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State of the Sector

A Report on the Data Maturity of Social Purpose Organisations in India

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Special thanks to all the experts who gave us their valuable time and inputs, which helped us design the questionnaire and refine our thinking about how to take this project forward.

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Foreword



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India's development sector stands at a pivotal moment. Never before has it had access to such a wealth of digital tools, diverse datasets, and opportunities for collaboration across stakeholders. Yet, the potential of these resources remains only partially realised. For Social Purpose Organisations (SPOs), the ability to collect, manage, analyse, and apply data effectively has become essential for driving meaningful and scalable change.

Despite the abundance of data, SPOs face persistent barriers: restrictive or unclear data access policies, outdated formats, siloed databases, and a shortage of data talent. The Centre for Data Science and Social Impact (CDSSI) at the Indian School of Development Management presents in this report a comprehensive assessment of these challenges, the sector's current state of data maturity, and opportunities to advance data-driven decision-making for stronger outcomes.

From my experience leading a program to build data maturity in SPOs—through skill-building and year-long talent placements—we observed two approaches. Some organisations recognised the strategic value of investing in talent, embedding them in environments where mentorship, domain expertise, and organisational vision converged. These organisations made a cultural shift toward working with data as an integral part of decision–making. Others treated the placement of data talent as short–term project–based resource, without fully leveraging the opportunity to make data capacity a sustained organisational asset. Across more than 30 SPOs, we found that the journey varies widely: grassroots organisations often face steeper learning curves due to limited awareness of tools, entrenched cultural barriers to adopting a data mindset, and funding constraints.

Data-driven decision-making goes beyond reporting or monitoring. It shapes intervention design, improves research frameworks, and optimises resource allocation. Structured data models strengthen donor engagement, enabling SPOs to present transparent, evidence-based narratives that build trust and credibility. SPOs that treat data maturity as a strategic investment strengthen relationships with stakeholders, improve program relevance, and foster a culture of learning and accountability.

Understanding the sector's current position is the first step to collective progress. With a shared commitment to building data capabilities, stakeholders can unlock more efficient, effective, and impactful use of resources. This will enhance the sector's ability to improve lives through informed policy, and accelerate poverty alleviation. I commend CDSSI for taking up this important piece of work at this time and look forward to seeing how the ISDM CDSSI DMA becomes a seminal resource for the sector on its journey toward greater efficiency and meaningful change.

Foreword



Trisha Varma
Director, Global
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of Development
Management

Good data practices can be a gamechanger for the Indian social sector. They help us see our work clearly, learn faster, and make better decisions at an organisational level. Yet, our experience at the Centre for Data Science and Social Impact (CDSSI) at the Indian School of Development Management (ISDM) has shown that while many Social Purpose Organisations (SPOs) in India recognise the value of data, most are not using it to its full potential.

We created the Data Maturity Assessment (DMA) to help bridge this gap. Leveraging data is not just about embracing new technology. It is about building an organisation-wide capacity to collect, understand, and act on information such that [it] becomes part of everyday work. In a world where communities are increasingly digitally-connected, and donors expect greater transparency, SPOs need to be agile, credible, and responsive. And this can only happen when decisions are not based on incomplete or unreliable information—in other words, through strong data practices.

CDSSI works within and for the social sector—close enough to understand the ground realities, but positioned to offer an independent and sector-wide view. This gives us the ability to design theoretically sound tools that are also practical and relevant.

The DMA is the largest and most diverse study of data maturity in India's social sector. It is a milestone for CDSSI and ISDM, and we hope, a foundational step for the sector. It is not just a scorecard, but a mirror—reflecting where organisations are today, where their strengths lie, and what areas could benefit from attention. It covers both program and administrative data, recognising that it is equally essential to understand communities, as it is to run effective operations.

Our framework draws on respected global and national models, but is built specifically for Indian SPOs through deep consultations with leaders, practitioners, and data experts.

This report carries the voices of over 360 SPOs from across the country. It is by the sector, for the sector. We hope you will see your own organisation in its pages, and be inspired to take the next step on your data journey. The more SPOs that engage with, adapt, and evolve this framework, the stronger we all become.

Ultimately, this is about more than data. It is about stronger organisations, more effective programs, and deeper, lasting impact in the communities we serve. That is the future we at CDSSI are committed to building—together.

Abbreviations

1. SPO Social Purpose Organisation

In the context of this report, it refers to organisations engaged in program delivery.

2. DMA Data Maturity Assessment

3. FCRA licence Mandatory certification for organisations in India to legally receive foreign contributions

4. DPDPA Digital Personal Data Protection Act, 2023

5. MIS Management Information System

6. M&E Monitoring & Evaluation

7. GIS Geographic Information System

8. CRM system Customer Relationship Management

9. ROI Return on Investment

10. KIIS Key Informant Interviews

11. WASH Water, Sanitation and Hygiene

Glossary of Terms

- Data Life Cycle: The complete process of managing data, from how it is created and collected, to how it is stored, used, shared, and eventually archived or deleted.
- 2. Exploratory Sequential Mixed Methods Approach: A research design where the study begins with quantitative data collection and analysis to explore a phenomenon, followed by qualitative data collection to test or expand on the initial findings.
- **3. Data Maturity:** Measure of an organisation's ability to effectively collect, manage, analyse, and use data for decision-making.
- **4. Data proficiency:** The skill and competence of individuals or teams in working with data tools, analysis, and interpretation.
- **5. Foundation:** A non-profit entity (often grant-making) that funds and supports social impact initiatives.
- **6. Trust:** A legal structure for charitable activities in India, governed by a trust deed and state-specific trust laws.
- **7. Society:** A membership-based non-profit organisation in India registered under the Societies Registration Act, 1860, often focused on advocacy, research, or service delivery.

- **8. Section 8 company:** A non-profit company in India registered under the Companies Act, 2013, set up for promoting charitable objectives without distributing profits.
- **9. Traditional/Manual Methods (of Data Collection):** A way of gathering information without using digital tools, such as paper forms, in-person interviews, or handwritten records.
- **10. Open Data Policy:** A framework that mandates or encourages making data freely accessible, shareable, and reusable by anyone.
- **11. KoboToolbox:** An open-source suite of tools for data collection and analysis, especially in challenging field environments.
- **12. Airtable:** A cloud-based platform that combines spreadsheet and database features for organising and collaborating on data.
- **13. Power BI:** Microsoft's business analytics tool for visualising data, creating dashboards, and generating reports.
- **14. Data interoperability:** The ability of different data systems and software to exchange, interpret, and use information seamlessly.
- **15. Structured Monitoring Systems:** A set of organised, often digital, processes for tracking and recording data in a consistent way, making it easier to analyse and use for decision-making.



Executive Summary

In India's rapidly evolving social sector, data is central to credibility, responsiveness, and lasting impact. Social Purpose Organisations (SPOs) must move beyond using data solely for reporting, to embedding it in everyday decision-making and strategy. Yet, while awareness of data's value is growing, the sector still faces significant capability and cultural gaps.

To address this, the Centre for Data Science and Social Impact (CDSSI) at the Indian School of Development Management conducted the **first large-scale Data Maturity Assessment (DMA)** of 360 SPOs across India. The ISDM CDSSI DMA framework is unique because:

- It is tailored to Indian realities,
- It assesses both program/project data (beneficiaries, outcomes) and administrative data (HR, finance, operations), and
- It assesses data maturity across six dimensions: Culture and Mindset, HR Data Management, Project Data Collection, Storage and Accessibility, Analysis and Visualisation, and Training and Capacity Building.

With responses from a wide cross-section of SPOs varying in size, geography, thematic focus, and operational models, the DMA presents a comprehensive snapshot of the sector's data landscape. It offers benchmarks to help organisations locate

themselves on the maturity spectrum, identify capacity gaps, and learn from peer practices.

Key Findings of the ISDM CDSSI DMA:



Sector stuck in the middle: 70% of SPOs are in the Emerging or Progressing stages of data maturity. Only 4% qualify as Experts.



Awareness-action gap: While a majority use data for program design, far fewer use it for internal operations. Only a quarter integrate data across all core functions.



Barriers: Limited budgets, uncompetitive salaries for data talent, donor-driven compliance focus, and inadequate infrastructure.



Money matters: Higher annual expenditure strongly correlates with stronger data culture.



Thematic focus matters: Thematic diversification boosts maturity due to scale, accountability, and cross-cutting systems.

The report synthesises the findings into actionable insights. It offers SPOs a mirror to spark reflection, enable informed conversations, and guide investment into data capabilities that ultimately enhance impact.

To advance data maturity in the social sector, organisations must build a strong data culture. Leadership has a crucial role here, and must champion data use. Organisations must create cross-functional forums for reflection, and nurture internal data champions. Data should be used to inform strategy, fundraising, and communications, while staff at all levels should receive ongoing training, use open-source tools, and have access to funded learning opportunities.

It is essential to strengthen governance and compliance through simple data policies, security measures, and legal readiness. Inclusive and participatory practices such as collecting disaggregated data and integrating community feedback loops must be embedded in organisational culture. Data visualisation and storytelling can be improved through accessible dashboards and compelling narratives. Using shared infrastructure such as collaborative platforms, and standardised metrics, and advocating for long-term, flexible funding to sustain capacity and innovation, will also help.

By surfacing both gaps and opportunities in data use, the DMA invites SPOs, funders, and ecosystem actors to co-create the next stage of the sector's data journey. In doing so, it reinforces a shared commitment: that better use of data is not an end in itself, but a pathway to stronger organisations, smarter programs, and more resilient communities.

At a Glance

Data Maturity of the Social Sector

Data in Decision-Making



Less than half

of organisations have staff skilled in working with data

say they do not collect any data

66% consider data as extremely important. However, only

4%

have embedded data into their everyday decision-makina



34%

have invested in data training activities



10%

have either in-house data science teams or consulting support



Data for M&E

of organisations say that their M&E teams don't access project data



21%

of organisations do not have quantifiable monitoring indicators

16.7% of organisations engage in extensive data sharing with stakeholders/ partners

Digitisation of Data

28%



65%



of SPOs still use paper and pen for data collection

of SPOs rely on hard copy formats for data storage

Data Security







51%

of organisations do not have policies for data recovery and disaster mitigation

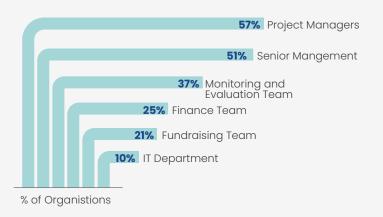
58%

of organisations do not know about DPDPA

26%

of organisations do not have any mechanism to secure data

Use of Project Data in SPOs



Context and Background

In recent years, the social sector has recognised that adopting a data-driven mindset and having strong data practices is critical for lasting impact. This goes beyond just using technology. It means building the capacity to collect, interpret, and use data effectively to inform decisions, improve programs, and demonstrate results. For Social Purpose Organisations (SPOs), many of which face resource constraints, developing these capabilities is essential to stay credible, responsive, and impactful in a rapidly changing environment.

Greater adoption of data-driven practices helps SPOs align their work with the needs of digitally-savvy communities who demand accountability and timely, personalised engagement, as well as evolving donor expectations. Without a strong data culture, organisations risk making decisions based on incomplete information, limiting their ability to learn, adapt, and grow.

A Data Maturity Assessment (DMA) is a vital tool to support this shift. It helps organisations understand where they stand in their

data journey—what they do well and where they need to improve in areas like data collection, analysis, governance, and usage. This clarity guides SPOs in building the right systems, skills, and processes to embed data into everyday decision–making and strategy.

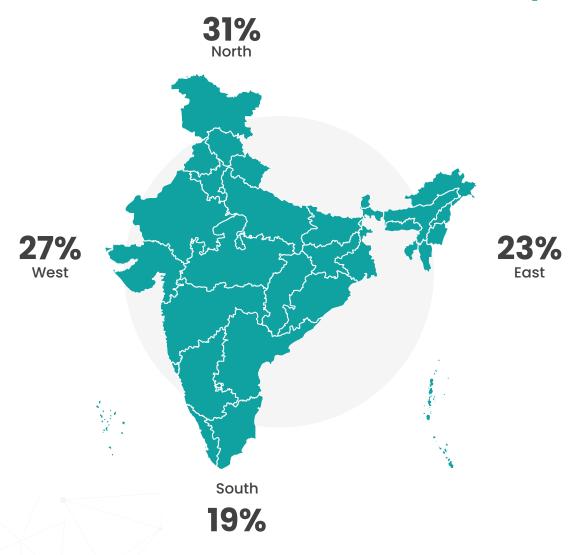
It is equally important to understand the sector's overall data maturity, because it helps SPOs benchmark themselves. This in turn reveals shared challenges and opportunities, which helps funders, leaders, and practitioners direct resources where they can make the biggest difference.

It is with this thought that we, at the Centre for Data Science and Social Impact (CDSSI) at the Indian School of Development Management, conducted the first large-scale DMA study of 360 SPOs across India. By providing a common framework and evidence base, this study empowers the sector to prioritise capacity building and move together toward higher data maturity—ultimately unlocking greater impact for the communities it serves.

The ISDM CDSSI DMA survey was conducted with **360 SPOs**, across the below thematic areas:

- Education and Literacy
- Nutrition
- · Health and Family Welfare
- Women's Development and Empowerment
- Climate, Environment and Forests
- Livelihood and Rural Development
- WASH
- Financial Inclusion
- · Skills and Training
- Art and Culture
- Rural Development and Poverty Alleviation
- Agriculture and Food Security
- Urban Governance
- · Drinking Water
- Human Rights
- Aged/Elderly
- Adolescent and Youth
- Child Protection and Early Child Development
- Mental Health
- Community Development
- Animal Husbandry
- Poultry

Distribution of SPOs in the DMA survey



Why is the ISDM CDSSI DMA Unique?

The ISDM CDSSI DMA framework offers a clear, practical way for SPOs to understand how they use data. It looks at two key types of data:

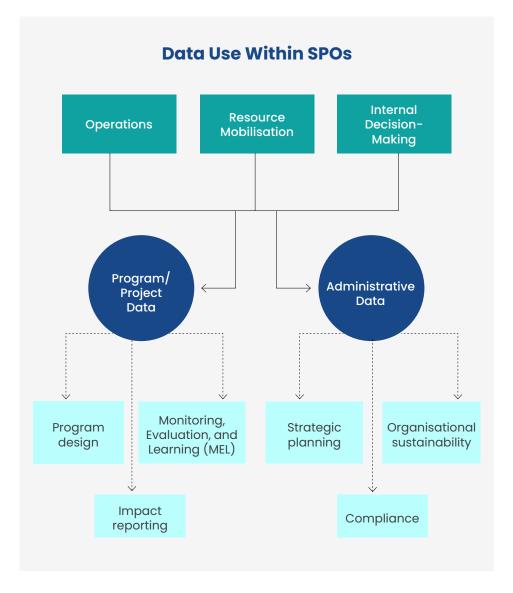
Program/Project Data: Information about beneficiaries, services, and measurable outcomes—essential for designing programs, monitoring progress, and demonstrating impact.

Administrative Data: Operational details like HR metrics and financials that support planning, organisational sustainability, and compliance.

The framework measures how well SPOs manage data across all areas—from fundraising and HR to program design and evaluation.

It builds on global and national models (like Deloitte's Insight-Driven Organisation Framework, IBM's Data Governance Maturity Model, and NITI Aayog's Data Governance Quality Index), but has been tailored specifically for Indian SPOs, making it unique.

Instead of a one-size-fits-all score, the ISDM CDSSI DMA acts as a mirror—helping organisations assess their current practices, spot strengths, and plan realistic improvements. The framework was carefully shaped through consultations with experts, field practitioners, and SPO leaders to reflect ground realities.



The ISDM CDSSI DMA is the largest and most diverse study of data maturity in the Indian social sector. It provides a comprehensive analysis of data management practices, organisational culture, capacities, and challenges faced by SPOs. The study captures varying levels of data maturity and offers targeted resources to support improvementsparticularly for resourceconstrained SPOs.

A foundational idea of the study is that culture and mindset are central to data maturity. Strengthening data practices is not just about adopting technology, but also about fostering a datainformed organisational culture. The study identifies practical steps for SPOs to embed data use into their systems and processes, addressing both infrastructure needs (such as tools and storage) and capacity requirements (such as skilled personnel and leadership commitment).



Project Data Analysis and Visualisation

Turning data into insights that inform decisions



Project Data Collection and Monitoring

Systems for gathering program data across its life cycle



Training and Capacity Building

Opportunities to improve data literacy and analytical capabilities

Culture and Mindset

How much an organisation values, promotes data use, and fosters a culture of learning



Six dimensions of the DMA framework



Human Resource Data Management

How well HR data is handled to support staffing decisions and workforce planning



Data Storage and Accessibility

Secure, organised data storage and easy access

Methodology

The ISDM CDSSI DMA adopts an Exploratory Sequential Mixed Methods Approach, in which the quantitative survey was followed by qualitative interviews. This approach facilitated a deeper exploration of the findings, and allowed for the validation of unexpected trends or patterns identified in the quantitative data.

We designed a detailed questionnaire to assess key areas such as Culture and Mindset, Leadership, Data use in Programs and Operations, Tools and Technologies, and Skills. SPOs were scored across five maturity levels—from Basic (0-20) to Expert (81-100)—to assess how deeply data is integrated into their work. We tested the questionnaire in a pilot to ensure it was clear, relevant, and reliable. Feedback from the pilot helped us improve the questionnaire by adding sections to better capture the complexities SPOs face, while keeping the core ISDM CDSSI DMA framework intact.

The NGO Darpan database, a comprehensive online registry of Non-Profit Organisations in India maintained by the National Institution for Transforming India (NITI Aayog), which has information on over 4,00,000 registered SPOs, served as the sampling frame for this study. Using quota sampling, we selected a representative group of 1,700 SPOs across India, of which 360 SPOs—covering different regions, budgets, sizes, and thematic areas—completed the survey (Fig. 1-4).

01

02

Ouantitative Tool Development

Based on ISDM CDSSI DMA Framework

Pre-testing and Refinement

- · Clarity, reliability tested
- Pilot-informed refinement
- Sections modified based on feedback

04

03

Quota Sampling

- Adjusted for region, size, sector
- · Ensured diversity and proportionality

Ouantitative Sampling

- NGO Darpan database
- ~16,000 SPOs contacted

05

Final Sample

1700 SPOs stratified by 06

Data Collection

360 SPOs completed the survey—either



The quantitative survey was followed by qualitative interviews that added depth and nuanced understanding to the quantitative insights. We conducted 17 in-depth interviews with senior leaders from a mix of organisations. These conversations revealed how SPOs think about and use data, as well as the challenges they face.

While the ISDM CDSSI DMA offers valuable insights into the data practices of SPOs in India, the study had a few limitations in terms of its methodology. The major challenge we faced during the study was a limited sample size, due to a lower response rate. To mitigate regional concentration and skewed distribution, we employed a combination of targeted outreach and snowball sampling to reach SPOs across diverse geographies and segments. This helped us in improving the representational spread of the survey.

A Detailed Look at the Sample Profile

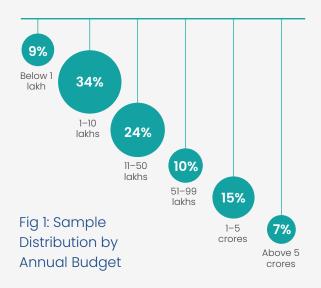


Fig 2: Sample Distribution by Size of Organisation

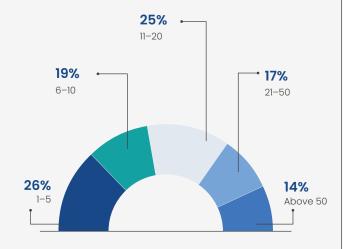


Fig 3: Sample Distribution by Age of Organisation

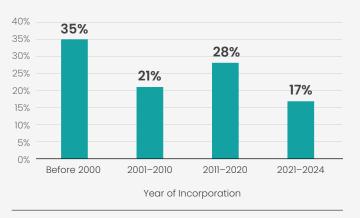
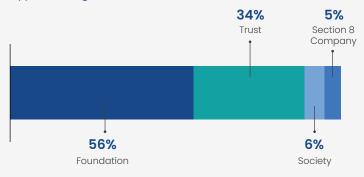


Fig 4: Sample Distribution by Type of Organisation



Of the organisations that participated in the survey, 31% are based in the Northern region, while 35% were established prior to the year 2000. 70% employ fewer than 50 staff members, and 68% reported an annual expenditure of less than ₹50 lakh.

Key Insights From the ISDM CDSSI DMA

1. The Majority of the Sector is Stuck in the Middle

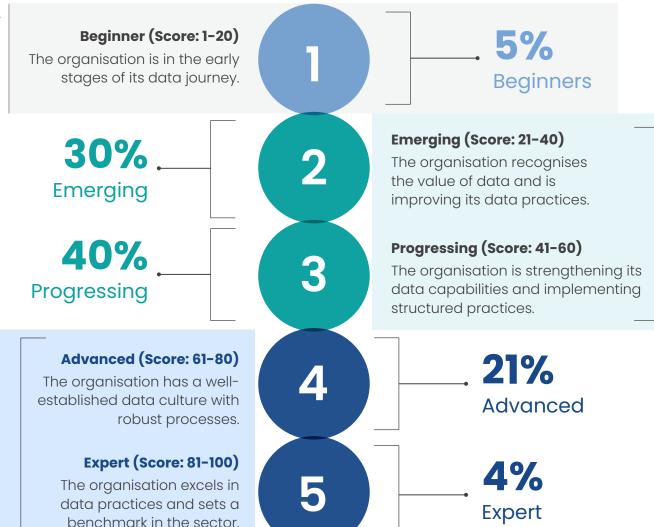
The ISDM CDSSI DMA classifies SPOs into five stages of data maturity based on their scores.

25%

of SPOs are in the Expert

categories.

and Advanced



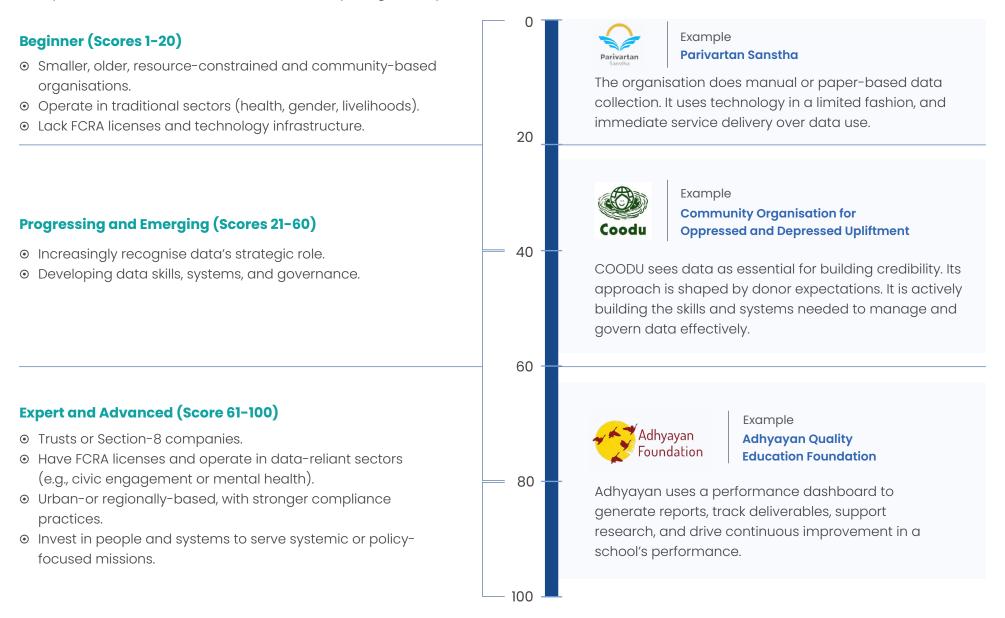
70%

of SPOs in the middle majority

in transition.

reflects a sector

The table below outlines the defining characteristics of SPOs at each level of data maturity, with examples of how data use varies across the spectrum—from foundational efforts to fully integrated systems:



2. From Awareness to Action, Data Integration Remains a Gap

Data Maturity

The sector-wide average

decision-making.

Data Operations

Challenges in accessibility, security, and staff training.

Talent Shortage

Uncompetitive salaries hinder the development of in-house expertise.

Donor-Driven Practices

Emphasis on compliance over strategic decisionmaking.

data maturity score stands at

48 out of 100. An analysis of across the six dimensions of the ISDM CDSSI DMA framework reveals a crucial insight: while the social sector increasingly recognises and values data, it has yet to embed it meaningfully into everyday ---> Limited Data Use

Focus on basic metrics restricts strategic insight and learning.

Insufficient Donor Support

Limited backing for long-term data capacity building.

Funding Constraints

Lack of resources for strong data systems and tools.

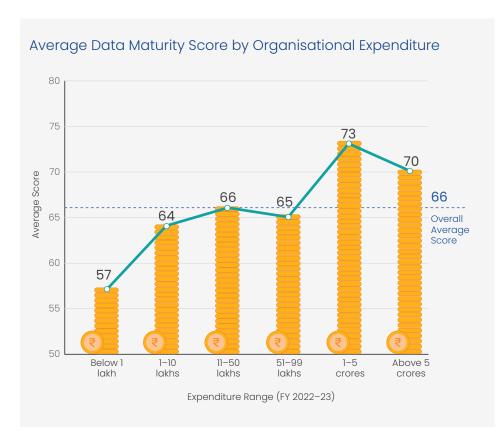
Encouragingly, over 80% of organisations report strong scores in aspects such as recognising data's importance, prioritising its use, and applying it internally. However, this growing awareness does not consistently translate into integrated, organisation-wide practices. For instance, while 84% of organisations use data for program or project design, a smaller percentage (59%) apply it to operational domains like fundraising, and even fewer (55%) apply it to financial management and HR services. Only 28% of organisations report using data comprehensively across all core functions, revealing a gap between intention and implementation.

is the average data maturity score of the sector

3. Financial Resources Drive Data Culture

An analysis of data maturity scores across different organisational facets reveals a clear trend: organisations with greater financial resources tend to have significantly stronger data cultures.

Those spending between ₹1 crore and ₹5 crore annually scored an average of 73, compared to an average score of just 57 for organisations spending less than ₹1 lakh.



This suggests that financial capacity enables investment in technology, systems, and staff training—key enablers of data maturity. Variables such as an organisation's age, workforce size, and region had little to no bearing on data maturity.

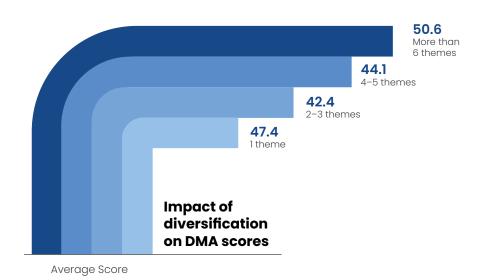
However, funding constraints remain a common challenge across organisations of all sizes in the social sector. Limited operational budgets restrict access to the infrastructure needed for effective data use in areas such as HR and financial systems. As a result, smaller organisations often rely on manual methods, prioritising immediate program delivery over long-term data strategies. For example, Dr Mahmood Alam Educational Trust shared that additional funding would directly improve their capacity, service quality, and data-informed decision-making. On the other hand, COODU (Community Organisation for Oppressed and Depressed Upliftment) was able to transition from traditional data collection methods to digital approaches over the past 20–30 years when aided by external guidance and funding. Funding helped the organisation invest in data systems to centrally monitor its data, improve data analysis, and make efforts to upskill its employees.

Ultimately, expenditure is a key lever to shape the mindset and culture needed to make data a strategic asset, not just a reporting tool.

4. Thematic Focus Shapes Data Practices

Thematic focus plays a significant role in shaping an organisation's data maturity. SPOs working in areas like education and nutrition tended to score higher on the DMA. In contrast, organisations working in areas such as animal husbandry, arts and culture, or rural livelihoods, tended to score lower. These fields are populated by smaller, resource-constrained organisations with lower data capacity.

Interestingly, the analysis also reveals that greater thematic diversification correlates with higher data maturity. Organisations operating in more than six thematic areas achieved a higher average score of 51 compared to those working across two to five themes (scores between 42 and 44).





These findings suggest that both the nature and breadth of an organisation's work meaningfully influence its ability to build and sustain data practices. In short, what you do—and how much of it—shapes how well you use data.

Recommendations

| Practical Tips For | Build a Strong Data Culture | Expand and Diversify Data Use |
|--------------------|--|---|
| | A culture that values data is foundational to widespread, routine use of data for decision-making, beyond siloed teams. | Data should inform decisions across functions, from program strategy to fundraising, not just compliance. |
| SPOs | Ensure leadership explicitly champions data use. Create cross-functional/cross-departmental forums to reflect and discuss data insights regularly. Designate internal data champions across departments. | Integrate data use into planning, budgeting, fundraising, and communications. Train teams to ask the right questions from their data. Move beyond outputs to outcome and impact tracking. |
| Funders | Include data culture indicators in grant assessments. Fund leadership workshops focused on building data-informed thinking. | Encourage reporting beyond output metrics. Support multi-use data systems that serve both internal learning and donor reporting. |
| Intermediaries | Develop case studies showing how strong data cultures drive impact. Design toolkits on internal change management around data. | Offer training on adaptive data use for decision-making. Share templates on how to embed data into strategic reviews. |

| | | (₹) |
|--------------------|---|--|
| Practical Tips For | Strengthen Capacity and Literacy | Invest in Governance and Compliance |
| | Without the right skills and understanding, even simple data systems remain underutilised. | Without data security, privacy, and governance, SPOs risk operational disruption and legal non-compliance. |
| SPOs | Invest in basic data literacy for all staff, not just M&E teams. Use free and open-source tools (e.g., KoBo, Airtable, Power BI) and build peer learning groups to reduce costs. Allocate small budgets annually for data-related training. | Create simple data protection policies (e.g., backups, user access, encryption). Build awareness of the Digital Personal Data Protection Act. Develop data-sharing and consent protocols with beneficiaries. |
| Funders | Offer optional or pooled capacity-building grants alongside core grants. Cover training and tool licensing in project budgets. | Include governance as part of due diligence. Fund expert guidance for small SPOs to strengthen compliance. |
| Intermediaries | Create modular courses on data skills by maturity level. Facilitate online or regional cohorts for hands-on learning. | Publish easy-to-use compliance toolkits and templates. Host workshops on ethical data use and rights-based data governance. |

| Practical Tips For | Enable Inclusive and Participatory Data Practices | Promote Innovation in Visualisation and Communication |
|--------------------|---|--|
| | Inclusive data captures diverse voices and leads to more equitable program decisions. | Data must be accessible and engaging to influence decisions across internal and external stakeholders. |
| SPOs | Collect and analyse disaggregated data by segments. Integrate feedback loops into program cycles. Train field teams on participatory data collection methods. | Train staff in data storytelling and dashboard creation. Use tools like Canva, Power BI, or Tableau for simple, shareable insights. |
| Funders | Incentivise reporting on inclusion metrics and community engagement. Fund tech or tools that enable intersectional analysis (e.g., GIS mapping, mobile surveys). | Support visualisation and reporting tools in grants. Accept innovative formats like dashboards, podcasts, or short videos. |
| Intermediaries | Develop guidelines on inclusive M&E. Develop participatory data tools and platforms. | Offer templates for dashboarding and reporting. Run competitions or showcases on data storytelling. |

| Practical Tips For | Foster Shared Infrastructure and Collaboration | Advocate for Long-Term Data Investments |
|--------------------|--|---|
| | Smaller SPOs struggle to build and sustain independent systems-shared infrastructure can increase reach and efficiency. | Without the right skills and understanding, even simple data systems remain underutilised. |
| SPOs | Join peer alliances and knowledge collaboratives that share data templates and systems. Collaborate with peers to align on metrics and share learnings. | Build long-term data goals into organisational strategy. Make the case to donors for recurring investment in data. |
| Funders | Invest in shared platforms (cloud storage, CRM systems, dashboards). Support collaborative efforts for data interoperability across the ecosystem. | Provide multi-year grants with flexible components for data systems. Embed data as a strategic component in organisational development. Fund positions like data managers or analysts directly. |
| Intermediaries | Create standardised indicators and protocols. Facilitate multi-stakeholder dialogues on data ecosystems. | Develop ROI models showing cost-effectiveness of data investments. Advocate for pooled grants or endowments for sector-wide strengthening of data maturity. |



Earlier, progress was tracked manually using pen and paper. Today, digital tools are used. However, the organisation still struggles with standardising tools and gaining acceptance among field staff.

43-year-old grassroots organisation working with marginalised communities in NE India



Funders and partners prefer organisations with robust data systems, as they provide confidence in accountability and impact delivery.

4-year-old social enterprise supporting women micro-entrepreneurs



Enhanced data quality directly correlates with organisational credibility and donor trust.

30-year-old organisation implementing programs in health and environment in Southern India

Data Practices Across Dimensions

Organisational Culture and Mindset

84.4% of organisations leverage data primarily for program or project design

57%

of organisations have leadership that invests in datadriven processes

27.8%

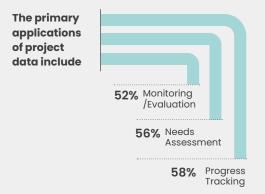
of organisations use data for all decision-making

Human Resource Data Management



Project Data Collection and Monitoring







Data Accessibility and Storage

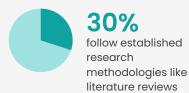


15%

of organisations conduct regular security audits and updates 53%

of organisations rely on ad hoc meetings and discussions for internal knowledge sharing

Project Data Analysis and Visualisation



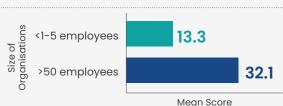


have dedicated inhouse data science teams and also consult external experts for specific projects

Data-related Training

34% of organisations report investment in data training

Organisational capacity strongly correlates with internal data competencies



Data Practices Across Functional Areas

Financial Management 41% 39% of organisations use data of organisations to identify use data to cost-saving assess donor opportunities preferences and behaviour 36% 35% of organisations of organisations use data to use data to assess financial analyse the risks and effectiveness of opportunities past fundraising

Internal Learning

of organisations use data extensively for internal learning and evaluation

of organisations encourage data-related discussions with peers and seniors

50% of organisations do not have an open data policy

of organisations use systematic knowledge-sharing tools: webinars, portals, databases

of organisations have frequent sessions conducted by team members for sharing knowledge and learnings

Organisations with higher data maturity scores have a plan for regular data training.

Program/Project Design and Implementation











Conclusion

The ISDM CDSSI DMA underscores a growing recognition of data's strategic value among Indian SPOs. But this awareness has not fully translated into the systems, skills, or governance mechanisms required for meaningful use. Targeted interventions that build foundational capabilities—particularly in under-resourced organisations—can significantly elevate sector-wide data maturity.

This research contributes a practical, replicable five-stage framework for assessing data readiness in the social sector. It offers both funders and practitioners a roadmap for capacity-building. It also provides a practical framework that can inform similar efforts globally. Other countries with resource-constrained social sectors can adapt these insights to assess and strengthen their own data ecosystems.

For funders and policymakers, the findings advocate for a sequenced approach: strengthen core competencies before advancing to predictive analytics or machine learning.

For SPOs, the framework serves as a diagnostic tool to assess their preparedness for deploying data science tools effectively.

The study opens up possibilities for comparative analysis, helping uncover what works, where, and why. It also lays the groundwork for building shared principles around data use, governance, and impact in the non-profit space.

It is our hope that this research contributes to a data transformation in the Indian social sector, enabling SPOs to become data-forward and Al-ready. We invite you to engage with us by taking the survey to gauge your own data readiness, or sharing insights, anecdotes and data that can further strengthen the sector's understanding of data maturity.

Appendix I — Detailed Methodology

We employed an Explanatory Sequential Mixed Methods Approach, where the quantitative survey was followed by qualitative interviews.

Quantitative Methodology

Quantitative Tool Development

We developed the questionnaire for the survey after several rounds of deliberations and discussions, both internally and with external experts, ensuring alignment with the ISDM CDSSI DMA framework throughout the process. Drawing from the principles of the framework, the tool was designed to comprehensively assess key aspects of data maturity, including culture and mindset, leadership, program and operations, data practices, tools and technologies, and skills. We then pre—tested the questionnaire in a pilot survey to ensure its clarity, relevance, and reliability. Feedback from the pilot phase was used to refine the survey questionnaire further, which led to the incorporation of additional sections to better capture the nuances of data maturity across different SPOs.

Quantitative Sampling

The NGO Darpan database, a comprehensive online registry of Non-Profit Organisations in India maintained by the National Institution for Transforming India (NITI Aayog), which has information on over 4,00,000 registered SPOs, served as the sampling frame for this

study. From the list, we contacted a random sample of 15,000—20,000 organisations for the survey. The calls were distributed proportionally across states to ensure that they reflect the real distribution of NGOs in India. Of the contacted organisations, about 1,700 consented to participate in the survey, forming the final sampling frame for the study. Using quota sampling, we selected a representative group of 360 implementing organisations across India, covering different regions, budgets, sizes, and thematic areas.

We identified regional imbalances, such as under-representation in the Southern region and over-representation in the Northern region. To address this, we used quota sampling to adjust the sample, ensuring accurate regional representation by targeting contacts from the ISDM registration campaign and the NGO Darpan database.

Similarly, we considered an organisation's employee size using quota sampling to ensure proportional representation of small, medium, and large organisations. We gave special attention to larger organisations (100+ employees) to understand their data systems, making adjustments to accurately reflect their presence in the final sample.

Additionally, sectoral representation posed a challenge, as many organisations reported working across multiple sectors during the

consent process, making it difficult to use as a sampling criterion. We made special efforts to ensure that the final sample of 360 NGOs/SPOs represented a diverse range of sectors.

Quantitative Data Collection

We collected responses from the participants of the quantitative survey through a combination of self-administered and telephonic interview methods to ensure a higher response rate and data quality. Interviews were conducted in both English and regional languages to accommodate respondents' comfort and comprehension. Trained enumerators facilitated the interviews, and they ensured clarity in understanding survey questions and accuracy in responses. For the self-administered component, we provided the participants with survey links and guidance on completing the questionnaire at their own convenience.

Qualitative Methodology

Qualitative Survey

Following the quantitative survey, we conducted a qualitative survey to validate and probe deeper into the insights that were generated. The qualitative survey questionnaire explored various aspects of data management and practices among SPOs, capturing insights into their organisational practices and the challenges they face across various key areas:

- Program Delivery and Challenges: Understanding challenges in program delivery, including gaps in resources, finance and data usage and identifying opportunities for improvement.
- Perceptions of Data and Information: Exploring how organisations view, value, and utilise data in operations and decision-making processes.
- Data Collection and Reporting: Examining the types of data collected, reporting mechanisms, and comparisons with sectorlevel practices to identify best practices.
- Data Management Challenges: Investigating issues in data collection, storage, and usage, and exploring potential solutions for enhancing data management systems.
- Client and Funder Perspectives: Assessing the role of data in interactions with stakeholders, particularly in securing funding or partnerships.
- Data Systems and Security: Understanding tools and technologies used for managing and securing data including adherence to data protection regulations.
- Sector-Level Improvements: Identifying opportunities for sectorwide advancements in data practices and exploring how organisations can contribute to driving these changes.

Qualitative Sampling

Of the 360 SPOs who participated in the quantitative study, 90 organisations consented to take part in the qualitative interview. These organisations were then contacted via telephone, with 17 SPOs participating in the final interviews.

Qualitative Data Collection

Seventeen Key Informant Interviews (KIIs) were conducted with respondents holding leadership roles within the organisations. They provided valuable insights into how organisations approach data usage for program delivery and decision-making. The interviews explored data management practices, including strengths, challenges, gaps, and strategies employed by the organisations, offering a deeper understanding beyond the quantitative data collected.

Data Validation

We validated the insights obtained from the quantitative survey by probing deeper into observed trends and anomalies, using the qualitative survey. This validation process strengthened the reliability of the overall quantitative analysis, ensuring a more accurate and comprehensive assessment of data practices and maturity across the surveyed organisations.

Scoring

The scoring assessed the data maturity of SPOs across five levels—Beginner (0–20), Emerging (21–40), Progressing (41–60), Advanced (61–80) and Expert (81–100).

Scores were assigned to each question and section in the DMA quantitative tool based on predefined criteria, allowing for a comprehensive assessment of data maturity of SPOs. There were single option and multiple choice questions. Each question was scored out of a maximum of 5 points with the following criteria applied based on question type:

| Question Type | Scoring Method | Logic |
|----------------------------------|--|--|
| Likert Scale Question | Weights assigned in ascending order | Increasing weights reflect increasing levels of data maturity. Each response option is assigned a weight based on its relevance. |
| Multiple Response Question | Two distinct scoring methods: 1. Equal Weights: All options assigned equal weights, scaled to score out of 5. 2. Weighted Options: Scores assigned based on importance of each selected option. | Equal weight method: All options treated equally, and total score scaled to 5. Weighted method: The importance of each option is considered for scoring. |
| Combination of Options | Specific combinations scored differently | Certain combinations of selected options have different scores to represent varying levels of data maturity. |

After scoring each individual question, we grouped the scores section by section. We then standardised the total score for each section to a scale of 100. To ensure that each section was properly weighted, we assigned a value to each based on its importance in assessing data maturity. To calculate the overall data maturity score, we multiplied the score of each section by its assigned weight and then summed these weighted scores. It then resulted in a final data maturity score out of 100, providing a comprehensive measure of the organisation's overall data maturity level.

Benchmarking

We used the mean scores of the SPOs to assess their data maturity by comparing their performance with sector standards. These benchmarks provided SPOs with a clear understanding of their position within the sector and helped them identify their strengths and areas of improvement.

Appendix II — Detailed Questionnaire

Organisation and Respondent's Profile

| Questions | Answers | Code | Skip |
|---|---|------|------|
| Name of the SPO (under which it is registered) (in words) | | | |
| Address (in words) | | | |
| Type of Non-Profit | Foundation | 01 | |
| Organisation (Single | Trust | 02 | |
| Response) | Society | 03 | |
| | Cooperative | 04 | |
| | Section 8 company | 05 | |
| | Others (specify) | 99 | |
| Location of the | Andaman and Nicobar Islands | 1 | |
| Headquarters (Single Response) | Andhra Pradesh | 2 | |
| | Arunachal Pradesh | 3 | |
| | Assam | 4 | |
| | Bihar | 5 | |
| | Chandigarh | 6 | |
| | Chhattisgarh | 7 | |
| | Dadra & Nagar Haveli and Daman & Diu | 8 | |
| | Delhi | 9 | |
| | Goa | 10 | |
| | Gujarat | 11 | |
| | Haryana | 12 | |
| | Himachal Pradesh | 13 | |

| Questions | Answers | Code | Skip |
|-----------|-------------------|------|------|
| | Jammu and Kashmir | 14 | |
| | Jharkhand | 15 | |
| | Karnataka | 16 | |
| | Kerala | 17 | |
| | Ladakh | 18 | |
| | Lakshadweep | 19 | |
| | Madhya Pradesh | 20 | |
| | Maharashtra | 21 | |
| | Manipur | 22 | |
| | Meghalaya | 23 | |
| | Mizoram | 24 | |
| | Nagaland | 25 | |
| | Orissa | 26 | |
| | Puducherry | 27 | |
| | Punjab | 28 | |
| | Rajasthan | 29 | |
| | Sikkim | 30 | |
| | Tamil Nadu | 31 | |
| | Telangana | 32 | |
| | Tripura | 33 | |
| | Uttar Pradesh | 34 | |
| | Uttarakhand | 35 | |
| | West Bengal | 36 | |

| Questions | Answers | Code | Skip |
|---------------------------|---|------|------|
| Other States of Operation | Andaman and Nicobar Islands | 1 | |
| (Multiple Responses) | Andhra Pradesh | 2 | |
| | Arunachal Pradesh | 3 | |
| | Assam | 4 | |
| | Bihar | 5 | |
| | Chandigarh | 6 | |
| | Chhattisgarh | 7 | |
| | Dadra & Nagar Haveli and Daman & Diu | 8 | |
| | Delhi | 9 | |
| | Goa | 10 | |
| | Gujarat | 11 | |
| | Haryana | 12 | |
| | Himachal Pradesh | 13 | |
| | Jammu and Kashmir | 14 | |
| | Jharkhand | 15 | |
| | Karnataka | 16 | |
| | Kerala | 17 | |
| | Ladakh | 18 | |
| | Lakshadweep | 19 | |
| | Madhya Pradesh | 20 | |
| | Maharashtra | 21 | |
| | Manipur | 22 | |
| | Meghalaya | 23 | |
| | Mizoram | 24 | |
| | Nagaland | 25 | |
| | Orissa | 26 | |
| | Puducherry | 27 | |

| Questions | Answers | Code | Skip |
|-----------------------|-------------------------------|------|------|
| | Punjab | 28 | |
| | Rajasthan | 29 | |
| | Sikkim | 30 | |
| | Tamil Nadu | 31 | |
| | Telangana | 32 | |
| | Tripura | 33 | |
| | Uttar Pradesh | 34 | |
| | Uttarakhand | 35 | |
| | West Bengal | 36 | |
| Thematic Area of Work | Education and Literacy | А | |
| (Multiple Responses) | Nutrition | В | |
| | Health and Family Welfare | С | |
| | Women's Development and | D | |
| | Empowerment | | |
| | Climate, Environment and | Е | |
| | Forests | | |
| | Livelihood and Rural | F | |
| | development | | |
| | WASH | G | |
| | Financial Inclusion | Н | |
| | Skills and Training | I | |
| | Art and Culture | J | |
| | Rural Development and Poverty | K | |
| | Alleviation | | |
| | Agriculture and Food Security | L | |
| | Urban Governance | М | |
| | Drinking Water | N | |
| | Human Rights | 0 | |
| | Aged/Elderly | Р | |

| Questions | Answers | Code | Skip |
|---|--|------|--------------------|
| | Adolescent and Youth | Q | |
| | Child Protection and Early Child Development | R | |
| | Mental Health | S | |
| | Community Development | Т | |
| | Animal Husbandry | U | |
| | Poultry | V | |
| | Others (Specify) | Z | |
| When was your organisation established? (in year) | | | |
| Do you have an FCRA | Yes | 1 | |
| license? (Single response only) | No | 2 | If H=02, Skip J |
| When did you receive your FCRA license? | (in years) | | |
| Total number of employees (in numbers) | Total Employees (excluding volunteers and interns) | | |
| | Volunteers and Interns | | |
| Annual expenditure of the organisation for the latest | 2022-23 () in Lakhs | | |
| year which you can provide (in INR and Lakhs) | 2023-24 () in Lakhs | | |

| Questions | Answers | Code | Skip | |
|---|-----------------------------|------|------|--|
| Respondent Profile | | | | |
| Contact Number (in numbers) | | | | |
| Name of the Respondent (full name in words) | | | | |
| Position/Role in the Organisation (in words) | | | | |
| Highest Education Level | Never attended school | 01 | | |
| (Single Choice) | Primary (till class V) | 02 | | |
| | Upper Primary (till VIII) | 03 | | |
| | Higher Secondary (till XII) | 04 | | |
| | Graduate | 05 | | |
| | Post-Graduate and Above | 06 | | |
| Official Email Id (in words) | | | | |
| How long have you been associated with the current organisation? (in years) | | | | |

Section B: Culture and Mindset

Please respond to the following questions based on your organisation's perspective for both project and administrative (HR and Financial) data. Project data refers to information related to the projects/programs that your organisation has conducted or is conducting which includes activities such as data collection, storage, analysis, and visualisation. Administrative data includes human resource/employee, financial and fundraising data of the organisation.

This section is a mix of both: single and multiple response-based questions.

| Questions | Answers | Code | Skip |
|---|---------|------|------|
| On a scale of 1 to 5, how important is | 1 | 01 | |
| data considered in your organisation? (Single Response) | 2 | 02 | |
| (Single Response) | 3 | 03 | |
| | 4 | 04 | |
| | 5 | 05 | |
| On a scale of 1 to 5, to what extent is data considered an organisational priority currently? (Single Response) | 1 | 01 | |
| | 2 | 02 | |
| | 3 | 03 | |
| | 4 | 04 | |
| | 5 | 05 | |
| On a scale of 1 to 5, to what extent | 1 | 01 | |
| is data used for internal learning, evaluation, and to identify needs and problems? (Single Response) | 2 | 02 | |
| | 3 | 03 | |
| | 4 | 04 | |
| | 5 | 05 | |

| Questions | Answers | Code | Skip |
|---|---|------|-----------------------|
| For what purposes does your organisation use data? (Multiple Responses) | Financial and HR Management | А | If Q4=A, ask Q5 |
| | Fundraising Management | В | If Q4=B, ask Q6 |
| | Program/Project Design | С | If Q4=C, ask Q7 |
| What role does data play in guiding financial and HR management decisions? (Multiple Responses) | To provide insights into revenue and expenditure, turnover, performance evaluation. | А | |
| | To identify cost-saving opportunities and recruitment optimisation | В | |
| | To assess financial risks and opportunities | С | |

| Questions | Answers | Code | Skip |
|--|--|------|------|
| What role does data play in informing fundraising decisions? (Multiple Responses) | To assess donor preferences and behaviour | А | |
| | To identify potential donors | В | |
| | To analyse the effectiveness of past fundraising | С | |
| | To estimate donation trends and forecasting future contributions | D | |
| What role does data play in informing program/project design decisions? (Multiple Responses) | To tailor programs to specific stakeholder needs (beneficiaries, and donors) | А | |
| | To evaluate past program performance for effectiveness | В | |
| | To allocate resources efficiently based on data insights. | С | |
| On a scale of 1 to 5, how often do | Never | 01 | |
| employees in your organisation | Rarely | 02 | |
| discuss about topics related to data (both project and administrative | Sometimes | 03 | |
| data) with their peers and senior | Most of the time | 04 | |
| management? (Single Response) | Always | 05 | |
| On a scale of 1 to 5, to what extent is | 1 | 01 | |
| data shared with other organisations/ | 2 | 02 | |
| external parties [(excluding donors) (to the extent allowable under privacy | 3 | 03 | |
| laws)]? | 4 | 04 | |
| | 5 | 05 | |

| Questions | Answers | Code | Skip |
|--|--|------|------------------------|
| On a scale of 1 to 5, to what extent are | 1 | 01 | |
| leaders proficient in understanding | 2 | 02 | |
| and interpreting data? | 3 | 03 | |
| (to be answered only if the respondent is a senior resource person/leader in the organisation) (Single Response) | 4 | 04 | |
| | 5 | 05 | |
| On a scale of 1 to 5, to what extent | 1 | 01 | If Q11=1 |
| are leaders willing to invest resources | 2 | 02 | or 2, |
| (time, money, effort) into data-driven | 3 | 03 | ask |
| practices and solutions? | 4 | 04 | Q12, |
| (to be answered only if the respondent is at mid-management level/someone who is not a leader/owner in the organisation) (Single Response) | 5 | 05 | else skip to Q13 |
| What factors might hinder the leaders' willingness to incorporate data-driven practices? (Multiple Responses) | Given that the available tools and technologies within the organisation are sufficient, there is no need to incorporate data-driven practices. | А | |
| | Lack of understanding of data analytics concepts | В | |
| | Resistance to change from employees | С | |
| | Insufficient financial resources for implementation | D | |
| | Don't Know | Х | |
| | Others (please specify) | Z | |

Section C: Human Resource/Employee Data

Please respond to the following questions based on your organisation's practice in handling human resource/employee data.

This section is a mix of both: single and multiple response-based questions.

| S. No | Questions | Answers | Code | Skip |
|-------|--|--|------|----------------|
| 13. | Which of the following best describes | We DON'T store and maintain human resource/employee data. | 01 | If Q13=01, |
| | your organisation's approach to managing human | We rely only on manual records to manage employee information (Paper based). | 02 | skip to Q17 |
| | resource/employee data? (Single Response) | We use excel spreadsheets to manage basic employee information (contact info. etc.). | 03 | |
| | | We outsource employee data management to third party HR service providers. | 04 | |
| | | We use spreadsheets/third party agencies to collect and manage detailed employee information (performance, compensation and benefits, training data, ratings. etc.). | 05 | |
| | | We maintain a software-based database where all employee information is stored and regularly updated. | 06 | |
| | | Don't Know | 98 | |

| S. No | Questions | Answers | Code | Skip |
|-------|---|--|------|----------------|
| 14 | For what purpose is the human resource/ | To assess employee job satisfaction and training needs | А | |
| | employee data analysed? | To evaluate the effectiveness of recruitment efforts | В | |
| | (Multiple Responses) | For retention and turnover analysis | С | |
| | | For workforce planning and strategic decision-making | D | |
| | | To track employees' attendance and payments | Е | |
| | | We DON'T analyse HR data | F | |
| | | Don't Know | Х | |
| 15 | How often is the | Not Analysed | 01 | If Q14=F, |
| | human resource/ employee data analysed? (Single Response) | Monthly | 02 | skip to Q16 |
| | | Quarterly | 03 | |
| | | Half Yearly | 04 | |
| | | Annually | 05 | |

| S. No | Questions | Answers | Code | Skip |
|-------|--|--|------|------|
| 16 | What factors determine your | For immediate operational needs and vacancies | 01 | |
| | organisation's hiring strategies? (Multiple Responses) | For current project requirements | 02 | |
| | (matapie respenses) | To meet mission goals and strategic objectives of the organisation | 03 | |

| S. No | Questions | Answers | Code | Skip |
|-------|-----------|---|------|------|
| | | Data-driven analysis and forecasting determine future hiring needs Feedback from stakeholders guide | 04 | |
| | | hiring strategies Others (please specify) | 98 | |

Section D: Project Data Collection and Monitoring

This section is a mix of both: single and multiple response-based questions.

| S. No | Questions | Answers | Code | Skip |
|-------|--|--------------------------------|------|----------------------|
| 17 | Who are generally involved in planning | We don't collect project data | А | If Q17=A, skip to |
| | and managing data collection in your organisation? | Project Managers | В | Q27 |
| | (Multiple Responses) | Community/Field Staff | С | |
| | | Monitoring and Evaluation Team | D | |

| S. No | Questions | Answers | Code | Skip |
|-------|-----------|------------------------|------|----------------------|
| | | Research Team | E | If Q17=A, skip to |
| | | Volunteers | F | Q27 |
| | | Interns | G | |
| | | External/Third Parties | Н | |
| | | Don't Know | X | |

| S. No | Questions | Answers | Code | Skip |
|-------|---|--|------|------|
| 18 | Does your organisation | NO skilled staff is present | 01 | |
| | have appropriate | Only a few staff are skilled | 02 | |
| | number of skilled staff for collecting and managing project data? (Single Response) Each option provides a different level of detail regarding the organisation's approach to data collection and management, from the absence of skilled staff to high-level staff | Dedicated person/team in charge is present (e.g., a data manager or senior administrator) | 02 | |
| | | Dedicated skilled analytics roles established, with multiple team members responsible for data collection, analysis, and interpretation to guide decisionmaking. | 03 | |
| | | High level staff commitment across senior, specialist, technical, and administrative levels, with clearly defined roles | 04 | |
| | | Don't Know | 98 | |
| 19 | Who uses project data | Senior Management | А | |
| | in your organisation? (Multiple Responses) | Monitoring and Evaluation Team | В | |
| | | Project Managers | С | |
| | | Finance Team | D | |
| | | Fundraising Team | F | |
| | | IT Department | G | |
| | | Don't Know | Х | |
| | | Others (Specify) | Z | |

| S. No | Questions | Answers | Code | Skip |
|-------|--|---|------|------|
| 20 | For what purposes are | For client reporting purpose | А | |
| | project data collected | For monitoring and evaluation | В | |
| | and used? (Multiple Responses) | For resource allocation | С | |
| | | For needs assessment | D | |
| | | For risk management | Е | |
| | | For evidence-based program planning to decide future course of action | F | |
| | | To track the progress | G | |
| | | To understand stakeholder/ beneficiary needs and preferences | Н | |
| | | Don't Know | Х | |
| 21 | Does your organisation collect and record | Only limited data is collected manually, mostly on paper. | 01 | |
| | data in consistent and efficient ways? (Single Response) | Data collection is both on paper and in digital form (includes emails, audio and video recordings), however not consistent and structured at this moment. | 02 | |
| | | Data collection is getting structured, and processes are being improved to make it more organised and integrated. | 03 | |
| | | Data is systematically and consistently collected via a range of methods. | 04 | |

| S. No | Questions | Answers | Code | Skip |
|-------|--|---|------|------|
| | | Staff are well trained in data collection and data is collected systematically and is automated, wherever possible. | 05 | |
| 22 | What are the | Internal program data only | 01 | |
| | sources of data your organisation uses? | Secondary research data only | 02 | |
| | (The source can be any primary research/ evaluation report and data or use of | Frequently uses both internal data and relevant external/secondary sources to corroborate findings. | 03 | |
| | publicly available external research, e.g., government, academic research, use of publicly available open data sets) (Single Response) | Data systematically collected through primary research, with data collection typically outsourced to specialised agencies. | 04 | |
| | | Utilises various data sources (internal, secondary, primary research) as per requirement in a methodological and systematic manner, with proficiency in leveraging open- source data. | 05 | |
| 23 | Has the organisation identified quantifiable | We DON'T have quantifiable monitoring indicators. | 01 | |
| | indicators for monitoring project/ program outcomes? (Single Response) | Organisation has started identifying clear monitoring goals. | 02 | |

| S. No | Questions | Answers | Code | Skip |
|---------------|---|--|------|------|
| | | The organisation insists on identifying monitoring indicators for each program goal. However, these indicators are not very consistent and reliable. | 03 | |
| | | Clear monitoring indicators are identified for each program goal, and they are reliable. | 04 | |
| | | Monitoring indicators are set for each goal at the onset of the program and are aligned to the theory of change. | 05 | |
| | | Don't Know | 98 | |
| 24 | How often is the | Weekly | 01 | |
| | project data collected | Fortnightly | 02 | |
| | and monitored? | Monthly | 03 | |
| | (Single Response) | Quarterly | 04 | |
| | | Half-Yearly | 05 | |
| | | Annually | 06 | |
| 25 | Are you able to collect the right or relevant | Never | 01 | |
| | data? ('Right data' i.e., relevant, meaningful and necessary) | Rarely | 02 | |
| | | Sometimes | 03 | |
| (Single Respo | (oii i gie kesponse) | Most of the time | 04 | |

| S. No | Questions | Answers | Code | Skip |
|-------|--|---------|------|------|
| | | Always | 05 | |
| 26 | Does your organisation have an "Open Data Policy"? | Yes | 01 | |

| S. No | Questions | Answers | Code | Skip |
|-------|---------------------|---------|------|------|
| | (Single Response) | No | 02 | |
| | Examples of | | | |
| | external data are | | | |
| | publicly available | | | |
| | external research | | | |
| | like government, | | | |
| | academic research, | | | |
| | use of publicly | | | |
| | available open data | | | |
| | sets. | | | |

Section E: Project Data Storage and Accessibility

This section is a mix of both: single and multiple response-based questions.

| S. No | Questions | Answers | Code | Skip |
|----------|--|---|------|---|
| 27 | What are the tools/ technologies for storing project data? (Multiple Responses) | We DON'T store data. | | If Q27= B, skip to Q40 If Q27=A, skip to Q34 |
| | | In the form of physical/hard copies, files, and documents | В | |
| | | Locally stored in computers | С | |

| S. No | Questions | Answers | Code | Skip |
|----------|---|--|------|------|
| | | Other devices (external hard drives, memory sticks, cameras, phones, etc.) | D | |
| | | Central shared storage (e.g., cloud-based storage, etc.) | E | |
| | | Don't Know | Χ | |
| 28 | How do you secure your data? (Multiple Responses) | We DON'T have any mechanism to secure data. | 01 | |

| S. No | Questions | Answers | Code | Skip |
|----------|---|--|------|------|
| | | By storing data on secure, password protected servers | 02 | |
| | | By enabling two factor authentication | 03 | |
| | | By implementing encryption methods to protect sensitive information | 04 | |
| | | By conducting regular security audits and updates | 05 | |
| 29 | How frequently is data | Never | 01 | |
| | security tested? (Single Response) | Monthly | 02 | |
| | (Single Response) | Half-Yearly | 03 | |
| | | Annually | 04 | |
| | | Whenever there is an issue | 05 | |
| 30 | Do you know about | Yes | 01 | |
| | Digital Personal Data Protection Act, 2023? (Single Response) | No | 02 | |
| 31 | Does the organisation have an internal knowledge sharing system in place for sharing information and learnings with everyone in the organisation? (Multiple Responses) | There is NO existing knowledge sharing system accessible for everyone. | А | |
| | | Knowledge sharing happens through ad hoc meetings and discussions occasionally. | В | |
| | | There are frequent sessions conducted by team members for sharing knowledge and learnings, but these are not systematised. | С | |

| S. No | Questions | Answers | Code | Skip |
|----------|---|--|------|------|
| | | Regular webinars/talks conducted to share learnings and best practices with the larger organisation. This is done in a systematic way. | D | |
| | | Knowledge portal/centralised database in place to share learnings where employees can access knowledge-based material as per their need. | E | |
| 32 | What is the strategy | There is NO such policy in place. | 01 | |
| | around deletion of obsolete Project data? (Single Response) By obsolete data we | There is awareness that obsolete data should be deleted. But it is NOT monitored or enforced. | 02 | |
| | mean the information which is no longer useful or relevant to the organisation's current needs. | Obsolete data is deleted often. However, some old unwanted data may still remain in the system. | 03 | |
| | canoni necas. | Systems are in place to ensure obsolete data is regularly deleted. | 04 | |
| | | System has automated process to detect and delete obsolete data on a regular basis. | 05 | |
| 33 | Does your organisation have policies and | Yes | 01 | |
| | procedures in place for data backup and disaster recovery? (Single Response) | No | 02 | |

Section F: Project Data Analysis and Visualisation

This section is a mix of both: single and multiple response-based questions.

| S. No | Questions | Answers | Code | Skip |
|----------|---|---|------|------|
| 34 | Do you have a data analysis team for Project data? (Single Response) | Yes No | 01 | |
| 35 | Does your organisation employ or externally consult with someone who has data analysis or data | Organisation does NOT employ or consult individuals with data science skills | 01 | |
| | science skills? (Single Response) Each option provides a different level of | Organisation uses people with very basic data science skills | 02 | |
| | a different level of detail regarding the organisation's approach to data science expertise, ranging from the absence of such skills to employing highly skilled individuals. | Few individuals in specific roles possess adequate expertise, with occasional ad hoc hiring of external consultants for project-specific needs. | 03 | |
| | | Organisation maintains a small in-house team of data science experts and frequently consults external experts for specific projects. | 04 | |

| S. No | Questions | Answers | Code | Skip |
|----------|--|---|------|---------------------------|
| | | Organisation employs highly skilled individuals for advanced data analytics roles | 05 | |
| | | Don't Know | 98 | |
| 36 | Does your organisation perform evaluation or | Yes | 01 | If Q36=02, skip to Q38 |
| | research studies? (Single Response) | No | 02 | |
| 37 | 37 How are the research studies planned and designed in your organisation? (Multiple Responses) | Based on immediate needs and available resources | 01 | |
| | | Developed on a project-by- project basis with minimal reference to past studies | 02 | |
| | | Following established guidelines and methodologies, including literature review | 03 | |
| | | Driven by detailed past data analysis and best practices, with a dedicated team | 04 | |

| S. No | Questions | Answers | Code | Skip |
|----------|--|---|------|--------------------------|
| | | Apart from data analysis and thorough review of past studies, additionally use inputs from stakeholders and external validation | 05 | |
| 38 | What tools/technologies are used for analysing | We DON'T Analyse Project data. | А | If Q38=A, skip to Q40 |
| | Project data? (Multiple Responses) | Microsoft Excel and Google Sheets | В | |
| | | R Programming | С | |
| | | Python | D | |
| | | ATLAS.ti/NVivo | Е | |
| | | SPSS/STATA/SAS | F | |
| | | Don't Know | X | |

| S. No | Questions | Answers | Code | Skip |
|----------|--|---|------|--------------------------|
| 39 | What tools/technologies are used for reporting and | We DON'T report and visualise Project data. | А | If Q39=A, skip to Q40 |
| | visualising Project data? (Multiple Responses) | Tableau/Microsoft Power BI Dashboards | В | |
| | | Microsoft Power Point | С | |
| | | Canva | D | |
| | | Excel | Е | |
| | | R Programming | F | |
| | | STATA | G | |
| | | Don't Know | Х | |

Section G: Data-Related Training Activities

Please respond to the following questions on the basis of data-related training activities in your organisation.

This section is a mix of both: single and multiple response-based questions.

| S. No | Questions | Answers | Code | Skip |
|----------|--|---------|------|---------------------------|
| 40 | Does your organisation invest in data training | Yes | 01 | If Q40=02, skip to Q43 |
| | activities? (Single Response) | No | 02 | |

| S. No | Questions | Answers | Code | Skip |
|----------|---|--|------|------|
| 41 | Do the staff receive regular training on data protection, | NOT aware of data protection and security | 01 | |
| | security and regulations? (Single Response) | Little awareness or training in data protection and security | 02 | |

| S. No | Questions | Answers | Code | Skip |
|----------|--|---|----------------|------|
| | By training on data protection, security and regulations, we mean training to handle data breaches, subject access | Staff have basic data protection and security training but lack experiencing handling data breaches. | 03 | |
| | requests, changes in preferences on personal data. | Learning to ensure responsible and ethical practices | 04 | |
| | | Awareness about ethics, openness and protection of data are embedded throughout the organisation with regular training. | 05 | |
| | | Don't Know | 98 | |
| 42 | 42 How frequently does your organisation conduct training sessions on datarelated topics? (Single Response) | Weekly Fortnightly | 01 | |
| | | Monthly Quarterly Annually | 03 04 05 | |
| 43 | What factors hinder the organisation's commitment to investing in data training activities? | Lack of budget or resources Limited understanding of the potential benefits of data literacy | АВ | |
| | (Multiple Responses) | Reluctance to adopt new technologies | С | |
| | | Disinterest from employees | D | |
| | | Don't Know | X | |
| | | Others (Please specify) | | |

| | To be filled by the investigator/enumerator | | | | | |
|---|---|---------------------|--|--|--|--|
| R | Date of Interview (This will be auto recorded) | // | | | | |
| S | Start time of interview (This will be auto recorded) | : | | | | |
| Т | End time of interview (to be completed at the end of the interview) | : | | | | |
| U | The outcome of the call | Completed | | | | |
| | (to be completed at the end of the | Partially completed | | | | |
| | interview) | Not available | | | | |
| | | Refused | | | | |

Indian School of Development Management, supported by Citi India's CSR efforts, launched the Centre for Data Science and Social Impact (CDSSI) to connect Social Purpose Organisations (SPOs) with funders, academia, tech enablers and other experts, to drive rapid adoption of data and data science by the social sector

