

University of Dundee

## Prescribing patterns and response to antihyperglycemic agents among novel clusters of type 2 diabetes in Asian Indians

Anjana, Ranjit Mohan; Siddiqui, Moneeza Kalhan; Jebarani, Saravanan; Vignesh, Mani Arun; Raj, Nithyanantham Kamal; Unnikrishnan, Ranjit

*Published in:*  
Diabetes Technology & Therapeutics

*DOI:*  
[10.1089/dia.2021.0277](https://doi.org/10.1089/dia.2021.0277)

*Publication date:*  
2022

*Document Version*  
Peer reviewed version

[Link to publication in Discovery Research Portal](#)

### *Citation for published version (APA):*

Anjana, R. M., Siddiqui, M. K., Jebarani, S., Vignesh, M. A., Raj, N. K., Unnikrishnan, R., Pradeepa, R., Panikar, V. K., Kesavadev, J., Saboo, B., Gupta, S., Sosale, A. R., Seshadri, K. G., Deshpande, N., Chawla, M., Chawla, P., Das, S., Behera, M., Chawla, R., ... Mohan, V. (2022). Prescribing patterns and response to antihyperglycemic agents among novel clusters of type 2 diabetes in Asian Indians. *Diabetes Technology & Therapeutics*, 24(3), 190-200. <https://doi.org/10.1089/dia.2021.0277>

### **General rights**

Copyright and moral rights for the publications made accessible in Discovery Research Portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

- Users may download and print one copy of any publication from Discovery Research Portal for the purpose of private study or research.
- You may not further distribute the material or use it for any profit-making activity or commercial gain.
- You may freely distribute the URL identifying the publication in the public portal.

### **Take down policy**

If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

1           **PRESCRIBING PATTERNS AND RESPONSE TO ANTIHYPERGLYCEMIC**  
2           **AGENTS AMONG NOVEL CLUSTERS OF TYPE 2 DIABETES IN ASIAN INDIANS**

3  
4  
5           Ranjit Mohan Anjana<sup>#\*1</sup>, Moneeza Kalhan Siddiqui<sup>\*2</sup>, Saravanan Jebarani<sup>1</sup>, Mani Arun  
6           Vignesh <sup>1</sup>, Nithyanantham Kamal Raj <sup>1</sup>, Ranjit Unnikrishnan<sup>1</sup>, Rajendra Pradeepa <sup>1</sup>,  
7           Vijay K Panikar<sup>3</sup>, Jothydev Kesavadev<sup>4</sup>, Banshi Saboo<sup>5</sup>, Sunil Gupta<sup>6</sup>, Aravind R.  
8           Sosale<sup>7</sup>, Krishna G Seshadri<sup>8</sup>, Neeta Deshpande<sup>9</sup>, Manoj Chawla<sup>10</sup>, Purvi Chawla<sup>10</sup>,  
9           Sidhartha Das<sup>11</sup>, Manoranjan Behera<sup>12</sup>, Rajeev Chawla<sup>13</sup>, Anant Nigam<sup>14</sup>, Arvind  
10          Gupta<sup>15</sup>, Rajiv Kovil<sup>16</sup>, Shashank R Joshi<sup>17</sup>, Sanjay Agarwal<sup>18</sup>, Sarita Bajaj<sup>19</sup>, Ewan R  
11          Pearson<sup>2</sup>, Alexander SF Doney<sup>2</sup>, Colin NA Palmer<sup>2</sup>, Viswanathan Mohan<sup>1</sup>

12  
13  
14          <sup>1</sup> Madras Diabetes Research Foundation and Dr.Mohan's Diabetes Specialities  
15          Centre, Chennai, India

16          <sup>2</sup> Division of Population Health & Genomics, School of Medicine, University of  
17          Dundee, Dundee, UK,

18          <sup>3</sup> Dr. Panikar's Speciality Care Centre, Mumbai, Maharashtra, India

19          <sup>4</sup>Jothydev's Diabetes and Research Centre, Kerala, India

20          <sup>5</sup>Diabetes Care & Hormone Clinic, Ahmedabad, Gujarat, India

21          <sup>6</sup> Sunil's Diabetic Care & Research Center, Nagpur, Maharashtra, India

22          <sup>7</sup> Diacon Hospital, Bangalore, Karnataka, India

23          <sup>8</sup> Chennai Diabetes and Endocrine Clinic, Chennai, Tamilnadu, India

24          <sup>9</sup> Belgaum Diabetes Centre, Belgaum, Karnataka, India

25          <sup>10</sup> Lina Diabetes Care, Mumbai, Maharashtra, India

26          <sup>11</sup> Prof.S.Das Clinic, Cuttack, Odisha, India

27          <sup>12</sup>SCB Medical College, Cuttack, Odisha, India

28          <sup>13</sup> North Delhi Diabetes Centre, Delhi, India

29          <sup>14</sup> Nigam Diabetes Centre, Jaipur, Rajasthan, India

30          <sup>15</sup>Rajasthan Hospital, Jaipur, Rajasthan, India

31          <sup>16</sup> Dr. Kovil's Diabetes Care Centre, Mumbai, Maharashtra, India

32          <sup>17</sup> Lilavati Hospital and Research Centre, Mumbai, Maharashtra, India

33          <sup>18</sup> Aegle Clinic – Diabetes Care, Pune, Maharashtra, India

34          <sup>19</sup> MLN Medical College, Allahabad, Uttar Pradesh, India

35  
36  
37          \* **Joint first author**

38          # **Corresponding author**

39  
40          **Running title:** Drug response in novel clusters of diabetes

41  
42          **Key words:** Type 2 diabetes, Drug response, Clusters, Antihyperglycemic agents,  
43          Indians

47 **ABSTRACT**

48 **AIM:** To assess the prescribing patterns and response to different classes of  
49 antihyperglycemic agents in novel clusters of type 2 diabetes (T2D) described in India.

50

51 **MATERIALS AND METHODS:** We attempted to replicate the earlier described  
52 clusters of T2D In 32,867 individuals with new-onset T2D (within 2 years of diagnosis)  
53 registered between October 2013 and December 2020 at 15 diabetes clinics located  
54 across India, by means of k-means clustering utilising six clinically relevant variables.  
55 Individuals who had followup HbA1c upto 2 years were included for the drug response  
56 analysis (n=13,247).

57

58 **RESULTS:** Among the 32,867 participants included in the study, 20779 (63.2%) were  
59 males. The average age at diagnosis was 45 years and mean HbA1c at baseline was  
60 8.9 %. The same four clusters described in India earlier were replicated. Forty percent  
61 of the study participants belonged to the Mild Age-Related Diabetes [MARD] cluster,  
62 followed by Insulin Resistant Obese Diabetes [IROD] (27%), Severe Insulin Deficient  
63 Diabetes [SIDD] (21%) and Combined Insulin Resistant and Deficient Diabetes  
64 [CIRDD] (12%) clusters. The most frequently used antihyperglycemic agents were  
65 sulphonylureas, metformin and dipeptidyl peptidase-4 inhibitors apart from insulin.  
66 While there were significant differences in HbA1c reduction between drugs across  
67 clusters, these were largely driven by differences in the baseline (pre-treatment)  
68 HbA1c.

69

70 **CONCLUSIONS:** In this new cohort we were able to reliably replicate the four  
71 subtypes of T2D earlier described in Asian Indians. Prescribing patterns show limited  
72 usage of newer antihyperglycemic agents across all clusters. Randomized clinical  
73 trials are required to establish differential drug responses between clusters.

74 **Introduction**

75 Type 2 diabetes (T2D) is a widespread metabolic disorder characterized by  
76 considerable heterogeneity in its pathophysiology, clinical manifestations and natural  
77 history [1]. Efforts have been made in various populations to identify distinct “clusters”  
78 of phenotypic characteristics and laboratory markers in individuals diagnosed with  
79 T2D, and to assess whether these clusters correlate with differential risk of diabetes  
80 complications [2,3]. A pioneering study by Ahlquist et al [2] in a Scandinavian  
81 population, led to the identification of five major subtypes of T2D termed as  
82 Severe Autoimmune Diabetes (SAID), Severe Insulin Deficient Diabetes (SIDDD),  
83 Severe Insulin Resistant Diabetes (SIRD), Mild Obesity-related Diabetes (MOD) and  
84 Mild Age-Related Diabetes (MARD). This was also reproduced in several other ethnic  
85 groups [4-6].

86

87 Asian Indians represent an ethnic group with high predilection for T2D, who have been  
88 shown to have certain differences with respect to clinical features compared to white  
89 Caucasians, such as onset of diabetes at younger ages and lower levels of obesity,  
90 as well more severe beta-cell insufficiency early in the disease course [7,8]. An earlier  
91 attempt at identifying clusters of T2D in the Asian Indian population replicated two of  
92 the clusters identified by Ahlquist et al viz. SIDDD and MARD, while two novel clusters  
93 termed CIRDD (Combined Insulin Resistant and Deficient Diabetes) and IROD  
94 (Insulin Resistant Obese Diabetes) were also described [9].

95

96 More than seven distinct categories of drugs are available for the management of T2D  
97 at present. There is little information available on the overall prescribing patterns of  
98 antihyperglycemic medications in India. However, there is considerable interest in

99 assessing how individuals in each of the clusters of T2D, respond to various classes  
100 of antihyperglycemic medications. In the present paper, we attempt to analyse the  
101 prescribing patterns for T2D in India, and assess the response to different classes of  
102 antihyperglycemic agents among the recently described “clusters” of Asian Indian  
103 patients with T2D by doing a retrospective analysis of medical records collected from  
104 multiple diabetes care practices across India.

## 105 **Materials and Methods**

### 106 ***Study population and inclusion criteria***

107 Retrospective data on individuals with new-onset T2D (within 2 years of diabetes  
108 diagnosis) who were registered between October 2013 and December 2020 was  
109 collected from 15 diabetes clinics located in 12 states and one Union Territory of India  
110 **(Figure 1)**. Diabetes was diagnosed if the fasting plasma glucose level was  $\geq 126$   
111 mg/dL (7.0 mmol/L) and/or 2-hour postload glucose level was  $\geq 200$  mg/dL (11.1  
112 mmol/L) and/or if the patient had been prescribed pharmacotherapy for diabetes by a  
113 physician [10], while T2D was diagnosed by absence of ketosis, good beta-cell reserve  
114 as shown by fasting C-peptide assay  $>0.6$  pmol/mL and absence of pancreatic calculi  
115 (on abdominal radiograph)[11]. In order to assign individuals into clinical clusters, data  
116 on age at diagnosis, body mass index (BMI), waist circumference, glycated  
117 hemoglobin (HbA1c), serum triglycerides and high density lipoprotein (HDL)  
118 cholesterol were collected from 32,867 individuals with new onset T2D. Height (in cms)  
119 was measured using a stadiometer, weight (in kg) was measured with an electronic  
120 weighing scale and waist circumference was measured using a nonstretchable  
121 measuring tape. BMI was calculated using the formula: weight (Kg)/(height in m)<sup>2</sup>.  
122 HbA1c, triglycerides and HDL-cholesterol were determined by standard  
123 methodologies followed in the respective labs in the 15 study centres (Serum

124 triglyceride analyzed by enzymatic method, HDL-cholesterol by direct method and  
125 HbA1c by high performance liquid chromatography).

126

127 Other data collected included medication use (if any) at the time of first visit and at the  
128 follow-up visits up to two years from first visit. Baseline drug information was available  
129 for 30,152 individuals and follow-up HbA1c (was available from 4.2 -7.4 months) was  
130 available in 14,240 individuals (**Figure 2**). Among these, 10,013 drug-naïve individuals  
131 who were prescribed medications at first visit were included in the analysis. Individuals  
132 who were not drug-naïve were included in the study if new medications were added at  
133 baseline (n=2941). Individuals in whom no changes were made to previous  
134 medications at baseline and who had a follow-up HbA1c available within 3 months  
135 were included (n=293), whereas those without a follow-up HbA1c (n=933) were  
136 excluded. Therefore, a total of 13,247 individuals were included in the drug response  
137 analysis. Written informed consent to use anonymized medical data was obtained  
138 from all study participants and approval was obtained from the Institutional Ethics  
139 Committee.

140

## 141 **Statistical analysis**

142

143 Baseline characteristics were described using means and standard deviations for  
144 normally distributed variables, and median and interquartile range for non-normally  
145 distributed variables. Clustering methodology has been previously described [9].  
146 Briefly, k-means clustering was done with a k value of 4 using k-means function (with  
147 a maximum iteration of 10,000) in R V.3.6.0. Cluster forming tendency of the data was  
148 validated by the Hopkins statistic value. The optimal number of clusters was

149 determined based on silhouette width. Cluster-wise stability was computed by Jaccard  
150 bootstrap method. Cluster analysis was performed on scaled and centered values.  
151 Cluster labels were assigned based on the phenotype characteristics of individual  
152 cluster mean values of the variables.

153 Sensitivity analysis was done using duration of diabetes <1 year, <3 years, and <5  
154 years. Clustering tendency of the three different duration groups had Hopkins statistics  
155 values of 0.14, 0.16 and 0.16 respectively indicating that there was no significant  
156 difference between the duration groups.

157 HbA1c response was presented as absolute difference in HbA1c and percentage  
158 difference  $[(\text{pre-treatment HbA1c} - \text{post treatment HbA1c})/\text{pre-treatment HbA1c}]$ . In  
159 our regression models, we presented the therapy as a predictor and HbA1c reduction  
160 as the dependent outcome. Both univariate models and multivariable models were  
161 presented. Multivariable models were adjusted for sex, time on treatment, BMI, waist  
162 circumference, triglyceride levels and HbA1c at baseline. Models are presented with  
163 the step-wise inclusion of variables. Model 1 is univariate (HbA1c reduction), Model 2  
164 is adjusted for sex, BMI, waist circumference and time on treatment, Model 3 is  
165 adjusted for sex, BMI, waist circumference, time on treatment, HDL-C and triglycerides  
166 and Model 4 is adjusted for sex, BMI, waist circumference, time on treatment, HDL-C,  
167 triglycerides and baseline HbA1c. We employed two sets of analyses using models  
168 described above. In the first analyses set, we demonstrated the within-cluster effect of  
169 medications when compared to metformin use. In the second set that was analyzed,  
170 we demonstrated the across-clusters effect of a therapy while using those prescribed  
171 the therapy in MARD as the reference. The second analyses was performed in order  
172 to demonstrate the bias in comparing individuals prescribed different therapies, who  
173 would intrinsically have differing disease severity. We also examined the HbA1c

174 reduction in response to therapies in the SIDD, IROD and CIRDD clusters in  
175 comparison to the MARD cluster taken as the reference. Models are presented in the  
176 same step-wise method as described above.

## 177 **Results**

178

179 Among the 32,867 participants included in the study, 20779 (63.2%) were males. The  
180 average age at diagnosis was 45 years and the baseline HbA1c was 8.9 %. Mean BMI  
181 in males and females was 26.7 kg/m<sup>2</sup> and 28.4 kg/m<sup>2</sup> respectively, while waist  
182 circumference was 94.8 cm in females and 96.6 cm in males. Mean HDL-C and  
183 triglycerides were 41 mg/dL and 159 mg/dL respectively.

184

185 **Table 1** describes the characteristics of individuals in the various clusters. Forty  
186 percent of the population belonged to the MARD cluster, followed by 27% who  
187 belonged to the IROD cluster, 21% to the SIDD and 12% to the CIRDD clusters. The  
188 distribution of other characteristics followed the expected distributions of the clusters  
189 in Asian Indians as previously described. The youngest age (36.1 years) of diagnosis  
190 was observed in the SIDD cluster, who also had the lowest BMI (23.3 kg/m<sup>2</sup>) and waist  
191 circumference (85.3 cm). The SIDD cluster also had the highest baseline HbA1c  
192 (11.2%). Individuals in the IROD cluster had the highest BMI (32.9 kg/m<sup>2</sup>) and waist  
193 circumference (108.9 cm), and systolic (130.4 mmHg) and diastolic (81.4 mmHg)  
194 blood pressure. The CIRDD cluster represents a combination of characteristics of  
195 SIDD and IROD, with a relatively young age of diabetes diagnosis (41.9 years), and  
196 high BMI (26.8 kg/m<sup>2</sup>) and HbA1c (9.4%) at baseline. They also had the lowest HDL-  
197 C and the highest triglycerides (365 mg/dL). The MARD cluster represented the  
198 mildest presentation of T2D and had the oldest average age at diagnosis (50.7 years),



199 lowest HbA1c at diagnosis (7.6%), highest HDL-C (43 mg/dL) and lowest triglycerides  
200 (139 mg/dL).

201 The proportion of prescriptions of antihyperglycemic agents is provided in **Table 2**.  
202 Overall, metformin and sulphonylureas (SU) (in combination followed by singly) were  
203 the most commonly prescribed oral antihyperglycemic agents as per our analyses.  
204 Among the newer agents, dipeptidyl peptidase-4 inhibitors [DPP4i] were the most  
205 frequently used drugs, mostly in combination with metformin. In the SIDD cluster,  
206 insulin was frequently used with or without oral antidiabetic drugs [OAD](48.4%). This  
207 was followed by a combination of metformin and SU (13.7%) and triple therapy of  
208 metformin, SU and DPP4i (10.2%). In the IROD cluster, the most frequently prescribed  
209 medication was a combination of metformin and SU (26.9%), followed by insulin and  
210 an oral agent (21.5%). In the CIRDD group, the most common prescription at  
211 presentation was a combination of insulin and an oral agent (34.3%), followed by dual  
212 therapy of metformin and SU(21.8%), and triple therapy of metformin, SU and DPP4i  
213 (19.4%). Finally, in the MARD cluster, the most common prescription was dual therapy  
214 of metformin and SU (32.6%), followed by metformin monotherapy (17%).

215

216 **Table 3** shows the mean difference in HbA1c between baseline and follow-up  
217 (unadjusted HbA1c response) and the average interval between HbA1c  
218 measurements. The treatments used reflect the baseline HbA1c, with insulin being  
219 initiated in those with the highest HbA1c. Consistent with this, the greatest absolute  
220 reduction with insulin as monotherapy or in combination with oral agents, were found  
221 in SIDD and CIRDD as compared to IROD and MARD. In the SIDD group, dual  
222 therapies with SU/DPP-4i, Metformin/ Sodium-glucose co-transporter-2 inhibitors  
223 [SGLT2i], Metformin/thiazolidinediones [TZD] also produced significant reductions in

224 HbA1c, while in CIRDD, triple therapy with metformin/SU/TZD and  
225 metformin/SU/DPP-4i was associated with greater HbA1c reduction compared to dual  
226 therapy. In IROD and MARD, the absolute reductions in HbA1c were lower with all the  
227 commonly used therapeutic combinations compared to the other clusters, probably on  
228 account of the lower baseline HbA1c in these two clusters. The interval between the  
229 HbA1c measurements in the various clusters ranged from 4.2 to 7.4 months.

230

231 **Table 4** shows the multivariable models of HbA1c response in comparison to  
232 metformin monotherapy within clusters. In the SIDD cluster, use of insulin was  
233 associated with greater reductions in HbA1c [13% (7%-18%) when used as  
234 monotherapy and 15% (10%-20%) in combination with oral agent] compared to  
235 metformin monotherapy even after adjusting for anthropometric measures, duration of  
236 treatment and lipid parameters (Model 3). Similarly, monotherapy with SU was  
237 associated with a 7% (0.3%-14%) greater reduction, and triple drug therapy with an  
238 8% (2%-13%) greater reduction in HbA1c compared to metformin monotherapy in the  
239 SIDD cluster.

240

241 Applying Model 3 to IROD, the insulin treated patients had the greatest reductions in  
242 HbA1c, when used as monotherapy 19% (8%-30%) or in combination with oral agent  
243 18% (14%-22%) compared to metformin monotherapy. Combination of metformin and  
244 SGLT2i was associated with 7% (1%-15%) greater reduction, and triple drug therapy  
245 with 10% (6%-16%) greater reduction in HbA1c in this cluster compared to metformin  
246 monotherapy.

247

248 The most impressive reductions in HbA1c in the CIRDD cluster were associated with  
249 insulin monotherapy [24% (15%-32%) greater reduction in HbA1c compared to  
250 metformin monotherapy] and combination of insulin with oral agents [20% (14%-25%)  
251 greater reduction in HbA1c compared to metformin monotherapy]. Dual therapy with  
252 metformin and SU was associated with 11% (6%-17%), and metformin and DPP-4i  
253 with 9% (2%-16%) greater reduction in HbA1c in this cluster compared to metformin  
254 monotherapy, whereas triple drug combinations of metformin and SU with TZD and  
255 DPP-4i were associated with 14% (4%-24%) and 12% (6%-18%) greater reductions  
256 in HbA1c respectively compared to metformin monotherapy.

257

258 In MARD, relative reductions in HbA1c were 5% (1%-9%) or lower for all drug  
259 categories and combinations studied in comparison to metformin monotherapy.

260

261 However, on applying Model 4 and after adjusting for baseline HbA1c, the differences  
262 lost statistical significance across all cluster categories.

263

264 **Table 5** shows multivariable models of HbA1c response in comparison with HbA1c  
265 response in the MARD cluster. Using MARD as reference, similar results were  
266 obtained with insulin showing best reductions of HbA1c in SIDD and CIRDD,  
267 secretagogues showing good reduction in SIDD, and CIRDD requiring early use of  
268 insulin and combination therapy.

## 269 **Discussion**

270 This study reports on the following findings: Firstly, in this independent cohort drawn  
271 from across India, we confirm the four subtypes of T2D namely SIDD, IROD, CIRDD  
272 and MARD as discussed in our original paper [9]. Earlier we had done the clustering

273 from a single diabetes clinic in India and the replication was done in a national  
274 epidemiological database. Being a cross-sectional study, the latter, despite being  
275 representative of India, lacked treatment details. In this study we have replicated our  
276 findings across 15 diabetes clinics in different parts of India. Secondly, based on  
277 observational data from these clinics, we describe, for the first time, patterns of drug  
278 prescriptions in new onset T2D in clinics across India. We also attempted to assess  
279 the relative efficacy of various drug categories across diabetes clusters, although the  
280 data were collected retrospectively.

281

282 The most commonly prescribed therapies in this newly diagnosed population with T2D  
283 in India were insulin, metformin, SU and DPP4i. Very few patients were prescribed  
284 newer antidiabetic agents such as SGLT2i and glucagon like peptide-1 receptor  
285 agonists (GLP-1RA). This pattern is similar to that reported in previous studies from  
286 India [12,13]. While many of the newer treatment options have been shown to have  
287 pleiotropic benefits in T2D, the limited use of these agents in India is most likely driven  
288 by affordability. It has also been shown even in the US that metformin, SU and insulin  
289 continue to be the most frequently used therapies for T2D [14]. However, in countries  
290 where patients do not have to pay for drug treatment out of pocket (such as the United  
291 Kingdom), newer agents such as DPP-4i and SGLT2i are fast replacing SU as the  
292 second most commonly prescribed oral antihyperglycemic medication [15]. The  
293 familiarity and comfort of the treating physician with tried and tested agents could also  
294 be a reason for the continued popularity of these older drugs in India. The fact that  
295 randomized controlled trials for efficacy of therapeutics are rarely conducted in non-  
296 white populations could further impact prescribing hesitancy.

297

298 Currently accepted therapeutic guidelines recommend initial monotherapy with  
299 metformin in individuals with T2D, unless the HbA1c is profoundly elevated [16-18].  
300 The choice of second line therapy if metformin alone fails to control hyperglycemia, is  
301 based on factors such as risk of hypoglycemia, need to avoid weight gain, renal and  
302 cardiovascular status and patient affordability and preferences. Studies in European  
303 populations suggest that patient phenotype could have a bearing on response to  
304 antihyperglycemic medications. For instance, it has been shown that individuals with  
305 markers of insulin resistance respond poorly to DPP-4i [19]. It has also been  
306 suggested that men with lower BMI respond better to SU whereas obese women  
307 respond better to TZD[20]. However, there is, as yet, little information available as to  
308 whether the clinical phenotype of the patient would influence the choice of first or  
309 second-line pharmacotherapy for T2D in Asian Indians.

310

311 The “Asian Indian phenotype” of T2D is uniquely characterised by young age of onset,  
312 occurrence at relatively low BMI, and relatively early onset of beta cell dysfunction  
313 [7].A recent meta-analysis has suggested that Asian Indians respond differentially to  
314 various classes of antihyperglycemic agents compared to white Caucasians; however,  
315 these findings did not consider the heterogeneity in clinical phenotypes among Asian  
316 Indians with T2D and was not intended to assess the efficacy of non-metformin  
317 therapies as first line agents for T2D [21].

318

319 In their landmark study, Ahlquist et al [2] suggested that, based on the presumed  
320 pathophysiology, individuals in different subcategories of T2D could be expected to  
321 respond differentially to various classes of antihyperglycemic medications. They  
322 postulated that individuals in the “insulin-deficient” clusters were more likely to respond

323 to insulin-providing therapies (including early initiation of exogenous insulin therapy)  
324 whereas those in the “insulin resistant” clusters would respond optimally to metformin.  
325 They also noted that a significant proportion of individuals were not being prescribed  
326 phenotype-appropriate medications, potentially leading to suboptimal glycemic control  
327 and potentially increased risk of complications. In our study, we found that some  
328 clinicians were indeed utilising phenotype-specific treatments in a fair proportion of  
329 their patients even in the absence of formal subclassification data to guide them. For  
330 instance, we find that insulin use was most frequent in the SIDD cluster, the subgroup  
331 with the most profound insulin deficiency, while use of metformin was highest in the  
332 insulin resistant, obese IROD cluster. This is likely because leaner patients tend to be  
333 preferentially put on insulin while overweight patients tend to be treated more with  
334 metformin. However, our results also suggest that there appears to be a large  
335 proportion of patients who are not on therapy appropriate to their phenotype and  
336 pathophysiology. We believe that knowledge of the subtypes of T2D would enable  
337 clinicians to fine-tune their management of diabetes so as to treat their patients more  
338 precisely and effectively.

339

340 As our study is a retrospective analysis of real-world data,our results reflect the  
341 diabetes management of physicians treating patients as per their clinical needs rather  
342 than randomising them to certain therapies as would be the case in a randomised  
343 controlled clinical trial. However, our results do provide some interesting clues as to  
344 the relative efficacy of different classes of antihyperglycemic agents in the various  
345 clusters of T2D. For instance, individuals with the insulin deficient SIDD and CIRDD  
346 phenotypes, who are younger and have difficult-to-control hyperglycemia, are likely to  
347 benefit from early initiation of insulin and insulin secretagogues and more widespread

348 use of combination therapies as opposed to metformin monotherapy. However,  
349 recommendations on cluster-specific management of T2D would need to await the  
350 completion of well-designed randomised controlled trials which are being planned.

351 The main limitation of our study lies in its retrospective nature. Most of the differences  
352 in drug response across clusters are driven by baseline HbA1c, precluding firm  
353 conclusions on the relative efficacy of these agents. There could be a small chance of  
354 including Maturity Onset Diabetes of the Young (MODY) or other types of diabetes  
355 as T2D. However, as the overall prevalence of other subtypes is very low, it is unlikely  
356 to influence the overall results. Another limitation is that we were unable to account for  
357 the effects of lifestyle modification on glycemic response; however, the participants  
358 have been recruited from specialist diabetes clinics with standardised treatment  
359 protocols, where all participants have been provided with diet and physical activity  
360 advice in addition to pharmacotherapy. A third limitation is that we were unable to  
361 assess the efficacy of drug categories such as GLP1 receptor agonists, alpha-  
362 glucosidase inhibitors and (to an extent) SGLT2i, on account of the small number of  
363 patients prescribed these drugs, perhaps a reflection of the cost of these agents as  
364 most patients pay out of pocket for medications in India. **The final limitation relates to**  
365 **the varying modes of data collection and standardisation of management across study**  
366 **centres; however, the majority of data (n=19002) comes from the electronic medical**  
367 **records of the coordinating centre, which is a single institution with branches all over**  
368 **India following standardized protocols. Data from the other study centres have been**  
369 **collected from manual as well as electronic medical records. However, all the sites**  
370 **follow common management guidelines from the Research Society for the Study of**  
371 **Diabetes in India (RSSDI).`**

372

373 **Conclusions**

374 In conclusion, our results confirm the four distinct clusters of T2D in a patient  
375 population derived from multiple diabetes clinics across India. Our findings suggest  
376 that traditional antihyperglycemic agents continue to enjoy wide popularity among  
377 prescribers in India, while newer agents are yet to gain ground. Identification of distinct  
378 phenotypes of T2D (using easily measurable variables) could help clinicians decide  
379 upon the most effective forms of therapy for the individual patient, an important first  
380 step towards precision and personalised diabetes care. Randomised controlled clinical  
381 trials are necessary to compare the efficiency of various classes of antihyperglycemic  
382 agents in different subtypes of T2D in India.

383 **Authors' Contributions:** RMA, VM, ERP and MKS conceived the study, and were  
384 involved in the interpretation of data and wrote the first and subsequent drafts of the  
385 manuscript. VM, RMA, VKP, JK, BS, SG, AS, KGS, ND, MC, PC, SD, MB, RC, AN,  
386 AG and RK provided data for analysis. SJ, MAK, NKR and MKS were involved in  
387 statistical analyses. RU, RP, CNAP, ASFD, SRJ, SA and SB were involved in the  
388 interpretation of data and provided comments on drafts of the manuscript. All authors  
389 contributed to revision of the manuscript and approved the final submitted version.  
390 RMA is the guarantor of this work, and as such had full access to all the data in the  
391 study and takes responsibility for the integrity of the data and the accuracy of the data  
392 analysis.

393

394 **Author Disclosure Statement:** No competing financial interests exist.

395



396 **Funding Information:** This research was funded by the National Institute for Health  
397 Research (NIHR) (INSPIRED 16/136/102) using UK aid from the UK Government to  
398 support global health research.

399

400

## 401 **References**

402 1. Galicia-Garcia U, Benito-Vicente A, Jebari S, Larrea-Sebal A, Siddiqi H, Uribe KB,  
403 Ostolaza H, Martín C. Pathophysiology of Type 2 Diabetes Mellitus. *Int J Mol Sci.*  
404 2020 ;21:6275.

405 2. Ahlqvist E, Storm P, Käräjämäki A, Martinell M, Dorkhan M, Carlsson A, et al. Novel  
406 subgroups of adult-onset diabetes and their association with outcomes: a data-  
407 driven cluster analysis of six variables. *Lancet Diabetes Endocrinol.* 2018 ;6:361-  
408 369.

409 3. Anjana RM, Baskar V, Nair ATN, Jebarani S, Siddiqui MK, Pradeepa R,  
410 Unnikrishnan R, Palmer C, Pearson E, Mohan V. Novel subgroups of type 2  
411 diabetes and their association with microvascular outcomes in an Asian Indian  
412 population: a data-driven cluster analysis: the INSPIRED study. *BMJ Open*  
413 *Diabetes Res Care.* 2020 Aug;8(1):e001506.

414 4. Zou X, Zhou X, Zhu Z, Ji L. Novel subgroups of patients with adult-onset diabetes  
415 in Chinese and US populations. *Lancet Diabetes Endocrinol.* 2019;7:9-11.

416 5. Bello-Chavolla OY, Bahena-López JP, Vargas-Vázquez A, Antonio-Villa NE,  
417 Márquez-Salinas A, Fermín-Martínez CA, et al; Metabolic Syndrome Study Group;  
418 Group of Study CAIPaDi. Clinical characterization of data-driven diabetes  
419 subgroups in Mexicans using a reproducible machine learning approach. *BMJ Open*  
420 *Diabetes Res Care.* 2020;8:e001550.

- 421 6. Duarte V, Ivo C, Veríssimo D, Silva J, Lopes L, Passos D, et al. Novel Clusters of  
422 Adult-Onset Diabetes in a Portuguese Population .Austin Journal of Endocrinology  
423 and Diabetes 2020;7:1076
- 424 7. Unnikrishnan R, Anjana RM, Mohan V. Diabetes in South Asians: is the phenotype  
425 different? Diabetes. 2014;63:53-5
- 426 8. Staimez LR, Deepa M, Ali MK, Mohan V, Hanson RL, Narayan KMV. Tale of two  
427 Indians: Heterogeneity in type 2 diabetes pathophysiology. Diabetes Metab Res  
428 Rev. 2019 ;35:e3192.
- 429 9. Anjana RM, Baskar V, Nair ATN, Jebarani S, Siddiqui MK, Pradeepa R, et al. Novel  
430 subgroups of type 2 diabetes and their association with microvascular outcomes in  
431 an Asian Indian population: a data-driven cluster analysis: the INSPIRED study.  
432 BMJ Open Diabetes Res Care. 2020 ;8:e001506.
- 433 10. Alberti KG, Zimmet PZ, Definition ZPZ. Definition, diagnosis and classification  
434 of diabetes mellitus and its complications. part 1: diagnosis and classification of  
435 diabetes mellitus provisional report of a WHO consultation. Diabet Med  
436 1998;15:539–53.
- 437 11. Mohan V, Shanthi Rani CS, Amutha A, et al. Clinical profile of long-term survivors  
438 and nonsurvivors with type 2 diabetes. Diabetes Care 2013;36:2190–7.
- 439 12. Singla R, Bindra J, Singla A, Gupta Y, Kalra S. Drug Prescription Patterns and  
440 Cost Analysis of Diabetes Therapy in India: Audit of an Endocrine Practice. Indian  
441 J Endocrinol Metab. 2019 ;23:40-45.
- 442 13. Mokta J, Mokta K, Ranjan A, Joshi I, Garg M. Diabetes Drug Prescription Pattern  
443 and Awareness Among Health Care Providers in Sub-Himalayan Region of India:  
444 A Population Based Study. J Assoc Physicians India. 2017;65:50-54.

- 445 14. Le P, Chaitoff A, Misra-Hebert AD, Ye W, Herman WH, Rothberg MB. Use of  
446 Antihyperglycemic Medications in U.S. Adults: An Analysis of the National Health  
447 and Nutrition Examination Survey. *Diabetes Care*. 2020 ;43:1227-1233
- 448 15. Wilkinson S, Douglas I, Stirnadel-Farrant H, et al. Changing use of antidiabetic  
449 drugs in the UK: trends in prescribing 2000–2017. *BMJ Open* 2018;8:e022768.
- 450 16. ICMR guidelines for management of type 2 diabetes 2018. Available at  
451 [https://main.icmr.nic.in/sites/default/files/guidelines/ICMR\\_GuidelinesType2diabe](https://main.icmr.nic.in/sites/default/files/guidelines/ICMR_GuidelinesType2diabetes2018_0.pdf)  
452 [tes2018\\_0.pdf](https://main.icmr.nic.in/sites/default/files/guidelines/ICMR_GuidelinesType2diabetes2018_0.pdf). Accessed on 10<sup>th</sup> August 2021
- 453 17. Chawla R, Madhu SV, Makkar BM, Ghosh S, Saboo B, Kalra S; RSSDI-ESI  
454 Consensus Group. RSSDI-ESI Clinical Practice Recommendations for the  
455 Management of Type 2 Diabetes Mellitus 2020. *Indian J Endocrinol Metab*.  
456 2020;24:1-122.
- 457 18. American Diabetes Association. 9. Pharmacologic Approaches to Glycemic  
458 Treatment: *Standards of Medical Care in Diabetes-2021*. *Diabetes Care*. 2021  
459 Jan;44(Suppl 1):S111-S124.
- 460 19. Dennis JM, Shields BM, Hill AV, Knight BA, McDonald TJ, Rodgers LR, et al;  
461 MASTERMIND Consortium. Precision Medicine in Type 2 Diabetes: Clinical  
462 Markers of Insulin Resistance Are Associated With Altered Short- and Long-term  
463 Glycemic Response to DPP-4 Inhibitor Therapy. *Diabetes Care*. 2018 ;41:705-712.
- 464 20. Dennis JM, Henley WE, Weedon MN, Lonergan M, Rodgers LR, Jones AG, et al;  
465 MASTERMIND Consortium. Sex and BMI Alter the Benefits and Risks of  
466 Sulfonylureas and Thiazolidinediones in Type 2 Diabetes: A Framework for  
467 Evaluating Stratification Using Routine Clinical and Individual Trial Data. *Diabetes*  
468 *Care*. 2018 ;41:1844-1853.

469 21. Gan S, Dawed AY, Donnelly LA, Nair ATN, Palmer CNA, Mohan V, et al. Efficacy  
470 of Modern Diabetes Treatments DPP-4i, SGLT-2i, and GLP-1RA in White and  
471 Asian Patients With Diabetes: A Systematic Review and Meta-analysis of  
472 Randomized Controlled Trials. Diabetes Care. 2020 ;43:1948–57.

473

474

475 **ADDRESS FOR CORRESPONDENCE #**

476 **Dr. R.M.ANJANA, MD., Ph.D, Dip Diab (UK), FACP., FICP.,FRCP., (Glasg, Edin., &Lond)**

477 VICE PRESIDENT

478 MADRAS DIABETES RESEARCH FOUNDATION

479 ICMR CENTRE FOR ADVANCED RESEARCH ON DIABETES&

480 MANAGING DIRECTOR & CONSULTANT DIABETOLOGIST

481 DR.MOHAN'S DIABETES SPECIALITIES CENTRE,

482 4, CONRAN SMITH ROAD, GOPALAPURAM,

483 CHENNAI - 600 086. INDIA

484 TEL NO: (9144) 4396 8888; FAX NO: (9144) 2835 0935

485 Email: [dranjana@drmohans.com](mailto:dranjana@drmohans.com)

486 Website :[www.drrohansdiabetes.com](http://www.drrohansdiabetes.com), [www.mdrf.in](http://www.mdrf.in)