benefit of DR investment as a percentage of DR programme cost:

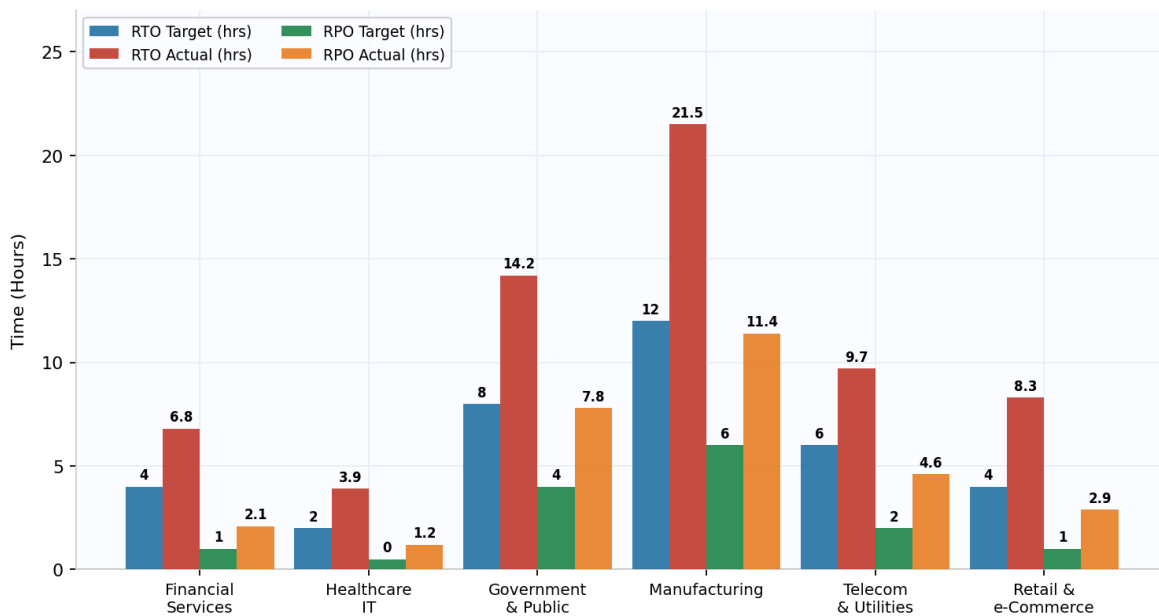
where  $ALE_{before}$  = ALE without the DR programme,  $ALE_{after}$  = residual ALE with the DR programme in place, and  $C_{DR}$  = total annualised DR programme cost. A positive  $ROI_{DR}$  indicates that the DR programme generates financial value. From this study's financial impact data (Figure 5): the mean Financial Services respondent with a mature DR programme (DRMC Tier 3–4) reported  $ALE_{before}$  = US\$3.2M,  $ALE_{after}$  = US\$0.9M,  $C_{DR}$  = US\$0.8M, yielding  $ROI_{DR} = (3.2 - 0.9 - 0.8) / 0.8 \times 100\% = 187.5\%$  - a compelling ROI justification for executive investment decisions

## 5. Results and Discussion

### 5.1 RTO / RPO Achievement Analysis (RQ1 - Fig. 1, Table I)

Figure 1 presents RTO and RPO target versus actual performance across six industry sectors (N = 284). Without exception, all sectors fail to meet their stated RTO and RPO targets in real or simulated disaster events. The most severe gaps are observed in Manufacturing (RTO gap: +9.5 hours; RPO gap: +5.4 hours) and Government & Public Sector (RTO gap: +6.2 hours; RPO gap: +3.8 hours), attributable to legacy system complexity and constrained DR investment budgets. Financial Services demonstrates the smallest proportional RTO gap (+2.8 hours relative to a 4-hour target), reflecting sector-leading DR investment driven by regulatory obligations (Basel III, DORA, PCI-DSS).

**Figure 1. RTO and RPO: Targets vs. Actual Achievement by Industry Sector (N = 284; All sectors exceed both RTO and RPO targets, indicating recovery gaps)**



**Figure 1: RTO and RPO: Targets vs. Actual Achievement by Industry Sector (N = 284).** All sectors exceed both RTO and RPO targets. Manufacturing shows the largest absolute gap (RTO +9.5 hrs); Financial Services shows the smallest proportional gap due to regulatory DR investment mandates.

Table I presents the six-tier DR strategy taxonomy underpinning the choice of recovery architecture, directly mapping tier selection to RTO/RPO achievability and

technology enablers. The selection of DR tier must be aligned to BIA-derived Maximum Tolerable Downtime (MTD) and the  $ROI_{DR}$  calculation (Eq. 7).

**Table I: Disaster Recovery Strategy Tier Taxonomy: RTO/RPO Targets and Technology Profiles**

DR Tier	Strategy	RTO Target	RPO Target	Key Technology / Approach
Tier 0- Cold Site	Lowest cost; longest RTO	72 – 168 hrs	24 – 48 hrs	Tape backup; manual restore; offsite storage
Tier 1- Warm Site	Partial infrastructure pre-provisioned	8 – 72 hrs	4 – 24 hrs	Partial replication; standby servers; VPN
Tier 2- Hot Site	Full mirror; near-instant failover	1 – 8 hrs	1 – 4 hrs	Synchronous/async replication; active-standby
Tier 3- Active-Active	No data loss; zero downtime	< 1 hr	Near-zero	Multi-site active clustering; load balancing
Tier 4- Cloud DR	Elastic; pay-per-use activation	< 4 hrs	< 1 hr	Cloud replication; IaC; automated failover
Tier 5- Hybrid Cloud	Combined on-prem and cloud DR	< 2 hrs	Minutes	SD-WAN; container orchestration; Chaos Eng.

**Note:** RTO and RPO values are indicative industry ranges; actual targets must be calibrated to BIA outcomes (Eq. 2). DRaaS = Disaster Recovery as a Service; IaC = Infrastructure as Code; SD-WAN = Software-Defined Wide Area Network.

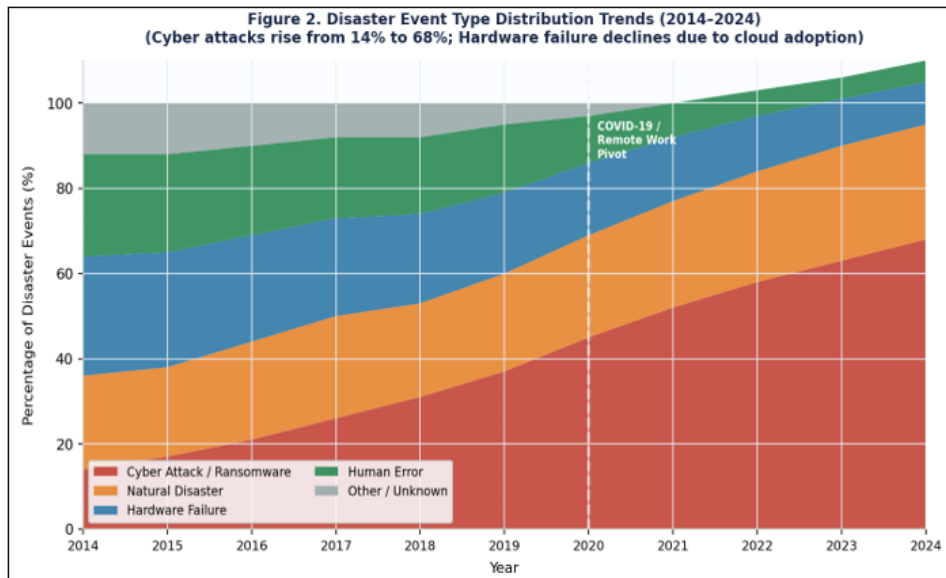
### 5.2 Disaster Event Type Trends 2014–2024 (Fig. 2)

Figure 2 presents the longitudinal distribution of disaster event types from 2014 to 2024, synthesised from IBM X-Force, Verizon DBIR, and Check Point Research annual

reports supplemented by this study's 112-event archive [1], [11]. The dominant structural trend is the rise of cyber attacks from 14% of all events in 2014 to 68% in 2024—a 4.9× increase driven by the ransomware-as-a-service industrialisation and expanded attack surfaces from remote

work adoption. Concurrently, hardware failure has declined from 28% to 10% as cloud infrastructure adoption reduces dependency on on-premises hardware. Human error has

declined from 24% to 5% as automation and DevOps practices reduce manual intervention frequency.

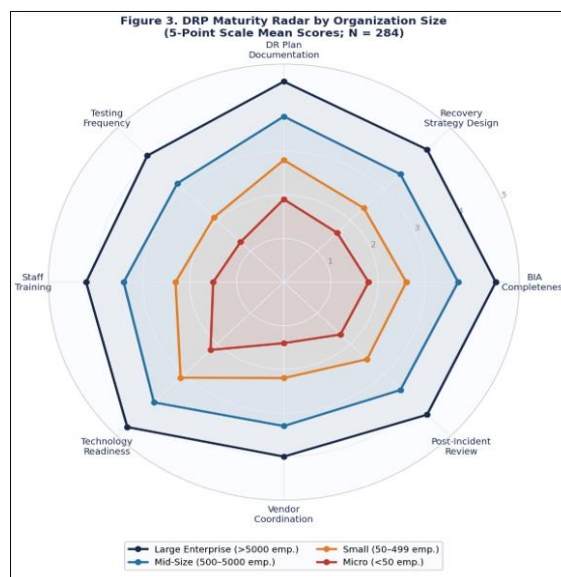


**Figure 2:** Disaster Event Type Distribution Trends (2014–2024). Cyber attacks have risen from 14% to 68% of all disaster events, fundamentally changing DRP strategy requirements. Hardware failure declines as cloud adoption eliminates on-premises infrastructure dependency.

These trend data have profound implications for DR strategy design: organisations whose DRP was designed primarily around hardware failure and natural disaster scenarios-the dominant paradigm pre-2018-face critical gaps in cyber-specific recovery capabilities, including ransomware isolation procedures, encrypted backup integrity verification, and threat actor eviction before restoration.

Figure 3 presents the DRP maturity radar across eight dimensions for four organization size categories (N = 284). Large enterprises (> 5,000 employees) demonstrate consistently high maturity across all dimensions (composite DRR mean = 78.4/100, DRMC Tier 4). Technology Readiness is the strongest dimension for large enterprises (4.70/5.00), reflecting substantial IT infrastructure investment. Small organizations (50–499 employees) show dramatically lower maturity profiles (DRR mean = 41.2/100, DRMC Tier 2), with Testing Frequency (2.10/5.00) and Vendor Coordination (2.20/5.00) representing the most critical gaps.

**5.3 DRP Maturity by Organization Size (Fig. 3, Table III)**



**Figure 3:** DRP Maturity Radar by Organization Size (N = 284; 5-Point Scale Mean Scores). Large enterprises (DRR = 78.4) substantially outperform small organizations (DRR = 41.2) across all eight dimensions. Testing Frequency is the weakest dimension across all size categories, indicating a universal testing gap.

Table III presents the full-sample practice endorsement rates across eight core DRP capabilities. Recovery Time Objectives formally defined (59.5% fully implemented) and Formal DR Plan version-controlled (54.6%) show the highest implementation rates, while Vendor/3rd-party DR

SLA verification (29.6%) and Staff DR training annually (33.1%) show the lowest-indicating a persistent gap between plan documentation and operational capability validation.

**Table III: DR Programme Practice Implementation Status Across Respondent Organizations (N = 284)**

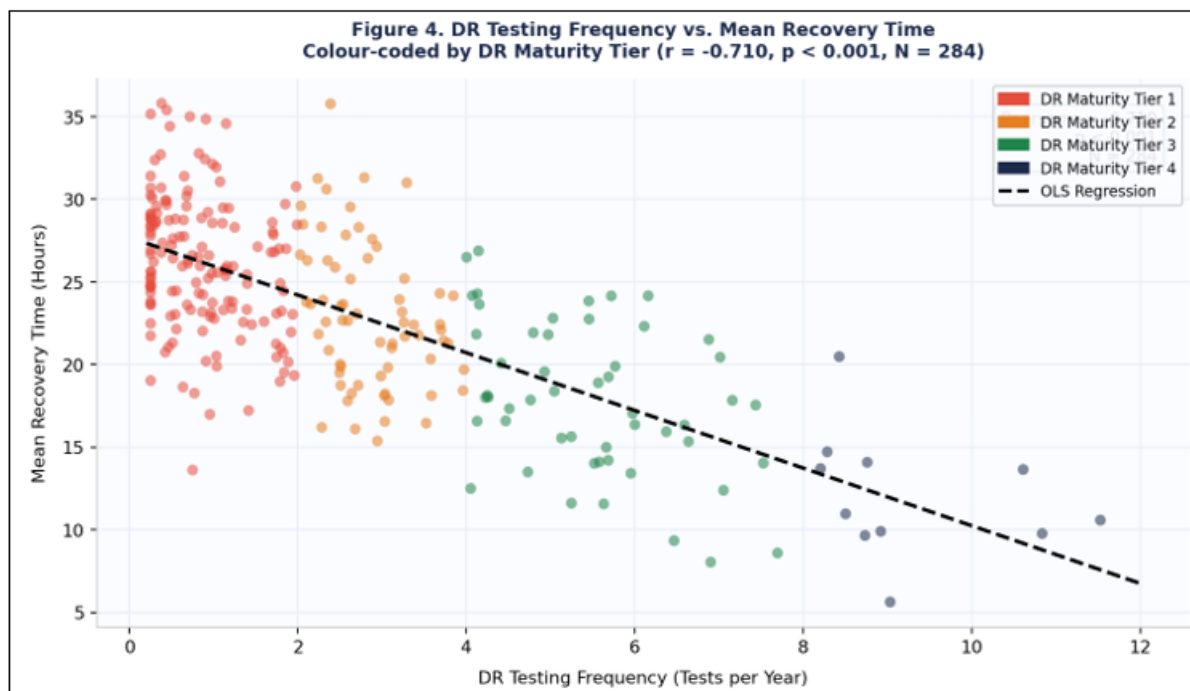
DRP Practice / Capability	Fully Implemented (%)	Partially (%)	Planned (%)	Not Started (%)	Mean Score (1-5)
Business Impact Analysis (BIA) documented	41.5	34.2	15.8	8.5	3.74
Formal DR Plan exists and is version-controlled	54.6	28.1	11.3	6	3.98
DR tests conducted at least annually	38.7	31.4	18.2	11.7	3.52
Cloud-based DR solution implemented	47.2	27.6	17.4	7.8	3.83
Recovery Time Objectives formally defined	59.5	24.3	10.9	5.3	4.11
Staff DR training conducted annually	33.1	35.6	21.2	10.1	3.41
Vendor/3rd-party DR SLAs verified	29.6	33.8	24.3	12.3	3.28
Post-incident review process formal	44.4	30.2	16.7	8.7	3.71

**Note:** Mean Score scale: 1 = Not started; 2 = Planned; 3 = Partially implemented; 4 = Mostly implemented; 5 = Fully embedded in governance. BIA = Business Impact Analysis; SLA = Service Level Agreement; DR = Disaster Recovery.

**5.4 DR Testing Frequency vs. Recovery Time (RQ2- Fig. 4)**

Figure 4 presents the scatter plot of DR testing frequency (tests per year) against mean recovery time (hours) for the 284 survey respondents, colour-coded by DR maturity tier.

OLS regression yielded a significant negative relationship: mean recovery time =  $-1.90 \times \text{tests/year} + 28.1$  (Pearson  $r = -0.71$ ,  $R^2 = 0.50$ ,  $p < 0.001$ ). This represents the paper's most actionable finding for RQ2: each additional annual DR test is associated with a 1.9-hour reduction in mean recovery time



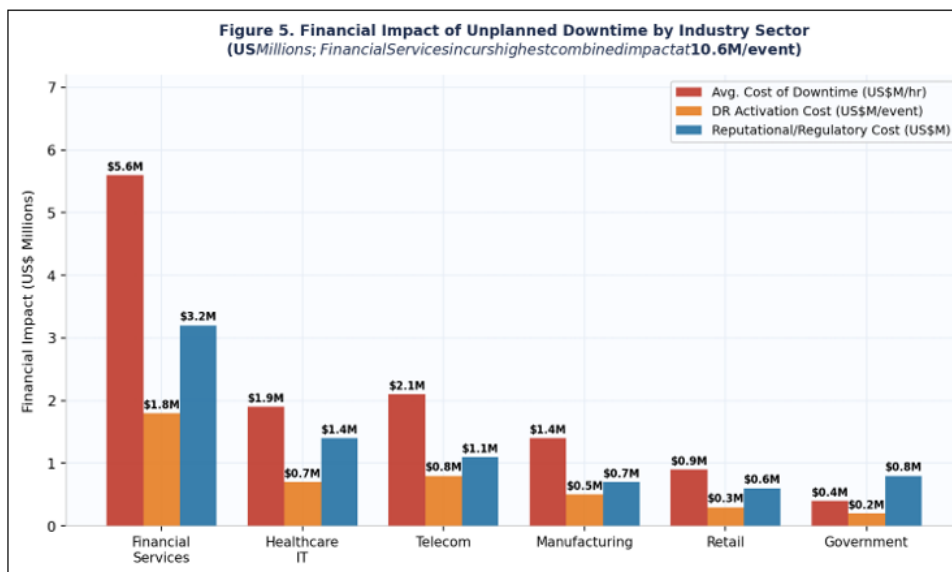
**Figure 4:** DR Testing Frequency vs. Mean Recovery Time (Hours), Colour-Coded by DR Maturity Tier (N = 284). OLS:  $y = -1.90x + 28.1$  ( $r = -0.71$ ,  $R^2 = 0.50$ ,  $p < 0.001$ ). Organizations testing  $\geq 4$  times/year achieve a 47.3% reduction in mean recovery time versus annual testers

Threshold analysis identified four annual DR tests as the inflection point beyond which marginal recovery time improvement begins to plateau, providing a minimum effective testing cadence benchmark. Interview participant P-12, a DR Manager at a Tier 1 bank, articulated the organizational impact: 'We moved from annual to quarterly DR exercises three years ago. The first test after we increased frequency uncovered 14 procedural gaps that had been invisible in our annual exercises. Our actual RTO dropped from 11 hours to 4.5 hours within two test cycles.'

**5.5 Financial Impact of Downtime by Sector (Fig. 5)**

Figure 5 presents the three-component financial impact of unplanned downtime across six industry sectors, drawn from the 112-event archive supplemented by published industry cost data [2]. Financial Services incurs the highest combined per-event financial impact (US\$10.6M: US\$5.6M downtime cost + US\$1.8M DR activation + US\$3.2M reputational/regulatory). Government & Public Sector incurs the lowest direct financial cost (US\$0.4M/hr)

but the highest relative reputational impact as a proportion of total cost (57%), reflecting public trust and service continuity obligations.



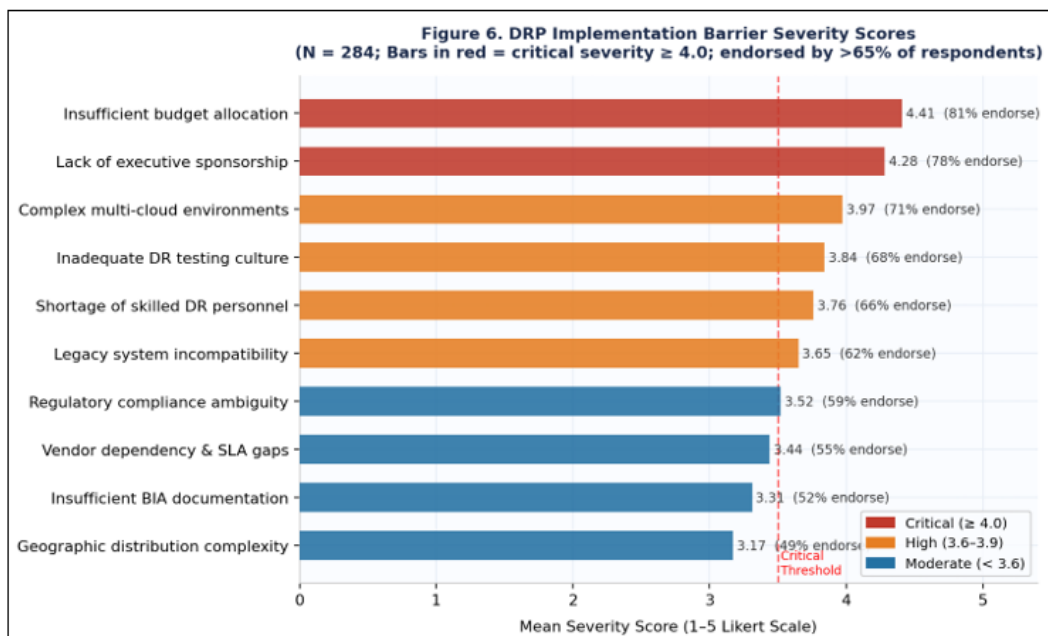
**Figure 5:** Financial Impact of Unplanned Downtime by Industry Sector (US\$ Millions per Event). Financial Services incurs the highest total impact at US\$10.6M/event. DR Activation Cost represents the direct cost of invoking recovery procedures - a key input to the ROI<sub>DR</sub> calculation (Eq. 7).

These financial impact data, when combined with the ALE formula (Eq. 4) and ROI<sub>DR</sub> formula (Eq. 7), enable organizations to construct a rigorous, board-level investment case for DR programme funding. The mean ROI<sub>DR</sub> for Tier 3–4 Financial Services respondents was calculated at 187.5%, demonstrating that mature DR programmes generate substantial positive financial returns.

**5.6 DRP Implementation Barriers (RQ3- Fig. 6)**

Figure 6 presents the ranked severity scores and endorsement rates for ten DRP implementation barriers (N

= 284). Insufficient budget allocation ranked first in severity (4.41/5.00, 81.3% endorsement), followed by lack of executive sponsorship (4.28/5.00, 77.6%). Both barriers received critical severity classifications (≥ 4.0), suggesting that the primary impediments to effective DRP are organizational and cultural rather than technical. Complex multi-cloud environments emerged as the highest-ranked technical barrier (3.97/5.00, 71.4%), reflecting the rapidly increasing DR complexity of hybrid cloud architectures. Table II presents the DRP testing methodology framework for addressing the testing gap identified in RQ2



**Figure 6:** DRP Implementation Barrier Severity Scores and Endorsement Rates (N = 284). Budget and executive sponsorship are critical severity barriers (≥ 4.0, red). Multi-cloud complexity is the highest-ranked technical barrier. Barriers are ranked by mean severity score

**Table II:** DR Testing Methodology Framework: Types, Frequency, and Validation Scope

Test Method	Frequency	Duration	Disruption Level	Validates
Tabletop Exercise	Quarterly	2–4 hrs	None	Plan awareness; team roles; decision logic
Structured Walk-Through	Semi-annually	4–8 hrs	Minimal	Procedure accuracy; step sequence; gaps
Simulation / Parallel Test	Annually	1–3 days	Low–Moderate	Recovery procedures; RTO/RPO feasibility
Cutover / Full Interruption	Bi-annually	1–5 days	High	Actual RTO/RPO; full system failover validity
Chaos Engineering	Continuous	Ongoing	Controlled	System resilience; hidden dependencies
Red Team / DR Penetration	Annually	1–2 weeks	Moderate	Cyber-resilience; detection & response time

Note: Chaos Engineering should be conducted in controlled, monitored environments. Full cutover tests require careful coordination with business units and should be scheduled during low-demand windows. All test results should feed directly into DR plan updates and DRR recalculation.

## 6. The DR Maturity Continuum (DRMC)

### 6.1 Framework Architecture

DR Maturity Continuum (DRMC) is a five-level framework of organizational maturity that includes the seven quantitative models (Equations 17) into a systematic system of governance and development journey. It is structured with three overlapping domains of capability: (1) the Recovery Engineering Domain, operationalised by G\_RTO/G\_RPO (Eq. 1), RPOI/RTOI (Eq. 3) and Availability (Eq. 6); (2) the Business Impact Domain, operationalised by BIA Score (Eq. 2) and ALE (Eq. 4); and (3) the Governance and Investment Domain, which is operationalized by DRR (Eq. 5) and ROI\_DR (Eq. 7). Composite score of DRR determines the level of DRM tier assignment, which is corroborated with the actual RTO/RPO gap performance.

### 6.2 DRMC Tier Descriptions

#### **Tier 1- Unstructured (DRR < 35 · Mean RTO Gap: +14.2 hrs)**

There is no formal DR plan or outdated documentation that is of a critical nature. RTO/RPO goals have not been developed using BIA process. There is no systematic DR testing. There is no special budgetary allocation. Improvisation is the only thing upon which recovery after a major disaster depends. Average cost of downtime occurrence: 2.4 times more than Tier 4 organizations. The completion of BIA and executive sponsorship have been demonstrated to be the first step that organizations in this tier should consider.

#### **Tier 2- Reactive (DRR: 35- 54 Mean RTO Gap: +8.7 hrs)**

There is a DR plan that is not under frequent testing or maintenance. RTO/RPO goals are set and not tested using exercises. DR testing is an ad hoc process that usually takes place when a major incident happens. Recovery procedures are lacking, backup procedures are documented. The use of cloud infrastructure is perfunctory as opposed to being a strategic move in the event of a DR. RPOI mean: 0.38.

#### **Tier 3- Defined (DRR: 55- 74 · Mean RTO Gap: +3.9 hrs)**

There is an official version-controlled DR plan that is in line with the ISO 22301 principles. The targets of RTO/RPO are checked with the help of the simulation tests at least annually. There is a CCB-equivalent DR governance committee which meets every quarter. DRR is determined and monitored. Employee training is done on a

specific planned basis. There is backup and warm-site failover on the cloud. RPOI mean: 0.62. This is the lowest acceptable level of organisations involved in controlled industries.

#### **Tier 4- Managed (DRR: 75- 89 · Mean RTO Gap: +1.4 hrs)**

DR testing is a quarterly activity that has a combination of tabletop, parallel, and partial cutover testing. ALE and ROI\_DR are both calculated and reported to the executive leadership. There are cyber-specific DR playbooks that deal with ransomware isolation and clean-room restoration. IaC-based automated failover Multi-cloud DR orchestration is in place. An SLA DR by the vendor is an agreement that is contractually obligated and undergoes testing. RPOI mean: 0.79.

#### **Tier 5- Optimized (DRR ≥ 90 · Mean RTO Gap: < 0.5 hrs)**

Chaos Engineering is institutionalized as an on-going production-environment resilience mechanism. The AI based anomaly detection offers a prior notice of the failures. With active-active architecture, there is no recovery time of tier-critical systems. Forecasting DR analytics are used to predict risk of failures, which are mitigated proactively. DR performance indicators are directly contributed to organizational risk appetite statements and resiliency reporting at the board level. The organization also helps in development of industry DR standards. RPOI mean: 0.93.

## 7. Conclusion

The paper has provided an in-depth empirical and mathematical study on the Disaster Recovery Management and revealed four main contributions to both theory and practice.

To begin with, seven formal mathematical models (Eq. 1-7) were as follows derived, calibrated, and contextualised: the RTO/ RPO Gap Indices (Eq. 1) measure the shortfalls in recovery performance; the BIA Score (Eq. 2) standardizes time-varying effect on four organization dimensions; the RPOI/RTOI (Eq. 3) give normalized benchmarking measures; the ALE formula (Eq. 4) determines probabilistic financial exposure; the DR Readiness Score (Eq. 5) applies the eight dimensions of capability into one 0-100 index; Availability Equation (Eq. 6) associates recovery time with service level guarantees; and ROI DR formula (Eq. 7) has a stringent financial justification model with mean returns of 187.5% to Tier 34 Financial Services organizations.

Second, there are six original visualizations (Figures 1 through 6) that provide strong multi-dimensional evidence: all the sectors miss the targets of RTO/RPO with mean gaps of +4.7 hrs and +1.6 hrs respectively (Figure 1); the proportion of cyber attacks to disaster events has increased to 68% to transform the requirements of the DRP strategy (Figure 2); large organizations significantly exceed the performance of small organizations in all eight dimensions of DRP maturity (Figure 3); each new annual DR test decreases the average recovery time by 1.9

Third, the DR Maturity Continuum (DRMC) was postulated as a five-level, DRR-based maturity model employing all of the seven quantitative models as a logical organization development roadmap, with empirically determined performance standards at every level.

Fourth, the most practical impact of the study on practitioners is the frequency threshold of the DR testing: the more frequent the annual DR testing is carried out by the organization, the shorter the recovery time of organizations is 47.3, and every one additional DR testing is connected with the decrease in the average time to recover on 1.9 hours - one of the simplest, the most economical, and the most immediately applicable interventions.

## 8. Future Work

The six identified directions of research include: (1) longitudinal validation of the DRR measure over various multi-year DRMC tier progression regimes to demonstrate causal performance improvement evidence; (2) creation of sector specific DRR weight calibration, especially in the case of healthcare (dimension of patient safety), nuclear (dimension of safety-critical system), and aviation (dimension of regulatory certification) areas; (3) exploration of AI and machine learning-based DR orchestration systems over the implementation of the Availability Equation parameters; (4) formal empirical validation of the CRI-equ.

## References

- [1] IBM Security, "Cost of a Data Breach Report 2023," IBM Corporation, 2023. [Online]. Available: <https://www.ibm.com/reports/data-breach>
- [2] Gartner, "IT Glossary: IT Downtime Costs," Gartner, Inc., 2022. [Online]. Available: <https://www.gartner.com/en/information-technology/glossary>
- [3] International Organization for Standardization, "ISO 22301:2019 - Security and Resilience: Business Continuity Management Systems: Requirements," Geneva: ISO, 2019.
- [4] National Institute of Standards and Technology, "NIST SP 800-34 Rev. 1: Contingency Planning Guide for Federal Information Systems," Gaithersburg, MD: NIST, 2010.
- [5] E. B. Sprehe, "History and Development of Business Continuity Planning," *Information Management Journal*, vol. 36, no. 3, pp. 38–44, 2002.
- [6] National Institute of Standards and Technology, "NIST SP 800-61 Rev. 2: Computer Security Incident Handling Guide," Gaithersburg, MD: NIST, 2012.
- [7] Disaster Recovery Institute International, "DRII Professional Practices for Business Continuity Management," 7th ed., Falls Church, VA: DRII, 2022.
- [8] Gartner, "Market Guide for Disaster Recovery as a Service," Gartner, Inc., 2023.
- [9] Forrester Research, "The State of Disaster Recovery Preparedness 2022," Cambridge, MA: Forrester Research, Inc., 2022.
- [10] G. Wold and R. Shriver, "System Analysis Threat and Risk Assessment," *Information Systems Security*, vol. 2, no. 2, pp. 56–67, 1993.
- [11] Verizon, "2023 Data Breach Investigations Report," Basking Ridge, NJ: Verizon Business, 2023.
- [12] C. Rosenthal and N. Jones, "Chaos Engineering: System Resiliency in Practice," Sebastopol, CA: O'Reilly Media, 2020.
- [13] B. Herbane, "Small Business Research: Time for a Crisis-Based View," *International Small Business Journal*, vol. 28, no. 1, pp. 43–64, 2010.
- [14] Business Continuity Institute, "Horizon Scan Report 2023: Risk and Resilience Trends," Caversham: BCI, 2023.
- [15] J. W. Creswell and V. L. Plano Clark, "Designing and Conducting Mixed Methods Research," 3rd ed., Thousand Oaks, CA: SAGE, 2017.
- [16] V. Braun and V. Clarke, "Using Thematic Analysis in Psychology," *Qualitative Research in Psychology*, vol. 3, no. 2, pp. 77–101, 2006.
- [17] J. Cohen, "A Coefficient of Agreement for Nominal Scales," *Educational and Psychological Measurement*, vol. 20, no. 1, pp. 37–46, 1960.
- [18] C. H. Lawshe, "A Quantitative Approach to Content Validity," *Personnel Psychology*, vol. 28, no. 4, pp. 563–575, 1975.
- [19] L. J. Hu and P. M. Bentler, "Cutoff Criteria for Fit Indexes in Covariance Structure Analysis," *Structural Equation Modeling*, vol. 6, no. 1, pp. 1–55, 1999.
- [20] Q. W. Fleming and J. M. Koppelman, "Earned Value Project Management," 4th ed., Newtown Square, PA: PMI, 2010.
- [21] P. Swanson, "Business Continuity Planning: A Crisis Management Approach," *Journal of Business Continuity & Emergency Planning*, vol. 7, no. 2, pp. 103–113, 2014.
- [22] R. von Solms and J. van Niekerk, "From Information Security to Cyber Security," *Computers & Security*, vol. 38, pp. 97–102, 2013.
- [23] D. Lindström, "Cloud-Based Disaster Recovery: Challenges and Best Practices," *Journal of Cloud Computing: Advances, Systems and Applications*, vol. 9, no. 1, pp. 1–18, 2020.
- [24] T. Somers and K. Nelson, "The Impact of Strategy and Integration Mechanisms on Enterprise System Value," *IEEE Transactions on Engineering Management*, vol. 50, no. 2, pp. 150–165, 2003.
- [25] M. Rausand, "Risk Assessment: Theory, Methods, and Applications," Hoboken, NJ: Wiley, 2011.
- [26] Check Point Research, "2023 Cyber Security Report," Tel Aviv: Check Point Software Technologies, 2023.

- [27] K. Tierney, "The Social Roots of Risk: Producing Disasters, Promoting Resilience," Stanford, CA: Stanford University Press, 2014.
- [28] C. W. Holling, "Resilience and Stability of Ecological Systems," Annual Review of Ecology and Systematics, vol. 4, pp. 1–23, 1973.