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Natural History of Type 2 Diabetes in Indians: Time to Progression

K.M. Venkat Narayan, Dimple Kondal, Howard H. Chang, Deepa Mohan, Unjali P. Gujral, Ranjit Mohan Anjana, Lisa R. Staimez, Shivani A. Patel, Mohammed K. Ali, Dorairaj Prabhakaran, Nikhil Tandon, and Viswanathan Mohan

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Natural History of Type 2 Diabetes in Indians: Time to Progression
Bidirectional Markov model: Normoglycemia ↔ IIFG → Diabetes Aim: **Annual transition probability** To understand what is the natural **iIFG** Normoglycemia **Annual transit ion probability** 7.50 (see) Average Time Spent,
Years (95% CI)
9.7 (8.4, 11.4) verage Time Spent,
Years (95% CI)
40.3 (34.6, 48.2) **Diabetes** 8.6% (7.3, 9.8) history of diabetes in Indians. Data: 22.8% (19.6, 26.6) 0.41% $(0.34, 0.49)$ Prospective data from adults **Annual transition prol Annual transit ion probability** Bidirectional Markov model: Normoglycemia ↔ IGT → Diabetes aged >20 years from the Chennai **Annual transit ion probability** site of the Centre for Cardiometabolic 5.1% (4.4, 5.7) **IGT** Normoglycemia Diabetes 5.1% (4.4, 5.7) Average Time Spent,
Years (95% CI) **Annual transit ion probability** verage Time Spent<mark>,</mark>
Years (95% CI) Risk Reduction in South Asia 13.9% (12.0, 15.9) 34.5 (29.5, 40.8) $6.1(5.3, 7.1)$ 15.4% (11.9.19.4) (CARRS) <u>15.44, 19.4, 19.4, 19.4</u> annual transition prob **Annual transit ion probability** ilFG, isolated impaired fasting glucose; IGT, impaired glucose tolerance

Continuous-time Markov models to estimate annual transition probabilities through multiple progressive states across two pathways (normoglycemia \leftrightarrow iIFG \rightarrow diabetes; normoglycemia \leftrightarrow IGT \rightarrow diabetes), allowing for regression from prediabetes to normal but assuming diabetes as an absorbing state.

Conclusion: Individuals reside in normoglycemic states for, on average, 35–40 years, thus providing time to investigate ways to prevent prediabetes.

ARTICLE HIGHLIGHTS

Why did we undertake this study?

We conducted this study to understand the natural history of diabetes in Indian populations.

What is the specific question(s) we wanted to answer?

We wanted to describe the natural history of diabetes in Indians.

What did we find?

Progression to diabetes among the Indian population is rapid once an individual has prediabetes (6 years for impaired glucose tolerance and 9.7 years for isolated impaired fasting glucose). Regression to normoglycemia from prediabetes is three times more likely than a progression from normoglycemia to prediabetes.

What are the implications of our findings?

In the Indian population, individuals reside in normoglycemic states for, on average, 35–40 years, which provides time to investigate ways to prevent prediabetes.

Natural History of Type 2 Diabetes in Indians: Time to Progression

K.M. Venkat Narayan, $1,2$ Dimple Kondal, $3,4$ Howard H. Chang, 1 Deepa Mohan,⁵ Unjali P. Gujral, $1/2$ Ranjit Mohan Anjana, 5 Lisa R. Staimez, 1,2 Shivani A. Patel, $1,2}$ Mohammed K. Ali, $1,2}$ Dorairaj Prabhakaran, 2,3,4 Nikhil Tandon, ⁶ and Viswanathan Mohan⁵

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OBJECTIVE

To describe the natural history of diabetes in Indians.

RESEARCH DESIGN AND METHODS

Data are from participants older than 20 years in the Centre for Cardiometabolic Risk Reduction in South Asia longitudinal study. Glycemic states were defined per American Diabetes Association criteria. Markov models were used to estimate annual transition probabilities and sojourn time through states.

RESULTS

Among 2,714 diabetes-free participants, 641 had isolated impaired fasting glucose (iIFG), and 341 had impaired glucose tolerance (IGT). The annual transition to diabetes for those with IGT was 13.9% (95% CI 12.0, 15.9) versus 8.6% (7.3, 9.8) for iIFG. In the normoglycemia \leftrightarrow iIFG \rightarrow diabetes model, mean sojourn time in normoglycemia was 40.3 (34.6, 48.2) years, and sojourn time in iIFG was 9.7 (8.4, 11.4) years. For the normoglycemia \leftrightarrow IGT \rightarrow diabetes model, mean sojourn time in normoglycemia was 34.5 (29.5, 40.8) years, and sojourn time in IGT was 6.1 (5.3, 7.1) years.

CONCLUSIONS

Individuals reside in normoglycemia for 35–40 years; however, progression from prediabetes to diabetes is rapid.

Indian people are at heightened risk of type 2 diabetes (1,2). However, it is unclear how Indians transition through the natural history of diabetes (e.g., from normoglycemia to prediabetes [impaired fasting glucose [IFG] or impaired glucose tolerance [IGT]), reversal to normoglycemia, and from prediabetes to diabetes), and how long people reside in each state (sojourn time).

RESEARCH DESIGN AND METHODS

We used data from the Chennai site of the longitudinal Centre for Cardiometabolic Risk Reduction in South Asia (CARRS) study (2010–2012) (3,4), up to the fourth follow-up (2016–2017) and implemented a continuous-time Markov model (5).

Inputs

Glycemic states (normoglycemia, IGT, iIFG, diabetes) were defined according to American Diabetes Association criteria, and distributions and characteristics of each glycemic state and the rates of progression from one stage to the next were obtained from the CARRS study (3,4) (details of definitions and cohort are provided in the [Supplementary Material\)](https://doi.org/10.2337/figshare.25213064) (6).

Among 5,961 participants with glucose measurements at baseline, 3,475 were free of diabetes. Of these, 2,714 participants had had a complete oral glucose tolerance test and

 1 Emory Global Diabetes Research Center, Woodruff Health Sciences Center and Emory University, Atlanta, GA

² Rollins School of Public Health, Emory University, Atlanta, GA

 3 Public Health Foundation of India, New Delhi, India

⁴ Centre for Chronic Disease Control, New Delhi, India

⁵Madras Diabetes Research Foundation & Dr. Mohan's Diabetes Specialties Centre, Chennai, India

⁶All India Institute of Medical Sciences, New Delhi, India

Corresponding author: K.M. Venkat Narayan, knaraya@emory.edu

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DIo, Oral Disposition Index; FPG, fasting plasma glucose; IQR, interquartile range; IR, insulin resistance. *Normoglycemia defined as FPG <5.6 mmol/L (100 mg/dL) and 2-hour postload glucose (2h-PG) <7.8 mmol/L (140 mg/dL) and no medication. †IFG defined as FPG between 5.6 and 6.9 mmol/L (100–125 mg/dL) and 2h-PG <7.8 mmol/L (140 mg/dL) and no medication. ‡IGT defined as FPG <7.0 mmol/L (126 mg/dL) and 2h-PG between 7.8–11.0 mmol/L (140–199 mg/dL) and no medication. §Significant P value for normoglycemia versus IGT. ¶Significant P value for normoglycemia versus iIFG. ⊥Significant P value for iIFG versus IGT. #Diabetes defined as FPG ≥7.0 mmol/L (126 mg/dL) or 2-h PG ≥11.1 mmol/L (200 mg/dL) or $HbA1c \ge 6.5%$ (48 mmol/mol) or on medication. **HOMA-IR = fasting insulin ($\mu U/L$) × fasting glucose (nmol/L)/22.5 or (I0($\mu U/mL$) × G0 (mmol/L)/22.5). $\text{+HOMA-B} = (20 \times \text{insulin})/(glucose - 3.5) \text{ or } (20 \times 10(\mu\text{U}/\text{ml})/G0 \text{ (mmol/L)} - 3.5). \text{~\#Dlo} = (\Delta I0 - 30/\Delta G0 - 30) \times (1/\text{fasting insulin}).$ §§Insulinogenic Index = (Δ I0-30/ Δ G0-30) where Δ I0-30 = insulin at 30 min minus fasting insulin; Δ G0-30 = glucose at 30 min minus FPG.

at least one follow-up assessment before 2017 to estimate changes in glycemia. The characteristics of the 2,714 participants included in the final analyses [\(Supplementary](https://doi.org/10.2337/figshare.25213064) [Fig. 3](https://doi.org/10.2337/figshare.25213064)) were similar to those of the overall sample [\(Supplementary Table 1](https://doi.org/10.2337/figshare.25213064)). Because only 187 people had combined IFG and IGT, they were classified as having IGT.

Multistate Analysis

We used multistate Markov models (7,8) to calculate annual transition probabilities for each state specified in [Supplementary](https://doi.org/10.2337/figshare.25213064) [Fig. 1.](https://doi.org/10.2337/figshare.25213064) The mean sojourn time was also calculated. For every participant, the outcome of interest was iIFG, IGT, or diabetes. Time was estimated from date of interview to the time of outcome

diagnosis, last date of visit, or death, whichever came first. We fitted two models: 1) normoglycemia to iIFG and regression to normoglycemia or progression to diabetes (i.e., normoglycemia \leftrightarrow iIFG \rightarrow diabetes); and 2) normoglycemia to IGT and regression to normoglycemia or progression to diabetes (normoglycemia ↔ $IGT \rightarrow$ diabetes). In the base case analysis, we assumed bidirectional change in states, allowing regression from prediabetes to normoglycemia ([Supplementary Fig. 1](https://doi.org/10.2337/figshare.25213064)). In a sensitivity analysis, we examined unidirectional progression, which assumes people cannot move back from prediabetes to normoglycemia [\(Supplementary Fig. 2\)](https://doi.org/10.2337/figshare.25213064). We performed stratified analyses by age (\leq 40 years vs. $>$ 40 years), sex, and BMI $(<$ 23 kg/m² vs. \geq 23 kg/m²) (9) to estimate the annual transition probabilities for each set of models. The data were analyzed using the msm package in R software, version 3.2.4, and Stata 16.0/MP.

RESULTS

As shown in Table 1, iIFG was nearly twice as frequent as IGT, and iIFG was more frequent in women (71.5%), and IGT in men (53.1%). Those with normoglycemia, followed by iIFG, and then IGT, had the lowest mean age (normoglycemia vs. iIFG vs. IGT: 37.6 vs. 40.9 vs. 43.7 years, respectively), weight (61.6 vs. 65.5 vs. 66.3 kg, respectively), BMI (25.0 vs. 27.2 vs. 27.0 kg/m^2 , respectively), waist circumference (81.6 vs. 85.7 vs. 88.1 cm, respectively), triglyceride levels (107.0 vs. 124.0 vs. 137.0 mg/dL, respectively), and total cholesterol level (180.4 vs. 187.4 vs. 189.9 mg/dL, respectively). The insulin levels at 0, 30, 120 min (iIFG: 7.7, 48.2, and 40.0 vs. IGT: 8.5, 53.1, and 59.6, respectively) and $HOMA-B$ (iIFG vs IGT: 69.3 vs. 91.5 mIU/mL/mmol/L, respectively) were lower in those with iIFG compared with participants with IGT, thus indicating iIFG is a more insulin-deficient state of prediabetes.

Markov Model Transition Probabilities

Normoglycemia ↔ iIFG → Diabetes Pathway Model

The estimated mean annual probability of remaining in normoglycemia and in iIFG were 92.1% (95% CI 91.2, 92.9) and 68.6% (65.2, 71.8), respectively (Table 2). The annual probability of conversion from

Table 2—Normoglycemia↔ iIFG →diabetes multistate Markov model annual probability of transition across states (overall and stratified by age, sex, and BMI)

normoglycemia to iIFG was 7.5% (6.7, 8.3) and from iIFG to normoglycemia was 22.8% (19.6, 26.6). Thus, it is about three times more likely for iIFG to revert to normoglycemia than it is for normoglycemia to progress to iIFG. The annual probability of conversion from iIFG to diabetes was 8.6% (5.7, 11.9). The estimated mean

sojourn times were 40.3 (34.6, 48.2) years and 9.7 (8.4, 11.4) years in normoglycemia and iIFG, respectively (Fig. 1).

The annual probability of transition for normoglycemia to iIFG was greater for those age >40 (9.9% [95% CI 8.1, 11.9]) years compared with those age \leq 40 (6.4% [5.5, 7.4]) years (Table 2). The

 $\mathsf{Normoglycemia}\longleftrightarrow \mathsf{IGT}\longrightarrow \mathsf{Diabetes}$

Figure 1—Multistate Markov models. The annual probability of remaining in the same state or transitioning to the next state.

annual probability of transition from iIFG to diabetes was higher in female participants than male participants (female: 7.9% [6.8, 9.0]; male: 6.8% [5.6, 8.1]). The annual probability of transition from normoglycemia to iIFG was greater for those with BMI \geq 23 kg/m² (8.9% [7.7, 10.2]) compared with BMI $<$ 23 kg/m² (5.5% [4.1, 7.1]).

Normoglycemia \leftrightarrow IGT \rightarrow Diabetes Pathway Model

The estimated annual probability of remaining in normoglycemia and IGT were 94.5% (95% CI 93.8, 95.2) and 70.1% (67.1, 73.8), respectively (Table 3). The annual probability of transition from normoglycemia to IGT was 5.1% (4.4, 5.7) and from IGT to normoglycemia was 15.4% (11.9, 19.4). Thus, it is about three times more likely for IGT to revert to normoglycemia than it is for normoglycemia to progress to IGT. The annual probability of transition from IGT to diabetes was 13.9% (12.0, 15.9). The estimated mean sojourn times were 34.5 (29.5, 40.8) years and 6.1 (5.3, 7.1) years for normoglycemia and IGT, respectively (Fig. 1).

The annual probability of transition from normoglycemia to IGT was higher among those aged >40 years (7.6% [95% CI 6.2, 9.1]) as compared with those aged \leq 40 years (3.9% [3.3, 4.7]) (Table 3). The annual transition probability from normoglycemia to IGT was higher in men (6.8%

[5.7, 8.0]) than women (3.9% [3.1, 4.6]). The annual probability of transition from normoglycemia to IGT was higher among those with BMI \geq 23 kg/m² (6.2% [5.2, 7.1]) as compared with those with BMI $<$ 23 kg/m² (3.5% [2.6, 4.5]).

Sensitivity Analysis (Unidirectional Models)

In a sensitivity analysis, the probabilities of transition based on unidirectional models (i.e., not allowing regression from prediabetes to normoglycemia) were similar to those of base case bidirectional models ([Supplementary Tables 1](https://doi.org/10.2337/figshare.25213064)–[4](https://doi.org/10.2337/figshare.25213064)).

CONCLUSIONS

In an urban Indian population aged \geq 20 years, progression to diabetes is rapid once an individual has prediabetes. On average, people reside 35–40 years in normoglycemic states, and only 9.7 years in iIFG or 6.1 years in IGT before advancing to diabetes (assuming bidirectional transition from normoglycemia to prediabetes). Prediabetes represents a fragile state, with a nearly three times likelihood of either iIFG or IGT reverting to normoglycemia than normoglycemia progressing to prediabetes. However, at the onset of prediabetes, and after accounting for reversibility, the rate of progression from prediabetes to diabetes was rapid, and those with IGT progressed to diabetes faster (13.9% per annum) than those with iIFG (8.6% per annum).

Similar to our findings, several studies have reported a high incidence of diabetes and prediabetes in Indians (1,10–12). In addition, we found a higher rate of conversion from normoglycemia to iIFG than to IGT, suggesting reduced insulin secretion (lower HOMA of β -cell function $[HOMA- β]$) as an early defect (13). However, among those with prediabetes, those with IGT had a more rapid conversion to diabetes, suggesting poorer insulin sensitivity as a key factor at later stages in those already susceptible, and iIFG and IGT being potentially different phenotypes with differences in pathophysiology (13–15). Similar to previous reports (16), we also found that iIFG is the more frequent (almost two-thirds) prediabetes manifestation in Indians, iIFG is the common phenotype in women, and IGT the more common phenotype in men. Although lifestyle interventions are effective in reducing the incidence of diabetes among adults with IGT (17,18), these interventions seem not effective in individuals with iIFG (19).

The strengths of our study include data from a representative sample, high response and retention rates, multiple time points of follow-up, and objective measures of glycemia derived from three-step oral glucose tolerance tests. To our knowledge, no previous study of diabetes in Indians has estimated time to progression or time spent

Table 3—Normoglycemia ↔ IGT→ diabetes multistate Markov model annual probability of transition across states (overall and stratified by age, sex, and BMI)

in each glycemic state. Given the fragility of the prediabetes state, we conservatively assumed bidirectional transition from normoglycemia to iIFG or IGT (in separate models), but we also performed sensitivity analyses to explore the effect of unidirectional transitions and other assumptions.

The results for transition probabilities across states were robust regardless of assumption of bidirectional or unidirectional progression, but the estimates of time in each state were substantially longer under assumption of bidirectionality. Last, we performed stratified analyses by

age, sex, and BMI. Our study has some limitations, including that data are from one city; however, diabetes incidence across urban India is quite similar, and the sample reflects the age and sociodemographic distribution of populations of cities in India.

In conclusion, we found a high rate of conversion from normoglycemia to IFG or IGT in Indians, and once an individual has prediabetes, the conversion to diabetes is rapid. On the hopeful side, people at risk for diabetes reside in normoglycemic states for an average of 35–40 years, and those transitioning through the more frequent iIFG stage reside there for an average of 9.7 years (as opposed to 6.1 years in IGT). These findings suggest the need to test interventions to prevent the occurrence of prediabetes.

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References

1. Chiu M, Austin PC, Manuel DG, Shah BR, Tu JV. Deriving ethnic-specific BMI cutoff points for assessing diabetes risk. Diabetes Care 2011;34: 1741–1748

2. Oza-Frank R, Narayan KMV. Overweight and diabetes prevalence among US immigrants. Am J Public Health 2010;100:661–668

3. Nair M, Ali MK, Ajay VS, et al. CARRS surveillance study: design and methods to assess burdens from multiple perspectives. BMC Public Health 2012;12:701

4. Kondal D, Patel SA, Ali MK, et al. Cohort profile: the Center for cArdiometabolic Risk Reduction in South Asia (CARRS). Int J Epidemiol 2022;51:e358–e371

5. Marshall G, Jones RH. Multi-state models and diabetic retinopathy. Stat Med 1995;14:1975–1983 6. American Diabetes Association. 2. Classification and diagnosis of diabetes: Standards of Medical Care in Diabetes-2018. Diabetes Care 2018;41 (Suppl. 1):S13–S27

7. Kalbfleisch JD, Lawless JF. The analysis of panel data under a Markov assumption. J Am Stat Assoc 1985;80:863–871

8. Kay R. A Markov model for analysing cancer markers and disease states in survival studies. Biometrics 1986;42:855–865

9. World Health Organization Expert Consultation. Appropriate body-mass index for Asian populations and its implications for policy and intervention strategies. Lancet 2004;363:157–163

10. Mohan V, Deepa M, Anjana R, Lanthorn H, Deepa R. Incidence of diabetes and pre-diabetes in a selected urban South Indian population (CUPS-19). J Assoc Physicians India 2008;56:152– 157

11. Gujral UP, Narayan KMV, Kandula NR, Liu K, Kanaya AM. Incidence of diabetes and prediabetes and predictors of glycemic change among South Asians in the USA: the MASALA study. BMJ Open Diabetes Res Care 2020;8:e001063

12. Fazli GS, Moineddin R, Bierman AS, Booth GL. Ethnic differences in prediabetes incidence among immigrants to Canada: a populationbased cohort study. BMC Med 2019;17:100

13. Abdul-Ghani MA, Tripathy D, DeFronzo RA. Contributions of b-cell dysfunction and insulin resistance to the pathogenesis of impaired glucose tolerance and impaired fasting glucose. Diabetes Care 2006;29:1130–1139

14. Faerch K, Borch-Johnsen K, Holst JJ, Vaag A. Pathophysiology and aetiology of impaired fasting glycaemia and impaired glucose tolerance: does it matter for prevention and treatment of type 2 diabetes? Diabetologia 2009;52:1714–1723

15. Nathan DM, Davidson MB, DeFronzo RA, et al.; American Diabetes Association. Impaired fasting glucose and impaired glucose tolerance: implications for care. Diabetes Care 2007;30:753– 759

16. Anjana RM, Unnikrishnan R, Deepa M, et al.; ICMR-INDIAB Collaborative Study Group. Metabolic non-communicable disease health report of India: the ICMR-INDIAB national cross-sectional study (ICMR-INDIAB-17). Lancet Diabetes Endocrinol 2023;11:474–489

17. Weber MB, Ranjani H, Staimez LR, et al. The stepwise approach to diabetes prevention: results from the D-CLIP randomized controlled trial. Diabetes Care 2016;39:1760–1767

18. Ramachandran A, Snehalatha C, Mary S, Mukesh B, Bhaskar AD; Indian Diabetes Prevention Programme (IDPP). The Indian Diabetes Prevention Programme shows that lifestyle modification and metformin prevent type 2 diabetes in Asian Indian subjects with impaired glucose tolerance (IDPP-1). Diabetologia 2006;49:289–297

19. Sathish T, Khunti K, Narayan KMV, et al. Effect of conventional lifestyle interventions on type 2 diabetes incidence by glucose-defined prediabetes phenotype: an individual participant data meta-analysis of randomized controlled trials. Diabetes Care 2023;46:1903–1907