

1 **Charting uneven progress of sustainability: A multi-dimensional assessment of the SDGs**
2 **in Northeast India**

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13
14 **Abstract**

15
16 Sustainable Development Goals (SDGs) are used to assess progress in the fields of the
17 environment, economy, and society. Although assessments conducted at national and
18 international levels are popular, subnational research, especially on India, is less common.
19 Using 84 accessible indicators (2021–2022), a comprehensive study of 15 SDGs was conducted
20 across 103 districts in eight states in the northeastern region (NER) of India. Pearson’s
21 correlation, hierarchical clustering, network analysis, input-oriented data envelopment analysis
22 (DEA), inequality (using the Theil and Atkinson indices), and relative SDG scoring were
23 performed using ‘R’. Four SDG groupings (environmental, social, economic, and
24 socioeconomic) and individual SDGs at the overall, district, and state levels were analysed.
25 The results showed that the SDGs were asynchronous and had significant inequalities among
26 the NER districts, with SDG 13 showing the highest disparity and SDG 2 the lowest. Many
27 districts performed worse socioeconomically, even when they had higher environmental
28 scores. This study provided the first comprehensive multi-dimensional assessment of SDG
29 progress across Northeast India at the district level, revealing critical disparities and complex
30 interactions between environmental and socioeconomic goals. Finally, limitations in mitigating
31 the drawbacks of the NER SDG framework were discussed, coupled with policy suggestions
32 for environmental, societal, and economic aspects. The findings offered valuable insights for
33 policymakers in designing targeted interventions, promoting balanced development, and
34 addressing regional inequalities to achieve sustainable development in this ecologically
35 sensitive and socioeconomically diverse region.

36
37 **Keywords:** sustainable development goals; clustering; regional inequality; evenness;
38 subnational; India;

39 **Introduction**

40 Sustainability is a complex concept that extends beyond resource management. This
41 requires a careful balance that promotes economic growth, social equality, and environmental
42 preservation for both current and future generations. The United Nations' SDGs, which provide
43 a roadmap for achieving sustainability, tackle global challenges such as poverty, health,
44 education, and climate change through 17 interconnected goals (2015-2030).

45 The global pursuit of the SDGs faces numerous challenges, including uneven progress
46 across regions, resource constraints, and the complex interplay between economic development
47 and environmental preservation. These challenges are particularly pronounced in developing
48 regions with diverse socio-economic and ecological landscapes. Northeast India exemplifies
49 these complexities, with its rich biodiversity, varied topography, and unique cultural tapestry,
50 presenting both opportunities and obstacles for sustainable development. The region's
51 remoteness, historical underdevelopment, and vulnerability to climate change make it a critical
52 case study for understanding the nuanced challenges of implementing SDGs in ecologically
53 sensitive and economically diverse areas. As such, examining the progress of SDGs in
54 Northeast India not only addresses local development needs, but also contributes to a broader
55 understanding of sustainable development challenges in similar contexts globally.

56 Nestled amidst the Himalayas, Northeast India boasts a tapestry of eight states:
57 Arunachal Pradesh, Assam, Manipur, Meghalaya, Mizoram, Nagaland, Sikkim, and Tripura.
58 They house 3.77% of India's population and constitute 7.98% of the country's land area. The
59 land is rich in natural resources, ranging from rivers and fertile plains to mineral deposits and
60 bamboo forests. These NER states are bordered by five neighbouring countries: Tibet (north),
61 Bangladesh (southwest), Nepal (west), and Bhutan (northwest).

62 Achieving the SDGs in Northeast India faces several unique roadblocks. Limited
63 connectivity restricts access to markets, healthcare, and education, thus hindering economic
64 growth and social progress (SDG 1, 8, and 10). Unsustainable practices, such as slash-and-burn
65 agriculture, threaten biodiversity (SDG 13, 15). Shifting cultivation, a traditional practice, is
66 under pressure owing to population growth and deforestation. Job creation and income
67 generation remain challenges, leading to poverty and migration (SDG 1, 8). Ethnic tensions
68 and insurgency movements have hampered development efforts. Empowering women and
69 ensuring inclusive growth for all sections of society remains a work in progress (SDG 5 and
70 16).

71 Despite these hurdles, there has been a growing movement towards sustainable
72 solutions. From promoting organic farming and bamboo-based industries to developing

73 ecotourism and utilising renewable energy, there is a collective will to chart a greener path. By
74 addressing the specific needs of the region and harnessing its unique strengths, Northeast India
75 can unlock its true potential and emerge as a model for sustainable development. This article
76 examines the intricate connection between sustainability in Northeast India and SDGs. This
77 study aims to inform policies, practices, and future research on sustainable development in the
78 region.

79

80 **Literature review**

81 There have been a handful of studies on the sustainability of northeast India over the
82 last few years. To date, some studies have focused on the SDGs in the Indian context
83 (subnational or national). Mitra (1998) studied the environmental sustainability of Arunachal
84 Pradesh. Chaudhuri and Roy (2017) analysed spatial inequality in WaSH (Water-Sanitation-
85 Hygiene) facilities using 2011 census data. None of the NER states are seriously lacking in
86 rural-urban equality. Bora and Saikia (2018) analysed neonatal mortality rate, NNMR and
87 under-5 mortality rate, U5MR (SDG 3) in Indian districts based on data from the National
88 Family Health Survey (NFHS 2015-16). Most districts of NER India are unlikely to achieve
89 NMR-Male by 2030; however, they are likely to achieve U5MR-Male by 2030. Singh (2018)
90 conducted a study on regional disparities in sustainable development in the NER States of
91 India. Roy & Pramanick (2019) assessed SDG 6 on a national scale, in the context of an
92 ecologically safe and socioeconomically just operating space framework. Chhetri (2020)
93 analysed the SDG for Sikkim. Tiwari and Krishna (2020) analysed the social, economic, and
94 environmental performance of 641 districts of India using a composite sustainability index
95 based on comparatively old data (2011 census). Chaudhary et al. (2022) composed a
96 subnational scale assessment of threats to Indian biodiversity via application of the species
97 threat abatement and restoration (STAR) metric for amphibians, birds, and terrestrial
98 mammals. Arunachal Pradesh, Assam, and Meghalaya are among the top nine states
99 contributing 80% of the national STAR score. Ghosh et al. (2022) have analysed WaSH
100 poverty in India at district level. Drinking water poverty is relatively more prevalent in NER
101 than sanitation and hygiene poverty. Anand et al. (2023) published a few studies on food
102 security, land degradation, gender equality, health, etc., focusing on NER India in their book.
103 Ghosh et al. (2023) have analysed spatial clustering of diarrhoea among children (<5 years) in
104 707 districts of India. The prevalence of NER is mixed. Roy et al. (2023) have assessed urban
105 sustainability of 56 prominent cities of India using 14 SDGs (from 77 indicators) for 2020–
106 2021. Subramanian et al. (2023) performed an SDG progress assessment of 707 Indian districts

107 based on 33 indicators (covering nine out of 17 UN-SDGs) sourced from the NFHS (2016,
108 2021). They concluded that four SDGs (viz., SDG 1-3, 5) require urgent attention. Studies that
109 have focused on the SDG of NEER India is very less. When a search was conducted on the Web
110 of Science for original articles published from 01-01-2015 to 30-04-2024 with NEER India-
111 related terms (see Supplementary file), only 35 results were found. This proves that even after
112 nine years of the commencement of the UN-SDG proposal and six years to accomplish (2024-
113 2030), very little research has been conducted focusing on this region. This study examines the
114 intricate connection between sustainability in Northeast India and SDGs. Specifically, this
115 study addressed the following research questions (RQs):

116

117 **RQ 1:** What are the achievements and shortfalls in NEER districts regarding SDG performance?

118 **RQ 2:** What are the interrelationships among NEER district features in terms of SDG progress?

119 **RQ 3:** How can NEER districts be grouped according to their environmental, social, and
120 economic characteristics?

121 **RQ 4:** What are the efficiencies in utilising environmental scores (env-SDGs) towards socio-
122 economic achievements (socio-econ-SDGs)?

123 **RQ 5:** What is the extent of spatial inequality among NEER districts in terms of SDG
124 performance?

125 **RQ 6:** How do Evenness and Mean Index Scores reveal disparities and guide equitable SDG
126 progress across the NEER districts?

127 **RQ 7:** What is the relative performance of NEER districts compared with state, national, and
128 global benchmarks?

129 By addressing these research questions, this study aims to inform policies, practices, and future
130 research on sustainable development in the region, provide a comprehensive assessment of
131 SDG progress in Northeast India, and offer insights for targeted interventions to achieve
132 balanced and sustainable development.

133

134 **Methodology**

135 The SDG scores (2021-22) of 120 districts (from eight Indian states) in Northeast India
136 were collected from the NITI Aayog (NITI Aayog, 2024). Two SDGs (viz., SDGs 14 and 17)
137 were not included because their overall scoring was not available in the dataset. Owing to the
138 unavailability of data, only 103 districts were considered. The sample of 103 districts
139 represented 85.83% of the total districts in the NEER, providing a comprehensive dataset for our
140 analyses. The sample size exceeded the minimum requirements for our statistical methods, thus

141 ensuring statistically significant results. The 84 indicators included in this study covered
142 various topics related to local sustainability. The dataset used in this study has significant
143 coverage of the official districts of the eight NER states. Share (%) of official districts included
144 in study were Tripura (100%), Meghalaya (91%), Arunachal Pradesh (89%), Assam (77%),
145 Mizoram (72%), Nagaland (68%), Sikkim (66%), and Manipur (56%) (Figure S.1 in the
146 Supplementary File 1). The state-wise distribution of 103 study districts were Assam (26%),
147 Arunachal Pradesh (24%), Meghalaya (10%), Nagaland (11%), Manipur (9%), Tripura (8%),
148 Mizoram (8%), and Sikkim (4%) (Figure S.2).

149 Pearson's correlation coefficient was selected to investigate the interrelationships
150 between SDGs owing to its capacity to quantify the strength and direction of linear associations
151 between continuous variables, which is appropriate for SDG score data. It addressed our RQ2.
152 Our use of Pearson's correlation aligns with recent SDG studies, such as Pradhan et al. (2017),
153 who employed this technique to analyse synergies and trade-offs between SDGs on a global
154 scale. For the assembly of correlation between various SDG scores for all of the 8 states
155 included in this study, using the 'metan' (v1.18.0) package with 'R' (4.1.5).

156 To uncover more complex, non-linear relationships and provide a more comprehensive
157 perspective of SDG interactions in NER, network analysis was employed at both the goal and
158 indicator levels. This supplemented RQ2. Our network analysis approach builds on the work
159 of Le Blanc (2015), who used network analysis to map the interactions between SDGs at a
160 global level. This methodology is particularly advantageous in this context, as it facilitates the
161 elucidation of complex, non-linear relationships that might not be discernible through
162 correlation analysis alone, thus providing a more comprehensive perspective of SDG
163 interactions in the Northeast region. 'igraph' (v2.0.3) package in 'R' has been used for this. In
164 this network analysis, absolute correlation values and the Fruchterman-Reingold algorithm
165 (suitable for undirected graphs) were used.

166 The application of Hierarchical Clustering Analysis (HCA) in this study is particularly
167 valuable, as it enables the identification of patterns and similarities among districts without
168 pre-defining the number of clusters, which is crucial given the heterogeneous nature of NER
169 districts. This clustering approach allows the discovery of groups of districts with similar SDG
170 performance profiles, potentially informing targeted policy interventions and resource
171 allocation strategies tailored to the specific needs and challenges of each cluster. We employed
172 HCA to group districts based on their environmental and socioeconomic SDG features. The
173 application of HCA in our study was similar to the approach used by others (Wang et al. 2020;
174 Wu et al., 2022), who used clustering to group Chinese provinces based on their SDG

175 performance. It addressed our RQ3. This analytical technique is appropriate for our study
176 because it enables the identification of patterns and similarities among districts without pre-
177 defining the number of clusters, which is crucial given the heterogeneous nature of NER
178 districts. The within-cluster sum of squared (WSS) method was used to find cluster numbers
179 via the silhouette method through the Euclidean distance using a single linkage. The silhouette
180 method determines how well each point fits into its cluster and measures clustering quality.
181 The length of a line segment connecting two locations in Euclidean space is called the
182 Euclidean distance. The ‘cluster’ (v2.1.6), ‘dendextend’ (v1.17.1), and ‘factoextra’ (v1.0.7)
183 packages with ‘R’ have been used.

184 We applied Data Envelopment Analysis, specifically an input-oriented DEA with a
185 slack-based model (Tone, 2001) and variable return to scale assumption, to ascertain the
186 relative efficiency of districts in converting environmental SDG inputs into socioeconomic
187 SDG outputs. It addressed our RQ4. The use of DEA in our study is comparable to that of Guo
188 et al. (2024), who employed DEA to evaluate the efficiency of SDGs in Organisation for
189 Economic Co-operation and Development (OECD) countries. This methodology is particularly
190 appropriate for our study, as it facilitates the simultaneous comparison of multiple inputs and
191 outputs, which is essential given the multi-dimensional nature of SDGs. The efficiency of a
192 district can be assessed by comparing three environmental SDG inputs (SDG 6, 13, and 15)
193 and 12 socioeconomic SDG outputs (SDG 1-5, 7-12, 16). Moreover, if $\lambda \text{ sum} = 1$, DMU
194 is in the CRS subzone; if $\lambda \text{ sum} > 1$, DMU is in the DRS subzone; and if $\lambda \text{ sum} <$
195 1 , DMU is in the IRS subzone (Seiford and Zhu, 1999). The number of DMUs should be ≥ 2
196 times the sum of the inputs and outputs to achieve a sufficient discriminating power (Banker
197 et al. 1989). Another stipulation is that the number of DMUs \geq the sum of the input and output
198 variables. There are three input variables and the output variables are twelve in this study, and
199 103 DMUs meet both criteria, culminating in a model with sufficient discriminating power.
200 For these analyses, the ‘deaR’ (v1.4.1) package in ‘R’ was used.

201 To quantify the geographical variance in SDG performance across NER districts,
202 multiple inequality indices were employed: the Theil, Atkinson, and Gini indices. It addressed
203 our RQ5. Our application of inequality indices aligns with the approach of Chaudhuri and Roy
204 (2017), who employed inequality measures to study spatial disparities in water and sanitation
205 facilities (WaSH; SDG 6). The use of multiple indices provides a more robust assessment of
206 inequality. The Theil index was particularly valuable because of its decomposability property,
207 enabling the examination of inequality both within and between districts. Population-weighted
208 indices could not be calculated because of the unavailability of official annual population data

209 (2020-2021). For this analysis, the ‘REAT’ (v3.0.3) package in ‘R’ was used. Given the
210 existence of many Theil inequality measures, this study employs Stoermann's (2009)
211 formulation. For the Atkinson and Gini indices, the formulations of Portnov and Felsenstein
212 (2010) and Doersam (2004) were used, respectively.

213 Using the Evenness (EIS) and Mean Index Score (MIS) is essential for gaining a
214 comprehensive understanding of development. MIS provides a clear measure of overall SDG
215 performance, while EIS highlights imbalances across different goals, ensuring that progress is
216 not only effective, but also equitable. Together, these indices guide policymakers in identifying
217 underperforming areas, enabling targeted interventions to achieve holistic and inclusive growth
218 aligned with the SDG 2030 Agenda. It addressed our RQ6. The methodology of the EIS and
219 MIS calculations was derived from recent SDG studies (Liu et al. 2021, 2024; Qi et al. 2024).
220 EIS and MIS were calculated at the individual district, state, and SDG levels. All the
221 calculations have been made using ‘dplyr’ package (v1.1.4) in ‘R’.

222 To address RQ7, we developed a comparative index for the relative SDG performance
223 of the NER districts. This approach facilitates both intranational and extranational comparisons
224 and provides a comprehensive assessment of how these districts perform relative to their state,
225 national, and global counterparts. This methodology is crucial in our context as it enables the
226 identification of areas where NER is excelling or lagging in SDG progress. The relative
227 performance of any district in the NER on the SDGs was calculated by dividing the score by
228 the score of another district. This ratio can also be multiplied by 100 to convert the performance
229 scale to a %. This procedure was performed for all 103 NER districts.

230 Because of the nature of the available district SDG dataset (single point, single year) of
231 Northeast India, most of the usual advanced analyses, such as different types of regressions
232 and future projections, could not be applied in this study.

233

234 **Results**

235

236 *Achievements and shortcomings*

237 The SDGs in 103 NER districts were organised into three categories based on the
238 'wedding cake' framework (Folke et al. 2016). The first category comprised econ-SDGs (8-10
239 and 12) embedded within soc-SDGs (1-5, 7, 11, and 16). The second category depends on env-
240 SDGs (6, 13, and 15).

241 The most environmentally prosperous districts in NER (e.g. Karimganj, W-Jaintia Hills
242 [here, W = West], E-Garo Hills [here, E= East], Mamits, and Chirang) have led to the

243 implementation of environmental sustainability measures, including effective waste
244 management, conservation efforts, and climate action initiatives. Conversely, the worst
245 performing districts were Barpeta, Mon, SW-Garo Hills [SW = Southwest], Darrang, and
246 Tawang. Among the NER districts, 50.5% (n = 52) were below the regional (district) average
247 (70.14). The achievement gap (28.3) between the best (82.6, Karimganj of Assam) and worst
248 (54.3, Barpeta of Assam) performing districts in the env-SDGs suggests that, while some
249 districts have made significant progress, others are significantly lagging behind. Only 22%
250 (n=23) of the NER districts performed below the Indian national (state-level) average (66.03)
251 on the env-SDGs. This suggests that most NER districts are performing at or above the national
252 average in terms of environmental sustainability.

253 Socially best-performing districts in NER (e.g. Champhai, E-Sikkim, Aizawl, Serchhip,
254 and W-Tripura) have excelled in achieving social equity, quality education, and good health
255 and wellbeing. However, districts with the poorest social performance (e.g. N-Garo Hills [here,
256 N=North], S-Garo Hills [here, S=South], W-Jaintia Hills, Tuensang, and E-Garo Hills) may
257 face challenges, such as a lack of resources, infrastructure, or awareness that hinders their social
258 development efforts. Nearly half (49.5%, n=51) of the NER districts were below the regional
259 (district) average (64.05). The difference (26.12) in soc-SDG performance between the top-
260 (75.87, Champhai in Mizoram) and the lowest-performing districts (49.75, N-Garo Hills in
261 Meghalaya) suggests that while some districts have made significant progress, others are
262 significantly lagging behind. 61% (n=63) of the NER districts performed below the Indian
263 national (state-level) average (66.35) in the soc-SDGs. This suggests that a significant number
264 of NER districts need to improve their social development efforts to match the national average.

265 The most economically prosperous NER districts were Unakoti, Gomati, S-Tripura, W-
266 Tripura, and N-Tripura. On the other hand, the economically weakest districts were Shi Yomi,
267 Zunheboto, Kiphire, Leparada, and Kamle. 43% (n=45) of the NER districts performed below
268 the regional (district) average (67.84), suggesting uneven progress within the region. The
269 disparity between the best (86.25, Unakoti of Tripura) and worst (39.5, Shi Yomi of Arunachal
270 Pradesh) performing districts in the econ-SDGs was 46.75. Only 29% (n=30) of the NER
271 districts scored below the Indian national (state-level) average (62.34).

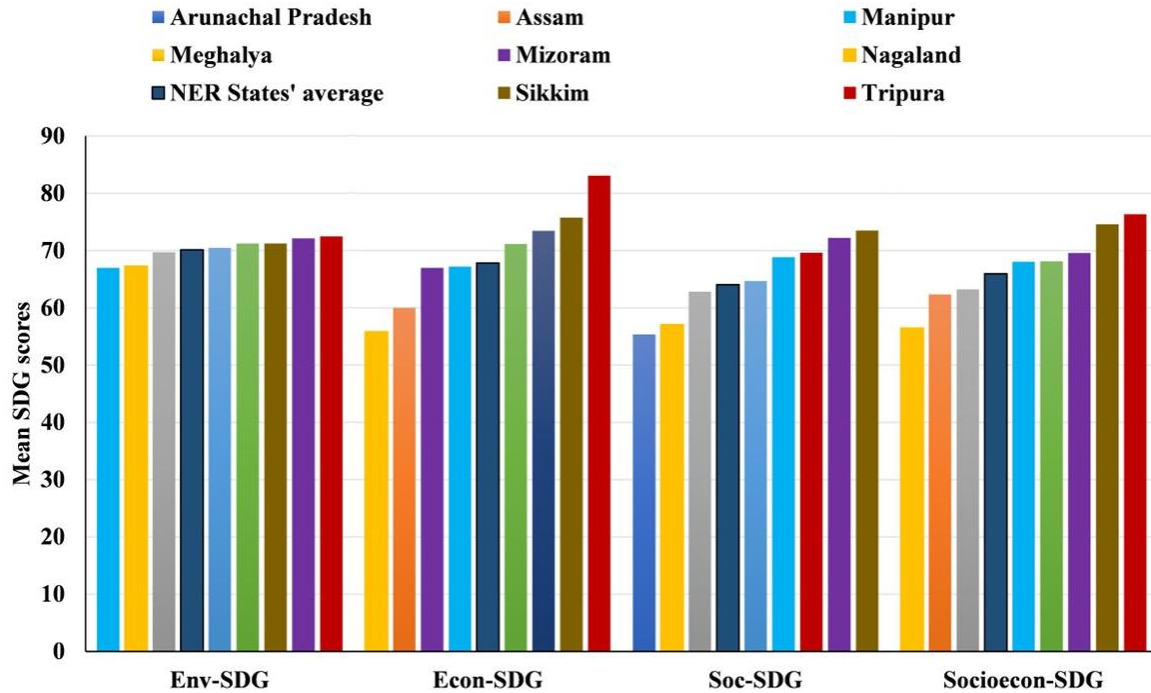
272 The most affluent districts, as indicated by the composite SDG, (e.g. E-Sikkim, Gomati,
273 N-Tripura, W-Tripura, and Serchhip) have excelled in achieving balanced development across
274 all SDGs. Conversely, the districts with the poorest composite SDG performance are Kiphire,
275 Zunheboto, Kra Daadi, Tuensang, and Mon. The achievement gap between the top-performing
276 district (75.87, E-Sikkim in Sikkim) and the lowest-performing district (53, Kiphire in

277 Nagaland) was 22.87. 45% (n=47) of the NER districts performed below the regional (district)
278 average (66.28) in the composite SDG. The order of districts with lower performance across
279 the individual SDGs was as follows: SDG 13 > 7 > 4 > 8 > 9 > 3 > 6 > 11 > 16 > 10 > 12 > 15
280 > 5 > 1 > 2. This suggests that the greatest disparities exist in SDG 13 (Climate Action), and
281 the least in SDG 2 (Zero Hunger). The disparity between the top and bottom performing
282 districts in the individual SDGs was as follows: SDG
283 7>13>12>11>10>9>15>5>1>16>6>4>2>3>8> composite SDG. This indicates that the
284 disparities are more pronounced in specific SDGs than in overall development.

285 The relationship between societal and economic development is well-established and
286 can be combined to achieve socioeconomic development. In the NER districts, W-Tripura,
287 Gomati, S-Tripura, E-Sikkim, and N-Tripura are among the most socioeconomically
288 prosperous. On the other hand, Zunheboto, Kiphire, Shi Yomi, Tuensang, and Kamle are the
289 worst performing districts in terms of socioecon-SDGs. 53% (n=55) of NER districts
290 performed below the regional (district) average (65.94) of the socioecon-SDGs. The
291 performance gap between the best (79.37, W-Tripura) and worst performing districts (49.87,
292 Zunheboto) in the socioecon-SDGs was 29.5. Only 35.9% (n=37) of the NER districts
293 performed lower than the Indian national (state-level) average (64.34).

294 Based on the overall scores (Figure. 1), Tripura, Sikkim, and Mizoram were the top
295 performing states, whereas Meghalaya, Arunachal Pradesh, and Nagaland had the lowest
296 performances. In terms of env-SDGs, Tripura outperformed Mizoram and Sikkim, which in
297 turn surpassed Assam, Meghalaya, the average of the NER states, Arunachal Pradesh,
298 Nagaland, and Manipur. In terms of soc-SDGs, Sikkim, Mizoram, Tripura, Manipur,
299 Arunachal Pradesh, the average of the NER states, Assam, Nagaland, and Meghalaya were
300 ranked accordingly. For the econ-SDGs, Tripura, Sikkim, Assam, Meghalaya, the average of
301 the NER states, Manipur, Mizoram, Arunachal Pradesh, and Nagaland were ranked in that
302 order. Lastly, in terms of socioecon-SDGs, Tripura, Sikkim, Mizoram, Assam, Manipur, the
303 average of the NER states, Meghalaya, Arunachal Pradesh, and Nagaland were ranked in that
304 order.

305



306
307

Figure 1. Achieving four types of SDGs for 103 districts (aggregated into 8 respective states) of Northeast India. The SDG groups are environmental (Env-SDG), economic (Econ-SDG), social (Soc-SDG), and Socioeconomic SDGs (Socioecon-SDG).

308

309 When considering the proportion of underperforming districts compared with the
 310 average performance for each state, the order of environmental underachievement was as
 311 follows: Meghalaya> Mizoram> Tripura> Assam> Arunachal Pradesh> Manipur> Nagaland>
 312 Sikkim. The order of societal underachievement was Tripura, Nagaland, Assam, Meghalaya,
 313 Arunachal Pradesh, Sikkim, Manipur, and Mizoram. The order of economic underachievement
 314 was Meghalaya, Manipur, Nagaland, Tripura, Arunachal Pradesh, Mizoram, Assam, and
 315 Sikkim. The order of socioeconomic underachievement is Manipur> Nagaland> Arunachal
 316 Pradesh> Tripura> Mizoram> Assam> Meghalaya> Sikkim. The order of overall
 317 underachievement was Manipur, Meghalaya, Tripura, Mizoram, Nagaland, Assam, Arunachal
 318 Pradesh, and Sikkim. Descriptive statistics based on individual SDGs and indicator data were
 319 also calculated (Tables S1 and S2, respectively, Supplementary File 1).

320

321 *Interrelationships*

322 The interrelationships between individuals and SDG groups were inferred using
 323 Pearson's correlation. After analysing the individual SDG scores of the NER districts (Figure.

324 2a), a strong positive correlation (0.7) was observed between SDG 1 and 8. This finding
325 suggests that economic growth is accompanied by a reduction in poverty. Similarly, a strong
326 positive correlation (0.63) exists between SDG 3 and 4, indicating that improved education
327 levels are associated with better health outcomes in the region. Additionally, SDG 6 and 7
328 exhibited a strong positive correlation (0.61), suggesting that access to clean water and
329 sanitation goes hand in hand with access to affordable and clean energy. Several SDGs exhibit
330 moderate positive correlations, indicating that improvements in one area tend to accompany
331 improvements in another. These include: SDG 1 and 10 (0.57), SDG 3 and 5 (0.57), SDG 4
332 and 5 (0.56), SDG 11 and 6 (0.51), and SDG 11 and 7 (0.5). Weaker positive correlations were
333 found between other SDGs, indicating a weaker relationship between improvements in one
334 SDG and the other. These include: SDG 1 and 13 (0.35), SDG 2 and 3 (0.28), and SDG 9 and
335 10 (0.23). Furthermore, a weak negative correlation was observed between SDG 13 and 8 (-
336 0.12), suggesting that economic growth in NER may occur at the expense of environmental
337 degradation. Pearson's correlation for individual SDG performance in NER districts grouped
338 by state (Figure. S.3) was also analysed.

339 The analysis of the grouped SDG scores for the NER districts (Figure. 2b) showed that
340 the strongest positive correlation (0.87) existed between the socioecon-SDGs and econ-SDGs.
341 This indicates that economic development is strongly linked to improvements in social
342 indicators, such as education, health, and gender equality. A strong positive correlation (0.7)
343 was observed between env- and soc-SDGs. This suggests that environmental well-being goes
344 hand in hand with social progress in the region. Districts that perform well in terms of
345 environmental sustainability also tend to score better on social indicators. A moderate positive
346 correlation (0.63) was observed between the socioecon-SDGs and the composite SDG. This
347 suggests that progress in socioeconomic goals contributes significantly to the overall SDG
348 achievement in the region. There was a moderate positive correlation (0.59) between econ- and
349 composite SDG. The env-SDGs had a moderate positive correlation (0.51) with the composite
350 SDG. There was a weak positive correlation (0.31) between soc-econ and env-SDGs. This
351 finding suggests a less clear link between socioeconomic development and environmental well-
352 being in this region. No negative correlations were observed between the SDG groups.
353 Pearson's correlation of the grouped SDG performance of NER districts grouped by state
354 (Figure. S.4) was also analysed.

355 Pearson's correlation of indicator level data for NER districts (after removing the data
356 gaps) have been composed. An analysis of the 13 indicators of the env-SDGs in the NER
357 districts (Figure. S.5) showed that the strongest positive correlation (0.72) existed between

358 forest cover (%) and change in forest cover. This suggests that districts with a higher % of
359 forest cover experienced an increase in forest cover, indicating successful afforestation in these
360 regions. A positive correlation (0.57) was observed between the % of area under forest cover
361 and number of forest fires. Although this may seem counterintuitive, it could be due to several
362 reasons, such as districts with larger forest cover having more forest areas and some forest fires
363 being controlled for forest management purposes. A moderate positive correlation (0.55) was
364 found between the % of forest cover and stage of forest growth, indicating that districts with a
365 higher % of forest cover also had forests in a more mature stage of growth. A moderate positive
366 correlation (0.5) was observed between the % of area under forest cover and availability of
367 toilets with toilet facilities. However, a weak positive correlation (0.18) existed between forest
368 cover and the % of households using LPG as cooking fuel, suggesting a weak link between
369 forest conservation efforts and the adoption of clean cooking fuels in the region. A negative
370 correlation (-0.57) was observed between the % of forest cover and the stage of forest
371 degradation, which is a positive finding as it indicates that districts with a higher % of forest
372 cover tend to have lower levels of forest degradation. Finally, a weak negative correlation (-
373 0.21) was found between the % area under forest cover and the availability of clean cooking
374 fuel.

375 After analysing 55 indicators of soc-SDGs in the NER districts ([Figure. S.6](#)), a strong
376 negative correlation (-0.71) was observed between the head count ratio, as per the multi-
377 dimensional poverty index (MPI), and the number of hospital beds empanelled under PMJAY
378 (per 10,000 eligible population). This indicates that, as the poverty index decreases, the number
379 of hospital beds increases, which is a positive outcome. Additionally, a positive correlation
380 (0.29) was observed between hospital beds empaneled under PMJAY and the % of affordable
381 houses completed against sanctions (rural and urban) under PMJAY. This finding suggests that
382 areas with more hospital beds tended to have a higher % of affordable housing. Finally, a strong
383 negative correlation (-0.78) was observed between beneficiaries covered under the National
384 Food Security Act (NFSA) (%) and hospital beds empanelled under the PMJAY. This suggests
385 that areas with more hospital beds tend to have fewer beneficiaries covered under the NFSA.

386 Analysis of the 16 econ-SDG indicators in the NER districts ([Figure. S.7](#)) shows that
387 the strongest positive correlation (0.74) exists between the surface area, presumably of roads,
388 and the % of workers engaged in agriculture. This suggests that districts with a higher % of
389 agricultural workers tend to have more developed road networks. This could be due to the fact
390 that agriculture often relies on efficient transportation to bring its products to market. A strong
391 positive correlation (0.7) was found between the number of informal micro-, small-, and

392 medium-sized enterprises (MSMEs) and the % of workers in the primary sector. While this
393 may seem counterintuitive at first, it could be because many rural areas with a high number of
394 primary-sector workers also have a significant number of small informal businesses that cater
395 to the needs of the local population. A moderate positive correlation (0.56) was observed
396 between the surface area and annual rainfall, indicating that areas with higher rainfall may
397 prioritise the construction of roads owing to transportation challenges during monsoons. A
398 moderate positive correlation (0.53) was found between the number of informal MSMEs and
399 Gross State Domestic Product (GSDP) per capita, suggesting that districts with a higher
400 number of informal businesses tend to have a higher GDP per capita, potentially reflecting
401 greater economic activity. Additionally, a moderately positive correlation (0.52) is observed
402 between the % of agricultural workers and the % of households with bank accounts. This could
403 be due to various reasons, including government initiatives that promote financial inclusion in
404 rural areas. However, a weak positive correlation (0.22) was found between annual rainfall and
405 the GSDP per capita, and a weak positive correlation (0.19) was observed between the % of
406 workers in agriculture and the per capita availability of power. Pearson's correlation for each
407 individual SDGs at the indicator level for the NER districts was also calculated (Figure. S.8-
408 9).

409

410 *Networks*

411 Network analysis of individual SDGs and composite SDG (n=16) (Figure. S.10), with
412 the highest degree (5), the composite SDG appeared to be the most connected in the network,
413 suggesting its relevance as an overall measure of sustainable development in the region. SDG
414 1 and 4 both had a degree of 3, indicating significant connectivity within the network. SDGs
415 3, 9, 13, 6, and 2 have a moderate degree of 2, suggesting some level of interconnectedness.
416 SDGs 8, 7, and 16, with a degree of 1, appear to be the least connected in the network. Some
417 of the strongest positive correlations are found between Composite SDG and 1 (0.804),
418 Composite SDG and 9 (0.704), SDG 4 & 6 (0.632). Some of the strongest negative correlations
419 were found between SDG 13 and 2 (-0.542) and SDG 13 and 6 (-0.592). This suggests potential
420 trade-offs between climate initiatives and goals related to food security and water management
421 in the region. SDG 1 shows strong positive correlations with the composite SDG (0.804), 9
422 (0.579), and 3 (0.577). This suggests that poverty reduction is central to sustainable
423 development in the northeastern states. SDG 4 demonstrates positive correlations with SDG 6
424 (0.632), 2 (0.565), and 8 (0.501). This highlights the potential role of SDG 4 as a catalyst for
425 progress in other areas, particularly in water and sanitation (SDG 6) and economic growth

426 (SDG 8). The low connectivity of SDGs 7, 8, and 16 may indicate areas that require more
427 integrated approaches.

428 Network analysis of individual SDG indicators (n=84; [Figure. 2c](#)), most indicators in
429 the top tier (degree ≥ 60) were social. This suggests that social factors play a crucial role in the
430 sustainable development of the NER districts. The highest-degree indicator (80) is related to
431 education: "Percentage of trained teachers at the secondary level (Class 9-10)". Women's
432 empowerment features prominently, with "Exclusive women SHGs in bank-linked SHGs"
433 having the second-highest degree (74). Food security and agriculture are also significant, as
434 evidenced by "Productivity of fruits and vegetables (kg/ha)" (degree 70) and "Percentage of
435 Fair Price Shops (FPS) covered under online transaction system for PDS in the district" (degree
436 66). Economic indicators generally have lower degrees than social indicators do. The highest-
437 ranking economic indicators (degree 64) are related to employment and micro-, small-, and
438 medium-sized enterprises (MSMEs). Environmental indicators appear less frequently, and
439 typically have lower degrees. The highest-ranking environmental indicator is "Forest cover as
440 a percentage of the total geographical area" (degree 52). Indicators with the lowest degrees (2-
441 4) include a mix of social, economic, and environmental factors. Notably, some critical
442 indicators, such as "Infant Mortality Rate" and "Sex ratio at birth" had very low degrees (2 and
443 4, respectively). Health-related indicators formed a significant cluster of varying degrees (14-
444 58). Education-related indicators also featured prominently, with degrees ranging from 18 to
445 80. Gender-related indicators appear across various SDGs, highlighting the crosscutting nature
446 of gender issues. Indicators related to infrastructure and technology (e.g. Internet connectivity
447 and mobile network coverage) had moderate degrees (38-52), suggesting their growing
448 importance in sustainable development.

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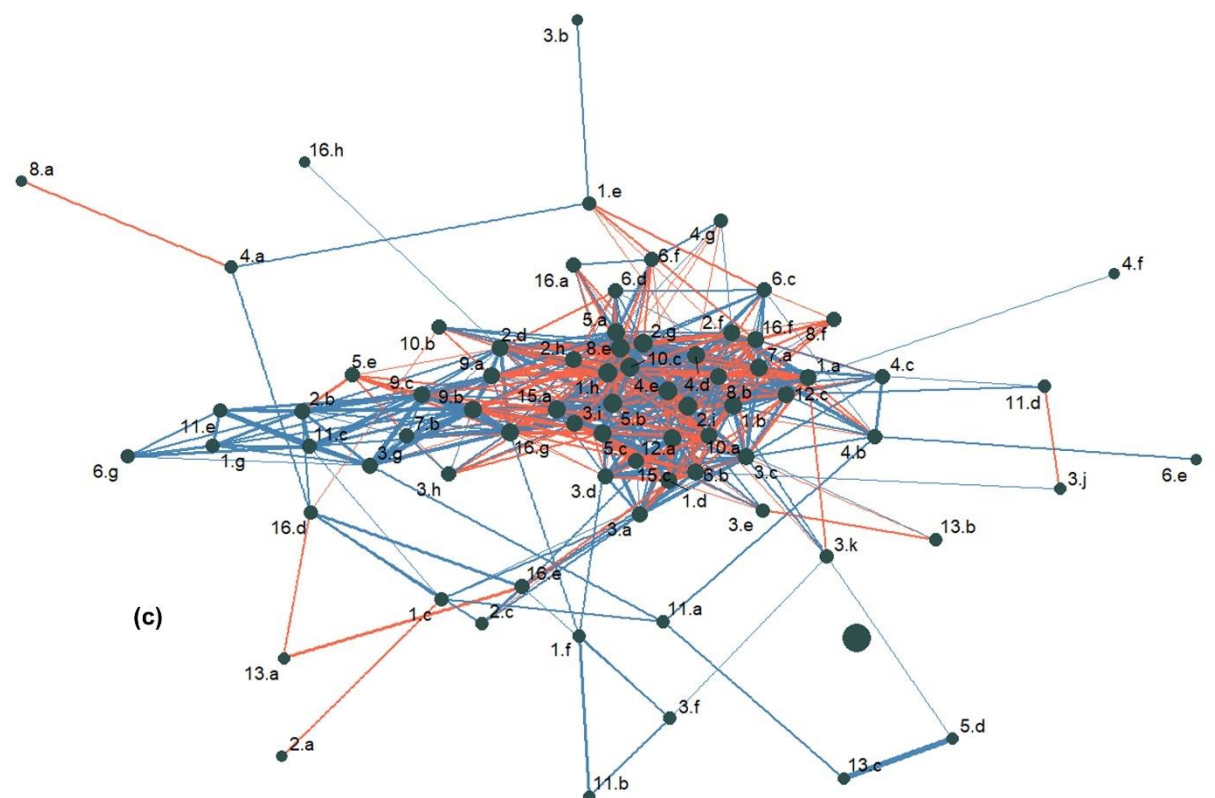
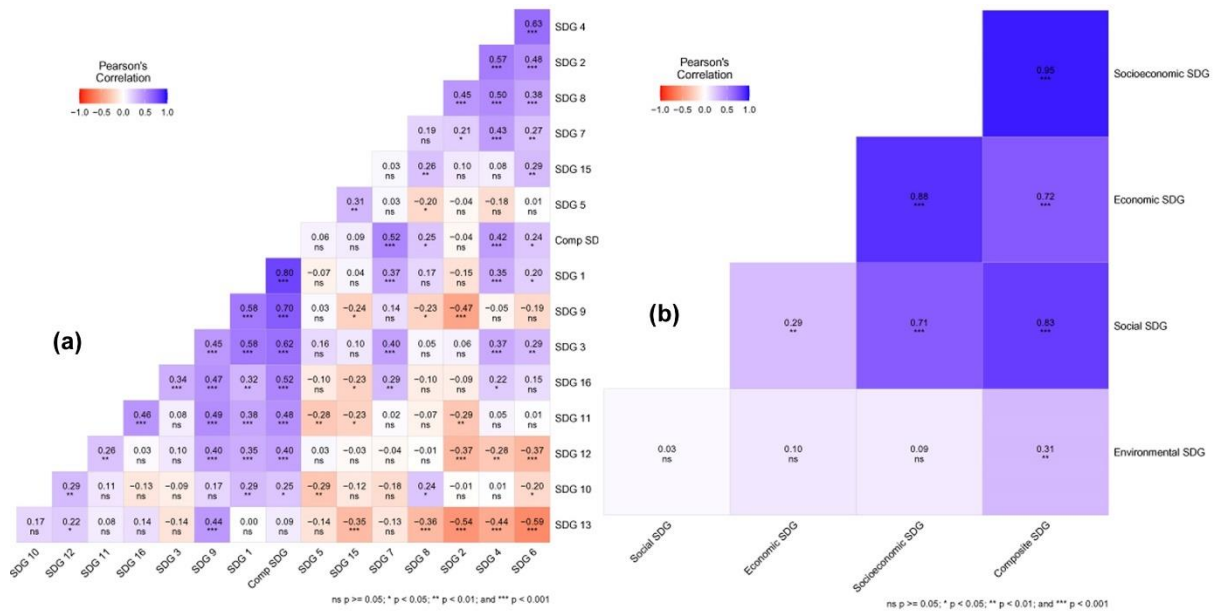


Figure 2. Interrelationships among SDGs for NER districts in India. (a) Pearson’s correlation of individual SDGs; (b) Pearson’s correlation of composite, env-, soc-, econ-, and socioeconomic SDGs; (c) Network analysis of SDG indicators (n=84). Node points were shown in dark slate grey, and edges were shown in steel blue & tomato colour.

450
451

452
453

454 *Clusters*

455 Hierarchical clustering analysis (HCA) was used for the two sets of data. First, the econ-
456 and soc-SDGs were used as outputs, whereas the env-SDGs were used as inputs (Figure. S.11).
457 Second, a similar methodology was applied, with the individual SDG scores of the env-SDGs
458 serving as inputs and those of the econ- and soc-SDGs serving as outputs (Figure. 3). The
459 author concluded that the five clusters were ideal in both situations via an analysis of optimal
460 clusters. Here, as we moved up the ladder, the scores improved on both sides. As per the env-
461 SDGs, the five clusters were composed of two (blue), 11 (red), 51 (green), 24 (orange), and 15
462 (violet) districts. This means that, from the perspective of env-SDGs, NER districts fall into
463 various categories with varying numbers of members. The five clusters on the right side are
464 composed of 14 (blue), 10 (red), 57 (green), 6 (orange), and 16 (violet) districts. This also
465 shows that socio-economically, NER districts fall into various categories with varying numbers
466 of members. When individual SDGs are considered rather than the average grouping scores
467 (i.e. the second HCA), a similar situation is observed in the second technique. These results
468 indicate three types of occurrences. For only a few of them, an almost equal status is achieved
469 for env-SDGs, soc-, and econ-SDGs. This means that districts such as Dibrugarh and Phek
470 showed similar performance on both sides. For the remaining districts which are the most
471 numerous, two types can be seen. Those with higher env-SDG performance did not reflect
472 better performance in the econ- and soc-SDGs. These districts are Nalbari, Udalguri, Leparada,
473 Zunheboto, Kiphire etc. Conversely, districts with better performance in the econ- and soc-
474 SDGs did not reflect their env-SDGs. These districts are Kohima, Dimapur, Serchhip, W-Khasi
475 Hills, SW Khasi Hills, and others. These findings suggest that improved performance in the
476 econ- and soc-SDGs in NER districts is not always correlated with improved performance in
477 env-SDGs. Similarly, districts that performed better in the econ- and soc-SDGs may not have
478 attained the same level of success in the env-SDGs. A similar picture emerged for the other
479 HCA, where individual SDGs were considered.

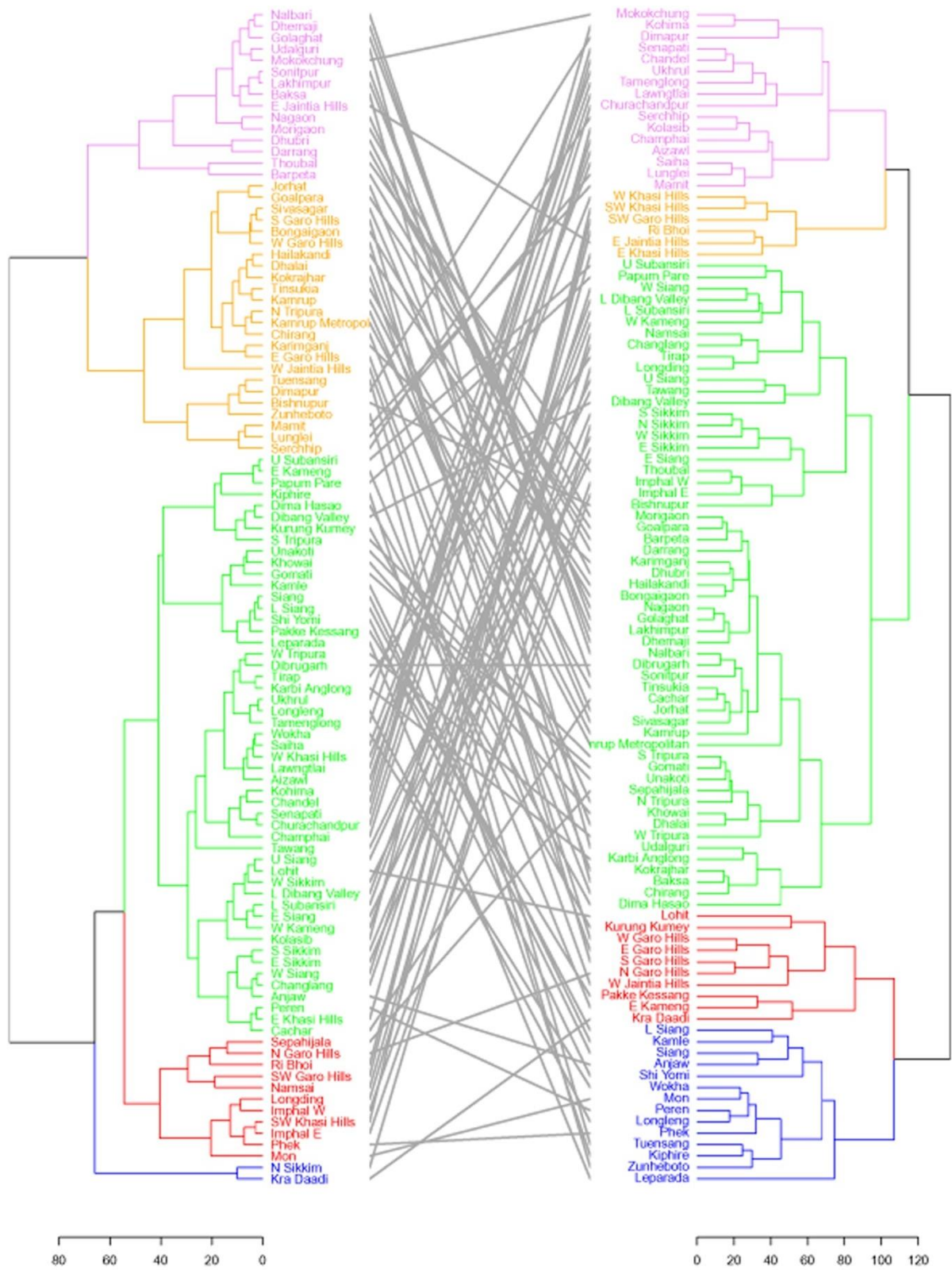


Figure 3. Bundling among SDGs for NER districts in India. Hierarchical clustering analysis, using individual SDG scores of SDG 6, 13 and 15 (left) and SDG 1–5, 7–12, and 16 (right).

480

481

482

483 *Efficiency*

484 Next, data envelopment analysis (DEA) is used to evaluate the efficiency of the 103
485 NER districts (Figure. 4). This technique for measuring performance was employed to evaluate
486 the comparative effectiveness of decision-making units (DMUs). The goal of this investigation
487 was to determine how well NER districts translate improved environmental characteristics into
488 socioeconomic opportunities.

489 There were 26 non-efficient districts (25.24%) in the NER (See Supplementary File 2).
490 The distribution of non-efficient districts for each NER state was 12% (Arunachal Pradesh),
491 40% (Assam), 11% (Manipur), 27% (Meghalaya), 12% (Mizoram), and 63% (Nagaland). None
492 of the districts in Sikkim and Tripura were inefficient.

493 To create improvement targets, a set of indicator values for comparable districts was
494 combined in a linear fashion. Improvement objectives indicate the changes that need to be
495 made to increase the efficiency of inefficient DMU's. Since reference districts are seen to
496 adhere to best practices, inefficient districts should make every effort to model their subsequent
497 actions. Only a handful (n = 18; 17.47%) of the NER districts acted as peers ≥ 3 times. Top ten
498 them, along with the times of appearance as references, were Tawang and N-Sikkim (14, both),
499 Barpetta (13), W- and S-Tripura (11, both), SW-Garo Hills and Morigaon (8, both), Papum Pare
500 (5), Namsai (5), and Mon (5).

501 DEA divides effective DMUs into three separate zones based on the returns to scale
502 (RTS) concept. DMUs can raise their outputs (in this case, soc- and econ-SDGs) at a faster
503 pace than their inputs (in this case, env-SDG) in the increasing returns to scale (IRS) zone; that
504 is, a greater increase in socioecon-SDG can be achieved with a relatively smaller increase in
505 env-SDG. The input: output ratio (in this case, the env-SDG/soc- and econ-SDG ratio) of
506 DMUs is continuously maintained in a constant return to scale (CRS) zone. Greater reductions
507 in DMUs' inputs (here, the env-SDG) occur with somewhat smaller shrinkages in their outputs
508 (here, the soc- & econ-SDGs) in the decreasing returns to scale (DRS) zone. Only five districts
509 (4.85%) belong to the IRS subzone. These districts are Karimganj (Assam), Baksa (Assam),
510 Tamenglong (Manipur), Ri Bhoi (Meghalaya), and Dimapur (Nagaland). Only six districts
511 (5.82%) belong to the DRS subzone. These districts are Kamle (Arunachal Pradesh), Kokrajhar
512 (Assam), Golaghat (Assam), Cachar (Assam), Lawngtlai (Mizoram), and Peren (Nagaland).
513 All the remaining districts (n=92 or 89.32%) belonged to the CRS subzone.

514

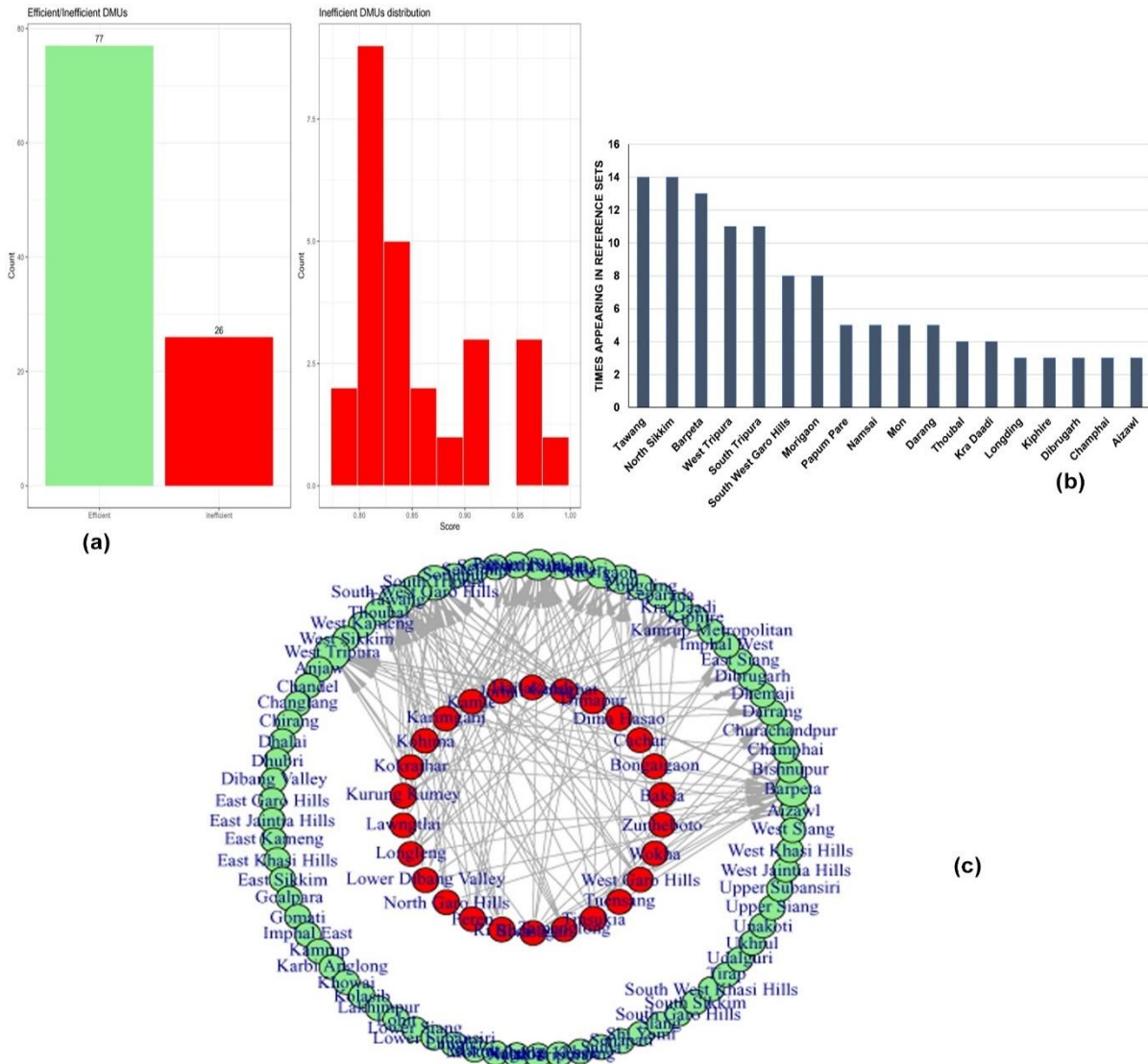


Figure 4. The efficiency of converting environmental SDGs into socioeconomic SDGs for NER districts in India. (a) Grouping of efficient and non-efficient DMUs (NER districts, $n = 103$); and Distribution of efficiency score for non-efficient DMUs ($n=26$ NER districts). (b) Ranking of efficient DMUs (NER districts) acting as peers (≥ 2 times) in reference sets, (c) non-efficient DMUs (26 NER districts, red, inner circle) and their respective reference efficient DMUs (77 NER districts, green, outer circle).

515

516

517

518 *Inequality*

519 A high inequality index indicates significant inequality in how different districts within
 520 the NER are progressing towards the SDGs. This means that some districts are far ahead of
 521 others, creating an uneven development landscape. The findings from the individual SDGs
 522 (Figure. 5a) of all NER districts (from the Theil index) indicate that the order of inequality is

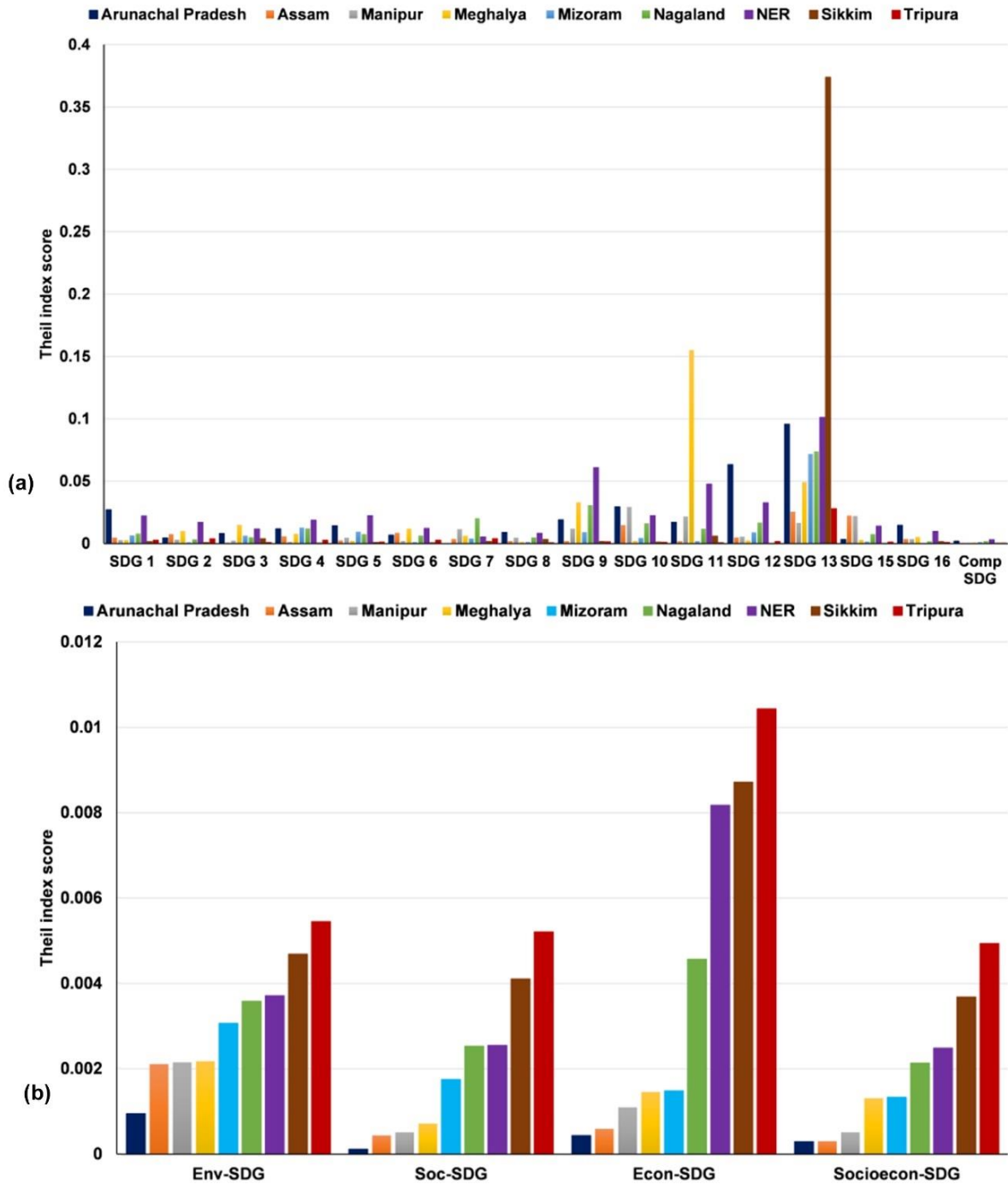
523 SDG 13 > 7 > 9 > 11 > 12 > 10 > 5 > 1 > 4 > 2 > 15 > 6 > 3 > 16 > 8 > composite SDG. SDG
524 13, 7, 9, and 11 showed the most significant inequalities among districts. This suggests uneven
525 progress in tackling climate change, energy access, infrastructure development, and sustainable
526 urbanisation across the NER. SDG 13 (0.095663) had the highest Theil index among all SDGs.
527 NER is rich in biodiversity and is ecologically sensitive. Uneven progress in climate-change
528 mitigation and adaptation strategies across districts could explain this high inequality. The
529 range of inequality in SDG 11 (range: 0.063253–0.155033) is significant. This suggests that
530 urban areas in some districts might be much farther ahead in terms of sustainable development
531 than others. There is considerable variation in the Theil index across districts for SDG 7 (range:
532 0.008896–0.020231). This could be due to factors such as differing levels of access to
533 renewable energy sources or hydropower potentials.

534 SDGs 1, 4, and 2 showed a lower degree of inequality. This might indicate a more
535 balanced effort across the districts in these areas. However, this does not necessarily mean that
536 all districts are doing well, just that the gaps are smaller. The range of Theil index values in
537 SDG 1 (range: 0.001465–0.026989) was relatively low compared with other SDGs. This might
538 indicate a more even distribution of progress in poverty reduction across districts, possibly due
539 to government initiatives or the region's agricultural potential. The NER is generally well-
540 watered. This low range of Theil index values in SDG 6 (range: 0.006706–0.012716) suggests
541 more equitable access to clean water and sanitation facilities across most districts.

542 In terms of grouped SDGs (Figure 5b), the degree of inequality was in the order econ-
543 SDG > soc-SDG > socioecon-SDG > env-SDG. Districts in the NER appear to have the most
544 significant inequality in achieving the socioecon-SDG (0.004934). This could indicate a gap
545 between progress in social development goals such as poverty reduction or education (SDG 1
546 and 4), and economic development goals such as decent work and industry (SDG 8). The Theil
547 index for env-SDGs (0.003582) was lower than that for socioecon-SDGs, but higher than that
548 for soc-SDGs (0.005206). This suggests moderate inequality in environmental progress across
549 districts. There might be variations in how districts address climate change (SDG 13) or protect
550 biodiversity (SDG 15). The Soc-SDGs (0.005206) appear to have a slightly lower level of
551 inequality than the env-SDGs. This suggests that districts might be making more even progress
552 on social development goals, such as poverty reduction, health (SDG 3), and gender equality
553 (SDG 5). This suggests uneven economic development across districts, which could be linked
554 to factors, such as access to markets, resources, and infrastructure. However, env-SDGs, which
555 include goals such as Climate Action and Life on Land, show the least disparity, suggesting a
556 more uniform response to environmental challenges across the districts.

557 When examining the disparities among individual SDGs at the intra-state level, it is
558 evident that SDGs 13 and 15 in Arunachal Pradesh, SDGs 13 and 3 in Assam, SDGs 10 and 4
559 in Manipur, SDGs 11 and 8 in Meghalaya, SDGs 13 and 16 in Mizoram and Nagaland (both),
560 SDGs 13 and 12 in Sikkim, and SDGs 13 and 11 in Tripura exhibited the highest and lowest
561 inequalities, respectively. Arunachal Pradesh, Meghalaya, Mizoram, and Nagaland had Theil
562 indices higher than the average for soc-SDGs. NEER generally performed well on some social
563 indicators, such as literacy rates. The high social inequality index in these states could be due
564 to uneven progress across different social goals (health, education, and gender equality) or
565 disparities between rural and urban areas. Assam, Sikkim and Tripura have a lower Theil index
566 for socioecon-SDG compared to the average. This might indicate more balanced progress
567 between social development (such as poverty reduction) and economic development (such as
568 decent work) in these districts. Assam has historically faced insurgency challenges, so its lower
569 socioeconomic inequality could be due to recent targeted initiatives.

570 Nonetheless, regarding the disparity in the SDGs grouped together, it is evident that the
571 env-SDGs display the greatest disparity in Assam, Meghalaya, Mizoram, Sikkim, and Tripura.
572 Conversely, the econ-SDGs exhibited the highest disparity in Arunachal Pradesh, Manipur,
573 and Nagaland. The Atkinson and Gini indices for the individual and group SDGs were also
574 calculated ([Figure. S.12-13](#)).



575

576

Figure 5. Distribution of scores of inequality analysis based on Theil index. Inequality in (a) individual and (b) grouped SDG score among NER districts of India.

577

578 *Evenness*

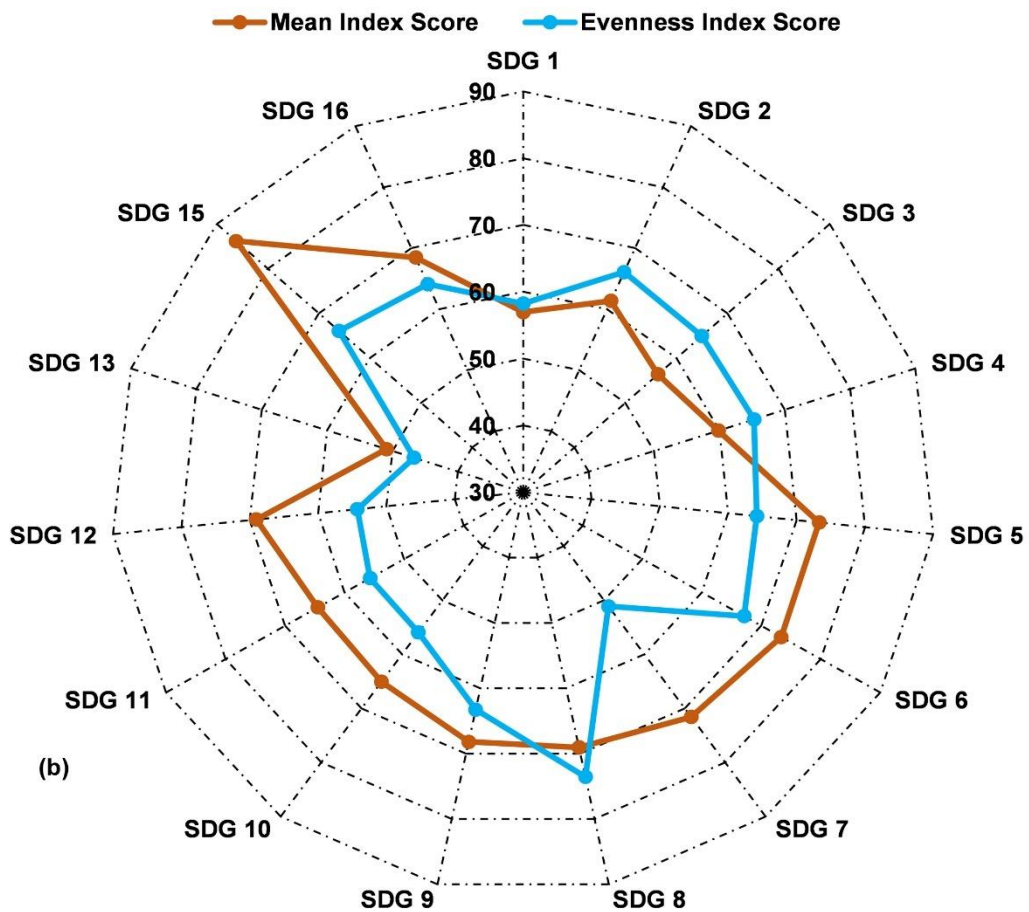
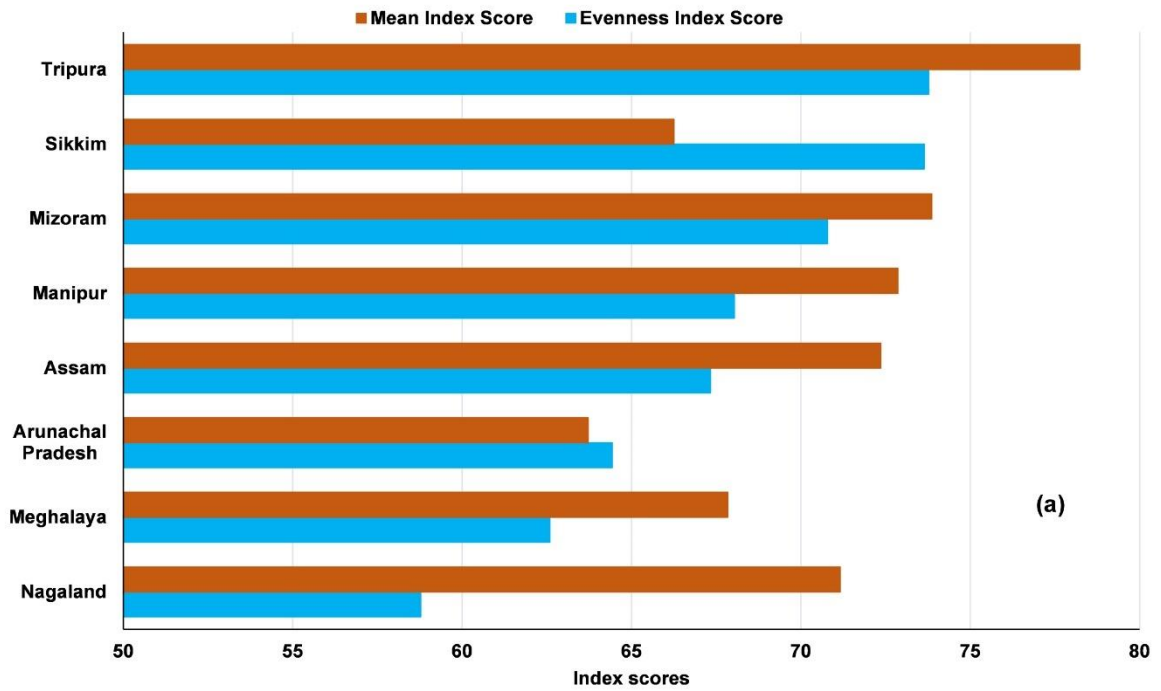
579 From the analysis of EIS and MIS at NER state level (Figure 6a), the EIS range from
 580 58.78 in Nagaland to 73.77 in Tripura, indicating significant disparities in SDG achievement.
 581 The increasing order of EIS was: Nagaland < Meghalaya < Arunachal Pradesh < Assam <

582 Manipur < Mizoram < Sikkim < Tripura. This suggests that, while some states perform well in
583 equitable resource distribution, others, particularly Nagaland, show less balance. Tripura and
584 Mizoram have the highest EIS (73.77 and 70.79, respectively), indicating a more equitable
585 distribution of resources. Tripura also had the highest MIS (78.23), reflecting both evenness
586 and overall performance in the SDG metrics. Assam, Manipur, and Arunachal Pradesh exhibit
587 moderate EIS (67.33 to 68.04), suggesting reasonable equity in SDG achievements, although
588 they do not reach the higher benchmarks of Tripura and Mizoram. Nagaland and Meghalaya
589 were the lower performers, with Nagaland scoring 58.78, indicating significant imbalances in
590 SDG achievements. A lower MIS suggests challenges in meeting the SDGs effectively. The
591 increasing order of MIS was: Arunachal Pradesh < Sikkim < Meghalaya < Nagaland < Assam
592 < Manipur < Mizoram < Tripura. These findings highlight the need for targeted policy
593 interventions in states like Nagaland and Meghalaya to improve equity and the overall SDG
594 performance. Successful strategies by Tripura and Mizoram could serve as models for
595 enhancing development outcomes in other states.

596 From the analysis of EIS and MIS at individual SDG levels ([Figure 6b](#)), SDG 15 had
597 the highest MIS (86.18) and commendable EIS (66.07), indicating effective environmental
598 sustainability. The increasing order of EIS for SDGs was: 13 < 3 < 1 < 4 < 2 < 11 < 10 < 9 <
599 16 < 12 < 8 < 7 < 5 < 6 < 15. SDG 5 and 6 also show high MIS (73.30 and 73.35%, respectively)
600 but exhibit some disparities in resource distribution, as reflected in their EIS (64.18 and
601 67.12%, respectively). SDG 1 and 3 have a lower MIS (56.95 and 56.39) and indicate
602 significant challenges in addressing these issues, necessitating targeted interventions. SDG 7
603 had the lowest EIS (51.11), whereas SDG 13 showed the lowest MIS (50.89) and EIS (46.71),
604 highlighting the critical challenges in these areas. The increasing order of MIS for SDGs was:
605 13 < 7 < 12 < 11 < 10 < 1 < 9 < 16 < 5 < 3 < 4 < 2 < 15 < 6 < 8. Overall, although certain SDGs
606 have been successful, significant disparities and challenges remain, particularly in poverty
607 alleviation, health, clean energy, and climate action. The

608 From the analysis of EIS and MIS at the NER district level, Tripura and Sikkim showed
609 strong MIS, with Tripura's districts scoring between 71.93 and 75.73, and Sikkim ranging from
610 71.87 to 75.87. The top five NER districts with the highest MIS were E-Sikkim, N-Tripura,
611 Gomati, W-Tripura, and Serchhip. The NER districts with lowest MIS were Kiphire,
612 Zunheboto, Kra Daadi, Tuensang, and Mon. Notably, Tripura's districts, such as Dhalai and
613 Gomati, have an EIS exceeding 81, indicating an equitable resource distribution. Mizoram also
614 performed well, with MIS values of 63.4 to 74.87. Lunglei stands out with an EIS of 85.9,
615 reflecting a balanced approach to achieving the SDGs. Nagaland faces significant challenges,

616 particularly in districts like Kiphire (MIS of 53) and Zunheboto (EIS of 54.53), highlighting
617 the need for targeted interventions. Arunachal Pradesh and Assam exhibited variability in
618 performance, with some districts performing well, while others, such as Kra Daadi (MIS of
619 55.6), lagged significantly. This indicates a disparity in SDG achievements. The top five NER
620 districts with the highest EIS were Lunglei, Mokokchung, Dhalai, Serchhip, and Gomati. The
621 NER districts with lowest EIS were E-Kameng, Kra Daadi, Kiphire, Shi Yomi, and Pakke
622 Kessang. Overall, while Tripura, Sikkim, and Mizoram demonstrate notable success in SDG
623 achievement, Nagaland and certain districts in Arunachal Pradesh and Assam require focused
624 efforts to improve performance and equity in resource distribution.



625
626

Figure 6. Comparative performance of evenness index score (EIS) and mean index score (MIS) for (a) NER states (n=8) and (b) individual SDGs.

627 **Relative scoring**

628 We must assess whether the performance of the districts in NER India with respect to
629 SDGs is superior or inferior to that of other comparable entities.

630

631 **a) Intranational**

632 The relative performance of NER districts with their respective states and nations (i.e.
633 India) (Figure. 7a-7d, with the Tripura district as an example).

634 Districts that performed better in the composite SDG than their respective states were
635 28 (Assam), 16 (Arunachal Pradesh), and 3 (Manipur & Sikkim). This suggests uneven
636 progress across districts in these states. This means that > 50% of the districts perform equally
637 or better than their respective states' composite SDG. This is a positive sign, as it indicates that
638 a significant number of districts are not just keeping pace with their state's progress, but are
639 also potentially leading the way in sustainable development. The top districts that have
640 outperformed their states for each individual SDG are: Kamrup Metropolitan (SDG 1), W-Garo
641 Hills (SDG 2), E-Sikkim (SDG 3), Kohima (SDG 4), Karbi Anglong (SDG 5), L-Dibang
642 Valley (SDG 6) [here, L=Lower], W-Garo Hills (SDG 7), Imphal-W (SDG 8), Serchhip (SDG
643 9), Peren (SDG 10), SW-Khasi Hills (SDG 11), Udalguri (SDG 12), N-Tripura (SDG 13),
644 Kolasib (SDG 15), and Leparada (SDG 16). It is necessary to identify and address the needs of
645 those who lag behind their state average.

646 In terms of env-SDG scores, 56 NER districts performed better, and three performed
647 equally well (i.e. total >57%) with the state's score. The top three districts were Karimganj
648 (127%), N-Tripura (124%), and Mamit (123%). While >79% of the NER districts (n=82)
649 scored higher than their state's soc-SDG score, this is the lowest % compared to the econ- and
650 env-SDGs. The top three such districts were L-Subansiri (128%), Tawang (127%), and W-
651 Siang (125%). A staggering 99 districts (>96%) outperformed their state's econ-SDGs. The
652 top three such districts were Nalbari (140%), Udalguri (139%), and Dibrugarh (138%).

653 Only four NER districts (3.88%, viz. W-Tripura, Gomati, and N-Tripura in Tripura and
654 Serchhip in Mizoram) performed better than the national composite SDG. 71% of the NER
655 districts performed equally or better than the national env-SDG. This is a positive sign, and
656 shows that these districts are making good progress in terms of environmental sustainability.
657 The top three such districts were Karimganj (122%), W-Jaintia Hills (121.6%), and E-Garo
658 Hills (121.1%). While >43% of the districts (44 better and one equally performing) scored
659 higher than the national average on the soc-SDGs, this is a lower % compared than the env-
660 and econ-SDGs. The top three such districts were Champhai (115%), E-Sikkim (114%), and

661 Serchhip (113%). Approximately 67% of the NER districts performed better or equally well
662 with the national econ-SDGs. The top three such districts were Unakoti (134%), Gomati
663 (132%), and S-Tripura (131%). The relative intranational SDG performance for individual
664 SDGs was calculated (for a comparison of NER districts with their respective states, see [Figure.](#)
665 [S.14](#). For India, see [Figure. S.15](#)). Intranational relative SDG performance for grouped SDGs
666 was also calculated (for a comparison of NER districts with their respective states, see [Figure.](#)
667 [S.16](#) for a comparison of the NER districts with India (see [Figure. S.17](#)).

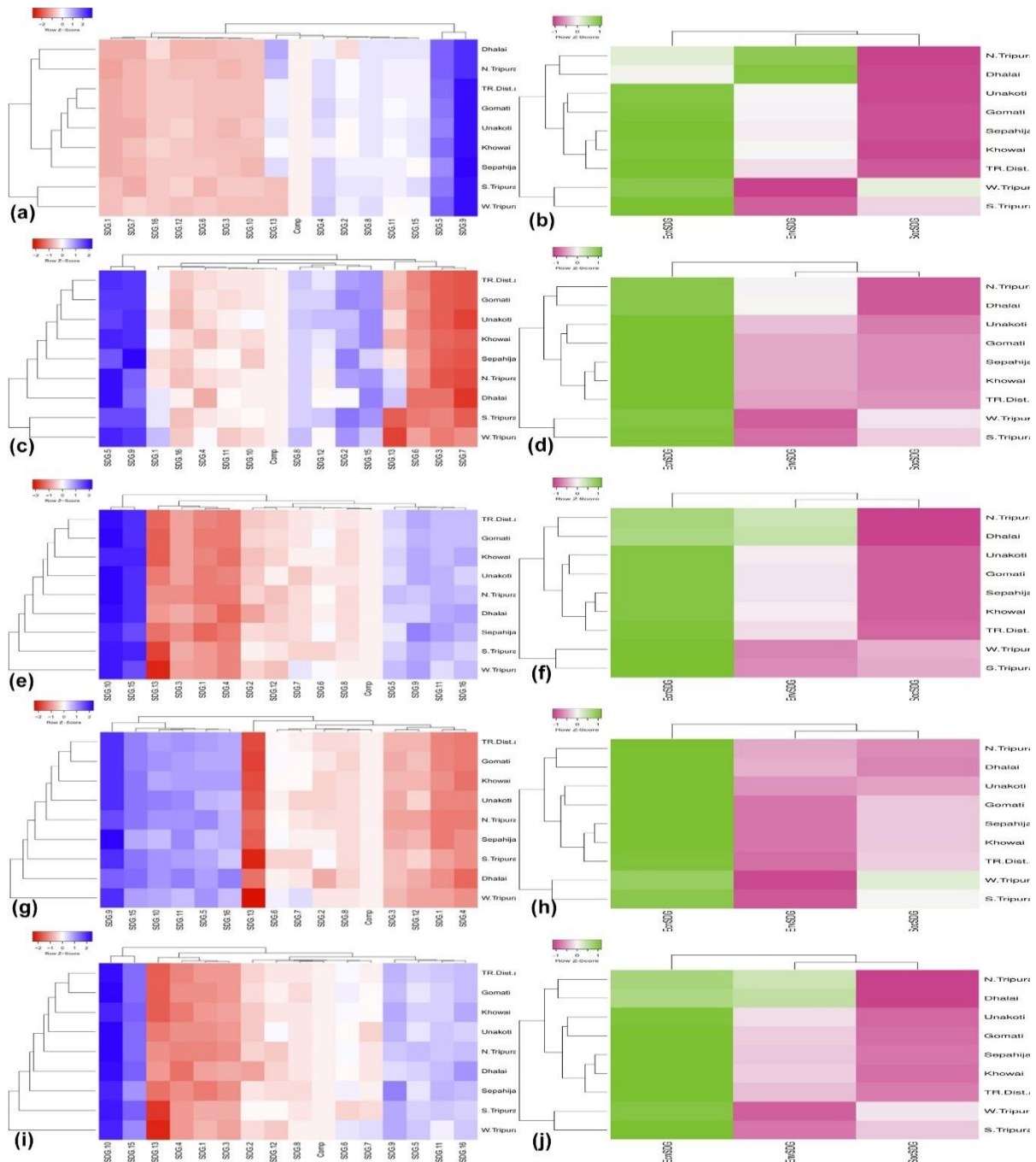
668

669 ***b) Extranational***

670 For this purpose, the relative performances of NER districts with their respective regions
671 (East and South Asia, ESA), income groups (lower-middle income, LMI), and global (world)
672 scores were considered ([Figure. 7e-7j](#), for the district of Tripura).

673 Better-performing NER districts had a higher score on the env-SDGs (78%) than on the
674 soc- (19%) and econ-SDGs (59%). This suggests that these districts are doing relatively well
675 in terms of environmental sustainability compared to their counterparts in the ESA. When
676 compared to the LMI group score, the NER districts performed similarly in the env-SDGs and
677 econ-SDGs (60%). However, they outperformed the LMI group in terms of soc-SDGs (68%).
678 This indicates that these districts are making significant strides towards achieving social equity,
679 quality education, and good health and well-being. The NER districts' performance in the env-
680 SDGs (72%) is comparable to the global score, suggesting that these districts are on par with
681 the global standards for environmental sustainability. However, the scores for the soc-SDGs
682 (32%) and econ-SDGs (58%) were lower than the global scores, indicating the need for focused
683 efforts in these areas to catch up with global progress. The relative extranational SDG
684 performance for individual SDGs was calculated (for a comparison of the NER districts with
685 the ESA: [Figure. S.18](#), for the LMI group: [Figure. S.19](#), the global score: [Figure. S.20](#)). The
686 extranational relative SDG performance for grouped SDGs was also calculated (for comparison
687 of NER districts with ESA: [Figure. S.21](#), the LMI group: [Figure. S.22](#), the global score: [Figure.](#)
688 [S.23](#)).

689



690

691

Figure 7. Relative SDG performance of 8 districts of Tripura (TR). Left side: for individual SDGs, in comparison to (a) Tripura state, (c) India, (e) East and South Asia (ESA), (g) Lower middle-income (LMI) economies, and (i) Global score. Right side: for grouped SDGs, in comparison to (b) Tripura state, (d) India, (f) East and South Asia (ESA), (h) Lower middle-income (LMI) economies, and (j) Global score.

692

693

694 **Discussions**

695 This research offers a thorough examination of the UN SDGs across NER India's
696 districts, uncovering significant variations in the area's advancement towards sustainability.
697 The results highlight the intricate nature of sustainable development in NER, which is
698 influenced by a distinctive combination of environmental, societal, and economic elements.
699 Addressing RQ1, our analysis revealed significant disparities in SDG achievement across the
700 NER districts. The Env-SDGs, particularly SDG 13, showed the highest variation, with a 28.3-
701 point gap between the best- and worst-performing districts. This highlights the region's
702 vulnerability to climate change and the uneven application of mitigation measures. The
703 fluctuating performance across the SDGs emphasises the region's varied challenges and
704 prospects, indicating that a uniform approach to sustainability is insufficient.

705 A notable finding of this research is the significant disparity in SDG achievement across
706 NER districts. The environmental objectives, particularly SDG 13, exhibited the greatest
707 variation, highlighting the area's susceptibility to climate change and unequal application of
708 measures to mitigate its effects. This is consistent with previous studies (Yadav, 2013; Dikshit
709 & Dikshit, 2014; Singha, 2018), indicating that the topography and ecological sensitivity of the
710 NER amplify the impact of climate change, necessitating region-specific adaptation measures.
711 Addressing RQ5, our inequality analysis using the Theil index revealed that the highest
712 disparities were in SDG 7, 9, and 13. This spatial inequality underscores the need for targeted
713 interventions in the lagging districts to ensure balanced regional development. The results
714 indicate that, while certain areas have achieved notable improvements, others have fallen
715 behind considerably, potentially widening regional inequalities if not tackled through focused
716 interventions.

717 In response to RQ6, our analysis of the Evenness Index Scores (EIS) and Mean Index
718 Scores (MIS) highlighted significant disparities among NER states and districts. For instance,
719 Tripura showed both high EIS (73.77) and MIS (78.23), indicating balanced and strong overall
720 performance, whereas Nagaland had the lowest EIS (58.78), suggesting significant imbalances
721 in SDG achievements. The econ-SDGs showed considerable variation, with certain districts
722 displaying strong economic outcomes, whereas others struggled financially. This unequal
723 economic progress could be linked to various factors, including inadequate infrastructure,
724 restricted access to markets, and disparate levels of industrial development across different
725 districts. These findings align with the existing literature (Barua, 2020; De, 2021; Roy et al.,
726 2022; Das and Dutta, 2024) that points to infrastructural and connectivity challenges as major

727 impediments to economic growth in the region. The findings of this research shed light on the
728 complex interplay between the env- and econ-SDGs, underscoring the conflict between
729 progress and ecological preservation. In response to RQ4, our data envelopment analysis
730 showed that only 26 of 103 districts (25.24%) were non-efficient in converting environmental
731 SDG inputs into socioeconomic SDG outputs. This suggests that while many districts are
732 effectively balancing environmental and socioeconomic progress, there is room for
733 improvement in approximately a quarter of the region's districts. In some regions, economic
734 advancement may be achieved at the expense of environmental health, illustrating the ongoing
735 challenge of balancing development and sustainability.

736 In response to RQ2, our correlation and network analyses revealed strong positive
737 relationships between several SDGs, notably between SDG 1 and 8, SDG 3 and 4, and SDG 6
738 and 7. These findings indicate that advancements in social welfare are interrelated, with
739 progress in one domain potentially catalysing improvements in others. The region exhibited a
740 varied social landscape, with some areas demonstrating excellence in healthcare, educational
741 attainment, and gender parity, while others lagged. The robust links noted between SDGs 1, 3,
742 and 4 indicate that advancements in social welfare are interrelated, with progress in one domain
743 potentially catalysing improvements in others. Nevertheless, the uneven distribution of soc-
744 SDGs highlights the need for more encompassing policies tailored to meet the unique
745 requirements of the most underprivileged districts. This is particularly vital for promoting
746 gender equality (SDG 5), as certain areas continue to face considerable obstacles.

747 Addressing RQ3, our hierarchical clustering analysis revealed that districts with strong
748 environmental performance did not always excel in econ- or soc-SDG indicators, and vice
749 versa. This decoupling of environmental and socioeconomic progress suggests that policy
750 interventions must be carefully calibrated to avoid trade-offs that may undermine long-term
751 sustainability. The analysis of district groupings based on their SDG achievements showed that
752 areas with strong environmental performance did not always excel in econ- or soc-SDG
753 indicators; the reverse was also true. This decoupling of environmental and socioeconomic
754 progress suggests that policy interventions must be carefully calibrated to avoid trade-offs that
755 may undermine long-term sustainability (Barua, 2020; Kokho, 2021; Jain et al., 2022).
756 Addressing RQ7, our relative performance analysis revealed that only four NER districts
757 (3.88%) performed better than the national composite SDG score. However, 71% of the NER
758 districts performed equally or better than the national environmental SDG score, indicating
759 strong environmental sustainability efforts in the region.

760 Our findings suggest a need for targeted policy interventions that address the specific
761 challenges of each district. These include capacity building for low-performing districts;
762 establishing knowledge-sharing networks; implementing performance-based incentives; and
763 developing tailored strategies for climate action, social development, and economic growth.
764 This research underscores the critical importance of adopting a nuanced approach to
765 sustainable development in NER, considering the distinct geographical, environmental, and
766 socioeconomic characteristics of each district. It is imperative for policymakers to focus on
767 targeted initiatives that address existing disparities in SDG achievement, particularly in
768 underdeveloped areas. Additionally, promoting inter-district collaboration and information
769 exchange could contribute to more balanced and sustainable growth across NER. Subsequent
770 investigations should concentrate on gaining a more detailed understanding of the underlying
771 causes of SDG inequalities and formulating strategies that combine the env-, econ-, and soc-
772 SDGs to attain comprehensive and sustainable development in NER India.

773 This study had certain limitations. These limitations not only restrict a deeper
774 understanding of the region's sustainability but also present opportunities for future research
775 and policy improvements.

776 a) Data gaps and omissions: For each of the NER district's unique SDG indicators, the dataset
777 is deficient in relevant and trustworthy data. Significantly, two complete SDGs are missing:
778 Life Below Water (SDG 14) and Partnerships for the Goals (SDG 17). Although the authors
779 contest their apparent insignificance (as deemed 'not relevant' by NITI Aayog), these
780 objectives probably have important but indirect links to NER, particularly when considering
781 consumption-based effects (such as supply networks). A thorough investigation of possible
782 synergies, trade-offs, and feedback loops among all 17 goals is hampered by omitting these
783 SDGs, which is essential for a thorough sustainability assessment. In addition, data are
784 frequently absent at the indicator level, even for the selected 103 districts. The data custodian,
785 NITI Aayog, has to fix these omissions and gaps in the data.

786 b) Limited indicator coverage: The NER dataset features fewer indicators for the four major
787 goals (SDG 10, 5, 9, and 12) than other SDG datasets for India (e.g. The Indian City SDGs or
788 state-UT SDGs) (Figure. S.24). In addition, it contains only two indicators for Clean and
789 Affordable Energy (SDG 7). More indicators are typically required for thorough evaluation, as
790 it would give a complete and more complex picture of the progress made, particularly for those
791 with less coverage.

792 c) Absence of time-series data: There is only one snapshot of SDG performance available in
793 the current dataset. It is impossible to forecast future trends or evaluate progress over time

794 because of the lack of time-series data. Evaluating India's chances of achieving the SDGs for
795 NER by 2030 is difficult in the absence of this temporal factor.

796 d) Incomplete district coverage: For 17 districts spread throughout the eight NER states, SDG
797 data are completely absent. Removing these districts diminishes the representativeness of the
798 dataset and restricts our comprehension of the sustainability landscape of the region as a whole.
799 To fully understand NER's progress towards the SDGs, data for every district must be included.

800 Our analysis has critical implications for sustainable development in Northeast India.
801 Significant variations in SDG achievement across districts, particularly in environmental goals,
802 underscore the need for tailored district-specific approaches. The decoupling of environmental
803 and socioeconomic progress in many districts highlights the challenge of balancing
804 development and ecological preservation, suggesting the need for green growth models. Strong
805 positive correlations between certain SDGs indicate that integrated approaches to social and
806 economic development could yield synergistic benefits. However, the high disparity in climate
807 action performance underscores a region's vulnerability to climate change impacts, implying
808 an urgent need for climate-resilient strategies. The varied social and economic landscapes
809 across districts suggest that development initiatives need to be more context-specific,
810 addressing unique local challenges. These implications collectively highlight the need for a
811 nuanced, multi-dimensional approach to sustainable development in NER, balancing
812 environmental conservation with socioeconomic progress, addressing regional disparities
813 through targeted interventions, and fostering inclusive growth that leaves no district behind.

814

815 **Policy suggestions**

816 This study examined the performance, interrelationship, and efficiency of the SDGs in
817 103 districts of NER India. A few suggestions for policymakers can be categorised into three
818 groups: environment, society, and economy (Figure 8). These policy recommendations provide
819 a structured approach towards achieving balanced regional development while addressing the
820 specific challenges identified through SDG performance analysis. The success of these
821 interventions relies heavily on coordinated implementation and continuous monitoring of
822 outcomes across all three dimensions of sustainability.

823

824 The environmental policy framework should encompass a comprehensive approach to capacity
825 enhancement and sustainable practices. This includes establishing targeted capacity-building
826 programs in environmentally vulnerable districts (e.g. Barpeta, and Mon) through systematic
827 training in waste management and climate action strategies. A regional knowledge-sharing

828 network was proposed to facilitate the transfer of best practices from high-performing districts
829 (e.g. Karimganj and West Jaintia Hills) for those requiring support. This framework advocates
830 performance-based incentivisation systems with transparent metrics for tracking environmental
831 SDG progress. Particular emphasis is placed on addressing climate action disparities in states
832 such as Arunachal Pradesh, Mizoram, and Nagaland through region-specific interventions
833 including renewable energy adoption and climate-smart agricultural practices.

834

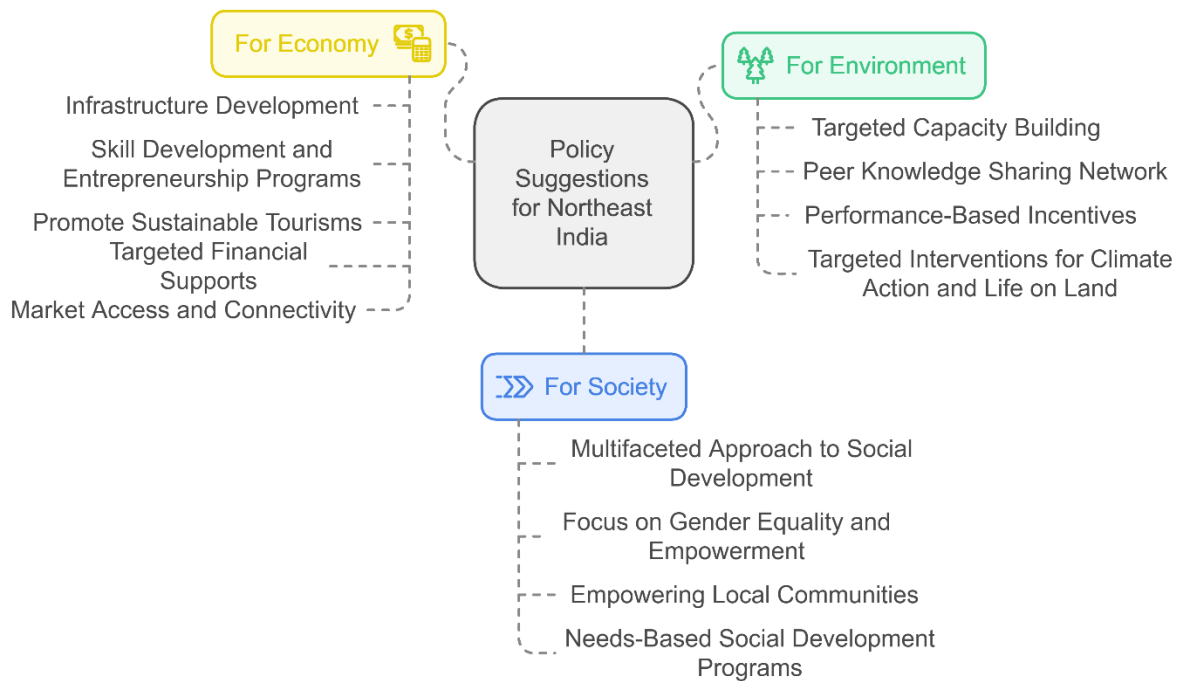
835 Social development strategies should adopt an integrated approach to address the
836 multidimensional challenges in underperforming districts. Priority interventions target districts
837 exhibiting significant social inequalities, such as the North Garo Hills and South Garo Hills,
838 through enhanced infrastructure and educational programs. The framework emphasises gender
839 mainstreaming across all social indicators, particularly in districts that demonstrate high social
840 SDG inequality. Community empowerment remains central to the strategy and advocates for
841 increased local participation and governance capacity-building. Additionally, needs-based
842 social development programs are proposed, exemplified by targeted educational initiatives in
843 Manipur, where SDG 4 disparities are pronounced.

844

845 The economic policy framework should prioritise infrastructural development in economically
846 disadvantaged districts such as Shi Yomi and Zunheboto, emphasising improved transportation
847 networks and market connectivity. A comprehensive approach to economic development
848 includes targeted skill enhancement and entrepreneurship programs tailored to the local
849 context. This strategy promotes sustainable tourism practices as an economic driver, while
850 preserving environmental integrity. Financial support mechanisms for micro, small, and
851 medium-sized enterprises (MSMEs) are proposed through grants and fiscal incentives.

852 Furthermore, the framework addresses market access limitations in states exhibiting high
853 economic SDG inequality, such as Arunachal Pradesh, Manipur, and Nagaland, through
854 strategic investments in the transportation infrastructure and digital connectivity enhancement.

855



856
857

Figure 8. Policy suggestions for Northeast India.

858

859 **Conclusion**

860 This study provides a comprehensive multi-dimensional assessment of SDG progress
 861 across Northeast India's districts, revealing critical disparities and complex interactions
 862 between environmental and socioeconomic goals. Our findings underscore the need for
 863 targeted, district-specific interventions to address developmental inequalities and promote
 864 balanced sustainable development in this ecologically sensitive, socioeconomically diverse
 865 region.

866 Key recommendations for policymakers include implementing tailored climate action
 867 strategies for districts lagging SDG 13, developing integrated socioeconomic policies that
 868 capitalise on strong positive correlations between related SDGs, establishing inter-district
 869 knowledge-sharing networks to disseminate best practices, and prioritising green growth
 870 models that balance environmental conservation with socioeconomic development. These
 871 targeted interventions aim to address the specific challenges faced by different districts while
 872 promoting balanced and sustainable development across the Northeast Region.

873 Future research should address this study's limitations by conducting longitudinal
 874 studies to track SDG progress over time, investigate the causes of intra-district inequalities,
 875 explore cross-border influences on sustainable development, and develop more comprehensive

876 indicators for underrepresented SDGs, particularly SDG 7, 14, and 17. These efforts will
877 provide a more nuanced understanding of the sustainable development challenges in Northeast
878 India and inform more effective targeted interventions.

879 By addressing these research gaps and implementing targeted policies, Northeast India
880 can make significant strides towards achieving the SDGs, potentially serving as a model for
881 sustainable development in other ecologically sensitive and diverse regions globally.

882

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