

1 Complex interlinkages, key objectives and nexuses
2 amongst the Sustainable Development Goals and
3 climate change

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15 **Summary**

16 *Background.* Global sustainability is underpinned by an enmeshed system of
17 complex socio-economic-ecological interactions. The objectives spelled out
18 by the United Nations' Sustainable Development Goals (SDGs) and Paris
19 Agreement are highly inter-dependent, making it difficult to establish policies
20 that are beneficial for both agendas.

21 *Methods.* We present a method to find interlinkages amongst the 17 SDGs
22 and climate change, using an extensive set of time series of 400 indicators
23 tracking the progress of the SDGs together with annual mean temperatures,
24 as the indicator for climate change. The method detects complex inter-
25 linkages amongst the objectives by computing significant partial distance
26 correlations—a nonlinear measure of conditional dependence that discounts

27 the effect of possible confounders. To capture the inter-related nature of the
28 objectives, we employ a network representation to determine the most impor-
29 tant objectives (highly central nodes in the network) and to obtain nexuses
30 of objectives (communities of nodes in the network).

31 *Findings.* We analysed the data of 35 groupings of countries reflecting var-
32 ious geographic and socio-economic criteria. We find that the significant
33 interlinkages, their centralities, and the nexuses of objectives vary greatly
34 across country groupings, yet *partnerships for the goals* (SDG 17) and *annual*
35 *temperatures* are highly central across many country groupings. Especially
36 in country groupings likely to be worst affected by climate breakdown, e.g.,
37 Africa, temperature shows a particularly strong inter-dependence with ur-
38 banisation, air pollution, and slum expansion (SDG 11). In many groupings
39 encompassing developing countries, a consistent nexus of objectives is formed
40 by poverty reduction (SDG 1), education (SDG 4), and economic growth (SDG
41 8), sometimes incorporating gender equality (SDG 5), and peace and justice
42 (SDG 16).

43 *Interpretation.* The large differences observed across groupings emphasise
44 the need to define policy priorities taking into account local circumstances.
45 Our analysis of centralities confirms the importance of multilateral agree-
46 ments and collaborations (SDG 17), which displays a strong interlinkage with
47 economic growth (SDG 8). Yet, our results also show that if the relationship
48 between SDGs 17 and 8 is left out, other SDGs, such as education (SDG 4)
49 and poverty reduction (SDG 1), become more central, thus suggesting routes
50 to more impactful international collaborations with a focus beyond economic

51 growth. The nexuses obtained in our analysis indicate that it is possible to
52 target multiple objectives simultaneously.

53 *Funding.* This work has been supported by the EPSRC under Grant EP/
54 N014529/1 funding the Centre for Mathematics of Precision Healthcare at
55 Imperial College London.

56 *Keywords:* SDGs, climate change, interlinkages, complex networks

57 **Introduction**

58 The current architecture of global sustainability is framed by two United
59 Nations' landmark agendas: the Paris Agreement and the 2030 Agenda for
60 Sustainable Development. Whilst the former focuses on strengthening the
61 global response to the climate crisis, the purpose of the latter with its 17
62 Sustainable Development Goals (SDGs) is to promote prosperity and social
63 justice by tackling inequality issues around poverty, hunger and gender, but
64 also environmental problems that include fostering biodiversity and increas-
65 ing resilience to a changing climate. Indeed, the objectives of both agendas
66 are inextricably linked [1, 2]. To date, interlinkages between the objectives of
67 both agendas have commonly been categorised as either counteracting (*trade-*
68 *offs*) or reinforcing (*synergies*) [3, 4, 5, 6, 7], e.g., by using a scale from −3
69 (“cancelling”) through 0 (“consistent”) to +3 (“indivisible”) [8]. Such a simple
70 categorisation based on a single number can be restrictive and misleading,
71 because it corresponds, in essence, to characterising the system of SDGs and
72 climate change as an ensemble of linearly related objectives [5, 9]. Thereby, it
73 neglects the known complexities of socio-economic(-ecological) systems, such
74 as *nonlinear* relationships between objectives. The presence of nonlinearity

75 opens the possibility that the interlinkage between two objectives may be
76 both counteracting *and* reinforcing, depending on the state of the objectives,
77 as exemplified in Figure 1.

78 In our analysis, we explicitly account for complex interactions between
79 objectives by calculating the *distance correlation*, a measure that can detect
80 arbitrary nonlinear relationships [10]. In addition, our method allows us to
81 control for shared (confounding) influences on each pairwise interlinkage by
82 conditioning on all possible subsets of the remaining objectives [11]. Choos-
83 ing the minimum distance correlation over these subsets, we measure the
84 direct strength of the relationship between any two objectives that cannot
85 be explained away by the influence of other objectives [12, Section 5.4.1], and
86 thus resolve the common problem of confounders in systems of more than
87 two variables. To the best of our knowledge, the issue of confounders has not
88 been considered in recent studies on SDG interlinkages [e.g., 13, 4, 5, 6].

89 We apply our approach to infer complex interlinkages using time series of
90 various indicators defined by the World Bank [14] to track progress towards
91 the 17 SDGs and, in addition, we introduce annual average temperatures as
92 an 18th variable to measure progress towards the Paris Agreement. This is
93 motivated by the fact that SDG 13, which is dedicated to climate action, is
94 focused on *adapting to* rather than *mitigating* the concurrent climate break-
95 down.

96 To integrate the collective information of all inferred interlinkages amongst
97 SDGs and mean temperatures, we turn to a network formulation to represent
98 the relationships the variables. We construct networks where the nodes cor-
99 respond to the 18 variables and the edges between them are weighted by

100 the minimum partial distance correlations. To capture the differences in
101 interlinkages across the world, we then compute such networks for various
102 regional, geopolitical, and economic groupings of countries. The resulting
103 networks allow us both to visualise our findings, and to carry out further
104 analyses; specifically, we find which of the SDGs are most central for the
105 different country groupings, and we obtain clusters of objectives (which we
106 denote ‘nexuses’) that show strong dependence to each other.

Latin America and the Caribbean

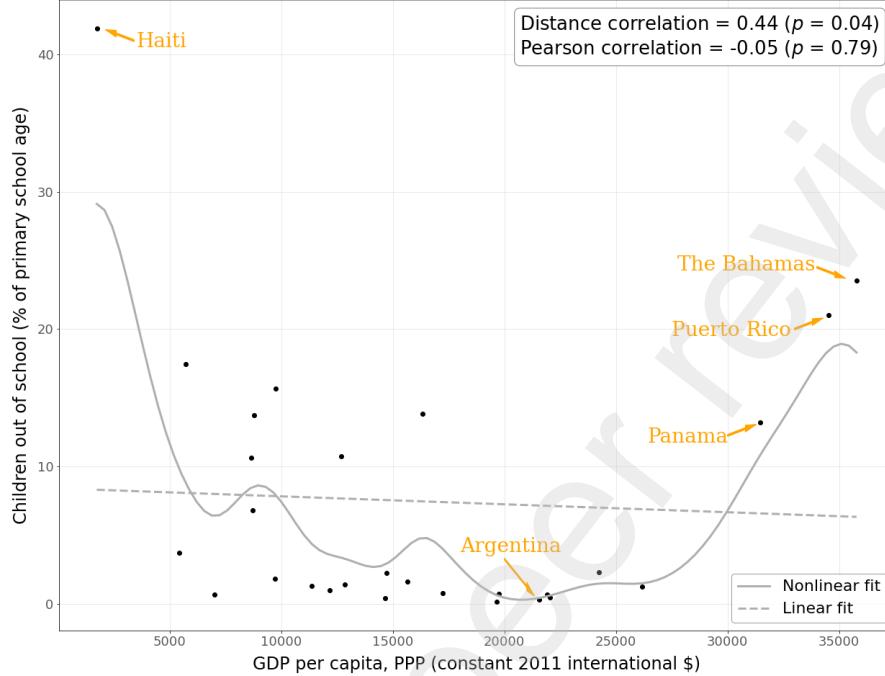


Figure 1: Example of a nonlinear relationship between two indicators: one of the indicators of SDG 4, children out of school (% of primary school age), and one of the indicators of SDG 17, gross domestic product (GDP) per capita, purchasing power parity (PPP) (constant 2011 international \\$) for countries in Latin America and the Caribbean. Each point corresponds to the latest available measurement of the two indicators for a country in this grouping. The dependence between the indicators is captured more accurately through a nonlinear (*solid line*) rather than linear (*dashed line*) relationship, as shown by a significant ($p < 0.05$) distance correlation versus a weak, non-significant linear (Pearson) correlation. The nonlinearity is reflected in a detrimental relationship at lower values of GDP that becomes synergistic at medium values, but turns detrimental again with higher values of GDP.

RESEARCH IN CONTEXT

Evidence before this study

Interlinkages amongst the SDGs have been studied, both qualitatively and quantitatively. Qualitative work has often focused on particular dependencies between a specific sustainability-related objective and some (other) SDGs, whereas quantitative methods have generally been used to analyse the SDGs from a more holistic perspective. Predominantly, quantitative studies have been based on linear or rank-based dependence measures (mostly Pearson or Spearman correlation), and have investigated pairwise relationships between SDGs without discounting the confounding influence of the remaining SDGs. Correlation coefficients so computed have then been used to classify relationships as synergies (positive coefficient) or trade-offs (negative coefficient). Taking correlation coefficients as edge weights in networks comprising the 17 SDGs, graph- and node-level metrics such as connectivity and degree or eigenvector centrality have sometimes been computed. Recent work has explored the contextual nature of the interlinkages by dis-aggregating countries into income-levels or regions, and has defined pairwise relationships for each grouping of countries separately instead of one generic global network.

Added value of this study

This analysis addresses three limitations of existing research, and develops a novel methodological framework for how interlinkages amongst the 17 SDGs plus climate change may be assessed in various regional,

geopolitical and economic contexts. Motivated by the holistic nature of policies addressing sustainability-related issues, we firstly view the 2030 Agenda for Sustainable Development and the Paris Agreement as *one interrelated system* of 18 temporal variables corresponding to the 17 SDGs and annual mean temperatures as an indicator of climate change. Secondly, we account for *arbitrary nonlinear relationships* between the 18 variables through a nonparametric measure of dependence, the partial distance correlation. Thirdly, we condition pairwise relationships on any possible subset of the remaining 16 nodes to control for shared influences and *isolate direct dependencies* between pairs of objectives. For each grouping of countries, we visualise and analyse the obtained dependencies as a network of 18 inter-related objectives. For each contextualised network, we compute the centrality, i.e., the relative importance, of each objective and apply community detection to find nexuses, i.e., groups of strongly connected objectives.

Implications of all the available evidence

Our analysis finds that the structure of the interlinkages amongst SDGs and climate change display strong differences across contexts. This supports that policies for sustainable development should be adjusted to the context in which they are executed. Primarily in developing countries, this is a challenging task due to the large foreign influence through investments with short-term economic interests. We show that foreign investment solely aimed at economic growth may slow or even hinder the progress of developing nations, and that such resources could

109

be spent more efficiently on other objectives such as poverty reduction or health, resulting in a general acceleration of progress towards meeting the SDGs and limiting global warming.

110 **Methods**

111 *Data*

112 Time series measuring progress on the SDGs are available from multiple
113 sources. Our data set is taken from the World Bank [14] and consists of
114 400 indicators measuring progress of the 17 SDGs over time from 2000 to
115 2019. We have chosen the World Bank data set because its indicators focus
116 on outcomes and impacts, rather than recording inputs and procedures as
117 those in the United Nations [15] data set. We also refrain from using the SDG
118 Index and Dashboard [16] data due to its more limited number of indicators
119 (85). For annual country-level average temperatures, we access data from
120 the Climatic Research Unit [17]. Since the World Bank data set for SDG 13
121 only had measurements for one year at the time of conducting this analy-
122 sis, we use instead indicators 13.1.1 and 13.1.3 of the United Nations [15]
123 data set for country-level time series of SDG 13, as was also done in Lusseau
124 and Mancini [5]. We remove the redundant indicator *net official develop-*
125 *ment assistance received (current US\$)* in SDG 17 since it already appears in
126 SDG 10. In summary, our data set consists of 400 indicators measured from
127 2000-2019 in 181 countries. Each indicator is associated with one of the 18
128 objectives (i.e., SDGs and climate change). Imputation of missing values is
129 carried out using a weighted average across those countries for which mea-

surements are available, with weights inversely proportional to the Euclidean distance between indicators. For our contextual analysis of results, countries are grouped regionally, geopolitically, and economically, resulting in a total of 35 different groupings (see Supplementary Material for full list of groupings and countries).

Notation

Let us consider a given grouping of n countries, and let us consider the countries as independent samples of the grouping. Hence the sample size of a grouping equals the number of countries in it. Let the variable X represent one of our 18 nodes (SDGs and climate change). Each variable X has an associated set of indicators I_X , and each indicator $i \in I_X$ has $T_{X,i}$ observations over time (some possibly imputed over the n samples). The variable X is constructed by appending all its indicators into a vector with dimensionality $d_X = \sum_{i=1}^{I_X} T_{X,i}$.

We now introduce the method to compute nonlinear dependencies between any two such variables X and Y , discounting the influence of the remaining 16 nodes.

Distance correlation

The *distance covariance* [10, 18] between $X \in \mathbb{R}^{d_X}$ and $Y \in \mathbb{R}^{d_Y}$, denoted by $\mathcal{V}^2(X, Y)$, is a measure of dependence between X and Y with the following important properties: (i) $\mathcal{V}^2(X, Y) \geq 0$, with equality if and only if X and Y are independent, i.e., it is a nonparametric measure that—unlike Pearson or Spearman correlation—is able to pick up non-monotonic dependencies (as the one seen in Figure 1 for example); (ii) $\mathcal{V}^2(X, Y) = \mathcal{V}^2(Y, X)$, i.e., it is

154 symmetric; and (iii) $\mathcal{V}^2(X, Y)$ is well-defined for $d_X \neq d_Y$. This last point
 155 makes it particularly well-suited for our setting since, due to the different
 156 numbers of indicators per SDG, dimensionality varies considerably between
 157 variables.

158 The *distance correlation* [10] $\mathcal{R}^2(X, Y)$ is the normalised version of the
 159 distance covariance and its properties include: (i) $0 \leq \mathcal{R}^2(X, Y) \leq 1$, i.e., it
 160 is a normalised quantity; and (ii) $\mathcal{R}^2(X, Y) = 1$ if and only if X and Y are
 161 linearly dependent. Since $\mathcal{V}^2(X, Y)$ and therefore also $\mathcal{R}^2(X, Y)$ are defined
 162 in terms of the underlying joint distribution of (X, Y) , denoted $P_{X,Y}$, which
 163 is usually unknown (see Supplementary Material for details), we require a
 164 way to estimate them from data.

165 *Unbiased estimator of the distance covariance*

166 Suppose that we have access to a sample of n pairs $(x_1, y_1), \dots, (x_n, y_n) \stackrel{\text{i.i.d.}}{\sim}$
 167 $P_{X,Y}$. We start by computing the *pairwise Euclidean distances* between sam-
 168 ples

$$a_{ij} := \|x_i - x_j\|, \quad b_{ij} = \|y_i - y_j\| \quad \forall i, j = 1, \dots, n. \quad (1)$$

169 Next, we define the *distance matrices* $A = (A_{ij})_{i,j=1}^n$ and $B = (B_{ij})_{i,j=1}^n$ as

$$A_{ij} = a_{ij} - \frac{1}{n-2} \sum_{l=1}^n a_{il} - \frac{1}{n-2} \sum_{k=1}^n a_{kj} + \frac{1}{(n-1)(n-2)} \sum_{k,l=1}^n a_{kl} \quad (2)$$

170 for $i \neq j$ and zero otherwise, and similarly for B_{ij} . An unbiased estimator of
 171 the *sample distance covariance* $\mathcal{V}_n^2(X, Y)$ is then given by

$$\mathcal{V}_n^2(X, Y) := \langle A, B \rangle = \frac{1}{n(n-3)} \sum_{i,j=1}^n A_{ij} B_{ij} \quad (3)$$

172 which converges to the population distance covariance $\mathcal{V}^2(X, Y)$ as $n \rightarrow \infty$
 173 almost surely [11].

174 *Partial distance correlation*

175 Pairwise dependence between X and Y may occur due to influences of the
 176 remaining 16 nodes. To establish the *direct* strength of dependence between
 177 any pair of nodes (X, Y) , we factor out shared effects from other nodes by
 178 conditioning on any subset of the remaining nodes: $\mathbf{Z} \subseteq \mathbf{V} \setminus \{X, Y\}$ where \mathbf{V}
 179 is the set of 18 nodes. The pairwise distances $c_{ij} := \|\mathbf{z}_i - \mathbf{z}_j\|$ are constructed
 180 from the vector of variables over the subset \mathbf{Z} , and the elements of the dis-
 181 tance matrix C_{ij} for \mathbf{Z} are computed analogously to (1) and (2). For any
 182 number $n \geq 4$ of samples $\{(x_i, y_i, \mathbf{z}_i)\}_{i=1}^n$ from (X, Y, \mathbf{Z}) , let $A(\mathbf{x})$, $B(\mathbf{y})$ and
 183 $C(\mathbf{z})$ be the distance matrices computed using the samples $\mathbf{x} = (x_1, \dots, x_n)$,
 184 $\mathbf{y} = (y_1, \dots, y_n)$, and $\mathbf{z} = (\mathbf{z}_1, \dots, \mathbf{z}_n)$, respectively. The projection $\mathcal{P}_{\mathbf{z}}(\mathbf{x})$ of
 185 $A(\mathbf{x})$ onto $C(\mathbf{z})$ and the complementary orthogonal projection $\mathcal{P}_{\mathbf{z}^\perp}(\mathbf{x})$ are
 186 defined by

$$\mathcal{P}_{\mathbf{z}}(\mathbf{x}) := \frac{\langle A(\mathbf{x}), C(\mathbf{z}) \rangle}{\langle C(\mathbf{z}), C(\mathbf{z}) \rangle} C(\mathbf{z}), \quad \mathcal{P}_{\mathbf{z}^\perp}(\mathbf{x}) := A(\mathbf{x}) - \mathcal{P}_{\mathbf{z}}(\mathbf{x}), \quad (4)$$

187 and similarly for the (orthogonal) projections of $B(\mathbf{y})$ onto $C(\mathbf{z})$, $\mathcal{P}_{\mathbf{z}}(\mathbf{y})$ and
 188 $\mathcal{P}_{\mathbf{z}^\perp}(\mathbf{y})$.

189 The *sample partial distance covariance* $\mathcal{V}_n^2(X, Y \mid \mathbf{Z})$ is then defined as
 190 the distance covariance between the orthogonal projections:

$$\mathcal{V}_n^2(X, Y \mid \mathbf{Z}) := \frac{1}{n(n-3)} \sum_{i \neq j}^n (\mathcal{P}_{\mathbf{z}^\perp}(\mathbf{x}))_{ij} (\mathcal{P}_{\mathbf{z}^\perp}(\mathbf{y}))_{ij}. \quad (5)$$

191 Hence, the partial distance correlation establishes the strength of the depen-
 192 dence between X and Y that cannot be explained away by \mathbf{Z} .

193 After normalising, we arrive at the *sample partial distance correlation*,

$$\mathcal{R}_n^2(X, Y \mid \mathbf{Z}) = \frac{\mathcal{V}_n^2(X, Y \mid \mathbf{Z})}{\sqrt{\mathcal{V}_n^2(X, X \mid \mathbf{Z}) \mathcal{V}_n^2(Y, Y \mid \mathbf{Z})}}. \quad (6)$$

194 Finally, as a measure of the direct dependence between X and Y we take
195 the *minimal partial distance correlation* over all possible subsets:

$$\mathcal{R}_n^2(X, Y) = \min_{\mathbf{Z} \subseteq V \setminus \{X, Y\}} \mathcal{R}_n^2(X, Y | \mathbf{Z}). \quad (7)$$

196 Since we pick the minimal partial distance correlation (7), this measure
197 reflects the strength of the dependence between X and Y that cannot be
198 explained by the subset \mathbf{Z} that influences both X and Y most.

199 In our case, there are 153 unique pairs of the 18 variables (SDGs and
200 climate change), and for each pair we compute the partial distance correla-
201 tion (7) by minimising over 65535 conditional sets \mathbf{Z} formed by the remain-
202 16 variables. This process is repeated for each of the 35 country groupings to
203 obtain a contextualised description of the dependencies amongst SDGs and
204 climate change.

205 *Statistical testing: significant partial distance correlations*

206 To determine which of the minimal partial distance correlations (7) are
207 statistically significant, we test against the null hypothesis of X being inde-
208 pendent of Y given \mathbf{Z} . To achieve this, we carry out a permutation test where
209 we produce 10000 random reorderings of the samples \mathbf{x} whilst the samples \mathbf{y}
210 and \mathbf{z} are unchanged [11]. We reject the null hypothesis if the correspond-
211 ing p -value is less or equal to a predetermined statistical significance level
212 $\alpha \in \{0.01, 0.05, 0.1\}$. When the minimal partial distance correlation between
213 two variables X and Y is found to be significant for a given level α , we say
214 there is an interlinkage between the two variables.

215 *Network analysis*

216 For any given grouping of countries, our procedure thus produces a set of
217 statistically significant interlinkages amongst the 18 variables, with strengths
218 given by (7). To further analyse the inter-related dependencies, we invoke
219 a network representation, where the variables are nodes and the significant
220 minimal partial distance correlations (7) constitute weighted edges.

221 Having thus constructed undirected weighted networks for every country
222 grouping, we conduct two further analyses. Firstly, we quantify the relative
223 importance of each of the 18 nodes using eigenvector centrality, a network-
224 theoretic measure that gives a high centrality score to nodes that are con-
225 nected to other similarly highly central nodes. Secondly, we apply community
226 detection by modularity maximisation to group nodes into *nexus*es, i.e., clus-
227 ters of objectives that have stronger partial distance correlations within the
228 nexus than across the other nexuses. Both methods are described in detail
229 in the Supplementary Material.

230 **Results**

231 We have performed our analysis for 35 groupings of countries reflecting
232 a variety of geographical and socio-economic factors. The analysis of all
233 country groupings is presented in full in the Supplementary Material where
234 all significant interlinkages, networks of objectives, eigenvector centralities,
235 and nexuses can be found, whereas we only point out a few key findings here.

236 An example of our results is presented in Figure 2, where we show the
237 network of the 18 variables and their eigenvector centralities for the country
238 grouping of Emerging Markets. In the network, the line type of the edges

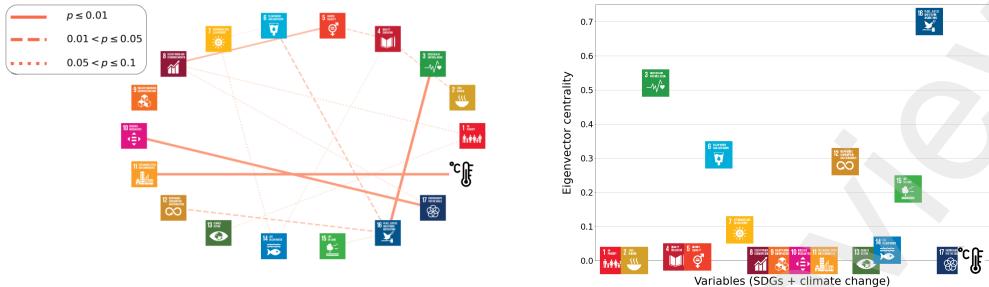


Figure 2: (*left*) The network of SDGs and climate change, illustrated by the symbols for the 17 SDGs and a thermometer, respectively, for the grouping Emerging Markets (BRICS + N-11). Darker and wider edges indicate larger partial distance correlation coefficients, and p -values of the corresponding conditional independence tests are visualised as different edge styles. (*right*) Normalised eigenvector centralities as a measure of the relative importance within the network for each of the 18 variables.

reflects the statistical significance level and the width is proportional to the minimal partial distance correlation (7). The eigenvector centralities are computed from the network with significance $\alpha = 0.1$.

As shown in the Supplementary Material, we find vastly different networks and eigenvector centralities across country groupings. Table 1 lists the three most central variables along with their eigenvector centrality for each of the 35 groupings. This differing picture across the contextualised networks supports the call to view the system of inter-related SDGs and climate change through jurisdictional and political lenses [19].

Most central objectives across different country groupings

We find that SDG 17, dedicated to enhancing national and international partnerships in the public and private sectors, is amongst the top three most

Table 1: The three most important variables (with their eigenvector centrality) in every grouping. Variables are named 1 to 17 for the SDGs and T for annual mean temperatures. The three variables that appear most often across all groupings are highlighted in bold.

Country grouping	Most central variables		
	1 st	2 nd	3 rd
World	7 (0.46)	17 (0.43)	9 (0.34)
Global North	17 (0.53)	9 (0.39)	7 (0.36)
Global South	4 (0.41)	17 (0.41)	8 (0.31)
Africa	17 (0.38)	T (0.34)	9 (0.33)
Northern and Western Africa	17 (0.62)	12 (0.55)	8 (0.33)
Middle Africa	T (0.66)	11 (0.55)	10 (0.44)
Southern and Eastern Africa	T (0.54)	11 (0.48)	3 (0.39)
Sub-Saharan Africa	11 (0.35)	4 (0.33)	10 (0.33)
Americas	17 (0.43)	7 (0.39)	1 (0.37)
North and Central America	6 (0.66)	16 (0.55)	7 (0.43)
Caribbean	11 (0.71)	T (0.64)	15 (0.30)
South America	11 (0.58)	12 (0.57)	3 (0.44)
Latin America and the Caribbean	17 (0.46)	1 (0.40)	11 (0.37)
Asia	17 (0.48)	8 (0.34)	7 (0.32)
Central and Eastern Asia	T (0.58)	3 (0.56)	11 (0.43)
South-eastern Asia	11 (0.58)	17 (0.56)	T (0.41)
Southern Asia	3 (0.60)	T (0.57)	2 (0.42)
Western Asia	T (0.58)	12 (0.41)	1 (0.39)
Europe	17 (0.47)	9 (0.42)	7 (0.35)
Northern Europe	11 (0.68)	10 (0.39)	17 (0.37)
Eastern Europe	17 (0.71)	9 (0.56)	T (0.43)
Southern Europe	7 (0.64)	5 (0.35)	4 (0.33)
Western Europe	T (0.71)	13 (0.71)	(all others)
Oceania	1 (0.57)	7 (0.57)	9 (0.36)
Oceania (excl. AUS + NZ)	2 (0.71)	1 (0.59)	13 (0.39)
High Income	17 (0.47)	7 (0.43)	9 (0.38)
Upper middle Income	T (0.45)	17 (0.41)	8 (0.30)
Lower middle Income	11 (0.41)	17 (0.38)	T (0.36)
Low Income	17 (0.55)	12 (0.40)	9 (0.33)
Least Developed Countries (LDC)	8 (0.48)	4 (0.46)	1 (0.43)
Land Locked Developing Countries (LLDC)	3 (0.53)	T (0.42)	11 (0.38)
Small Island Developing States (SIDS)	8 (0.48)	17 (0.42)	4 (0.35)
G20	7 (0.48)	17 (0.40)	2 (0.36)
Emerging Markets (BRICS + N-11)	16 (0.70)	3 (0.52)	6 (0.31)
OPEC	1 (0.52)	17 (0.51)	4 (0.44)

251 central SDGs for 20 out of the 35 groupings. The annual average temper-
252 ature variable appears central in country groupings that are likely to be
253 amongst the worst affected by, and least responsible for, climate change (e.g.,
254 Africa, Caribbean, Central America, Southern Asia, Western Asia) [20]. In
255 the majority of country groupings, efforts encompassing climate *adaptation*,
256 as measured by the indicators of SDG 13, are not as important as annual
257 average temperatures, our indicator for climate *mitigation*. We also find that
258 economic growth (SDG 8) is less important in most groupings, especially in
259 developed ones (e.g., High Income), although it is vital elsewhere (e.g., Least
260 Developed Countries and Small Island Developing States). Other variables
261 which are frequently amongst the three most important ones are SDG 11,
262 whose indicators track air pollution and urban population (e.g., Sub-Saharan
263 Africa, South-eastern Asia, Northern Europe); SDG 4 around primary, sec-
264 ondary and tertiary education (e.g., Global South); and SDG 1, considering
265 monetary and multidimensional poverty, and the coverage of social protection
266 schemes (e.g., Americas, Oceania).

267 *Comparison between Global North and Global South*

268 Table 2 summarises the comparison of the strength of the interlinkages
269 and the centralities of the objectives between the Global South and Global
270 North groupings. We find that SDG 17 (*partnerships for the goals*) is in-
271 volved in four out of the five strongest interlinkages in the Global North,
272 which results in it being the most important SDG by a considerable mar-
273 gin in that grouping. In contrast, SDG 17 only appears twice in the five
274 strongest interlinkages in the Global South, exactly as often as SDG 4 on ed-
275 ucation, which share the highest centrality in this grouping. The interlinkage

Table 2: (*left*) The five interlinkages with largest partial distance correlations and (*right*) the five objectives with the largest eigenvector centralities for the Global South (*top five rows*) and Global North (*bottom five rows*). The three interlinkages and objectives with the largest differences between Global South and Global North are highlighted.

Interlinkages				Objectives			
Edge	Partial distance correlation			Eigenvector centrality			Difference
	Global South	Global North	Difference	Node	Global South	Global North	
8–17	0.18	0.27	−0.09	17	0.41	0.53	−0.12
11–T	0.17	0.08	0.09	4	0.41	0.10	0.31
3–4	0.17	0.03	0.14	8	0.31	0.25	0.06
4–17	0.16	0.01	0.15	1	0.27	0.20	0.17
5–8	0.16	0.00	0.16	16	0.25	0.04	0.21
9–17	0.09	0.32	−0.23	17	0.41	0.53	−0.12
8–17	0.18	0.27	−0.09	9	0.21	0.39	−0.18
2–7	0.06	0.22	−0.17	7	0.25	0.36	−0.11
7–17	0.15	0.21	−0.07	2	0.21	0.29	−0.08
1–17	0.09	0.20	−0.11	T	0.16	0.26	−0.1

of SDG 17 to SDG 8 (*economic growth*) is the only one present amongst the five strongest interlinkages in both the Global South and Global North, although it is stronger in the latter. The largest discrepancy in the strength of the interlinkages between these two country groupings appears for 5–8, 9–17 and 2–7, and the largest discrepancy in the centralities of the objectives is observed in the importance of SDG 4 (*quality education*), 16 (*peace, justice and strong institutions*), and 9 (*industry, innovation and infrastructure*). Of these three, the former two are more central in the Global South and SDG 9 more in the Global North.

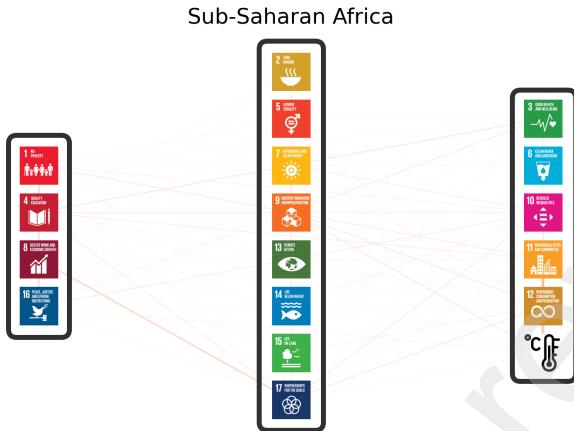


Figure 3: Nexuses in Sub-Saharan Africa

285 Nexuses of strongly interconnected objectives

286 We also find a varied set of nexuses of objectives across the different coun-
287 try groupings. One nexus, which is found across many groupings, especially
288 of developing countries, combines poverty reduction (SDG 1), education (SDG
289 4), and economic growth (SDG 8). In Sub-Saharan Africa, for instance, this
290 nexus additionally includes SDG 16 (see Figure 3), which is frequently seen
291 besides SDG 5 joining this triplet. In this country grouping, we can also
292 observe a strong bond between urbanisation (SDG 11), clean water and sani-
293 tation (SDG 6), health (SDG 3) and inequality (SDG 10), typical for developing
294 nations. Two other repeatedly observed nexuses are (i) around the connection
295 between SDG 7 and 9, which sometimes includes mean annual temperatures,
296 SDG 10 or 12 (e.g., in Latin America and the Caribbeans); and (ii) around
297 the conjunction of health (SDG 3) and clean water and sanitation (SDG 6),
298 e.g., in Europe.

299 Discussion

300 Our approach offers a principled method to quantify the strengths of
301 interlinkages amongst the 17 SDGs and climate change allowing for nonlin-
302 ear relationships between variables and controlling for unwanted effects of
303 confounding variables. Furthermore, our network analyses enable us to rank
304 variables according to their relative importance and group them into nexuses.

305 The stark differences observed across country groupings call for highly
306 contextualised policy prioritisation and resource allocation, and motivates
307 the inclusion of a diverse group of stakeholders in decision-making processes,
308 specifically stressing the importance of local expertise. For issues related
309 to climate breakdown, the benefits of including local knowledge have been
310 clearly communicated [21, 22], but such involvement must be expanded to the
311 larger set of social and economic objectives for an efficient implementation of
312 both the 2030 Agenda for Sustainable Development and the Paris Agreement.

313 The central role of SDG 17, which advocates for the enhancement of mul-
314 tilateral collaborations, and its strong interlinkage to SDG 8 on economic
315 growth appear in many country groupings, e.g., in the Global South and
316 Global North (Table 2). Given the fact that the indicators of SDG 17 in-
317 clude, e.g., foreign private investment, the strong association between SDGs
318 17 and 8 in the Global South may originate from the adoption of economic
319 growth as an essential objective for prosperity, arguably facilitated by eco-
320 nomic interests of organisations from the Global North [23]. However, if we
321 break the interlinkage between SDGs 17 and 8 in our network, SDG 8 be-
322 comes less central than SDGs 4, 1, 3, 7, 11 and 16 in the Global South. These
323 findings suggest that the association of foreign investment with local priori-

ties in poverty reduction, public health, access to (renewable) electricity, and peace and strong jurisdictions, could be strengthened by reducing the direct interlinkage of foreign investment with economic growth. Despite its wide acceptance as the panacea to humanitarian crises of all sorts in the current capitalocene, economic growth has been devastating for many livelihoods and natural habitats [24]. For the Global South, our analyses of centrality and nexuses of objectives indicate the need to strengthen the associations between SDG 17 and variables like SDG 4 [25] that are more central for both the 2030 Agenda for Sustainable Development and the Paris Agreement.

There are some methodological limitations to our analysis worth stating. Our time series have measurements at most over the last 20 years. This range is enough to discover many interlinkages, but some could be missed—most likely to happen for the environmental objectives of SDGs 14 and 15. Related research suggests dependencies of these objectives to health and wellbeing (SDG 3) [26], food (SDG 2) and water security (SDG 6), disaster risk reduction (SDG 13), and obviously climate change [27].

The traditional viewpoint that classifies interlinkages as *synergies* (positive correlation) and *trade-offs* (negative correlation) with Pearson [9, 28] and Spearman [3, 4, 6] correlations does not apply to our framework. As shown in Figure 1, our measure of dependence, the partial distance correlation, enables us to discover relationships of any nonlinear and nonmonotonic form, thus capturing the complex interplays characteristic of socio-economic(-ecological) systems. Indeed, by comparing the interlinkages found here to those obtained by Warchold et al. [3] using Spearman correlations, we find strong dependences for many pairs of variables (e.g., 7–9, 7–17, 8–17 for all countries) that

349 were labelled as ‘not classified’ under their rank-based analysis Warchold
350 et al. [3]. Our more accurate understanding of the complex dynamics at
351 play can lead to more successful interventions to reach the 2030 Agenda for
352 Sustainable Development and the Paris Agreement.

353 **Contributors**

354 F.L. conceived the study, F.L. and J.v.K. designed the statistical models,
355 F.L. carried out the analyses and analysed the results. All authors wrote the
356 manuscript.

357 **Declaration of interests**

358 The authors declare no competing interests.

359 **Data sharing**

360 Data used for our analyses are freely available from the sources cited in the
361 Methods section of this article. Web links to the corresponding data bases
362 are given in the references.

363 **Code sharing**

364 Code to reproduce these findings and visualisations may be found at <https://github.com/felix-laumann/SDG-networks>.
365

366 Acknowledgements

367 We would like to thank the World Bank, the United Nations, and the Climate
368 Research Unit for curating and collating the SDG indicator and temperature
369 data, and making them easily accessible. We would also like to express our
370 gratitude to Julien Cornebise, Carlos Ramos Carreño, and Mike Tennant for
371 insightful discussions.

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457 ing of communities in large networks, Journal of statistical mechanics:
458 theory and experiment 2008 (2008) P10008.

459 **Supplementary Material**

460 *Distance covariance*

461 Intuitively, the distance covariance assesses dependence by measuring to
462 what extent the distances between pairs of observations of X and Y are
463 correlated with each other. Formally, it is defined as

$$\mathcal{V}^2(X, Y) = \int |f_{X,Y}(t, s) - f_X(t)f_Y(s)|^2 w(t, s) dt ds \quad (8)$$

464 where $w(t, s) := (|t|_{d_X}^{1+d_X} |s|_{d_Y}^{1+d_Y})^{-1}$, and where the *characteristic function* f_Z
465 of a random variable Z is denoted as $f_Z(t) = \mathbb{E}[e^{itZ}]$ with $i^2 = -1$.

466 The *distance correlation* $\mathcal{R}^2(X, Y)$ is the normalised distance covariance

$$\mathcal{R}^2(X, Y) = \frac{\mathcal{V}^2(X, Y)}{\sqrt{\mathcal{V}^2(X, X)\mathcal{V}^2(Y, Y)}}, \quad (9)$$

467 if $\mathcal{V}^2(X, X)\mathcal{V}^2(Y, Y) > 0$, and zero otherwise.

468 *Comparison of Pearson, Spearman and distance correlation*

469 We compare the Pearson correlation coefficient r , which is a measure of
470 linear dependence between two random variables X and Y , and the Spearman
471 correlation coefficient ρ , a nonlinear monotonic measure of dependence, to
472 our utilised nonlinear nonmonotonic measure of dependence, the distance cor-
473 relation *dcor*. Figure 4 demonstrates that the distance correlation captures
474 the relationships more accurately than the Pearson and Spearman correlation
475 coefficients for (nonmonotonic) nonlinear dependencies such as $Y = X^2$ and
476 $Y = X^4$, and performs comparable to the Pearson correlation for $Y = X^3$.
477 All three measures perform equivalently for $Y = X$, obviously.

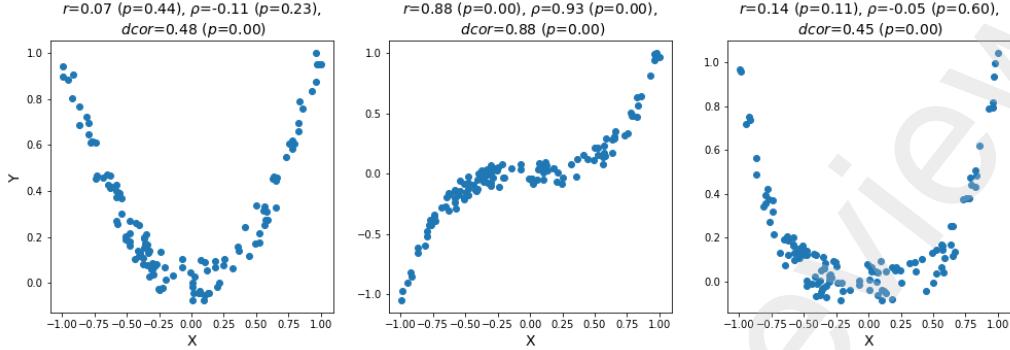


Figure 4: Pearson correlation r compared to distance correlation $dcor$ for, from left to right, $Y = X^2$, $Y = X^3$, and $Y = X^4$. Random uniform noise $[-0.1, 0.1]$ is added to Y .

478 Eigenvector centrality

479 The eigenvector centrality is defined in terms of the weighted adjacency
480 matrix $(K_{vt})_{v,t \in \mathbf{V}}$ with K_{vt} equal to the weight of the edge between nodes v
481 and t (i.e., the minimum partial distance correlation over all subsets). The
482 eigenvector centrality q_v for node v is then given by the v^{th} component of the
483 normalised eigenvector \mathbf{q} corresponding to the largest eigenvalue λ_{\max} of K ,
484 i.e.,

$$q_v = \frac{1}{\lambda_{\max}} \sum_{t \in \mathbf{V}} K_{vt} q_t, \quad (10)$$

485 where λ_{\max} is the largest eigenvalue in the eigenvector equation $K\mathbf{q} = \lambda\mathbf{q}$,
486 subject to $\mathbf{q} \neq 0$.

487 Community detection

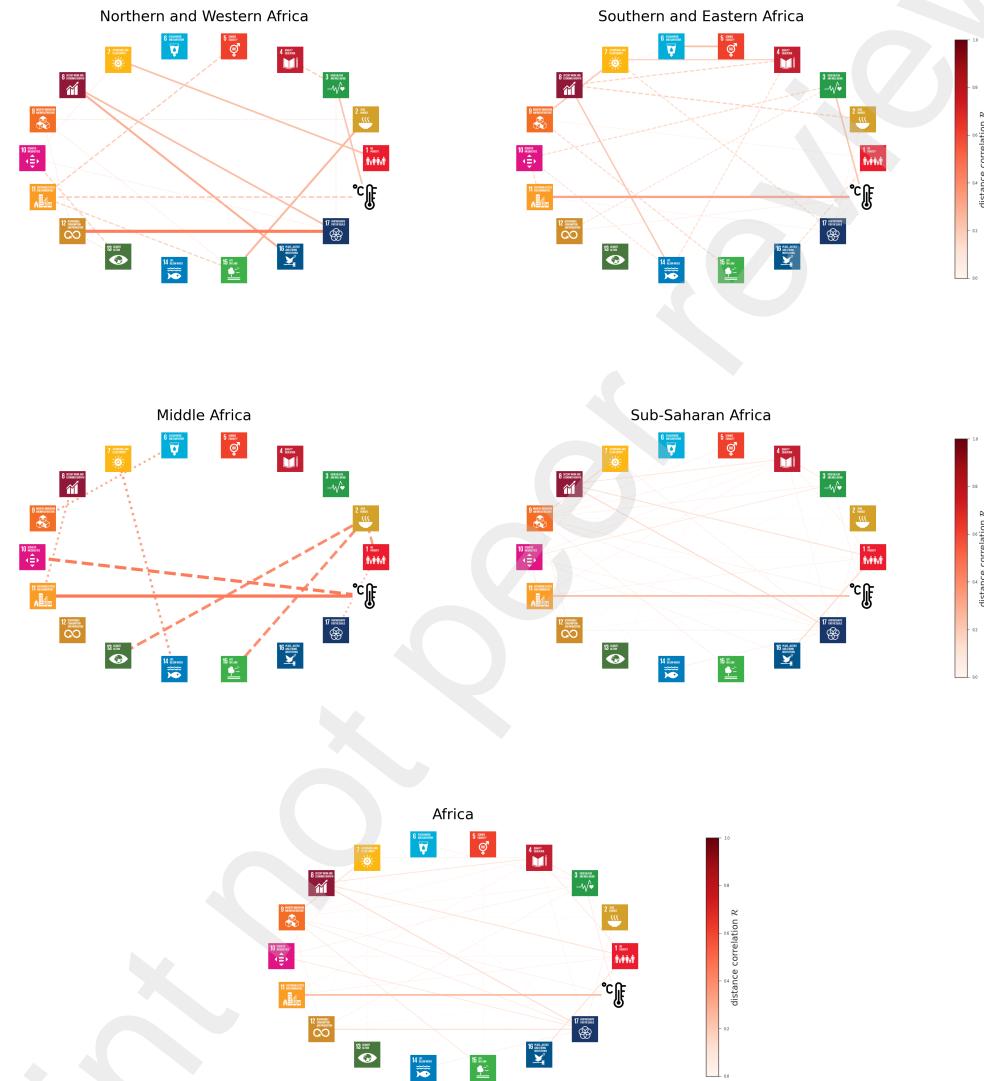
488 For any network of vertices and edges, $\mathcal{G} := (\mathbf{V}, \mathbf{E})$, denote by $\mathcal{P} =$
489 $\mathcal{P}_1, \dots, \mathcal{P}_p$ a partition of vertices \mathbf{V} , and call it one clustering of \mathcal{G} with
490 nonempty clusters \mathcal{P}_i . The modularity \mathcal{M} is a measure that describes the
491 strength of any such clustering by comparing the sum of its intra-cluster

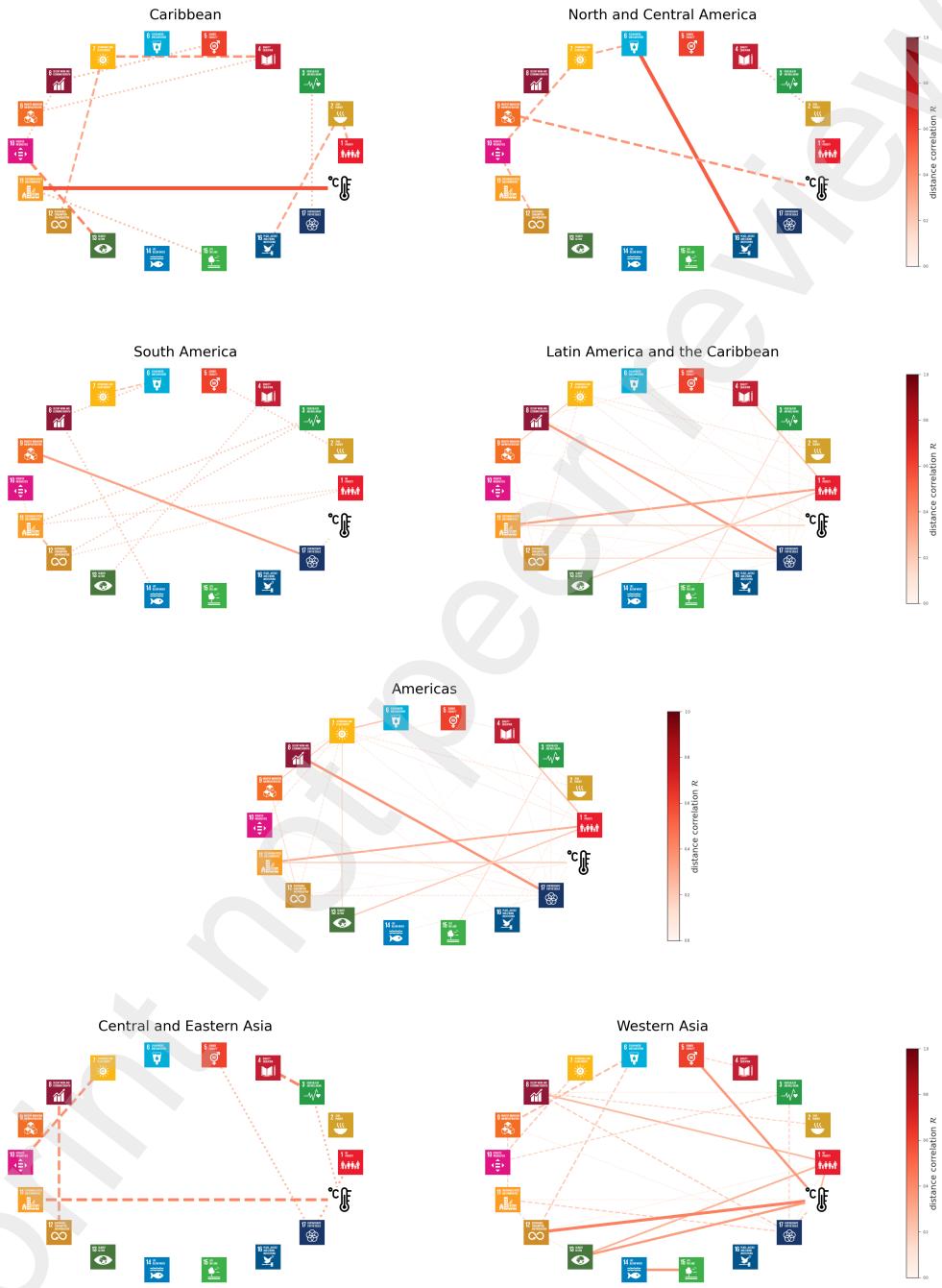
492 weighted edges against the sum of its inter-cluster ones. It is computed as
493 the fraction between the expected number of randomly distributed edges sub-
494 tracted from the actual number of edges within clusters and the total weights
495 of edges $m = |\mathcal{E}|$. It is defined as

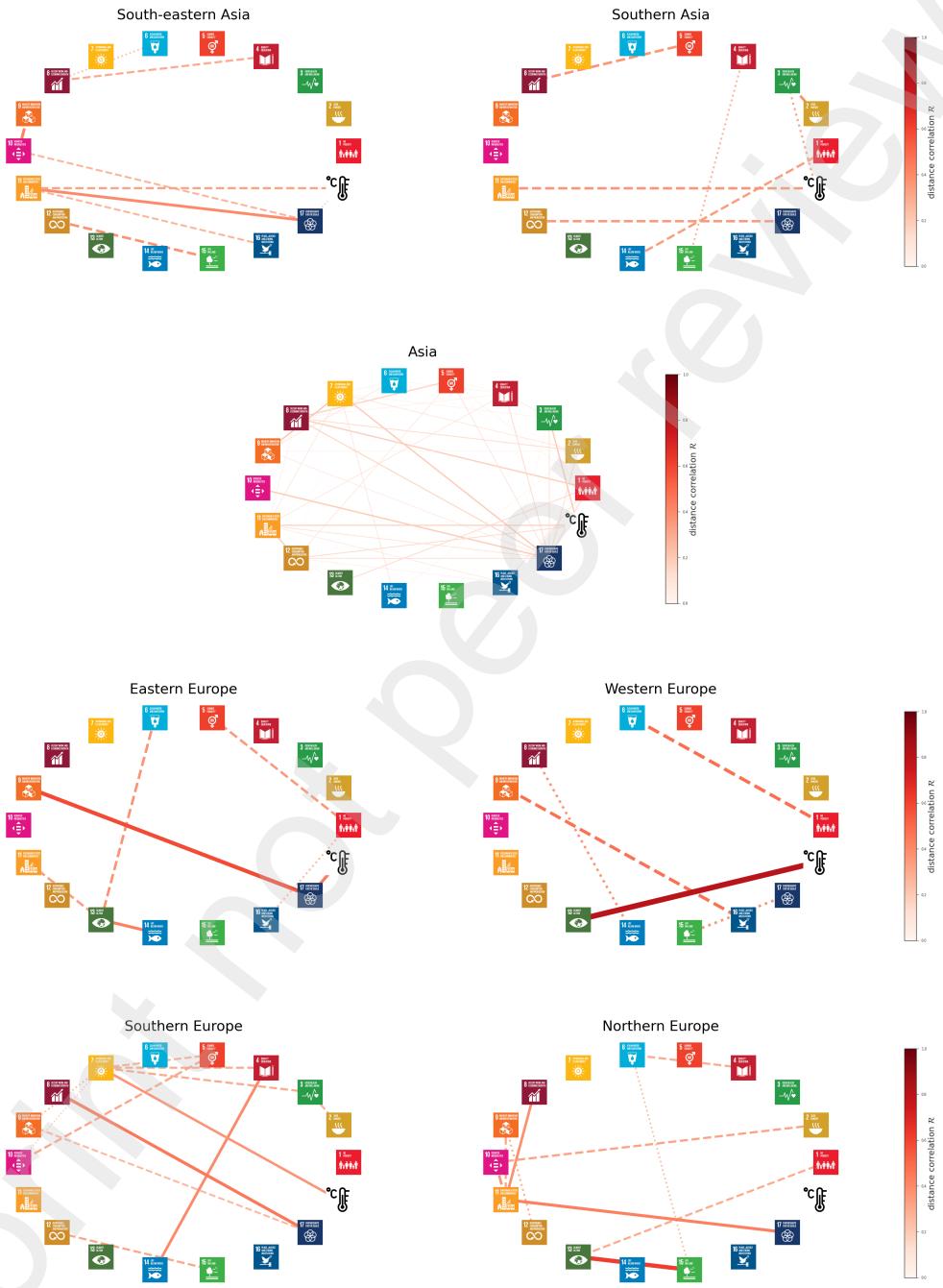
$$\mathcal{M}(\mathcal{P}) := \sum_{\mathcal{P}_i \in \mathcal{P}} \left[\frac{|E(\mathcal{P}_i)|}{m} - \left(\frac{|E(\mathcal{P}_i)| + \sum_{\mathcal{P}_j \in \mathcal{P}} |E(\mathcal{P}_i, \mathcal{P}_j)|}{2m} \right)^2 \right]. \quad (11)$$

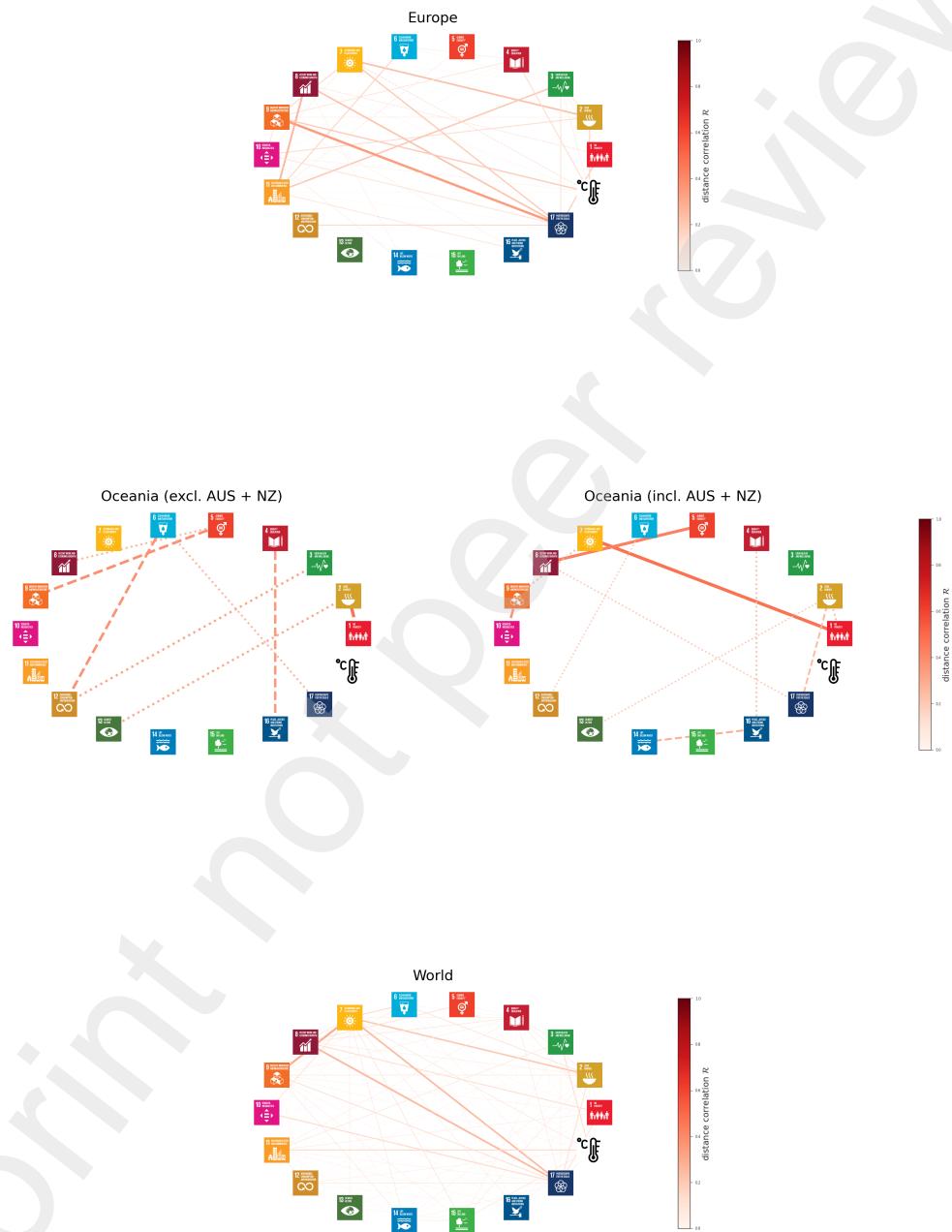
496 Note that the total number of clusters p does not have to be pre-defined
497 and emerges naturally with the maximisation of the modularity \mathcal{M} . We use
498 the Louvain algorithm [29] and randomise the node evaluation order and
499 the community evaluation order 100 times to search over the space of candi-
500 date clusterings efficiently and find the candidate with maximum modularity
501 almost surely.

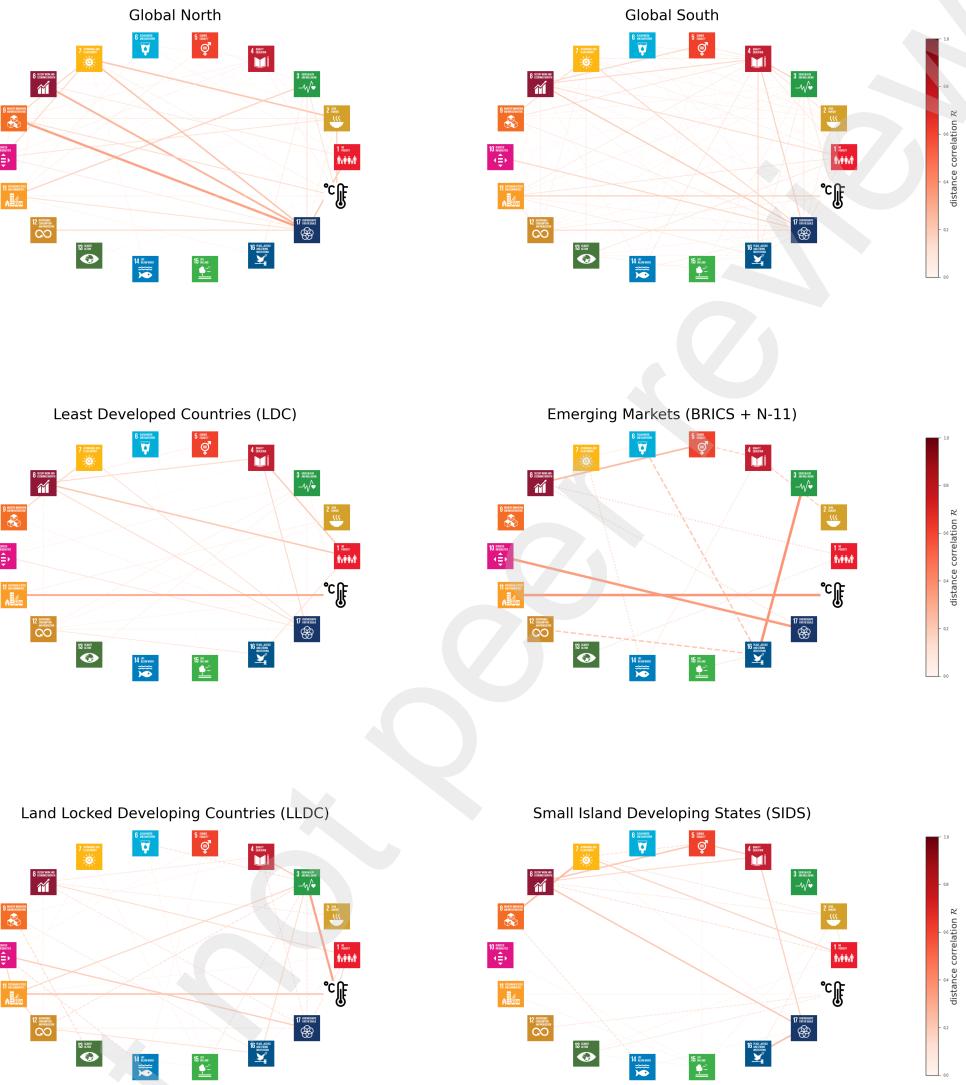
502 *Networks of groupings*





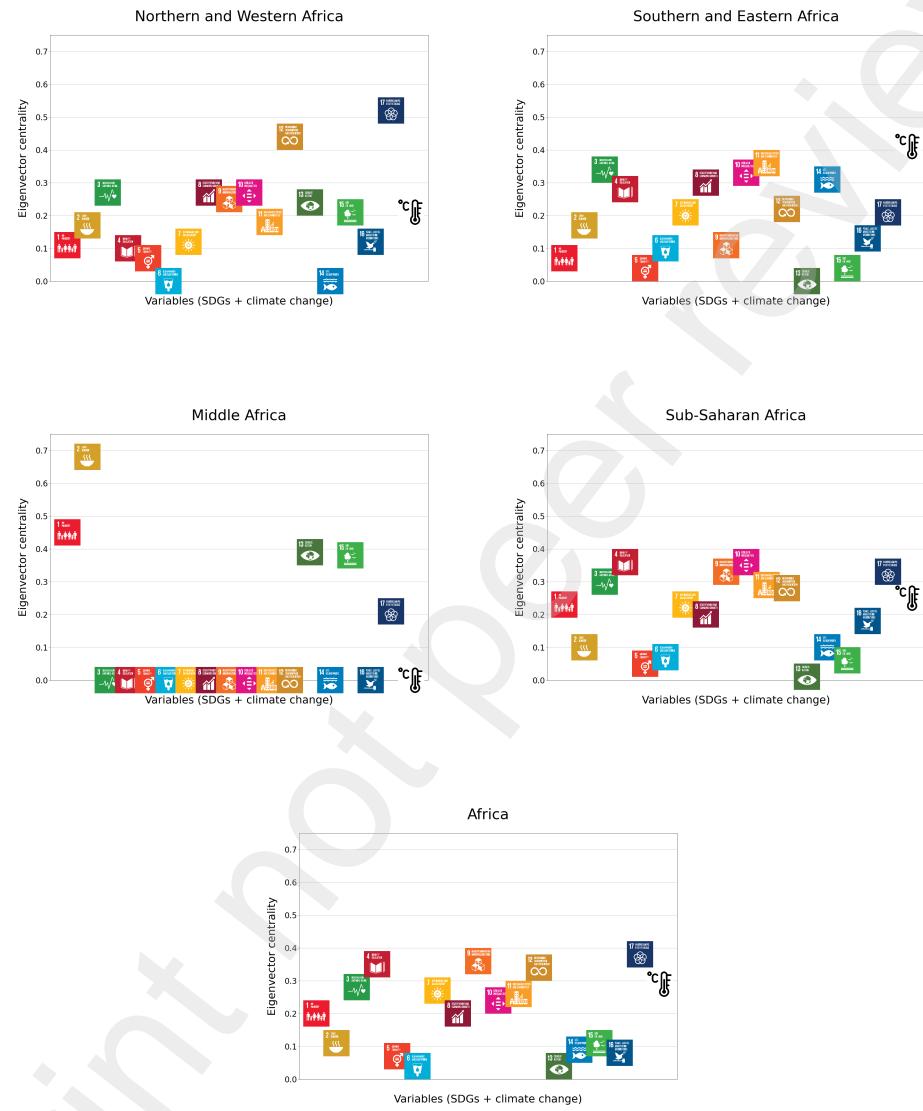


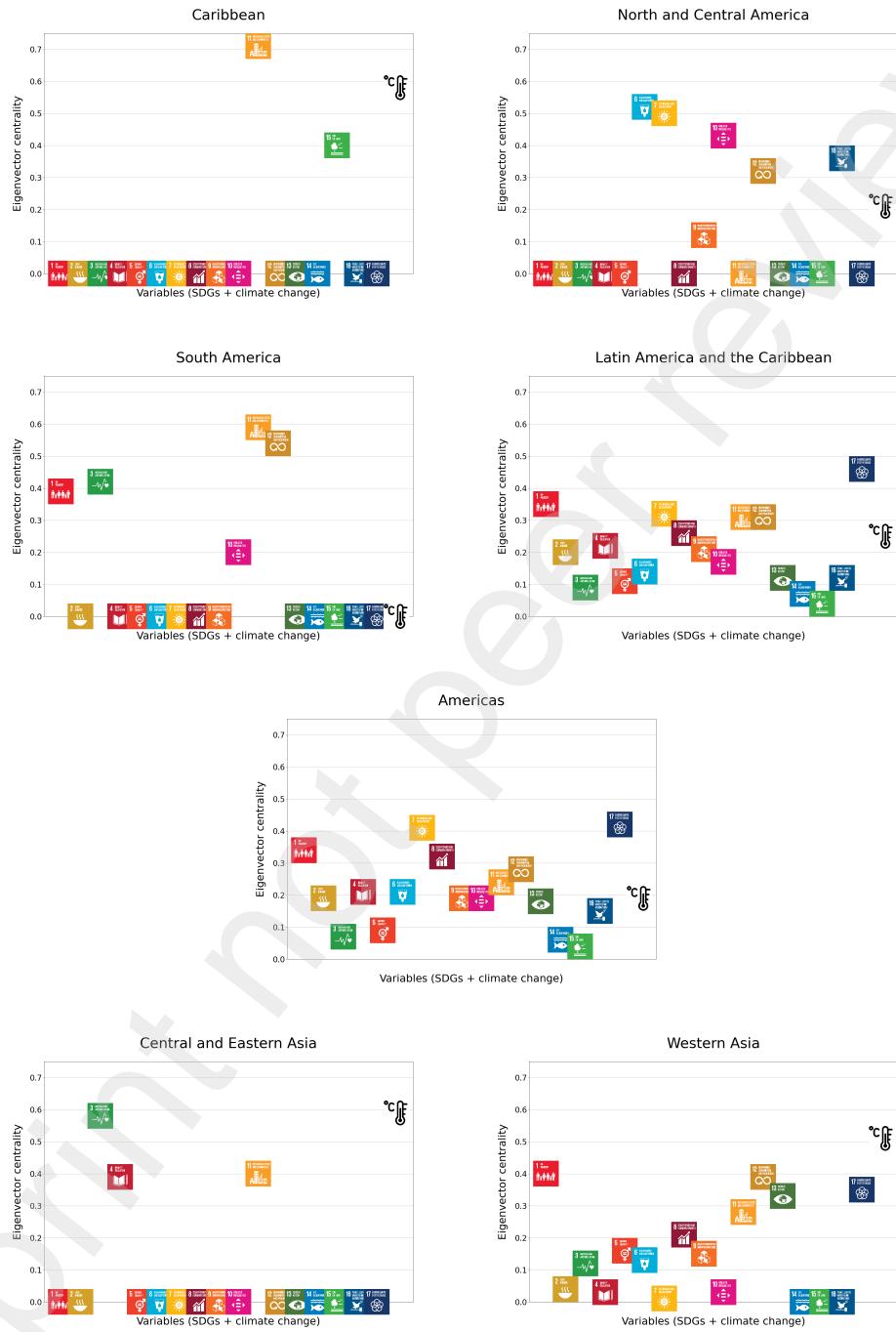






503 *Eigenvector centralities of groupings*

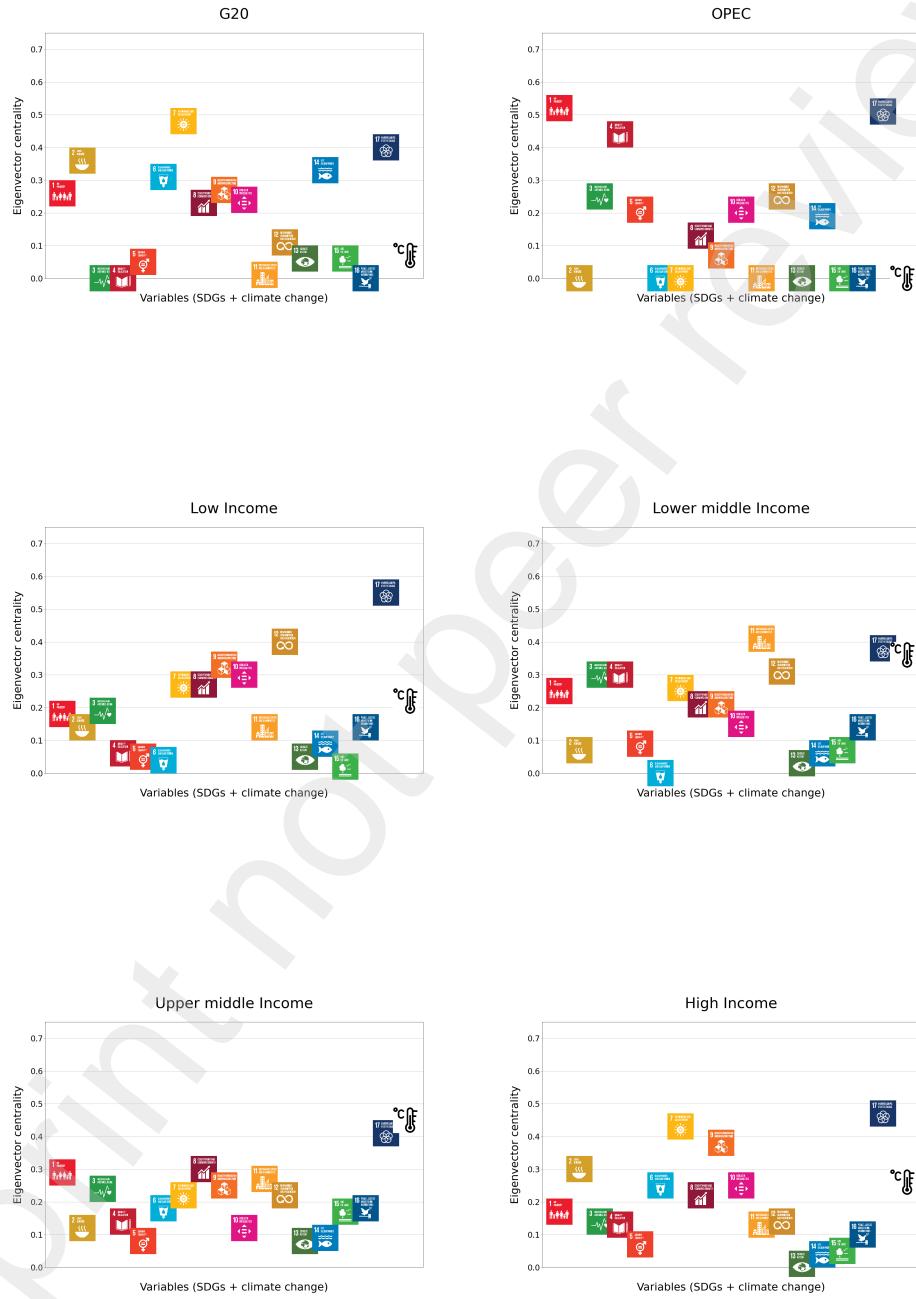






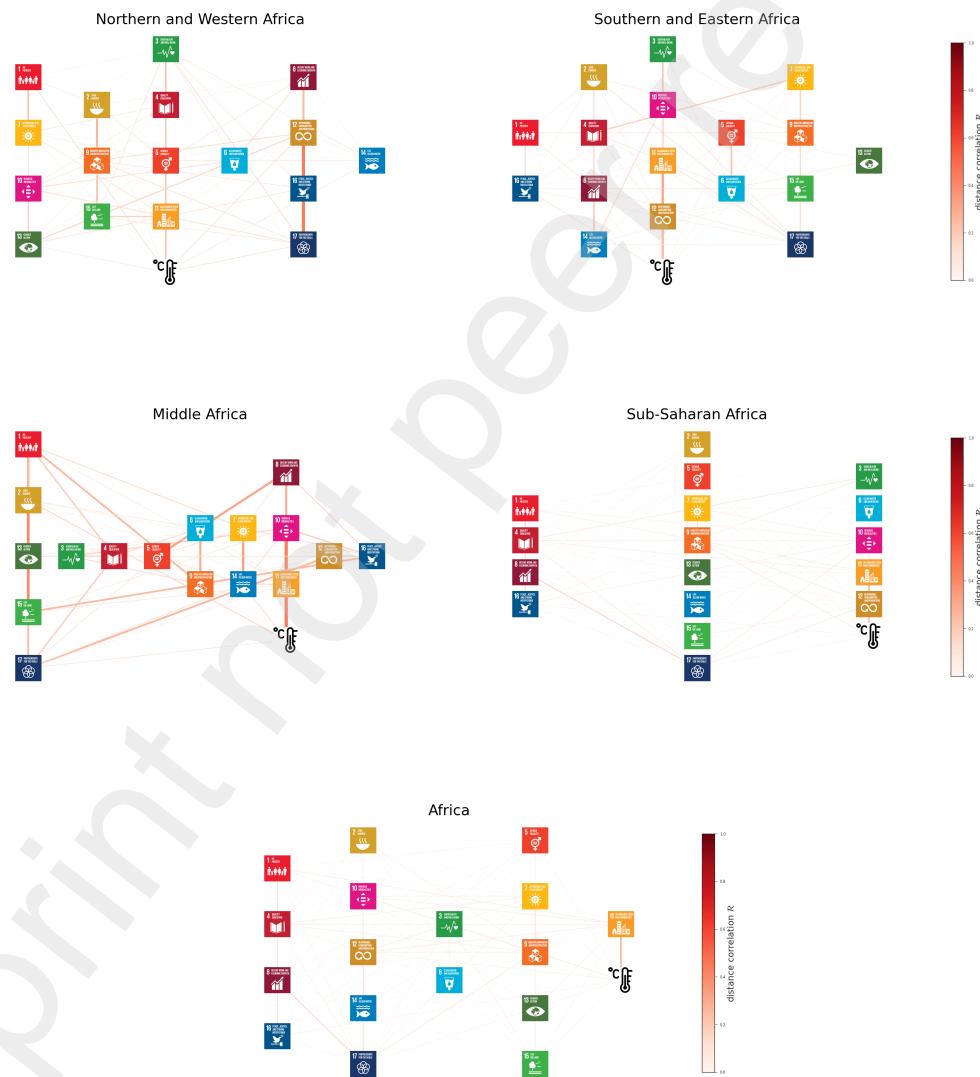


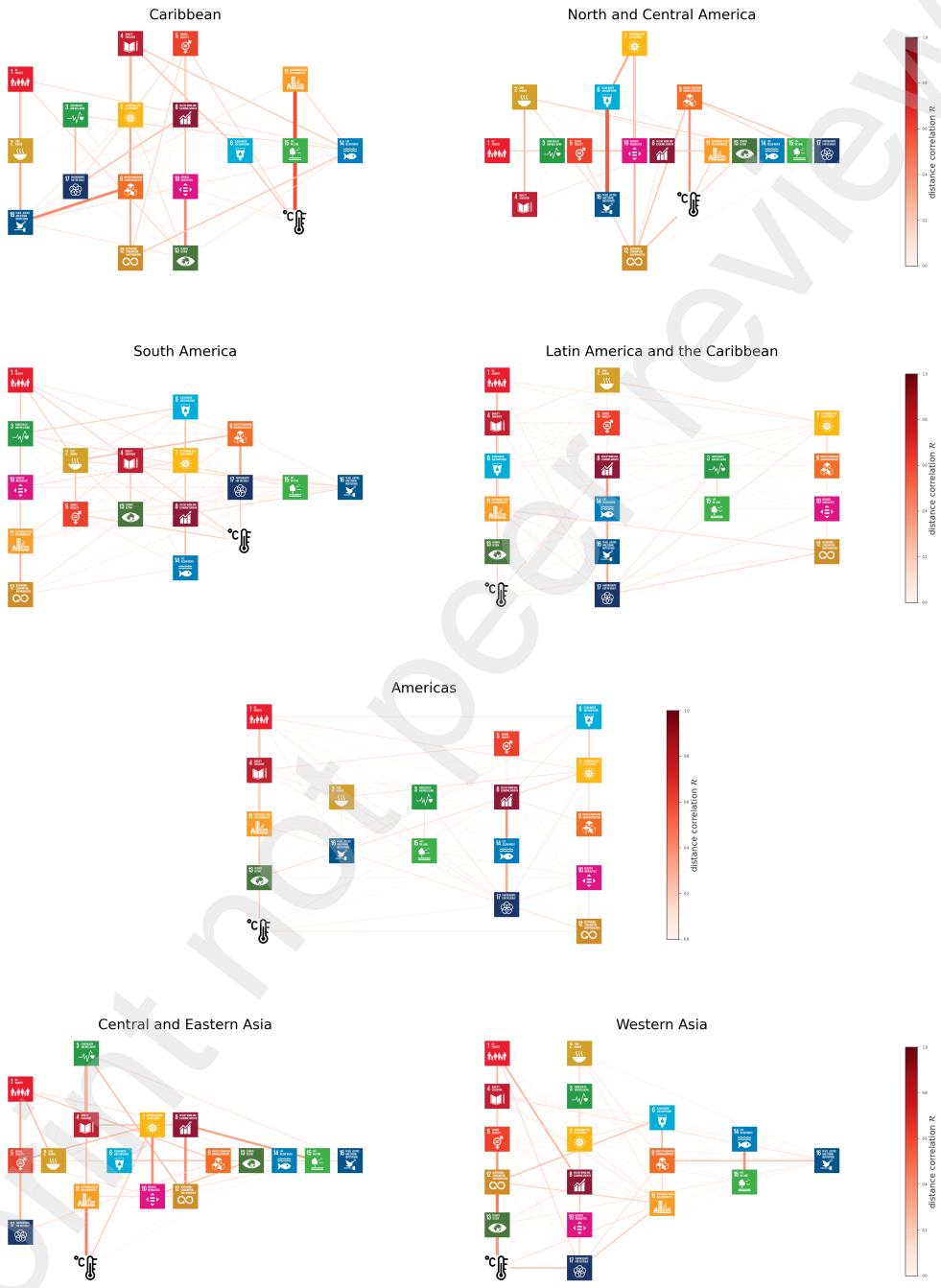


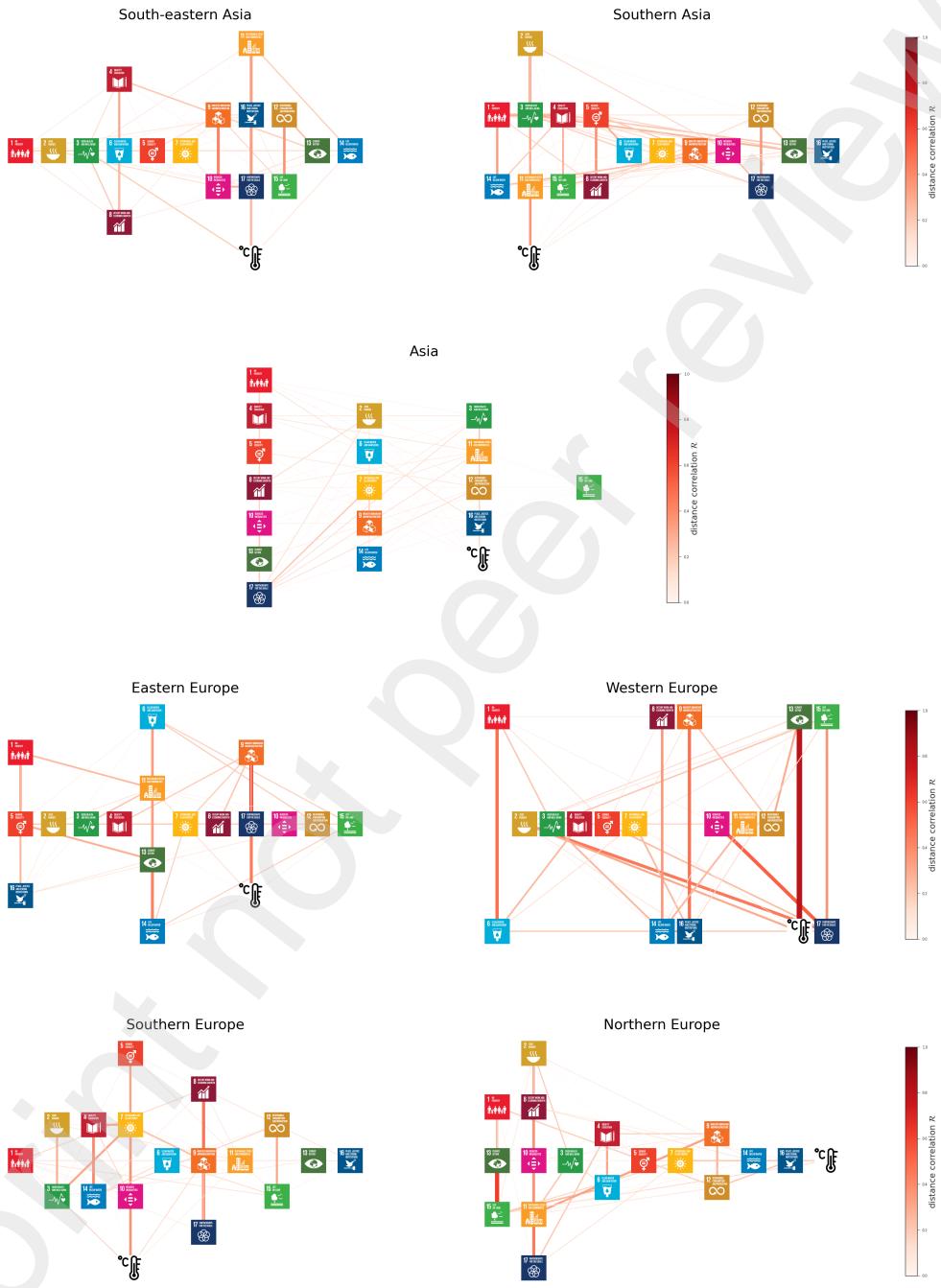


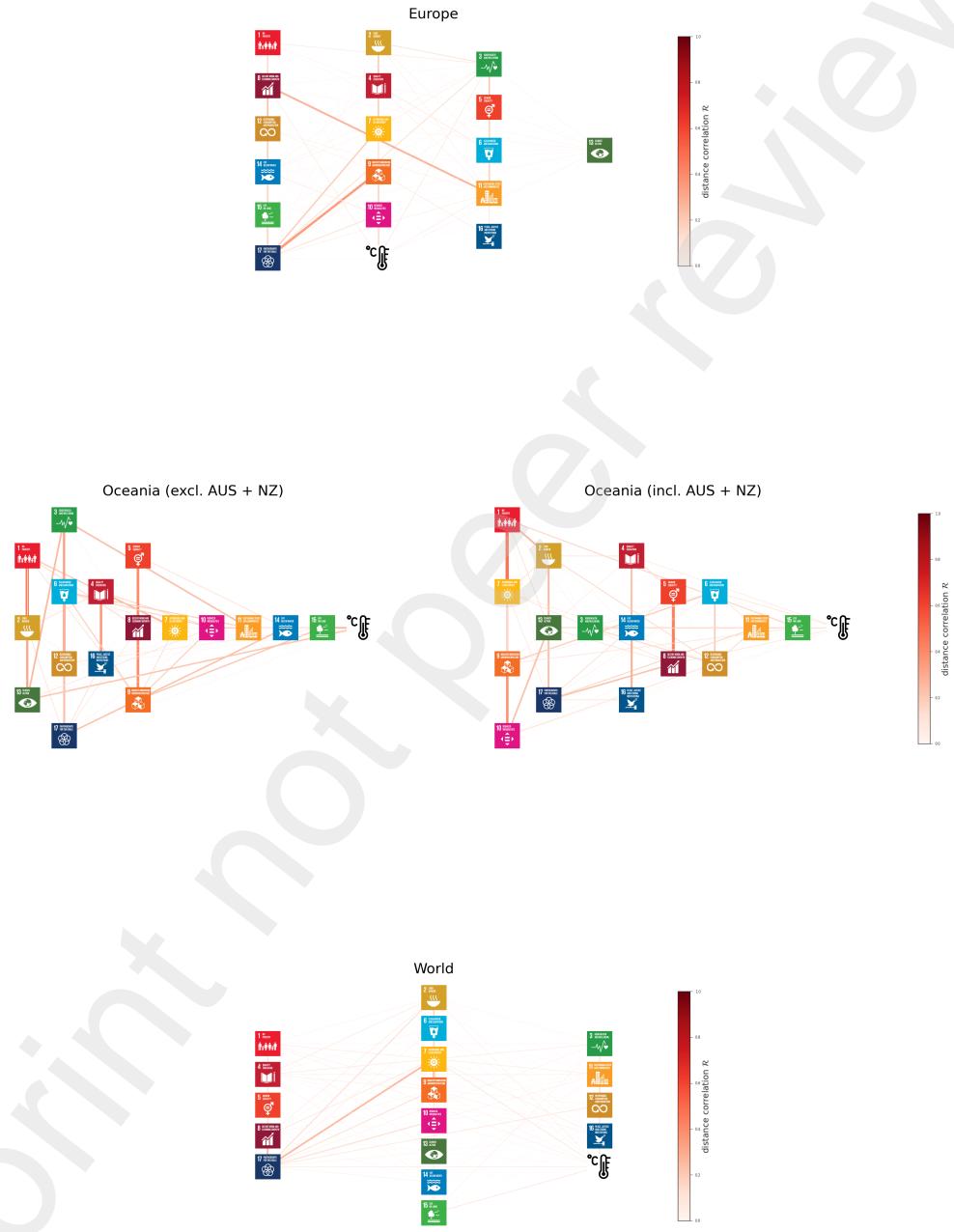
504 *Nexus of highly interconnected nodes*

505 In the subsequent plots, vertically aligned nodes belong to one nexus. In
 506 Northern and Western Africa, for example, SDGs 1, 7, 10, and 13 are one
 507 nexus, SDGs 2, 9, 15 are another, SDGs 3, 4, 5, 11, mean annual temperature
 508 are again another, and so forth.

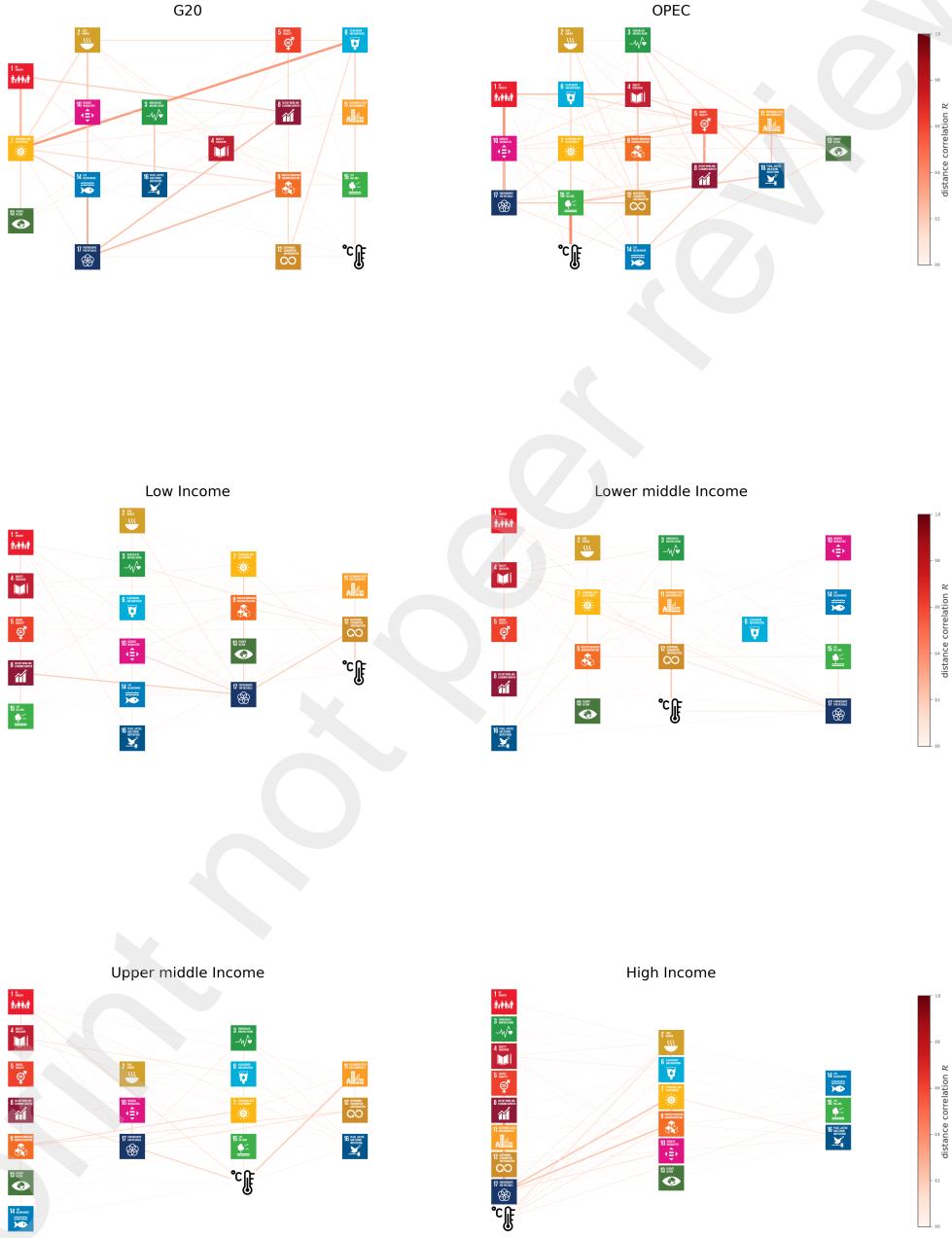












509 *Partial distance correlations with conditional subsets*

510 The subsequent tables show the minimal partial distance correlation (**min**
511 **dcor**) between any two variables **Var X** and **Var Y**, which are displayed in
512 the first two columns (T stands for our 18th variable, annual mean temper-
513 ature). The minimal partial distance correlation and its *p*-value are given
514 in the third and fourth column. In the rightmost column, the conditional
515 subset that resulted in the minimum partial distance correlation over any
516 subset of the remaining 16 variables is given.

Table 3: Northern and Western Africa

Table 4: Southern and Eastern Africa

Table 5: Middle Africa

Table 7: Africa

Table 6: Sub-Saharan Africa

Var X	Var Y	min_dcor	p-value	condition
1	0.0002	0.2593	(33', 36', T')	
1	3	0.025	0.7263	(2', 35')
1	4	0.0002	0.2593	(33', 36', T')
1	5	0.0296	0.1738	(2', 3', 8', 9', 10', 11', 13', 14', 15', 16', T')
1	6	0.051	0.3296	(7', 11', 13', 16')
1	7	0.0225	0.3296	(7', 11', 12', 13', 14', 16', 17', T')
1	8	0.12996	0.0009	(4', 5', 10', 11', 12', 13', 14', 16', 17', T')
1	9	0.0010	0.4488	(2', 3', 5', 7', 8', 10', 11', 12', 13', 15', 16', 17', T')
1	10	0.0027	0.2593	(33', 36', T')
1	11	0.06745	0.0287	(4', 5', 6', 8', 9', 10', 12', 16', 17', T')
1	12	0.0002	0.2593	(33', 36', T')
1	13	0.00673	0.2337	(4', 5', 7', 12', 14', 16')
1	14	0.051	0.3247	(5', 7', 8', 10', 14', 16')
1	15	0.0002	0.2593	(33', 36', T')
1	16	0.14533	0.0001	(4', 5', 10', 11', 12', 17', T')
1	17	0.00178	0.0228	(2', 3', 7', 8', 9', 10', 11', 12', 15', 14', 16', T')
1	18	0.0002	0.2593	(33', 36', T')
2	3	0.0	0.6785	(4', 5', 9', 10', 12', 14', 16', T')
2	4	0.0332	0.2337	(4', 5', 8', 9', 10', 11', 12', 13', 15', 17', T')
2	5	0.0205	0.2647	(4', 5', 8', 9', 10', 11', 12', 13', 15', 16')
2	6	0.00548	0.0498	(3', 7', 9', 11', 17', T')
2	7	0.0002	0.2593	(33', 36', T')
2	8	0.03661	0.1269	(4', 5', 7', 9', 11', 14', 16')
2	9	0.0002	0.2593	(33', 36', T')
2	10	0.0002	0.2593	(33', 36', T')
2	11	0.0205	0.2337	(4', 5', 9', 10', 12', 17', T')
2	12	0.0002	0.2593	(33', 36', T')
2	13	0.0	0.6836	(4', 11', 16', 17', T')
2	14	0.0297	0.0218	(5', 9', 10', 17', T')
2	15	0.0002	0.2593	(33', 36', T')
2	16	0.00316	0.2545	(1', 13', T')
2	17	0.0634	0.0007	(4', 5', 8', 9', 10', 11', 12', 14', 16', T')
2	18	0.0002	0.2593	(33', 36', T')
3	4	0.00007	0.2593	(1', 2', 7', 9', 10', 11', 12', 13', 15', 16', T')
3	5	0.0002	0.2593	(33', 36', T')
3	6	0.05472	0.0257	(7', 11', 13', 16', T')
3	7	0.0002	0.2593	(33', 36', T')
3	8	0.07755	0.3096	(4', 5', 7', 9', 10', 12', 13', 16', T')
3	9	0.0002	0.2593	(33', 36', T')
3	10	0.0002	0.2593	(33', 36', T')
3	11	0.0002	0.2593	(33', 36', T')
3	12	0.04899	0.0789	(4', 5', 9', 10', 11', 16', T')
3	13	0.0002	0.2593	(33', 36', T')
3	14	0.04417	0.0215	(T')
3	15	0.0002	0.2593	(33', 36', T')
3	16	0.0002	0.2593	(33', 36', T')
3	17	0.0	0.6836	(1', 2', 7', 9', 10', 11', 12', 13', 14', 16', T')
3	18	0.0002	0.2593	(33', 36', T')
4	6	0.00007	0.2593	(1', 2', 7', 9', 10', 11', 12', 13', 16', T')
4	7	0.0002	0.2593	(33', 36', T')
4	8	0.0002	0.2593	(33', 36', T')
4	9	0.0002	0.2593	(33', 36', T')
4	10	0.0002	0.2593	(33', 36', T')
4	11	0.05822	0.0653	(5', 10', 12', 13', 15', 16')
4	12	0.0002	0.2593	(33', 36', T')
4	13	0.0002	0.2593	(33', 36', T')
4	14	0.0	0.6785	(2', 3', 6', 7', 13', 16', T')
4	15	0.02049	0.2708	(1', 2', 3', 9', 10', 11', 13', 16')
4	16	0.0002	0.2593	(33', 36', T')
4	17	0.06442	0.0286	(1', 2', 3', 7', 9', 10', 11', 12', 14', 15', 17', T')
4	18	0.0002	0.2593	(33', 36', T')
5	6	0.00108	0.0345	(7', 11', 12', 13', 17', T')
5	7	0.0	0.6715	(1', 2', 3', 7', 9', 10', 11', 12', 13', 16', T')
5	8	0.0002	0.2593	(33', 36', T')
5	9	0.0002	0.2593	(33', 36', T')
5	10	0.0002	0.2593	(33', 36', T')
5	11	0.0002	0.2593	(33', 36', T')
5	12	0.0002	0.2593	(33', 36', T')
5	13	0.0002	0.2593	(33', 36', T')
5	14	0.0002	0.2593	(33', 36', T')
5	15	0.0002	0.2593	(33', 36', T')
5	16	0.0002	0.2593	(33', 36', T')
5	17	0.02214	0.07673	(7', 10', 12', 17', T')
5	18	0.0002	0.2593	(33', 36', T')
6	T	0.00033	0.8411	(7', 10', 12', 13', 17', T')
6	7	0.0002	0.2593	(33', 36', T')
6	8	0.0002	0.2593	(33', 36', T')
6	9	0.0002	0.2593	(33', 36', T')
6	10	0.0002	0.2593	(33', 36', T')
6	11	0.0002	0.2593	(33', 36', T')
6	12	0.0002	0.2593	(33', 36', T')
6	13	0.0002	0.2593	(33', 36', T')
6	14	0.0002	0.2593	(33', 36', T')
6	15	0.0002	0.2593	(33', 36', T')
6	16	0.0002	0.2593	(33', 36', T')
6	17	0.0002	0.2593	(33', 36', T')
6	18	0.0002	0.2593	(33', 36', T')
7	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
7	2	0.0002	0.2593	(33', 36', T')
7	3	0.0002	0.2593	(33', 36', T')
7	4	0.0002	0.2593	(33', 36', T')
7	5	0.0002	0.2593	(33', 36', T')
7	6	0.0002	0.2593	(33', 36', T')
7	7	0.0002	0.2593	(33', 36', T')
7	8	0.0002	0.2593	(33', 36', T')
7	9	0.0002	0.2593	(33', 36', T')
7	10	0.0002	0.2593	(33', 36', T')
7	11	0.0002	0.2593	(33', 36', T')
7	12	0.0002	0.2593	(33', 36', T')
7	13	0.0002	0.2593	(33', 36', T')
7	14	0.0002	0.2593	(33', 36', T')
7	15	0.0002	0.2593	(33', 36', T')
7	16	0.0002	0.2593	(33', 36', T')
7	17	0.0002	0.2593	(33', 36', T')
7	18	0.0002	0.2593	(33', 36', T')
8	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
8	2	0.0002	0.2593	(33', 36', T')
8	3	0.0002	0.2593	(33', 36', T')
8	4	0.0002	0.2593	(33', 36', T')
8	5	0.0002	0.2593	(33', 36', T')
8	6	0.0002	0.2593	(33', 36', T')
8	7	0.0002	0.2593	(33', 36', T')
8	8	0.0002	0.2593	(33', 36', T')
8	9	0.0002	0.2593	(33', 36', T')
8	10	0.0002	0.2593	(33', 36', T')
8	11	0.0002	0.2593	(33', 36', T')
8	12	0.0002	0.2593	(33', 36', T')
8	13	0.0002	0.2593	(33', 36', T')
8	14	0.0002	0.2593	(33', 36', T')
8	15	0.0002	0.2593	(33', 36', T')
8	16	0.0002	0.2593	(33', 36', T')
8	17	0.0002	0.2593	(33', 36', T')
8	18	0.0002	0.2593	(33', 36', T')
9	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
9	2	0.0002	0.2593	(33', 36', T')
9	3	0.0002	0.2593	(33', 36', T')
9	4	0.0002	0.2593	(33', 36', T')
9	5	0.0002	0.2593	(33', 36', T')
9	6	0.0002	0.2593	(33', 36', T')
9	7	0.0002	0.2593	(33', 36', T')
9	8	0.0002	0.2593	(33', 36', T')
9	9	0.0002	0.2593	(33', 36', T')
9	10	0.0002	0.2593	(33', 36', T')
9	11	0.0002	0.2593	(33', 36', T')
9	12	0.0002	0.2593	(33', 36', T')
9	13	0.0002	0.2593	(33', 36', T')
9	14	0.0002	0.2593	(33', 36', T')
9	15	0.0002	0.2593	(33', 36', T')
9	16	0.0002	0.2593	(33', 36', T')
9	17	0.0002	0.2593	(33', 36', T')
9	18	0.0002	0.2593	(33', 36', T')
10	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
10	2	0.0002	0.2593	(33', 36', T')
10	3	0.0002	0.2593	(33', 36', T')
10	4	0.0002	0.2593	(33', 36', T')
10	5	0.0002	0.2593	(33', 36', T')
10	6	0.0002	0.2593	(33', 36', T')
10	7	0.0002	0.2593	(33', 36', T')
10	8	0.0002	0.2593	(33', 36', T')
10	9	0.0002	0.2593	(33', 36', T')
10	10	0.0002	0.2593	(33', 36', T')
10	11	0.0002	0.2593	(33', 36', T')
10	12	0.0002	0.2593	(33', 36', T')
10	13	0.0002	0.2593	(33', 36', T')
10	14	0.0002	0.2593	(33', 36', T')
10	15	0.0002	0.2593	(33', 36', T')
10	16	0.0002	0.2593	(33', 36', T')
10	17	0.0002	0.2593	(33', 36', T')
10	18	0.0002	0.2593	(33', 36', T')
11	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
11	2	0.0002	0.2593	(33', 36', T')
11	3	0.0002	0.2593	(33', 36', T')
11	4	0.0002	0.2593	(33', 36', T')
11	5	0.0002	0.2593	(33', 36', T')
11	6	0.0002	0.2593	(33', 36', T')
11	7	0.0002	0.2593	(33', 36', T')
11	8	0.0002	0.2593	(33', 36', T')
11	9	0.0002	0.2593	(33', 36', T')
11	10	0.0002	0.2593	(33', 36', T')
11	11	0.0002	0.2593	(33', 36', T')
11	12	0.0002	0.2593	(33', 36', T')
11	13	0.0002	0.2593	(33', 36', T')
11	14	0.0002	0.2593	(33', 36', T')
11	15	0.0002	0.2593	(33', 36', T')
11	16	0.0002	0.2593	(33', 36', T')
11	17	0.0002	0.2593	(33', 36', T')
11	18	0.0002	0.2593	(33', 36', T')
12	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
12	2	0.0002	0.2593	(33', 36', T')
12	3	0.0002	0.2593	(33', 36', T')
12	4	0.0002	0.2593	(33', 36', T')
12	5	0.0002	0.2593	(33', 36', T')
12	6	0.0002	0.2593	(33', 36', T')
12	7	0.0002	0.2593	(33', 36', T')
12	8	0.0002	0.2593	(33', 36', T')
12	9	0.0002	0.2593	(33', 36', T')
12	10	0.0002	0.2593	(33', 36', T')
12	11	0.0002	0.2593	(33', 36', T')
12	12	0.0002	0.2593	(33', 36', T')
12	13	0.0002	0.2593	(33', 36', T')
12	14	0.0002	0.2593	(33', 36', T')
12	15	0.0002	0.2593	(33', 36', T')
12	16	0.0002	0.2593	(33', 36', T')
12	17	0.0002	0.2593	(33', 36', T')
12	18	0.0002	0.2593	(33', 36', T')
13	1	0.0002	0.2593	(1', 2', 3', 7', 17', T')
13	2	0.0002	0.2593	(33', 36', T')
13	3	0.0002	0.2593	(33', 36', T')
13	4	0.0002	0.2593	(33', 36', T')
13	5	0.0002	0.2593	(33', 36', T')
13	6			

Table 8: Caribbean

Table 9: North and Central America

Table 10: South America									
Var	X	Y	Y min	door	pos	condition			
1	2	0.1653	0.0882		(7, 11, 13)				
1	3	0.9	0.8765		(11, 12, 13*)				
1	4	0.13046	0.0876		(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
1	5	0.2266	0.0866		(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
7	0.12927	0.04256			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
9	0.0983	0.0625			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
10	0.08316	0.0267			(7, 12, 13, 17, 18)				
11	0.19285	0.0856			(7, 8, 9, 10, 11, 12)				
12	0.0821	0.0586			(7, 8, 9, 10, 11, 12, 14, 17)				
13	0.03018	0.0886			(4, 5, 7, 11)				
14	0.1095	0.0486			(4, 5, 7, 11, 12, 15)				
15	0.0509	0.0486			(4, 5, 7, 11, 12, 15)				
16	0.0	0.0478			(7, 8, 9, 10, 11, 12, 13, 17)				
17	0.07	0.0478			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17571	0.0479			(7, 8, 9, 10, 11, 12, 13, 17)				
2	3	0.9	0.9595		(4, 5, 7, 8, 9, 10, 14, 17)				
2	4	0.0209	0.0586		(7, 8, 9, 10, 11, 12, 13, 17)				
5	2.1709	0.0689			(7, 8, 9, 16)				
2	6	0.13142	0.1588		(4, 5, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
2	7	0.0509	0.0886		(4, 5, 7, 11, 12, 15)				
2	8	0.04539	0.0364		(7, 8, 9, 11, 13)				
10	0.21945	0.0554			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
11	0.0	0.0839			(7, 8, 9, 10, 11, 12)				
12	0.11	0.0742			(7, 8, 9, 12)				
14	0.0	0.0489			(7, 8, 9, 11, 12, 13, 15, 17)				
2	16	0.31078	0.0489		(4, 5, 7, 8, 9, 15, 17)				
2	17	0.03949	0.0316		(7, 8, 9, 12)				
7	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
3	4	0.04033	0.0695		(7, 11, 13, 14, 16, 17, 17)				
3	6	0.16663	0.0721		(7, 12, 13, 17)				
3	7	0.00941	0.0275		(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
3	8	0.06703	0.0675		(7, 11, 12, 13, 17)				
3	10	0.0	0.04865		(7, 8, 9, 12)				
11	0.22387	0.0622			(7, 8, 9, 10, 11, 12, 13, 17)				
12	0.2131	0.0620			(7, 8, 9, 11)				
13	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
14	0.04247	0.0487			(7, 8, 9, 11, 12, 13)				
15	0.1274	0.0723			(7, 11, 13, 17)				
17	0.0114	0.0622			(7, 8, 9, 11, 12)				
T	0.113	0.0265			(7, 12, 13, 14, 17)				
5	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
6	0.06812	0.0377			(7, 8, 9, 10, 11, 12, 13, 14, 15, 17)				
8	0.0	0.04671			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
9	0.04346	0.0464			(7, 8, 9, 11, 12)				
11	0.16162	0.0368			(7, 8, 9, 10, 11, 12, 13, 16, 17)				
12	0.06032	0.0164			(7, 8, 9, 16)				
14	0.07887	0.0821			(7, 8, 9, 11, 12, 15, 16, 17)				
15	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
16	0.0671	0.0283			(7, 8, 9, 10, 11, 12, 13)				
17	0.0	0.0507			(7, 8, 9, 10, 11, 12, 13, 15, 17)				
5	6	0.03018	0.0394		(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
7	8	0.04817	0.0366		(7, 8, 9, 10, 11, 12)				
9	0.0	0.05885			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
10	0.0	0.05786			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
12	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
14	0.0	0.0587			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
15	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
17	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	12	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
13	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
14	0.08367	0.0763			(7, 8, 9, 11, 17)				
15	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 16)				
16	0.0	0.05095			(7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)				
17	0.0	0.05097			(7, 8, 9, 10, 11, 12)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	13	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
14	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
15	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
16	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
17	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	14	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
15	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
16	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
17	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	15	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
16	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
17	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	16	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
17	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	17	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
18	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	18	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
19	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	19	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
20	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	20	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
21	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	21	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
22	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	22	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
23	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	23	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
24	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	24	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
25	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	25	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
26	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	26	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
27	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	27	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
28	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	28	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
29	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	29	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
30	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	30	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
31	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	31	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
32	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	32	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
33	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	33	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
34	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)				
T	0.17079	0.0487			(7, 8, 9, 10, 11, 12, 13, 17)				
5	34	0.0	0.0509		(7, 8, 9, 10, 11, 12, 13, 17)				
35	0.0	0.0509			(7, 8, 9, 10, 11, 12, 13, 17)		</td		

Table 11: Latin America and the Caribbean

Var X	Var Y	min door	p-value	condition
1	2	0.08171	0.5409	(4, 5, 6, 7, 11, 13, 16, 17, T)
1	3	0.08171	0.5409	(4, 5, 6, 7, 11, 12, 13, 15, 16, 17, T)
1	4	0.21448	0.0001	(2, 5, W, 7, 11, 17)
1	5	0.07859	0.46750	(2, 3, 4, 10, 11, 14, 16, 17)
1	7	0.08848	0.043	(5, 6, 9, 11, 17, T)
1	8	0.07885	0.46865	(4, 5, 6, 7, 9, 11, 12, 13, 14, 16, 17, T)
1	9	0.07885	0.46870	(4, 5, 6, 7, 9, 10, 11, 12, 13, 14, T)
1	10	0.0	0.68682	(3, 4, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, 17, T)
1	11	0.02033	0.0003	(3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, T)
1	12	0.0	0.67335	(2, 3, 4, 5, 7, 9, 10, 11, 12, 13, 14, 17)
1	13	0.20818	0.0006	(2, 4, 5, 6, 7, 11)
1	14	0.0	0.67335	(2, 3, 4, 5, 7, 8, 10, 11, 12, 13, 14, 17)
1	15	0.0	0.64035	(3, 4, 5, 6, 7, 10, 13, 16, 17)
1	16	0.0	0.64035	(3, 4, 5, 6, 7, 10, 13, 16, 17)
1	17	0.08136	0.0009	(2, 5, W, 7, 8, 10, 11, 12, 13, 14, 17)
T	1	0.02408	0.29137	(2, 4, 5, 6, 7, 9, 11, 12, 13, 14, 17)
2	4	0.09054	0.0328	(5, 6, 7, 11, 16, 17, T)
2	5	0.06656	0.0603	(4, 7, 8, 11, 12, 13, 16)
2	6	0.07885	0.46865	(4, 5, 6, 7, 8, 11, 12, 13, 14, 17)
2	7	0.12071	0.0128	(5, 6, 9, 11, 12, 15, 16, 17, T)
2	8	0.0	0.47255	(4, 7, 8, 11, 13, 14, 15, 16, 17, T)
2	9	0.0	0.47255	(4, 7, 8, 11, 13, 14, 15, 16, 17, T)
2	10	0.02385	0.0006	(2, 4, 5, 6, 7, 11)
2	11	0.0	0.47255	(4, 7, 8, 11, 13, 14, 15, 16, 17, T)
2	12	0.0	0.47255	(4, 7, 8, 11, 13, 14, 15, 16, 17, T)
2	13	0.02409	0.29487	(5, 6, 7, 8, 11, 15, 16, 17, T)
2	14	0.03030	0.13740	(1, 4, 5, 6, 7, 16, T)
2	15	0.02523	0.28787	(3, 5, 6, 7, 11, 14)
2	16	0.08191	0.06673	(1, 4, 5, 7, 8, 11, T)
2	17	0.02403	0.0003	(2, 4, 5, 6, 7, 8, 11, 12, 13, 15, 16, T)
3	4	0.0	0.58861	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 17)
3	5	0.0	0.72235	(3, 5, T)
3	6	0.0644	0.9721	(2, 4, 5, 6, 7, 14, 16, T)
3	7	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	8	0.0	0.48615	(2, 4, 5, 6, 7, 15, 17, T)
3	9	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	10	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	11	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	12	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	13	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	14	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	15	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	16	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
3	17	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	5	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	6	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	7	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	8	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	9	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	10	0.03326	0.73663	(2, 5, 11)
4	11	0.02587	0.29262	(2, T)
4	12	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	13	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	14	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	15	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	16	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
4	17	0.0	0.67335	(2, 4, 5, 6, 7, 15, 17, T)
T	5	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	6	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	7	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	8	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	9	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	10	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	11	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	12	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	13	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	14	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	15	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	16	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
T	17	0.00597	0.67389	(3, 5, 6, 7, 8, 11, 12, 13, 14, 15, 16, T)
5	6	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	7	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	8	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	9	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	10	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
5	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	7	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	8	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	9	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	10	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
6	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	8	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	9	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	10	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
7	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	9	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	10	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
8	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	10	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
9	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	11	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
10	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	12	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
11	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
12	13	0.0	0.67335	(2, 4, 5, 6, 7, 11)
12	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
12	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
12	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
12	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
13	14	0.0	0.67335	(2, 4, 5, 6, 7, 11)
13	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
13	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
13	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
14	15	0.0	0.67335	(2, 4, 5, 6, 7, 11)
14	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
14	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
15	16	0.0	0.67335	(2, 4, 5, 6, 7, 11)
15	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
16	17	0.0	0.67335	(2, 4, 5, 6, 7, 11)
T	1	0.03447	0.85765	(6, 7, 12, 13, 14, 15, 16, T)
T	2	0.03447	0.86239	(2, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	3	0.03447	0.87306	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	4	0.03447	0.88211	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	5	0.03447	0.88744	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	6	0.03447	0.89158	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	7	0.03447	0.89572	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	8	0.03447	0.90172	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	9	0.03447	0.90782	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	10	0.03447	0.91492	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	11	0.03447	0.92202	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	12	0.03447	0.92912	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	13	0.03447	0.93622	(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	14	0.03447	0.94332	(0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)
T	15	0.03447	0.95042	(1, 2,

Table 13: Central and Eastern Asia

Table 14: Western Asia

Table 15: South-eastern Asia

Table 16: Southern Asia

Table 17: Asia

Table 18: Eastern Europe

Var X	Var Y	min	decr	p-value	condition
1	0.0	0.8515	(2, 4, 5, 9, 12, 14, 17)		
1	3	0.0	0.9645	(2, 7, 8, 11, 12, 14, 16)	
1	4	0.0	0.9196	(2, 4, 7, 8, 10, 15, 16, 17)	
1	5	0.0	0.5083	(2, 4, 7, 8, 10, 15, 16, 17)	
1	6	0.0	0.8489	(2, 7, 10, 16, T)	
1	7	0.0	0.8186	(2, 3, 4, 5, 7, 10, 13, 17)	
1	8	0.0	0.4903	(2, 3, 9, 12, 13, 16, 17)	
1	9	0.0	0.9584	(2, 8, 13, 14, 15, 16, T)	
1	10	0.0	0.7057	(2, 3, 4, 5, 7, 10, 13, 17)	
1	11	0.0	0.2406	(2, 4, 5, 7)	
1	12	0.0	0.8054	(2, 4, 7, 8, 10, 11, 12, 13, 16, 17)	
1	13	0.0	0.8265	(2, 7, 10, 11, 13, 16)	
1	14	0.0	0.8803	(2, 4, 7, 8, 11, 12, 13, 16, 17, T)	
1	15	0.0	0.7043	(2, 4, 7, 8, 10, 11, 12, 13, 16, 17)	
1	16	0.0	0.2066	(2, 5, 10, 12)	
1	17	0.0	0.6769	(2, 5, 9, 10, 11, 12, 13, 15, 17)	
1	18	0.0	0.4903	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 15, 17)	
2	3	0.0	0.5045	(2, 7, 12, 13, 14, 15, 17, T)	
2	4	0.0	0.4916	(2, 3, 4, 6, 7, 10, 11, 12, 13, 16, T)	
2	5	0.0	0.8023	(2, 3, 5, 7, 8, 10, 14, 15, 16, 17)	
2	6	0.0	0.4764	(2, 5, 7, 8, 10, 12, 15, 17, T)	
2	7	0.0	0.7006	(2, 3, 4, 5, 7, 8, 10, 12, 15, 17)	
2	8	0.0	0.9295	(2, 4, 7, 8, 9, 10, 11, 12, 13, 15, 17)	
2	9	0.0	0.8281	(2, 7, 13, 14, 17)	
2	10	0.0	0.7018	(2, 4, 7, 8, 10, 11, 12, 13, 14, 17)	
2	11	0.0	0.1010	(2, 4, 7, 8, 10, 11, 12, 13, 14, 17)	
2	12	0.0	0.6728	(2, 4, 7, 8, 10, 11, 12, 13, 14, 17)	
2	13	0.0	0.7476	(2, 10, 11, 12, 13, 14, 17)	
2	14	0.0	0.4665	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 15, 17)	
2	15	0.0	0.4903	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 15, 17)	
2	16	0.0	0.6083	(2, 3, 5, 7, 8, 10, 12, 13, 14, 16, 17)	
2	17	0.0	0.1626	(2, 4, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17)	
2	18	0.0	0.4741	(2, 4, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
3	4	0.0	0.4936	(2, 7, 8, 10, 11, 12, 13, 14, 15, 17)	
3	5	0.0	0.5668	(2, 7, 10, 11, 12, 13, 14, 15, 17)	
3	6	0.0	0.4976	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 17)	
3	7	0.0	0.4919	(2, 6, 8, 11, 12, 13, 14, 16)	
3	8	0.0	0.6024	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
3	9	0.0	0.2404	(2, 7, 9, 11, 12, 13, 15, 17)	
3	10	0.0	0.4906	(2, 11, 14, 15, 16, 17)	
3	11	0.0	0.1088	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
3	12	0.0	0.4769	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
3	13	0.0	0.4903	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
3	14	0.0	0.4935	(2, 11, 12, 13, 15)	
3	15	0.0	0.4961	(2, 5, 6, 7, 8, 11, 17)	
3	16	0.0	0.4903	(2, 3, 4, 5, 6, 7, 8, 11, 17)	
3	17	0.0	0.5015	(2, 5, 9, 10, 11, 12, 13, 14, 15, 17)	
3	18	0.0	0.1198	(2, 4, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17)	
4	6	0.0	0.8255	(2, 7, 14)	
4	7	0.0	0.0775	(2, 9, 10, 11, 12, 13, 14, 15, 16, 17)	
4	8	0.0	0.1359	(2, 5, 7, 13, T)	
4	9	0.0	0.1888	(2, 5, 7, 13, T)	
4	10	0.0	0.1012	(2, 5, 7, 13, T)	
4	11	0.0	0.0911	(2, 10, 12, 14, T)	
4	12	0.0	0.0911	(2, 5, 7, 13, T)	
4	13	0.0	0.0655	(2, 10, 13, T)	
4	14	0.0	0.4816	(2, 6, 8, 10, 11, 12, 13, 14, 15, 17)	
4	15	0.0	0.4916	(2, 6, 8, 10, 11, 12, 13, 14, 15, 17)	
4	16	0.0	0.4936	(2, 6, 8, 10, 11, 12, 13, 14, 15, 17)	
4	17	0.0	0.4807	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	18	0.0	0.4916	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	19	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	20	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	21	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	22	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	23	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	24	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	25	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	26	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	27	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	28	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	29	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	30	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	31	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	32	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	33	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	34	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	35	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	36	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	37	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	38	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	39	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	40	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	41	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	42	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	43	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	44	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	45	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	46	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	47	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	48	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	49	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	50	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	51	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	52	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	53	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	54	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	55	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	56	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	57	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	58	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	59	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	60	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	61	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	62	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	63	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	64	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	65	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	66	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	67	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	68	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	69	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	70	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	71	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	72	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	73	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	74	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	75	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	76	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	77	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	78	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	79	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	80	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	81	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	82	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	83	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	84	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	85	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	86	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	87	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	88	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	89	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	90	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	91	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	92	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	93	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	94	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	95	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	96	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	97	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	98	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	99	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	100	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	101	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	102	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	103	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	104	0.0	0.4936	(2, 7, 9, 11, 12, 13, 14, 15, 17)	
4	105	0.0	0.4936	(2, 7, 9,	

Table 21: Northern Europe

Table 22: Europe

X	V	X-V	Mg-I	abs. mag.	phase	condition
1	3	0.07836	0.0205	(1, 2, 3, 9, 9, 12, 13, 14, 16)		
1	3	0.07836	0.0301	(2, '9, 10', 11', 12', 13', 14', 16')		
1	3	0.07836	0.0401	(2, '9, 10', 11', 12', 13', 14', 16')		
1	5	0.0765	0.0475	(2, 3, 7, 9, 12, 16')		
1	6	0.06333	0.0608	(2, 3, 4, 10', 11', 15')		
1	8	0.03393	0.0630	(2, 3, 4, 11', 13', 17')		
1	9	0.0	0.0605	(2, 3, 7, 10', 12', 15', 16')		
1	11	0.0	0.0605	(2, 3, 7, 12', 14', 15', 16')		
1	13	0.0	0.0638	(2, 3, 7, 12', 15')		
1	14	0.0	0.0638	(3, 9, 10', 11', 13')		
1	15	0.03406	0.0638	(3, 9, 10', 11', 13')		
1	16	0.01265	0.0357	(3, 9, 11')		
1	17	0.017144	0.0324	(2, 3, 7, 8, 9, 10', 12', 13', 16')	T	
1	17	0.017144	0.0324	(2, 3, 7, 8, 9, 10', 11', 12', 14', 17')	T	
2	3	0.1889	0.0600	(2, 3, 7, 10', 11')		
2	4	0.1479	0.0718	(1, 2, 3, 7, 9, 10', 11')		
2	5	0.09395	0.0600	(1, 2, 3, 7, 9, 10', 11')		
2	6	0.04393	0.1289	(1, 2, 3, 7, 9, 10', 11')	T	
2	7	0.02622	0.0600	(1, 2, 3, 7, 9, 10', 11')	T	
2	8	0.0	0.06435	(1, 2, 3, 7, 9, 10', 11')	T	
2	9	0.0895	0.0600	(2, 3, 7, 10', 11')	T	
2	10	0.0	0.0600	(2, 3, 7, 10', 11')	T	
2	11	0.0494	0.01629	(3, 4, 5, 7, 8, 9, 10', 12', 13', 14', 16')	T	
2	12	0.0	0.0600	(3, 7, 9, 10', 11', 15')	T	
2	13	0.0	0.0600	(3, 7, 9, 10', 11', 15')	T	
2	14	0.05648	0.0646	(3, 7, 9, 10', 11', 13', 17')	T	
2	15	0.01049	0.0244	(5, 7, 10', 12', 15')	T	
2	16	0.0	0.0600	(5, 7, 10', 12', 15')	T	
2	17	0.02811	0.1284	(3, 7, 9, 10', 11')	T	
2	T	0.1584	0.1284	(3, 7, 9, 10', 11')	T	
3	4	0.0	0.0625	(2, 3, 7, 10', 11')	T	
3	5	0.04123	0.1309	(2, 3, 7, 10', 11')	T	
3	6	0.0	0.0600	(2, 3, 7, 10', 11')	T	
3	7	0.0571	0.0749	(1, 2, 3, 7, 9, 10', 11')	T	
3	8	0.01632	0.1304	(1, 2, 3, 7, 9, 10', 11')	T	
3	9	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')	T	
3	10	0.01897	0.0267	(2, 3, 7, 9, 10', 11')	T	
3	11	0.02066	0.0600	(2, 3, 7, 8, 9, 10', 12', 13', 14', 15', 16')	T	
3	12	0.0	0.0600	(2, 3, 7, 8, 9, 10', 12', 13', 14', 15')	T	
3	13	0.03479	0.0718	(6, 7, 8', 11', 14')	T	
3	14	0.04087	0.0600	(6, 7, 8', 11', 14')	T	
3	15	0.0	0.0600	(6, 7, 8', 11', 14')	T	
3	16	0.0	0.0600	(6, 7, 8', 11', 14')	T	
3	17	0.0	0.0600	(6, 7, 8', 11', 14')	T	
4	T	0.13491	0.1309	(1, 2, 3, 7, 9, 10', 11')	T	
6	6	0.0612	0.06334	(1, 2, 3, 7, 9, 10', 11')		
6	7	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	8	0.0114	0.0633	(1, 2, 3, 7, 9, 10', 11')		
6	9	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	10	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	11	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	12	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	13	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	14	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	15	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	16	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
6	17	0.0	0.0600	(1, 2, 3, 7, 9, 10', 11')		
7	8	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
7	9	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
7	10	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
7	11	0.0824	0.06219	(2, 3, 7, 8, 9, 10', 11')		
7	12	0.0	0.0600	(2, 3, 7, 8, 9, 10', 11')		
7	13	0.02775	0.07013	(2, 3, 7, 8, 9, 10')		
7	14	0.0	0.0600	(2, 3, 7, 8, 9, 10')		
7	15	0.0	0.0600	(2, 3, 7, 8, 9, 10')		
7	16	0.0	0.0600	(2, 3, 7, 8, 9, 10')		
7	17	0.0	0.0600	(2, 3, 7, 8, 9, 10')		
8	9	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	10	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	11	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	12	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	13	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
8	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	10	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	11	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	12	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	13	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
9	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	11	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	12	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	13	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
10	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	12	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	13	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
11	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
12	13	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
12	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
12	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
12	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
12	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
13	14	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
13	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
13	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
13	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
14	15	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
14	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
14	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
15	16	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
15	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
16	17	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
17	18	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
18	19	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
19	20	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
20	21	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
21	22	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
22	23	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
23	24	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
24	25	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
25	26	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
26	27	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
27	28	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
28	29	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
29	30	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
30	31	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
31	32	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
32	33	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
33	34	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
34	35	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
35	36	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
36	37	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
37	38	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
38	39	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
39	40	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
40	41	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
41	42	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
42	43	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
43	44	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
44	45	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
45	46	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
46	47	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
47	48	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
48	49	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
49	50	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
50	51	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
51	52	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
52	53	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
53	54	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
54	55	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
55	56	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
56	57	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
57	58	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
58	59	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
59	60	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
60	61	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
61	62	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
62	63	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
63	64	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
64	65	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
65	66	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
66	67	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
67	68	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
68	69	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
69	70	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
70	71	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
71	72	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
72	73	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
73	74	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
74	75	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
75	76	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
76	77	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
77	78	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
78	79	0.0	0.0600	(2, 3, 7, 9, 10', 11')		
79	80	0.0	0.0600	(2, 3, 7, 9, 10', 11')		

Table 23: Oceania (excl. AUS + NZ) Table 24: Oceania (incl. AUS + NZ)

AUS + NZ)

Table 25: World

Table 26: Global North

Var	Yr	Min	dec	cond	condition
1	2	0.24	0.17938	(7, 7, 9, 10, 11, 12, 17, T)	
1	3	0.04682	0.05674	(7, 5, 10, 11, 12, 13, 16, T)	
1	4	0.04682	0.05674	(7, 5, 10, 11, 12, 13, 16, T)	
1	5	0.0476	0.0574	(7, 4, 14, 15, 17, T)	
1	6	0.05041	0.04435	(7, 4, 8, 9, 10, 11, 12, 13, 15, 17, T)	
1	7	0.05041	0.04435	(7, 4, 8, 9, 10, 11, 12, 13, 15, 17, T)	
1	8	0.05959	0.0009	(7, 3, 9, 13, 17, T)	
1	9	0.05959	0.0009	(7, 3, 9, 13, 17, T)	
1	10	0.03234	0.18378	(4, 5, 9, 10, 13, 16, T)	
1	11	0.0	0.4562	(7, 5, 9, 10, 13, 16, T)	
1	12	0.0	0.4292	(5, 4, 14, 17, T)	
1	13	0.0	0.4292	(5, 4, 14, T)	
1	14	0.0	0.4358	(7, 5, 9, 11, 13, 16, T)	
1	15	0.0	0.4358	(7, 5, 9, 11, 13, 16, T)	
1	16	0.00012	0.4409	(7, 3, 4, 8, 9, 10, 11, 12, 13, 17, T)	
1	17	0.05111	0.0001	(7, 3, 4, 8, 9, 10, 11, 12, 13, 14, 15, 16, T)	
2	1	0.00012	0.4409	(7, 3, 4, 8, 9, 10, 11, 12, 13, 17, T)	
2	3	0.04247	0.07689	(7, 10, 11, 12, 13, 14, 16, T)	
2	4	0.04247	0.07689	(7, 10, 11, 12, 13, 14, 16, T)	
2	6	0.05114	0.05341	(7, 1, 9, 10, 11, 13, T)	
2	7	0.05114	0.05341	(7, 1, 9, 10, 11, 13, T)	
2	8	0.0	0.4745	(7, 3, 4, 9, 11, 12, 13, 15, 16, T)	
2	9	0.0815	0.0075	(7, 1, 10, 11, 13, T)	
2	11	0.04243	0.23838	(7, 3, 5, 7, 9, 10, 12, 13, 14, 17, T)	
2	12	0.02183	0.23838	(7, 3, 5, 7, 9, 10, 12, 13, 16, T)	
2	13	0.0	0.4292	(7, 3, 5, 7, 9, 10, 12, 13, 16, T)	
2	14	0.0	0.4358	(7, 3, 5, 7, 9, 10, 11, 13, 15, T)	
2	15	0.0	0.4358	(7, 3, 5, 7, 9, 10, 11, 13, 15, T)	
2	16	0.0	0.4345	(7, 4, 8, 9, 10, 11, 15, T)	
2	17	0.07	0.0051	(7, 1, 9, 10, 11, 12, T)	
3	1	0.00012	0.4409	(7, 3, 4, 8, 9, 10, 11, 12, T)	
3	4	0.02699	0.18858	(7, 2, 5, 9, 10, 11, 13, 16, T)	
3	5	0.07032	0.0171	(7, 10, 11, 12, 13, 14, 16, T)	
3	6	0.0	0.4887	(7, 2, 5, 9, 10, 11, 13, 16, T)	
3	7	0.07144	0.25897	(7, 2, 5, 9, 10, 11, 13, 14, 16, T)	
3	9	0.0	0.4358	(7, 3, 5, 7, 9, 10, 11, 13, 15, T)	
3	10	0.03821	0.16543	(7, 2, 5, 7, 9, 11, 12, 16, T)	
3	11	0.0	0.4358	(7, 2, 5, 7, 9, 11, 12, 16, T)	
3	12	0.01673	0.25897	(7, 1, 5, 11, 15, 16, T)	
3	13	0.03167	0.13604	(7, 6, 7, 8, 11, 12, 14, 16, T)	
3	14	0.0	0.4358	(7, 2, 5, 7, 9, 11, 12, 14, 16, T)	
3	15	0.03019	0.14595	(7, 2, 5, 7, 9, 11, 12, 14, T)	
3	16	0.07729	0.009	(7, 1, 10, 11, 13, T)	
3	17	0.0	0.4358	(7, 2, 5, 7, 9, 10, 11, 13, 14, 16, T)	
T	1	0.11543	0.0009	(5, 7, 9, 10, 11, 12, 13, 14, 16, T)	
T	2	0.0	0.4358	(5, 7, 9, 10, 11, 12, 13, 14, 16, T)	
T	4	0.05239	0.03677	(7, 2, 5, 9, 10, 11, 13, 16, T)	
T	7	0.02161	0.23167	(7, 2, 5, 9, 10, 11, 14, 16, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, 16, T)	
T	9	0.06449	0.0317	(7, 2, 5, 9, 10, 11, 13, 16, T)	
T	10	0.06704	0.0317	(7, 2, 5, 9, 10, 11, 14, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, 16, T)	
T	12	0.0	0.4887	(7, 2, 5, 9, 10, 11, 13, 16, T)	
T	13	0.0	0.4358	(5, 6, 7, 8, 9, 10, 11, 12, 13, 15, T)	
T	14	0.03663	0.14595	(7, 2, 5, 9, 10, 11, 12, 13, 15, T)	
T	15	0.0	0.4676	(7, 2, 5, 9, 10, 11, 13, 16, T)	
T	16	0.0	0.4358	(7, 2, 5, 9, 10, 11, 13, 16, T)	
T	17	0.0	0.4358	(7, 2, 5, 9, 10, 11, 13, 15, T)	
T	1	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.01277	0.23167	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	4	0.06604	0.0387	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	7	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	9	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	10	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	12	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	13	0.00846	0.56671	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	14	0.00072	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	15	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	16	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	17	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, 14, T)	
T	1	0.00723	0.36919	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.03802	0.01001	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.03072	0.15358	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.00647	0.38726	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.0292	0.15787	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.0	0.4358	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.04682	0.05674	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	16	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	17	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	1	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	2	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	4	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	7	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	9	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	10	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	11	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	12	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	13	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	14	0.00012	0.4409	(7, 2, 5, 9, 10, 11, 12, T)	
T	15	0.00012	0.4409	(7, 2, 5, 9, 10,	

Table 27: Global South

Table 28: Least Countries (LDC)

(BRICS + N-11)

Table 29: Emerging Markets (BRICS + N-11) Table 30: Land Locked Developing Countries (LLDC)

<i>i</i>	<i>T</i>	<i>p</i>	<i>n</i>	<i>m</i>	<i>d</i>	<i>v</i>	<i>s</i>	<i>condition</i>
1	2	0.0067	0.61404	11	11	11	11	"
3	0	0.6795	0.47925	4	4	5	5	"
4	0.0016	0.61404	11	11	11	11	"	
5	0	0.68865	0.48174	3	3	4	4	"
7	0	0.68785	0.48174	3	3	4	4	"
8	0	0.68785	0.48174	3	3	4	4	"
10	0	0.67715	0.47925	4	4	5	5	"
11	0.02454	0.67359	11	11	11	11	"	
12	0.0005	0.61404	11	11	11	11	"	
13	0.00944	0.53885	5	5	5	5	"	
14	0.00504	0.62125	11	11	11	11	"	
15	0.0005	0.61404	11	11	11	11	"	
16	0.0076	0.66349	3	3	4	4	"	
17	0.12000	0.61404	11	11	11	11	"	
1	7	0.13007	0.60869	3	3	4	4	"
2	3	0.06069	0.60819	7	7	10	11	"
3	2	0.06069	0.60819	7	7	10	11	"
5	0.02446	0.66953	3	3	4	4	"	
6	0.02552	0.58267	5	5	5	5	"	
7	0.02552	0.66953	3	3	4	4	"	
8	0	0.6965	0.53287	3	3	4	4	"
9	0	0.6965	0.53287	3	3	4	4	"
10	0.00890	0.61404	11	11	11	11	"	
11	0	0.68925	0.48174	3	3	4	4	"
12	0.00050	0.61404	11	11	11	11	"	
13	5.045	0.45252	5	5	5	5	"	
14	0.03937	0.43836	7	7	10	11	"	
15	0.00050	0.61404	11	11	11	11	"	
16	0.00843	0.52665	5	5	5	5	"	
17	0.12000	0.61404	11	11	11	11	"	
1	17	0.02833	0.6386	3	3	4	4	"
2	17	0.02833	0.6386	3	3	4	4	"
3	4	0.10943	0.60869	7	7	10	11	"
5	3	0.06069	0.60819	7	7	10	11	"
6	3	0.13222	0.60819	7	7	10	11	"
7	0	0.65543	0.13139	1	1	2	3	"
8	0	0.65543	0.13139	1	1	2	3	"
9	0	0.64815	0.13139	1	1	2	3	"
10	0	0.67265	0.47925	4	4	5	5	"
11	0	0.66544	0.13139	1	1	2	3	"
12	0	0.65285	0.13139	1	1	2	3	"
13	0	0.65285	0.13139	1	1	2	3	"
14	0	0.65285	0.13139	1	1	2	3	"
15	0	0.65285	0.13139	1	1	2	3	"
16	0	0.65285	0.13139	1	1	2	3	"
17	0	0.65285	0.13139	1	1	2	3	"
1	3	0.03089	0.60801	7	7	10	11	"
2	3	0.03089	0.60801	7	7	10	11	"
3	4	0.06069	0.60819	7	7	10	11	"
4	6	0.03037	0.26048	3	3	4	4	"
5	6	0.03037	0.26048	3	3	4	4	"
6	7	0.03037	0.26048	3	3	4	4	"
7	9	0.07629	0.07019	3	3	4	4	"
8	10	0.03039	0.26048	3	3	4	4	"
9	11	0.03039	0.26048	3	3	4	4	"
10	12	0.03039	0.26048	3	3	4	4	"
11	13	0.03039	0.26048	3	3	4	4	"
12	14	0.03039	0.26048	3	3	4	4	"
13	15	0.03039	0.26048	3	3	4	4	"
14	16	0.03039	0.26048	3	3	4	4	"
15	17	0.03039	0.26048	3	3	4	4	"
16	18	0.03039	0.26048	3	3	4	4	"
17	19	0.03039	0.26048	3	3	4	4	"
1	5	0.00054	0.61404	11	11	11	11	"
2	6	0.00054	0.61404	11	11	11	11	"
3	7	0.00054	0.61404	11	11	11	11	"
4	8	0.00054	0.61404	11	11	11	11	"
5	9	0.00054	0.61404	11	11	11	11	"
6	10	0.00054	0.61404	11	11	11	11	"
7	11	0.00054	0.61404	11	11	11	11	"
8	12	0.00054	0.61404	11	11	11	11	"
9	13	0.00054	0.61404	11	11	11	11	"
10	14	0.00054	0.61404	11	11	11	11	"
11	15	0.00054	0.61404	11	11	11	11	"
12	16	0.00054	0.61404	11	11	11	11	"
13	17	0.00054	0.61404	11	11	11	11	"
14	18	0.00054	0.61404	11	11	11	11	"
15	19	0.00054	0.61404	11	11	11	11	"
16	20	0.00054	0.61404	11	11	11	11	"
17	21	0.00054	0.61404	11	11	11	11	"
18	22	0.00054	0.61404	11	11	11	11	"
19	23	0.00054	0.61404	11	11	11	11	"
20	24	0.00054	0.61404	11	11	11	11	"
21	25	0.00054	0.61404	11	11	11	11	"
22	26	0.00054	0.61404	11	11	11	11	"
23	27	0.00054	0.61404	11	11	11	11	"
24	28	0.00054	0.61404	11	11	11	11	"
25	29	0.00054	0.61404	11	11	11	11	"
26	30	0.00054	0.61404	11	11	11	11	"
27	31	0.00054	0.61404	11	11	11	11	"
28	32	0.00054	0.61404	11	11	11	11	"
29	33	0.00054	0.61404	11	11	11	11	"
30	34	0.00054	0.61404	11	11	11	11	"
31	35	0.00054	0.61404	11	11	11	11	"
32	36	0.00054	0.61404	11	11	11	11	"
33	37	0.00054	0.61404	11	11	11	11	"
34	38	0.00054	0.61404	11	11	11	11	"
35	39	0.00054	0.61404	11	11	11	11	"
36	40	0.00054	0.61404	11	11	11	11	"
37	41	0.00054	0.61404	11	11	11	11	"
38	42	0.00054	0.61404	11	11	11	11	"
39	43	0.00054	0.61404	11	11	11	11	"
40	44	0.00054	0.61404	11	11	11	11	"
41	45	0.00054	0.61404	11	11	11	11	"
42	46	0.00054	0.61404	11	11	11	11	"
43	47	0.00054	0.61404	11	11	11	11	"
44	48	0.00054	0.61404	11	11	11	11	"
45	49	0.00054	0.61404	11	11	11	11	"
46	50	0.00054	0.61404	11	11	11	11	"
47	51	0.00054	0.61404	11	11	11	11	"
48	52	0.00054	0.61404	11	11	11	11	"
49	53	0.00054	0.61404	11	11	11	11	"
50	54	0.00054	0.61404	11	11	11	11	"
51	55	0.00054	0.61404	11	11	11	11	"
52	56	0.00054	0.61404	11	11	11	11	"
53	57	0.00054	0.61404	11	11	11	11	"
54	58	0.00054	0.61404	11	11	11	11	"
55	59	0.00054	0.61404	11	11	11	11	"
56	60	0.00054	0.61404	11	11	11	11	"
57	61	0.00054	0.61404	11	11	11	11	"
58	62	0.00054	0.61404	11	11	11	11	"
59	63	0.00054	0.61404	11	11	11	11	"
60	64	0.00054	0.61404	11	11	11	11	"
61	65	0.00054	0.61404	11	11	11	11	"
62	66	0.00054	0.61404	11	11	11	11	"
63	67	0.00054	0.61404	11	11	11	11	"
64	68	0.00054	0.61404	11	11	11	11	"
65	69	0.00054	0.61404	11	11	11	11	"
66	70	0.00054	0.61404	11	11	11	11	"
67	71	0.00054	0.61404	11	11	11	11	"
68	72	0.00054	0.61404	11	11	11	11	"
69	73	0.00054	0.61404	11	11	11	11	"
70	74	0.00054	0.61404	11	11	11	11	"
71	75	0.00054	0.61404	11	11	11	11	"
72	76	0.00054	0.61404	11	11	11	11	"
73	77	0.00054	0.61404	11	11	11	11	"
74	78	0.00054	0.61404	11	11	11	11	"
75	79	0.00054	0.61404	11	11	11	11	"
76	80	0.00054	0.61404	11	11	11	11	"
77	81	0.00054	0.61404	11	11	11	11	"
78	82	0.00054	0.61404	11	11	11	11	"
79	83	0.00054	0.61404	11	11	11	11	"
80	84	0.00054	0.61404	11	11	11	11	"
81	85	0.00054	0.61404	11	11	11	11	"
82	86	0.00054	0.61404	11	11	11	11	"
83	87	0.00054	0.61404	11	11	11	11	"
84	88	0.00054	0.61404	11	11	11	11	"
85	89	0.00054	0.61404	11	11	11	11	"
86	90	0.00054	0.61404	11	11	11	11	"
87	91	0.00054	0.61404	11	11	11	11	"
88	92	0.00054	0.61404	11	11	11	11	"
89	93	0.00054	0.61404	11	11	11	11	"
90	94	0.00054	0.61404	11	11	11	11	"
91	95	0.00054	0.61404	11	11	11	11	"
92	96	0.00054	0.61404	11	11	11	11	"
93	97	0.00054	0.61404	11	11	11	11	"
94	98	0.00054	0.61404	11	11	11	11	"
95	99	0.00054	0.61404	11	11	11	11	"
96	100	0.00054	0.61404	11	11	11	11	"
97	101	0.00054	0.61404	11	11	11	11	"
98	102	0.00054	0.61404	11	11	11	11	"
99	103	0.00054	0.61404	11	11	11	11	"
100	104	0.00054	0.61404	11	11	11	11	"
101	105	0.00054	0.61404	11	11	11	11	"
102	106	0.00054	0.61404	11	11	11	11	"
103	1							

Table 31: Small Island Developing States (SIDS)

Var	Xar	Var	Yar	min	door	p-value	condition
1	2	0.0357	0.0333	(-1, 4, 7, 8, 10, 16)			
1	4	0.0196	0.0189	(2, 7, 8, 9, 11, 13, 16*, 17*, T)			
1	5	0.0	0.0486	(2, 7, 8, 9, 10, 11, 13, 16*, 17*, T)			
1	7	0.0606	0.0307	(2, 7, 8, 9, 13, 16*, T)			
1	8	0.0210	0.0189	(2, 7, 8, 9, 13, 16*, T)			
1	10	0.0309	0.0392	(2, 7, 8, 11*, 12*, 16*, T)			
1	12	0.0622	0.0551	(2, 7, 8, 13*, 14*, 15*, 16*)			
1	13	0.0797	0.0729	(2, 7, 8, 13*, 14*, 15*, 16*)			
1	15	4.05	0.4366	(7, 8, T)			
1	16	0.6395	0.1609	(2, 7, 8, 9, 11, 13, 16*, 17*, T)			
1	17	0.0	0.0389	(2, 7, 8, 9, 10, 13, 16*, T)			
1	T	0.0109	0.0786	(2, 7, 8, 9, 11, 15*, 17*)			
2	3	0.0048	0.0048	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, 16*)			
2	5	0.0284	0.0426	(2, 7, 8, 9, 10, 12*, 13*, 14*, 15*, 16*)			
2	7	0.1688	0.0217	(2, 7, 8, 9, 10, 12*, 13*, 14*, 15*, 16*)			
2	8	0.0929	0.0111	(2, 7, 8, 9, 12, 13*, 14*, 15*, 16*, T)			
2	10	0.0241	0.0282	(2, 7, 8, 12*, 13*, 15*, 16*, T)			
2	12	0.0021	0.0021	(2, 7, 8, 12*, 13*, 14*, 15*, 16*, T)			
2	13	0.0297	0.0367	(2, 7, 8, 15*, T)			
2	15	0.0751	0.0839	(2, 7, 8, 9, 10, 11, 13*, 14*, 16*, T)			
2	16	0.0931	0.0747	(2, 7, 8, 9, 12, 13*, 14*, 15*, 17*)			
2	T	0.0125	0.0545	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, 16*)			
3	4	0.0143	0.0143	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, 16*, T)			
3	6	0.0185	0.0184	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, T)			
3	8	0.0251	0.0251	(2, 7, 8, 9, 10, 12*, 13*, 14*, 15*, T)			
3	9	0.0382	0.0284	(2, 7, 8, 9, 10, 12*, 13*, 14*, 15*, T)			
3	11	0.0162	0.0367	(2, 7, 8, 9, 12*, 13*, 14*, 15*, T)			
3	12	0.0161	0.0247	(2, 7, 8, 9, 10, 11, 13*, T)			
3	13	0.0165	0.0247	(2, 7, 8, 9, 10, 11, 13*, T)			
3	14	0.0428	0.0362	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
3	15	0.0742	0.0667	(2, 7, 8, 9, 10, 11, 12*, 13*, 15*, T)			
3	17	0.0769	0.0313	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, T)			
4	5	0.1708	0.0835	(2, 7, 8, 9, 10, 11, 12*, 13*, 14*, 15*, T)			
4	6	0.0713	0.0284	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	8	0.1468	0.0616	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	9	0.0530	0.0179	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	11	0.0626	0.0289	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	13	0.0499	0.0595	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	14	0.0640	0.0709	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	16	0.0431	0.0198	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
4	17	0.1731	0.0121	(2, 7, 8, 9, 11, 12*, 13*, 14*, T)			
5	6	0.024	0.0583	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	7	0.0247	0.0173	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	9	0.0445	0.0696	(2, 7, 8, 9, 10, 11, 12*, T)			
5	11	0.0423	0.0368	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	12	0.0202	0.0297	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	13	0.0	0.0695	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	14	0.0	0.0695	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	15	0.0686	0.0289	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
5	17	0.0	0.0475	(2, 7, 8, 9, 10, 11, 12*, 13*, T)			
6	7	0.0667	0.0170	(2, 7, 8, 9, 10, 11, 12*, T)			
6	8	0.0347	0.0584	(2, 7, 8, 9, 10, 11, 12*, T)			
6	10	0.1593	0.0367	(2, 7, 8, 9, 10, 11, 12*, T)			
6	11	0.0781	0.0247	(2, 7, 8, 9, 10, 11, 12*, T)			
6	13	0.0277	0.0188	(2, 7, 8, 9, 10, 11, 12*, T)			
6	14	0.0062	0.0496	(2, 7, 8, 9, 10, 11, 12*, T)			
6	15	0.0405	0.0659	(2, 7, 8, 9, 10, 11, 12*, T)			
6	17	0.0	0.0475	(2, 7, 8, 9, 10, 11, 12*, T)			
7	6	0.0563	0.0425	(2, 7, 8, 9, 10, 11, 12*, T)			
7	8	0.0047	0.0495	(2, 7, 8, 9, 10, 11, 12*, T)			
7	9	0.0452	0.0279	(2, 7, 8, 9, 10, 11, 12*, T)			
7	10	0.0452	0.0425	(2, 7, 8, 9, 10, 11, 12*, T)			
7	11	0.0569	0.0279	(2, 7, 8, 9, 10, 11, 12*, T)			
7	13	0.0565	0.0238	(2, 7, 8, 9, 10, 11, 12*, T)			
7	14	0.0045	0.0496	(2, 7, 8, 9, 10, 11, 12*, T)			
7	15	9.05	0.0659	(2, 7, 8, 9, 10, 11, 12*, T)			
7	16	0.0	0.0475	(2, 7, 8, 9, 10, 11, 12*, T)			
7	17	0.0	0.0793	(2, 7, 8, 9, 10, 11, 12*, T)			
7	T	0.0271	0.0270	(2, 7, 8, 9, 10, 11, 12*, T)			
8	9	0.0731	0.0183	(2, 7, 8, 9, 10, 11, 12*, T)			
8	11	0.0	0.0475	(2, 7, 8, 9, 10, 11, 12*, T)			
8	13	0.0047	0.0426	(2, 7, 8, 9, 10, 11, 12*, T)			
8	14	0.0047	0.0426	(2, 7, 8, 9, 10, 11, 12*, T)			
8	15	0.0520	0.0218	(2, 7, 8, 9, 10, 11, 12*, T)			
8	16	0.0	0.0745	(2, 7, 8, 9, 10, 11, 12*, T)			
8	17	0.0	0.0793	(2, 7, 8, 9, 10, 11, 12*, T)			
8	18	0.0	0.0217	(2, 7, 8, 9, 10, 11, 12*, T)			
8	19	0.0	0.0255	(2, 7, 8, 9, 10, 11, 12*, T)			
8	20	0.0	0.0238	(2, 7, 8, 9, 10, 11, 12*, T)			
9	11	0.0	0.0183	(2, 7, 8, 9, 10, 11, 12*, T)			
9	13	0.0063	0.0158	(2, 7, 8, 9, 10, 11, 12*, T)			
9	15	0.0167	0.0425	(2, 7, 8, 9, 10, 11, 12*, T)			
9	16	0.0	0.0475	(2, 7, 8, 9, 10, 11, 12*, T)			
9	17	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
9	18	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
9	19	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
9	20	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	13	0.0234	0.0279	(2, 7, 8, 9, 10, 11, 12*, T)			
10	14	0.0520	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	15	0.0520	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	16	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	17	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	18	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	19	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
10	20	0.0	0.0520	(2, 7, 8, 9, 10, 11, 12*, T)			
11	14	0.0451	0.0345	(2, 7, 8, 9, 10, 11, 12*, T)			
11	15	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
11	16	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
11	17	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
11	18	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
11	19	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
11	20	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	13	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	14	0.0545	0.0251	(2, 7, 8, 9, 10, 11, 12*, T)			
12	15	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	16	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	17	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	18	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	19	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
12	20	0.0	0.0381	(2, 7, 8, 9, 10, 11, 12*, T)			
13	14	0.0362	0.0362	(2, 7, 8, 9, 10, 11, 12*, T)			
13	15	0.0	0.0397	(2, 7, 8, 9, 10, 11, 12*, T)			
13	17	0.0345	0.0345	(2, 7, 8, 9, 10, 11, 12*, T)			
13	18	0.0	0.0397	(2, 7, 8, 9, 10, 11, 12*, T)			
13	19	0.0	0.0397	(2, 7, 8, 9, 10, 11, 12*, T)			
13	20	0.0	0.0397	(2, 7, 8, 9, 10, 11, 12*, T)			
14	15	0.0	0.0397	(2, 7, 8, 9, 10, 11, 12*, T)			
14	16	0.0368	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
14	17	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
14	18	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
14	19	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
14	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
15	16	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
15	17	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
15	18	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
15	19	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
15	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
16	17	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
16	18	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
16	19	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
16	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
17	18	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
17	19	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
17	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
18	19	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
18	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			
19	20	0.0	0.0368	(2, 7, 8, 9, 10, 11, 12*, T)			

Table 32: G20

Table 33: OPEC

X	V	Var	Y	min	dec	p-value	condition
1	3	0.0241	0.4805	(Y, 7, 10, 11, 12, 13, 14, 15, 16, 17)			
1	3	3-05	0.8495	(Y, 7, 10, 13, 17)			
4	4	0.22158	0.4805	(Y, 7, 10, 11, 12, 13, 14, 16)			
1	6	0.07556	0.4805	(Y, 7, 10, 11, 12, 13, 14, 15, 16, 17)			
7	8	0.08	0.4805	(Y, 7, 9, 11, 13, 15, 16, 17)			
1	9	0.07338	0.71233	(Y, 7, 9, 11, 13, 15, 16, 17)			
1	10	0.08088	0.4805	(Y, 7, 9, 11, 13, 15, 16, 17)			
1	11	1e-05	0.4805	(Y, 7, 9, 12, 15, 16, 17)			
1	12	1e-05	0.4705	(Y, 7, 9, 10, 11, 13, 14, 16, 17)			
1	13	0.08	0.4805	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
1	14	0.02129	0.4805	(Y, 7, 9, 10, 11, 12, 13, 14)			
1	15	0.04622	0.6734	(Y, 7, 9, 10, 11, 12, 13, 14)			
1	16	0.04622	0.6734	(Y, 7, 9, 10, 11, 12, 13, 14)			
1	17	0.07093	0.6734	(Y, 7, 9, 10, 11, 12, 13, 14)			
2	3	0.04035	0.38187	(Y, 7, 9, 10, 11, 15, 16, 17)			
2	4	0.04144	0.4646	(Y, 7, 9, 12, 13, 16)			
2	6	0.17189	0.4805	(Y, 7, 9, 11, 13, 15, 16, 17)			
2	7	0.09224	0.4805	(Y, 7, 9, 13, 15, 17)			
2	9	0.01183	0.4826	(Y, 7, 9, 13, 15)			
2	10	1e-05	0.4826	(Y, 7, 9, 12, 13, 17)			
2	11	0.08	0.4826	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
2	12	0.12153	0.4826	(Y, 7, 9, 14, 16, 17)			
2	13	0.04826	0.4826	(Y, 7, 9, 11, 12, 13, 14)			
2	14	0.0532	0.6414	(Y, 7, 9, 12, 16, 17)			
2	15	0.087	0.2418	(Y, 7, 13, 17)			
2	16	0.04622	0.6734	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
2	17	0.05142	0.6734	(Y, 7, 9, 11, 12, 16, 17)			
2	T	6e-05	0.6734	(Y, 7, 9, 11, 12, 13, 14, 16, 17)			
3	3	0.05	0.4805	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
3	6	0.08	0.4805	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
3	8	0.12706	0.4129	(Y, 7, 9, 11, 12, 13, 14, 16, 17)			
3	10	0.0293	0.5724	(Y, 7, 9, 11, 16, 17)			
3	11	0.17444	0.7284	(Y, 7, 9, 11, 16, 17)			
3	12	0.04338	0.5724	(Y, 7, 9, 11, 16, 17)			
3	13	0.04338	0.5457	(Y, 7, 9, 11, 12, 14, 16, 17)			
3	14	0.0	0.4721	(Y, 7, 9, 10, 13, 15, 16, 17)			
3	15	0.08	0.4721	(Y, 7, 9, 10, 13, 14, 15, 17)			
3	16	0.12724	0.5724	(Y, 7, 9, 11, 12, 14, 16, 17)			
3	17	0.07576	0.5964	(Y, 7, 9, 11, 12, 15, 16, 17)			
3	T	1e-05	0.5964	(Y, 7, 9, 11, 12, 15, 16, 17)			
4	5	0.0817	0.4826	(Y, 7, 9, 13, 15, 17)			
4	6	1e-05	0.6734	(Y, 7, 9, 10, 11, 13, 14, 15, 16, 17)			
4	8	0.0177	0.4296	(Y, 7, 9, 11, 17)			
4	10	0.11232	0.6426	(Y, 7, 9, 11, 16, 17)			
4	11	0.1104	0.8789	(Y, 7, 9, 13, 15, 16, 17)			
4	12	0.04188	0.4805	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
4	13	0.01188	0.6426	(Y, 7, 9, 11, 12, 14)			
4	14	0.02887	0.4805	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
4	15	0.04188	0.6426	(Y, 7, 9, 11, 12, 14, 16, 17)			
4	16	0.00517	0.6993	(Y, 7, 9, 11, 12, 14, 17)			
4	17	0.00447	0.6993	(Y, 7, 9, 11, 12, 14, 16, 17)			
4	T	0.00447	0.6993	(Y, 7, 9, 11, 12, 14, 16, 17)			
5	6	0.0845	0.6426	(Y, 7, 9, 10, 11, 13, 14, 15, 17)			
5	7	0.0603	0.6643	(Y, 7, 9, 13, 16, 17)			
5	11	0.0603	0.6643	(Y, 7, 9, 13, 16, 17)			
5	12	0.04188	0.6643	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
5	13	0.04611	0.6643	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
5	14	0.0	0.6993	(Y, 7, 9, 10, 11, 13, 15, 17)			
5	15	0.01138	0.6643	(Y, 7, 9, 10, 11, 13, 14, 16, 17)			
5	16	0.12961	0.6643	(Y, 7, 9, 10, 11, 13, 14, 16, 17)			
5	17	0.08	0.6643	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
5	T	0.0843	0.6643	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
6	7	0.0838	0.6272	(Y, 7, 9, 12, 13, 17)			
6	8	0.05656	0.2967	(Y, 7, 9, 15, 16)			
6	10	0.0002	0.4595	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
6	11	0.0002	0.4595	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
6	12	0.02368	0.6956	(Y, 7, 9, 10, 12, 13, 14, 15, 17)			
6	13	0.00934	0.6956	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
6	14	0.01353	0.6956	(Y, 7, 9, 10, 11, 12, 14, 16, 17)			
6	15	0.01872	0.6956	(Y, 7, 9, 10, 11, 12, 13, 14, 16, 17)			
6	16	0.09	0.6956	(Y, 7, 9, 10, 11, 12, 14, 15, 17)			
6	17	0.02444	0.6956	(Y, 7, 9, 10, 11, 12, 14, 15, 17)			
6	T	0.02444	0.6956	(Y, 7, 9, 10, 11, 12, 14, 15, 17)			
7	16	0.02444	0.6956	(Y, 7, 9, 10, 11, 12, 14, 15, 17)			
7	T	0.15259	0.6734	(Y, 11, 12, 15, 17)			
8	9	0.07342	0.2913	(Y, 7, 9, 13, 16)			
8	10	0.08	0.7832	(Y, 7, 9, 10, 14)			
8	11	0.03272	0.6734	(Y, 7, 9, 13, 16)			
8	12	0.0056	0.6734	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	13	0.00556	0.6956	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
8	14	0.0	0.6956	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
8	15	0.01407	0.6956	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 17)			
8	16	0.0e-05	0.6985	(Y, 7, 9, 11, 17)			
8	17	0.011438	0.5853	(Y, 7, 9, 11, 12)			
8	18	0.00862	0.2348	(Y, 7, 9, 12, 14)			
8	19	0.00925	0.2948	(Y, 7, 9, 12, 16)			
8	20	0.00829	0.6574	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	21	0.00674	0.6574	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	22	0.02254	0.6449	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	23	0.0056	0.6449	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	24	0.02109	0.5539	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	25	0.06383	0.6023	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	26	0.0002	0.6023	(Y, 7, 9, 10, 11, 13, 15, 16, 17)			
8	27	0.014738	0.5922	(Y, 7, 9, 10)			
8	28	0.02424	0.5984	(Y, 7, 9, 16, 17)			
8	29	0.0058	0.5984	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
8	30	0.07	0.6705	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
8	T	0.05	0.4805	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	10	0.0829	0.6449	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	11	1e-05	0.6705	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	12	0.08	0.6705	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	13	0.02132	0.6705	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	14	0.11384	0.5578	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	15	0.0	0.5578	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	16	0.0305	0.6705	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	17	0.0	0.6416	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	18	0.007	0.4296	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	19	0.045	0.6449	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	20	0.0455	0.5724	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	21	0.005	0.5724	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	22	0.028642	0.5664	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	23	0.0867	0.6649	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	24	0.03978	0.6675	(Y, 7, 9, 10, 11, 12, 13, 14, 15, 16, 17)			
9	25	0.0001	0.6495	(Y, 7, 9, 10, 13, 15, 17)			
9	26	0.03832	0.5764	(Y, 7, 9, 10, 11, 13, 14, 16, 17)			

Table 34: Low Income

Table 35: Lower middle Income

X	Var	Y	unit	min	max	position	condition
1	0	0.0209	0.0041	-9.9	12.	'12, '13, '14, '15, '16, '17, "T"	
1	3	0.0134	0.3267	-7.4	8.	'9, '10, '11, '12, '13, '14, '15, '16, '17, "T"	
1	4	0.0767	0.0267	-5.0	7.	'8, '9, '10, '11, '12, '13, '17	
1	6	0.0268	0.3568	-9.	11,	'12, '13, '14, '16	
1	7	0.0134	0.0041	-8.0	11.	'10, '11, '12, '13, '14, '16, '17, "T"	
1	9	0.0374	0.1069	-4.0	9.	'4, '5, '6, '7, '8, '11, '12, '13, '14, '15, '16, '17, "T"	
1	11	0.0361	0.1234	-3.0	7.	'4, '5, '6, '7, '8, '11, '12, '13, '14, '15, '16, "T"	
1	12	0.0761	0.0267	-4.0	9.	'4, '5, '6, '7, '8, '11, '12, '13, '14, '15, "T"	
1	14	0.0364	0.1469	-3.0	7.	'6, '7, '8, '9, '11, '15, '17, "T"	
1	15	0.0204	0.7643	-7.	8.	"T"	
1	17	0.0511	0.0267	-4.0	9.	'4, '5, '6, '7, '8, '11, '12, '14, '16, "T"	
2	0	0.0765	0.0267	-4.0	9.	'4, '5, '6, '7, '8, '11, '12, '13, '14, '15	
2	3	0.0	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
2	4	0.0632	0.0267	-3.0	7.	'8, '9, '10, '11	
2	5	0.0	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
2	6	0.0149	0.3147	-4.0	9.	'7, '8, '10, '11	
2	7	0.0093	0.0267	-4.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
2	9	0.0	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
2	10	0.0143	0.0041	-4.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
2	12	0.0	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
2	14	0.0215	0.7683	-7.	8.	"T"	
2	15	0.0	0.0045	-5.0	7.	'5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '16	
2	17	0.0092	0.0267	-3.0	7.	'8, '9, '10, '11	
2	18	0.0224	0.6404	-7.	8.	"T"	
2	T	0.0194	0.7293	-7.	8.	"T"	
3	0	0.0025	0.0041	-4.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
3	5	0.0143	0.9151	-7.	8.	'9, '10, '11, '12, "T"	
3	6	0.0203	0.0041	-4.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
3	8	0.0	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
3	9	0.0002	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
3	10	0.0	0.0045	-5.0	7.	'5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '16	
3	11	0.0069	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15	
3	12	0.0001	0.0005	-8.0	9.	'10, '11, '12, '13, '14, '15, "T"	
3	13	0.0274	0.7748	-7.	8.	"T"	
3	14	0.0179	0.0005	-8.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
3	16	0.0	0.0005	-8.0	9.	'7, '8, '10, '11, '12, '13, '14, '15	
3	17	0.0219	0.0005	-8.0	9.	'7, '8, '10, '11, '12, '13, '14, '15, "T"	
3	18	0.0129	0.0005	-8.0	9.	'7, '8, '10, '11, '12, '13, '14, '15, "T"	
4	5	0.0054	0.0041	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	7	0.0324	0.1558	-2.0	7.	'3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	8	0.0074	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	9	0.0047	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	10	0.0	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	11	0.00719	0.0221	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	13	0.0245	0.1557	-2.0	7.	'3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	14	0.0174	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	15	0.0	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	16	0.0157	0.0224	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	17	0.0071	0.0736	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
4	18	0.0069	0.0267	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	8	0.0083	0.0147	-2.0	7.	"T"	
5	9	0.0005	0.0041	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	10	0.0009	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	11	0.0047	0.0055	-2.0	7.	"T"	
5	12	0.0016	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	13	0.0036	0.0076	-2.0	7.	"T"	
5	14	0.0174	0.3249	-2.0	7.	"T"	
5	15	0.0027	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	16	0.0074	0.0224	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	17	0.0018	0.0041	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	18	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	19	0.0074	0.0224	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
5	20	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
6	T	1.0000	0.0005	-2.0	7.	"T"	
7	0	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	6	0.0070	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	8	0.0053	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	10	0.0071	0.0267	-2.0	7.	"T"	
7	12	0.0068	0.0079	-2.0	7.	"T"	
7	13	0.0062	0.0079	-2.0	7.	"T"	
7	14	0.0034	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	15	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	16	0.0068	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	17	0.0071	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	18	0.0071	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	19	0.0071	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
7	20	0.0071	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	T	1.0000	0.0005	-2.0	7.	"T"	
8	9	0.00725	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	10	0.0063	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	12	0.0068	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	13	0.00673	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	14	0.00066	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	15	0.00125	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	16	0.00095	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	17	0.00475	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	18	0.00745	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	19	0.00745	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
8	20	0.00745	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	T	1.0000	0.0005	-2.0	7.	"T"	
9	13	0.0159	0.0041	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	15	0.0258	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	16	0.0086	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	17	0.00283	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	18	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	19	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
9	20	0.0005	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	T	1.0000	0.0005	-2.0	7.	"T"	
10	12	0.01843	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	14	0.0401	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	15	0.0045	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	16	0.00389	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	17	0.00434	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	18	0.00067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	19	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
10	20	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	T	1.0000	0.0005	-2.0	7.	"T"	
11	13	0.0184	0.00403	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	14	0.0258	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	15	0.0045	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	16	0.00389	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	17	0.00434	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	18	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	19	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
11	20	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	T	1.0000	0.0005	-2.0	7.	"T"	
12	13	0.0184	0.00403	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	14	0.0258	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	15	0.0045	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	16	0.00389	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	17	0.00434	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	18	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	19	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
12	20	0.0067	0.0005	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	
13	T	1.0000	0.0005	-2.0	7.	"T"	
13	14	0.0184	0.00403	-2.0	7.	'2, '3, '4, '5, '6, '7, '8, '9, '10, '11, '12, '13, '14, '15	

Table 36: Upper middle In-
come

Table 37: High Income

The grouping World encompasses all listed countries

LDC: Least Developed Countries

LLDC: Land Locked Developing Countries

SIDS: Small Island Developing States

Emerging Markets: BRICS + N-11