



REPUBLIC OF TÜRKİYE  
MINISTRY OF INDUSTRY  
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**TÜBİTAK**  
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# TÜBİTAK BİLGEM

Data-Driven Insights for Software Development Process Improvement:  
A Defect Analysis

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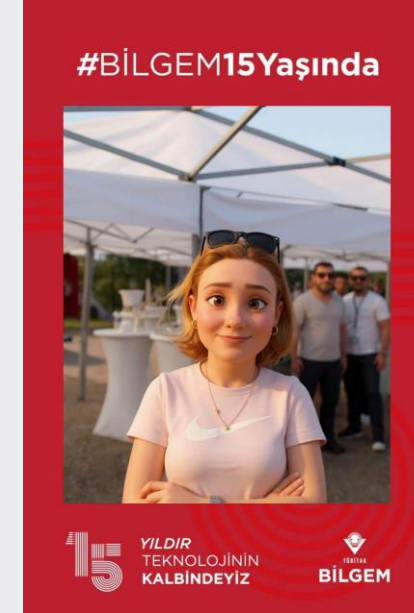


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- MSc in Industrial Engineer with 5 years of experience in software industry & public institutions
- Master Thesis: Self-Starting Control Charts for Software Development Projects, 2023.



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- MSc in Statistics with 10+ years of experience in the public sector.
- Master Thesis : Estimation of Variance Components in Measurement Systems Analysis, 2017.

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# 14%

Ph.D  
Degree

# 25%

Master's  
Degree



# 6

Institutes



# 8

Test Laboratories



# 200+

Products



# 230

Projects



# +1 Billion \$

Portfolio



# 15

Areas of  
Critical Research

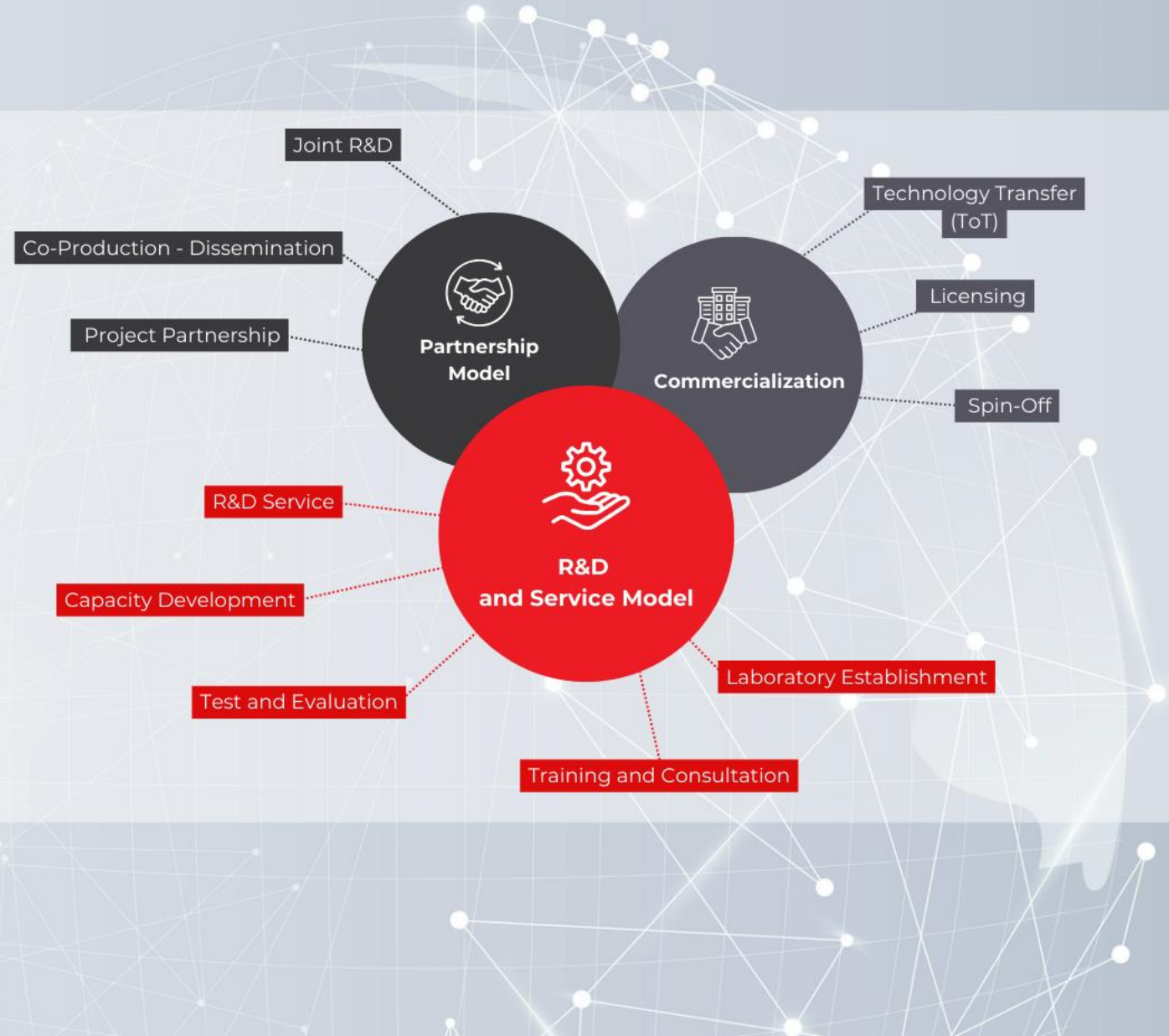
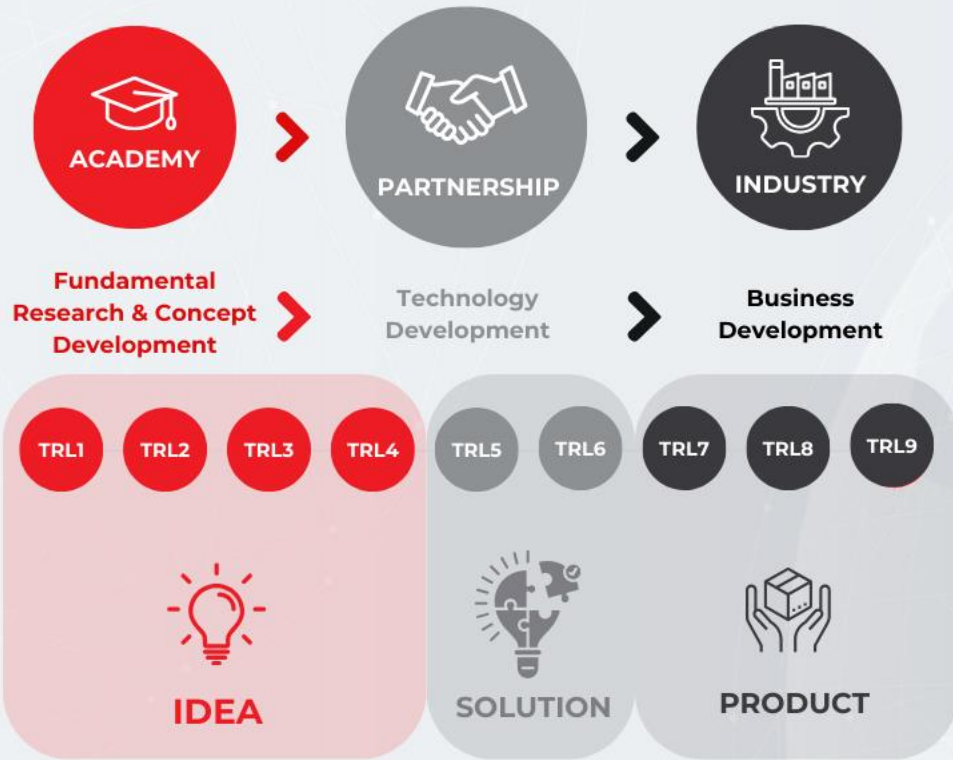


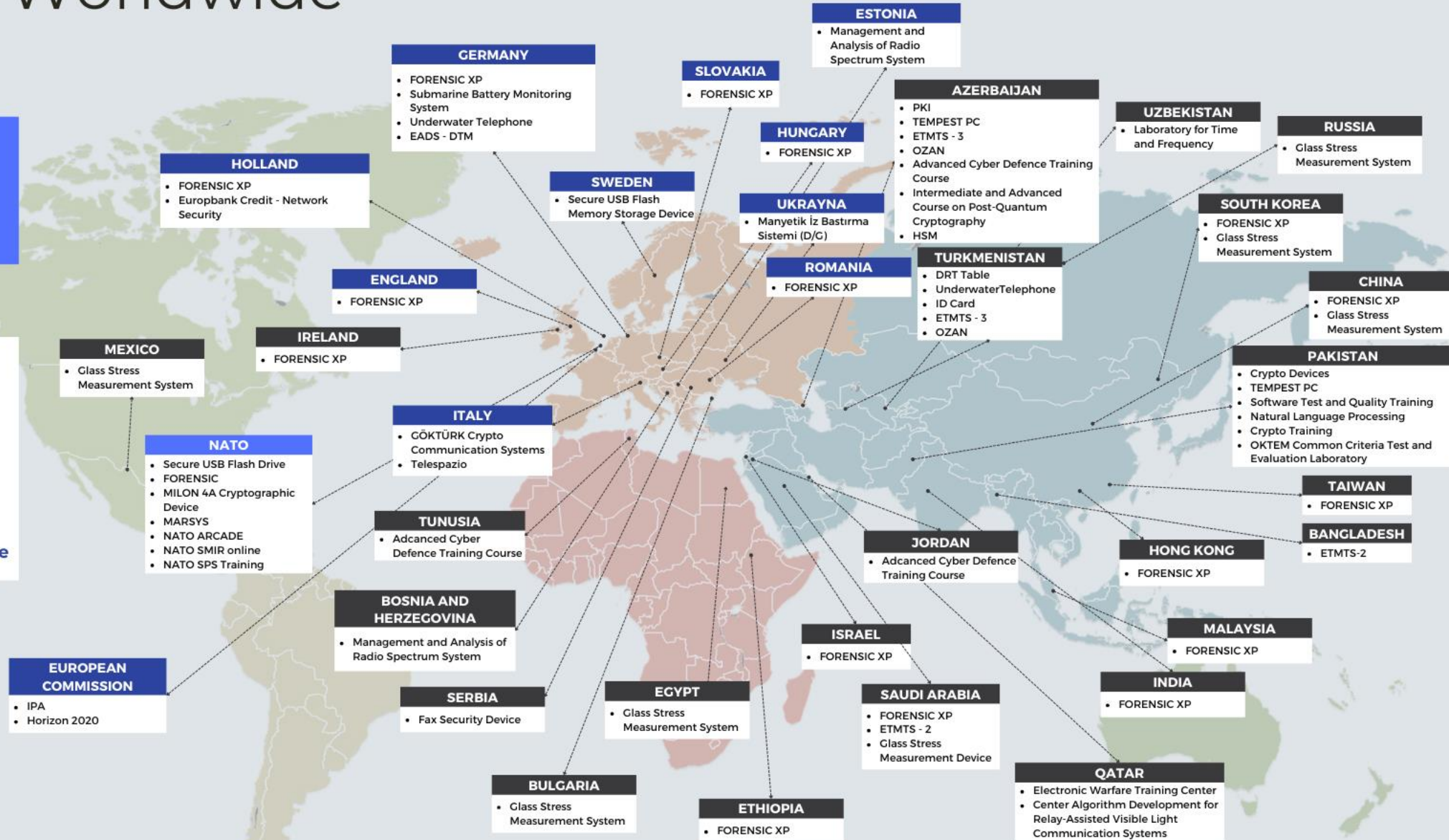
# Enabling Technologies and **Research Areas**





# Business and Collaboration Models



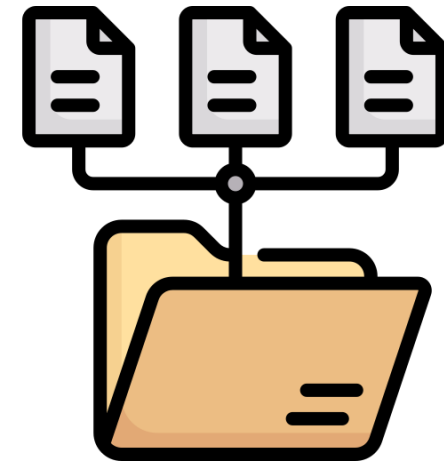




# INTRODUCTION



1. Related Work
2. Methodology
3. Analysis
4. Conclusions and Future Work



## Why Are Software Defects Important?

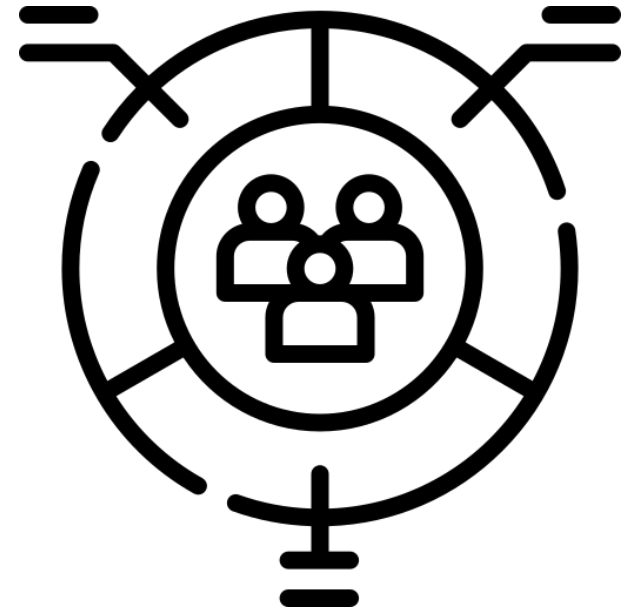
- Defects (bugs/errors) directly impact reliability, maintainability, and user satisfaction.
- Early detection & resolution reduce rework, cost, and organizational risk.
- Crucial in public-sector projects where credibility and compliance are critical.
- Defect analysis enables prevention by tracing root causes.





## What Makes the Public Sector Different?

- Greater **regulatory oversight** and **complex stakeholder** environments.
- **Longer** procurement and development **cycles**.
- Frequent **reuse** and **integration** across agencies → **higher defect** impact.
- Defects can **propagate** through **interconnected** systems.

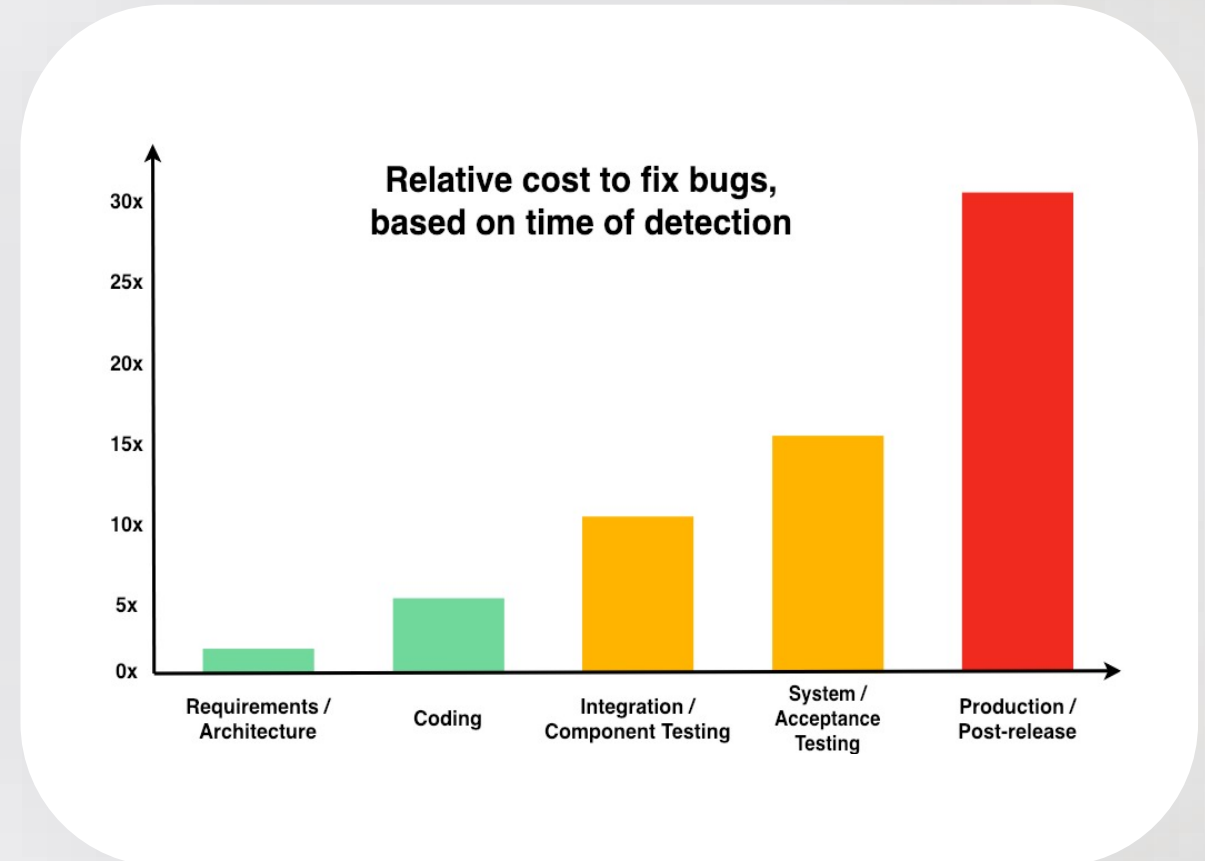


# RELATED WORK



## Why Defect Tracking Matters?

- Defect data is a critical management asset for process improvement (Grady, 1996).
- Defect frequency is a direct, quantifiable indicator of software quality.
- More defects = lower reliability, usability, efficiency.
- Early defect prevention is more cost-effective than late detection.



## From Detection to Prevention: Defect Causal Analysis (DCA)

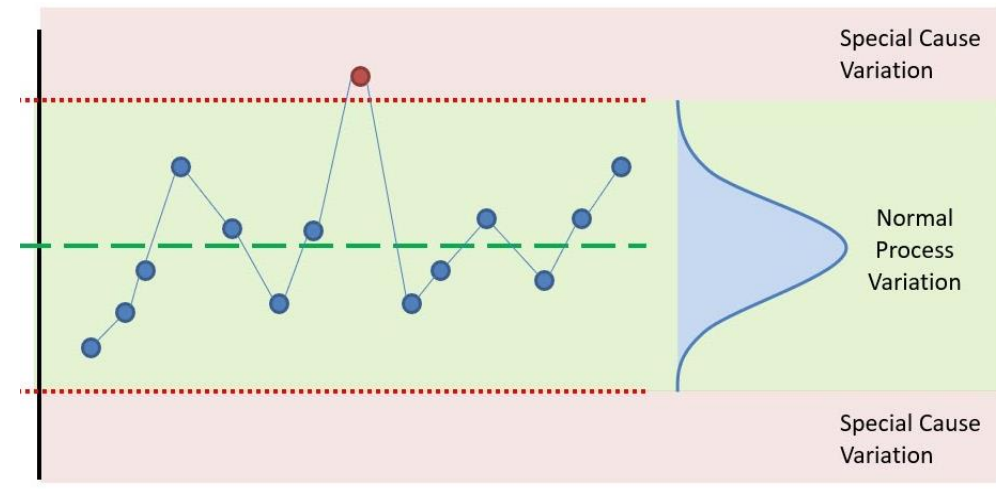
- DCA = structured method for **identifying root causes** of recurring defects.
- Uses **classification** tools like **Pareto** charts to **prioritize** frequent defect types.
- Enables targeted process improvements.
- Shifts focus from reactive fixing to proactive prevention.





## Using Statistical Process Control (SPC)

- Software processes exhibit variation: natural (common) vs. assignable (special) causes.
- SPC helps distinguish between the two and identify instability in the process.
- Control charts (e.g., defect density over time) help monitor process health.
- Timely action on SPC data leads to sustained process improvement.



# METHODOLOGY

## Why This Study?

- Triggered by a noticeable increase in production defects in a public-sector software project.
- Aimed to **analyze patterns**, timing, and root causes of defects.
- R Studio dashboards signaled **unusual trends** → deeper investigation initiated.
- Scope refined with input from technical lead & project manager for relevance.
- Goal is to support shift from reactive defect handling to proactive quality assurance, aligned with TQM and CMMI.



## Data Source and Structure

- Data pulled from a **task and issue management** platform used across the full SDLC.
- Platform captures **detailed metadata**: issue type, severity, component, status, etc.
- Dataset covers 147 resolved defects in production, from Jan 2024 to Apr 2025.
- Entries were validated & filtered in collaboration with project leadership.
- Focus: Only **confirmed production defects** (not test/staging environment issues).

Field	Value
Issue Key	143
Issue Type	Defect
Sprint Period	01.04.24
Severity	Medium
Defect Type	Coding
Component/s	A
Detected Activity	System Monitoring
Resolution	Done
Status	Closed



## Defect Variables Used in the Analysis

- **Severity:** Impact level of each defect (minor → major)
- **Detected Activity:** Phase where defect was found (e.g., Monitoring, Development, Test)
- **Component(s):** Affected modules/subsystems
- **Detected Sprint:** When the issue was logged (enables time-based analysis)
- **Defect Type:** Technical nature (e.g., coding, data, UI)



## Data Considerations and Limitations

- **Manual data entry** may affect accuracy/timeliness.
- **Subjectivity** in issue classification (defect vs. other). Sample size = 147 defects, limits generalizability.
- Despite limitations, dataset provides **real-world traceability** and reflects **day-to-day dev operations**.
- Supported by literature: **Better data → Better defect prediction** models [1].



## Expected Outcomes and Impact

- Identify conditions, components, and time periods with high defect incidence.
- Trace root causes of recurring issues using structured data.
- Reduce defect density → improves maintainability and user satisfaction.
- Support shorter release cycles, lower maintenance costs, and greater trust in public services.

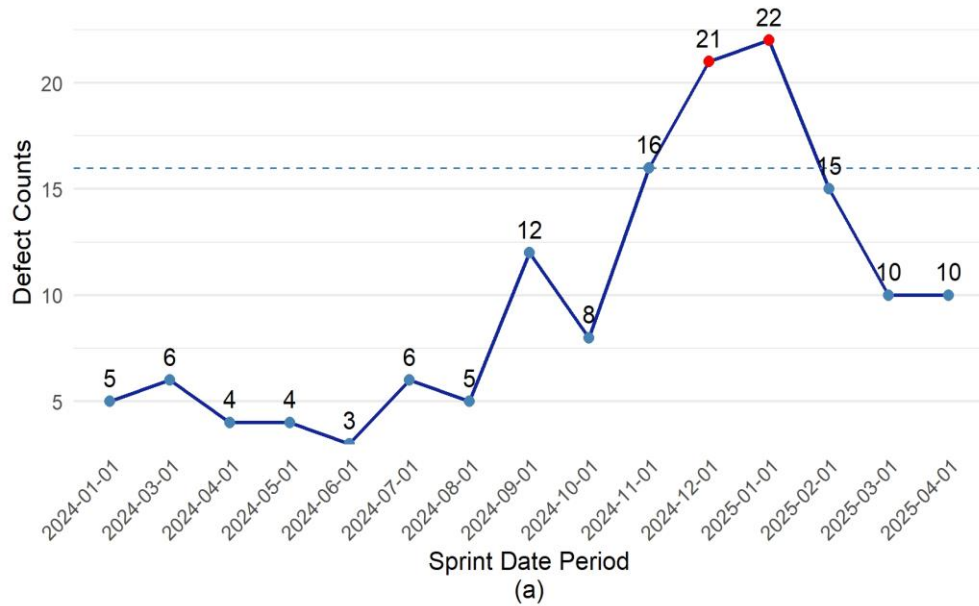


# ANALYSIS

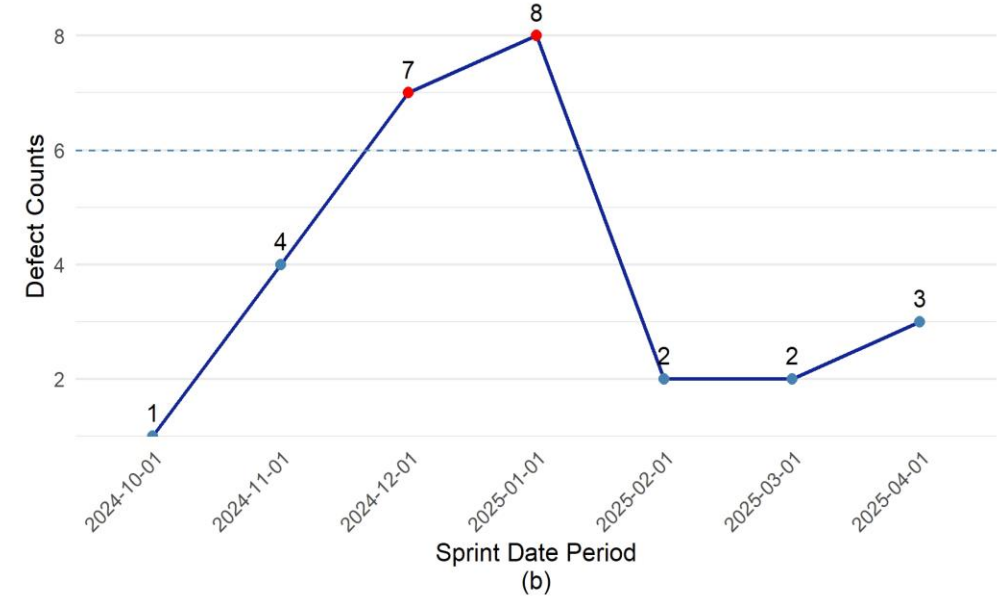
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## Monthly Defect Counts

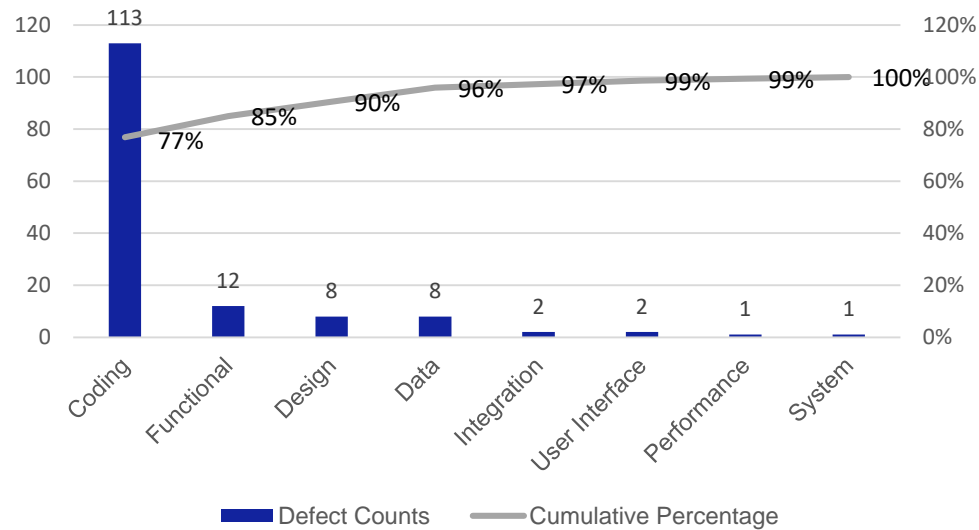


## Monthly Major Defect Counts

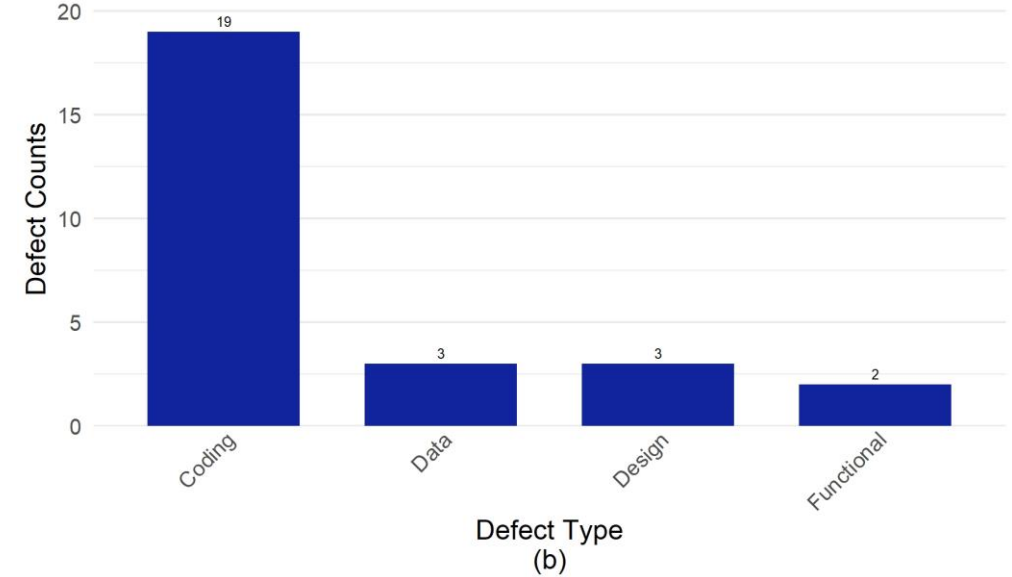


- Total and major defect counts increased significantly in some months.
- UCL (mean +  $1\sigma$ ) used to flag statistically significant anomalies.
- December 2024 and January 2025 exceeded control limits.
- Root cause: likely due to changes in reporting behavior, not true defect spikes

## Pareto Chart

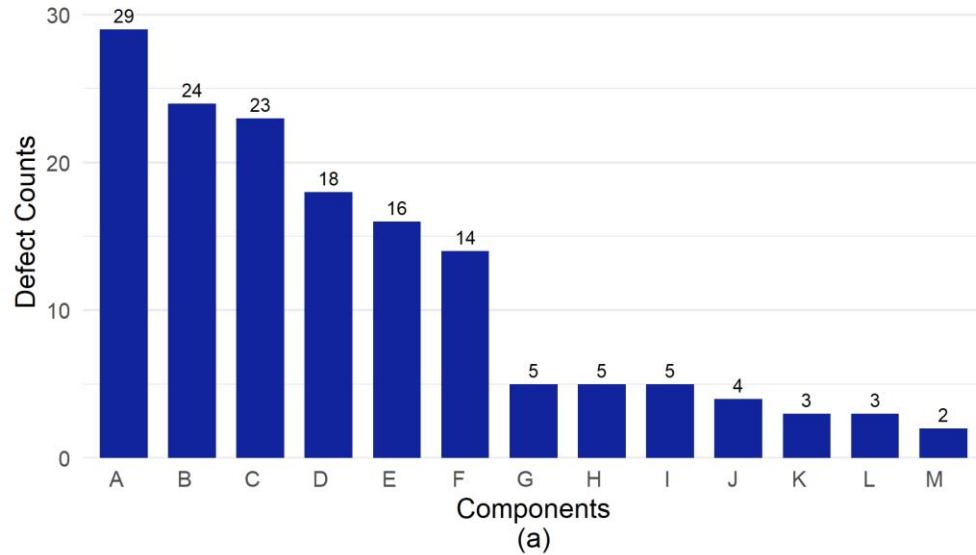


## Major Defect Counts By Defect Type

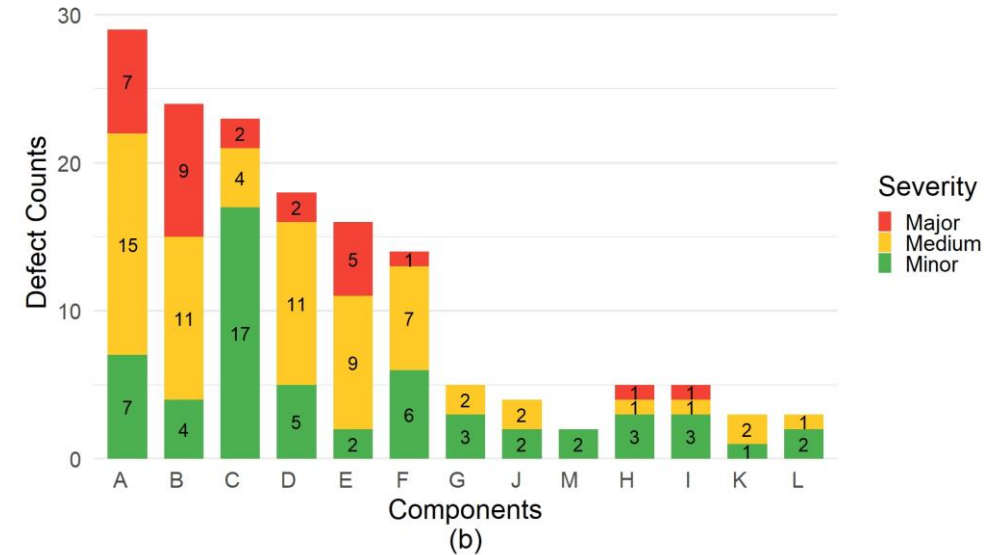


- Coding defects = 77% → largest opportunity for quality improvements.
- Other frequent types: functionality, architecture, data.
- Most major defects also stem from coding issues.
- Findings support focus on developer training and code reviews

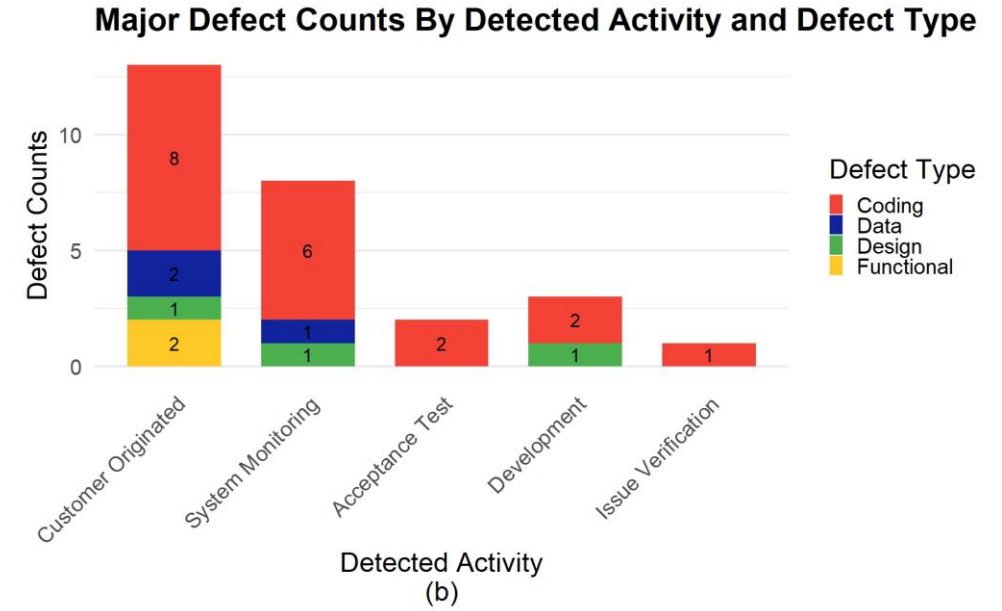
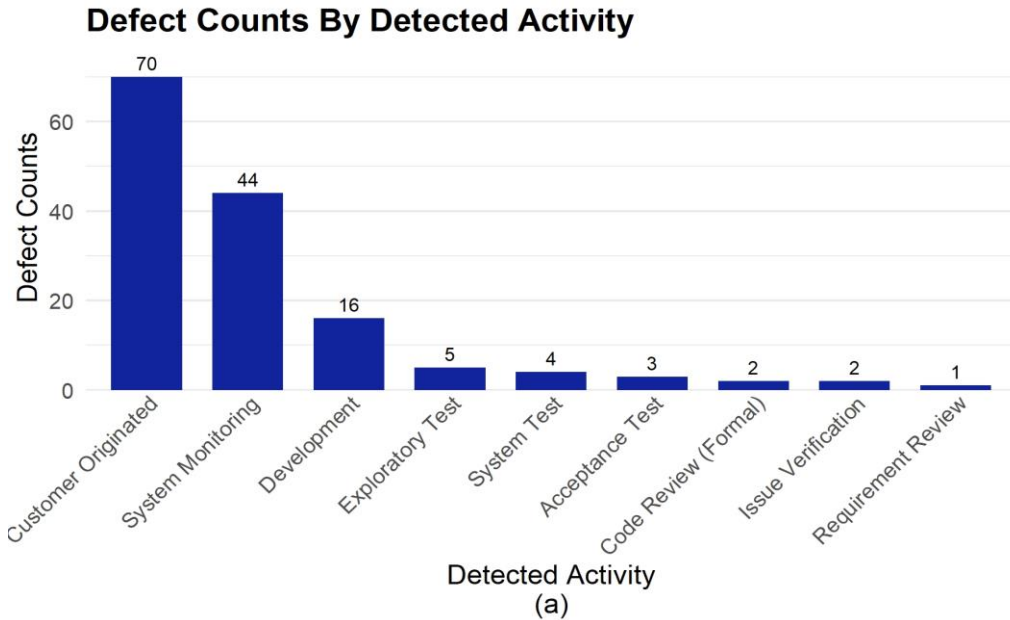
## Defect Counts By Component



## Defect Counts By Component and Severity



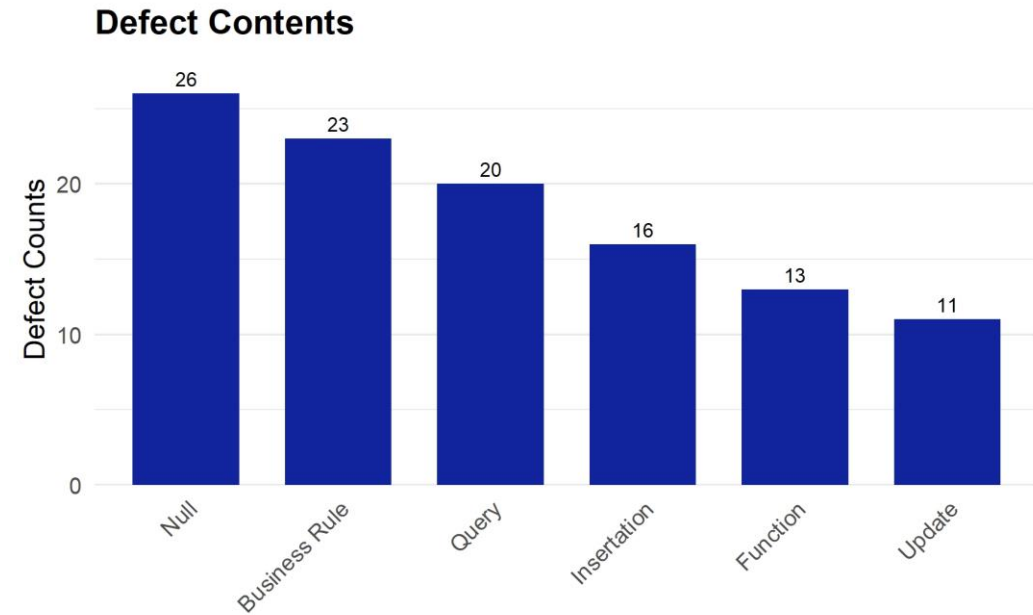
- Three components show notably high defect concentration.
- Component E: lower total defects but high % of major issues.
- Component C: most defects, but majority are minor.
- Insight helps prioritize quality actions per component



- Most defects detected post-deployment, especially by customers (70 cases).
- System monitoring tools caught 44 issues.
- Indicates room for improvement in early testing phases.
- Early detection reduces customer impact and operational risk



- Top defect types:
  - Null Pointer Exceptions (NPEs) – 26
  - Business rule violations – 23
  - Query issues – 20
- NPEs often caused by dynamic data sources (beyond static analysis).
- Suggest training, improved input validation, and business rule alignment.



- Coding defects dominate → prioritize code quality practices.
- High post-release detection → improve test coverage & early QA.
- Address reporting consistency through training or input validation.
- Analyze business rule violations for process misalignments



# CONCLUSIONS AND FUTURE WORK

## ***Key Conclusions***

- Coding defects are the dominant issue, especially related to Null Pointer Exceptions (NPEs).
- Most major defects were Customer Originated, highlighting late detection.
- Components A and B are the most defect-prone in both total and major categories.
- Frequent business rule violations suggest weaknesses in requirement analysis.

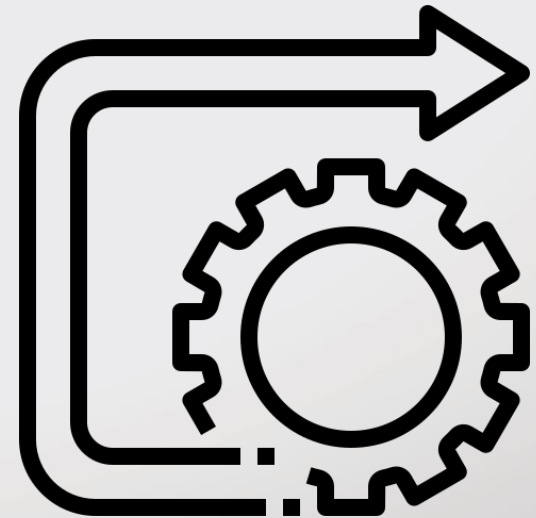
## ***Recommendations***

- Improve test coverage beyond happy paths, including edge cases (e.g. null handling).
- Conduct targeted developer training on code robustness and NPE prevention.
- Investigate root causes of business rule violations (customer vs. analyst errors).
- Increase analyst involvement in early project stages.





- Shift from reactive to proactive defect management.
- Integrate additional quality metrics for continuous monitoring.
- Develop a predictive model to anticipate high-risk defects pre-deployment.



# REFERENCES

- [1] Bosu, M.F., & MacDonell, S.G. (2013). Data quality in empirical software engineering: a targeted review. In: Proceedings of the 17th International Conference on Evaluation and Assessment in Software Engineering (EASE 2013), pp. 171–176.



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Thank you