Alterations in Biomarkers Related to Glycemia, Lipid Metabolism, and Inflammation up to 20 Years Before Diagnosis of Type 1 Diabetes in Adults: Findings From the AMORIS Cohort

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# OBJECTIVE

Type 1 diabetes is described to have an acute onset, but autoantibodies can appear several years preceding diagnosis. This suggests a long preclinical phase, which may also include metabolic parameters. Here we assessed whether elevations in glycemic, lipid, and other metabolic biomarkers were associated with future type 1 diabetes risk in adults.

# **RESEARCH DESIGN AND METHODS**

We studied 591,239 individuals from the Swedish AMORIS cohort followed from 1985–1996 to 2012. Through linkage to national patient, diabetes, and prescription registers, we identified incident type 1 diabetes. Using Cox regression models, we estimated hazard ratios for biomarkers at baseline and incident type 1 diabetes. We additionally assessed trajectories of biomarkers during the 25 years before type 1 diabetes diagnosis in a nested case-control design.

#### RESULTS

We identified 1,122 type 1 diabetes cases during follow-up (average age of patient at diagnosis: 53.3 years). The biomarkers glucose, fructosamine, triglycerides, the ratio of apolipoprotein (apo)B to apoA-I, uric acid, alkaline phosphatase, and BMI were positively associated with type 1 diabetes risk. Higher apoA-I was associated with lower type 1 diabetes incidence. Already 15 years before diagnosis, type 1 diabetes cases had higher mean glucose, fructosamine, triglycerides, and uric acid levels compared with control subjects.

# CONCLUSIONS

Alterations in biomarker levels related to glycemia, lipid metabolism, and inflammation are associated with clinically diagnosed type 1 diabetes risk, and these may be elevated many years preceding diagnosis.

Type 1 diabetes accounts for 99% of all diabetes in children and adolescents but can develop at any age. Notably, a recent study based on data from UK biobank

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© 2022 by the American Diabetes Association. Readers may use this article as long as the work is properly cited, the use is educational and not for profit, and the work is not altered. More information is available at https://www. diabetesjournals.org/journals/pages/license. estimated that 42% of all cases are diagnosed among patients between ages 30 and 60 years (1). Research regarding the pathogenesis of type 1 diabetes with adult onset is limited, but it has been shown that adult patients often present with only one autoantibody, whereas children tend to be positive for multiple autoantibodies. Patients with onset in adulthood are also described to have less severe decline in insulin secretion and a longer asymptomatic period compared with children and adolescents (2,3). Furthermore, it has been proposed that insulin resistance contributes to the development of autoimmune diabetes in adults, especially for the subgroup with latent autoimmune diabetes in adults (4,5).

Type 2 diabetes is known to develop over a long period of time through a prediabetes stage with elevated blood glucose levels. In line with this, we previously found that individuals who developed type 2 diabetes differed from those who did not in terms of several glycemiaand lipid-related biomarkers up to 25 years prior to diagnosis (6). Type 1 diabetes, on the other hand, is described to have an acute onset, and symptoms can develop quickly over a few weeks or even days. However, the first autoantibody indicating an autoimmune reaction typically occurs several years prior to manifestation of type 1 diabetes, suggesting a long preclinical phase (7). Similarly, it has been shown that adults who develop diabetes can have autoantibodies 10 years prior to diagnosis (8-10).

It is possible that other metabolic processes are operating in parallel, or perhaps even prior, to appearance of autoantibodies. Metabolic alterations, including lipid species and amino acids, have indeed been found in asymptomatic children who later develop type 1 diabetes (11,12). Corresponding studies in adults are, however, lacking. The aim of the current study was to describe alterations in biomarkers reflecting glycemia and lipid metabolism, as well as inflammatory markers, and the risk of adult-onset type 1 diabetes and to assess their temporal relationship with the development of type 1 diabetes. This knowledge could increase our understanding of metabolic changes that occur during the development of type 1 diabetes in adults.

# RESEARCH DESIGN AND METHODS

#### Study Population

The Swedish AMORIS (Apolipoproteinrelated MOrtality RISk) cohort has previously been described (13). In brief, this longitudinal cohort consists of 812,073 individuals who predominantly lived in Stockholm, Sweden, and who underwent health examinations through either occupational health checkups or outpatient visits between 1985 and 1996 (baseline period). The AMORIS cohort included  $\sim$ 35% of people living in the greater Stockholm area at the inclusion time, and sociodemographic characteristics were comparable with the general working population in Stockholm (13), whereas the mortality rate was lower than for the general population as a whole (6). Eligible for the current study were all individuals with at least one biomarker measurement of interest in the baseline period (n = 623, 152). We excluded all individuals with a documented diagnosis of diabetes at baseline in any of the registers described below, or suspicion of diabetes at baseline (glucose >7.8 mmol/L or fructosamine >2.3 mmol/L or HbA<sub>1c</sub> >6.4%) (n = 31,913), resulting in an analytical sample of 591,239 individuals. Using the unique Swedish personal identity number, study participants were followed until 31 December 2011 by linkage to the Swedish Cause of Death Register, the National Patient Register (PR), the National Diabetes Registry (NDR), and the Prescribed Drug Register (PDR). Country of birth and occupational status were ascertained from the Swedish Censuses from 1970 to 1990-whichever was closest to the first measurement. Gainfully employed individuals were classified as manual and lower nonmanual workers or higher nonmanual workers, representing low and high socioeconomic status, respectively.

This study complied with the Declaration of Helsinki and was approved by the regional ethics committee at Karolinska Institutet, Stockholm, Sweden (registration no. 2010/1:7).

#### **Biomarkers**

All laboratory analyses were performed on fresh blood samples at the accredited CALAB laboratory (Stockholm, Sweden) using standardized, automated, and welldocumented methodologies (6,13,14). In the current study, we included information on date of health examination (blood sampling), fasting status, and blood levels regarding glucose, fructosamine, triglycerides, total cholesterol, apolipoprotein (apo)A-I, and apoB and the apoB-toapoA-I ratio. In addition, we included information from haptoglobin, uric acid, albumin, C-reactive protein, leukocyte count, urea,  $\gamma$ -glutamyl transferase, and alkaline phosphatase. The majority of these biomarkers were part of a standard analysis package, which represented regular health screenings (13), and have previously been studied in relation to type 2 diabetes (6). Furthermore, we obtained information on BMI from health examinations for a proportion of the study population (25%). Information on HbA<sub>1c</sub> was only available for 13,811 (2.3%) of the cohort population and could not be used in the current study. Fructosamine was available for most study participants, has been shown to correlate well with HbA<sub>1c</sub> in subjects with type 1 diabetes, and does not require fasting samples (15).

# Identification of Incident Type 1 Diabetes

For identification of incident type 1 diabetes cases, we used information from NDR (16), PR (17), and PDR (18). In total, 2,534 individuals had a record of type 1 diabetes in NDR or PR during follow-up. Individuals with a type 2 diabetes record at any time, in either register, were not included. To additionally exclude any potentially misclassified type 2 diabetes cases, we specified that patients with type 1 diabetes have an age of diagnosis of <30 years, a record of insulin prescription, or a combination of these (n =1,122) (Supplementary Fig. 1). Further details can be found in Supplementary Material.

#### Statistical Analyses

Baseline characteristics for the full study population and for case and control subjects in the nested case-control setup are displayed as mean values and SD for continuous variables and as proportions for categorical variables.

In the prospective cohort analysis, we estimated hazard ratios (HR) with 95% CI for type 1 diabetes in relation to biomarkers using Cox proportional hazards regression models. Age was the underlying time scale to obtain flexible, nonparametric adjustment for age (19), and person-years were accumulated from age at baseline until age at diabetes diagnosis, death, or end of follow-up, 31 December 2011. These models were adjusted for sex, age (timescale), calendar time, fasting status (overnight fasting vs. nonfasting), socioeconomic status, and country of birth. We investigated the assumption of proportional hazards by applying the  $\chi^2$  test based on Schoenfeld residuals. If the assumption was violated regarding a confounder, we applied Cox models with interaction between the confounder and age in the model. To estimate and evaluate nonproportional effect and dependence on age, we additionally performed analysis stratified by median and quartiles of age at biomarker measurement.

In these analyses, we used the standardized, continuous variable values to allow that a 1-unit change in HR corresponds to the change of 1 SD in the standardized variable. We additionally dichotomized variables to estimate the HR associated with high compared with low levels of the biomarkers; cutoffs for the high levels were set according to published guidelines or previous publications on the AMORIS cohort (6,15,20). We performed a logarithmic transformation of the continuous variables triglycerides and BMI to approximate normally distributed values.

In addition, we conducted a nested case-control analysis within the cohort for all biomarkers. In this analysis, all incident type 1 diabetes cases were included, as well as 25 randomly selected control subjects per cases based on incidence density sampling through calendar year (21) and individual matching by sex and age-group (5-year interval). We defined the index date based on the date of type 1 diabetes diagnosis for case subjects and the date of selection for the control subjects. We performed conditional logistic regression and estimated associations between the biomarkers and type 1 diabetes with odds ratios (ORs) and 95% CI. All models were adjusted as described above.

To compare biomarker values between case and control subjects at given time points, we created population trajectories using mean biomarker values for each year, up to 25 years prior to diagnosis. These trajectories were based on a single measurement per individual and, hence, represent not alterations of biomarker levels for an individual over time but, rather, the mean value of case and control subjects, respectively; the methodology has previously been described (22). For each biomarker, we estimated weighted means and 95% confidence bands based on matching criteria as described above, for both case and control subjects and for each year prior to type 1 diabetes diagnosis.

All analyses were performed in R Studio with R, version 3.6.2; for details on packages used, see Supplementary Material.

#### Sensitivity Analysis

To exclude any potentially misclassification of subjects with type 2 diabetes, we repeated the main analyses including only individuals age <30 years and <40 years, respectively, at end of follow-up; 8.9% and 22.6% of type 1 diabetes case subjects had an age of onset  $\leq$ 30 and  $\leq$ 40 years, respectively. Furthermore, we performed additional analyses excluding individuals prescribed any oral glucoselowering therapy, as well as requiring insulin prescription for all case subjects within 1 year of diagnosis.

We additionally performed stratified analyses by running separate models for women and men, and younger and older individuals (with median age or age quartiles of age at blood sampling as cutoffs), to assess the potential variation by sex and age. For details on analysis both limited to fasting samples and adjusted for BMI, see Supplementary Material.

#### RESULTS

### **Cohort Analysis**

Overall, average age of cohort patients was 44 years at blood sampling. The cohort consisted of slightly more males than females, and the majority of individuals were nonmanual workers. During the follow-up period, 1,122 individuals were diagnosed with type 1 diabetes, with an incidence rate of 10.2 per 100,000 person-years.

Average age of type 1 diabetes case subjects was 53 years at disease onset (age ranging from 8 to 94 years), and 8.4% of all case subjects had a recorded disease onset of <30 years of age. Furthermore, there was a higher proportion of men and a higher proportion of nonmanual workers among type 1 diabetes cases compared with individuals without type 1 diabetes (Table 1). The average time from blood sampling to diagnosis was 11 years.

In using biomarkers as continuous variables, the glycemia-related markers glucose and fructosamine, as well as the markers of lipid metabolism triglycerides, apoB, and apoB-to-apoA-I ratio, were positively associated with risk of developing type 1 diabetes (Table 2). In addition, we found a positive association with type 1 diabetes risk for BMI, uric acid, haptoglobin, C-reactive protein, and leukocyte count, as well as the biomarkers y-glutamyl transferase and alkaline phosphatase. For apoA-I, on the other hand, we found a negative association with type 1 diabetes risk (Table 2). Overall, we found somewhat stronger associations in older individuals (Supplementary Table 1). Among these biomarkers, the associations for glucose, fructosamine, triglycerides, apoA-I, uric acid, and alkaline phosphatase were independent of age and sex (Table 3 and Supplementary Table 1). Sensitivity analysis using only biomarker values obtained from blood samples drawn after overnight fasting did not change the results (Supplementary Table 1).

To exclude potential misclassification of type 2 diabetes, we repeated the analyses including only individuals with age of onset <30 years. For this subgroup, the associations with type 1 diabetes risk for the biomarkers glucose, fructosamine, apoB-to-apoA-I ratio, and alkaline phosphatase remained (Table 3). Restricting the analysis to case subjects without any oral glucose-lowering therapy yielded similar associations for the majority of biomarkers (Supplementary Table 1). Finally, in repeating the analysis including only case subjects with insulin prescribed within 1 year of diagnosis (only possible for 28.3% of cases, diagnosed from 2005 onward when the PDR was started), the associations with glucose, triglycerides, and uric acid remained significant (Supplementary Table 1).

We obtained similar results when comparing high with low levels of biomarkers (Supplementary Table 2). Similar to biomarkers used as continuous variables, the associations for high levels of glucose, fructosamine, triglycerides, apoA-1, uric acid,  $\gamma$ -glutamyl transferase, and alkaline phosphatase were independent of age and sex of the individuals, and overall, we found

	Full stu	dy population	Type 1 dia	lbetes case subjects	Con	trol subjects	Difference case vs.
Variable	N	Mean (SD) or %	N	Mean (SD) or %	N	Mean (SD) or %	control subjects (P)
Age (years)	524,102	43.89 (14.39)	987 <sup>a</sup>	53.28 (15.67) <sup>b</sup>	24,675	53.22 (15.64) <sup>b</sup>	su
Follow-up (years)	524,102	18.87 (5.87)	987	11.26 (6.11)	24,675	11.03 (6.24)	ns
Sex (females)	247,828	48.25	369	37.39	9,225	37.39	ns
Country of birth (Sweden)	505,890	86.02	851	86.22	21,440	86.89	SU
Occupation Manual workers Nonmanual workers Unclassified/missing	124,871 279,572 117,804	48.85 35.77 14.25 <sup>c</sup>	278 512 194	28.17 51.87 19.66	6,178 13,569 4,848	25.08 54.99 19.65	0.0290 ns ns
BMI (kg/m <sup>2</sup> )	132,207	23.72 (3.72)	228	24.96 (4.09)	5,700	24.17 (3.56)	0.0068
Glucose (mmol/L)	524,102	4.83 (0.68)	987	5.24 (0.98)	24,675	4.80 (0.64)	< 0.0001
Fructosamine (mmol/L)	431,578	2.06 (0.19)	812	2.13 (0.20)	20,300	2.07 (0.19)	< 0.0001
Triglycerides (mmol/L)	542,834	1.08 (1.74)	1,040	1.66 (1.46)	26,000	1.28 (0.89)	< 0.0001
Total cholesterol (mmol/L)	544,827	5.53 (1.16)	1,040	5.55 (1.19)	26,000	5.51 (1.15)	ns
ApoB-to-apoA-I ratio	160,444	0.91 (0.31)	305	1.01 (0.37)	7,625	0.92 (0.31)	< 0.0001
ApoA-I (g/L)	184,163	1.43 (0.24)	348	1.34 (0.23)	8,700	1.42 (0.23)	< 0.0001
ApoB (g/L)	171,471	1.25 (0.36)	332	1.30 (0.37)	8,300	1.26 (0.35)	SU
Haptoglobin (g/L)	411,156	1.07 (0.34)	822	1.10 (0.35)	20,500	1.05 (0.32)	< 0.0001
Uric acid (µmol/L)	503,933	288.11 (73.45)	963	310.48 (79.25)	24,075	291.37 (69.88)	< 0.0001
Albumin (g/L)	487,892	43.25 (2.89)	930	43.36 (2.87)	23,250	43.47 (2.81)	ns
Urea (mmol/L)	383,174	5.13 (1.44)	744	5.21 (1.31)	18,600	5.16 (1.28)	ns
$\gamma$ -glutamyl transferase (µkat/L)	527,829	0.44 (0.69)	1,001	0.64 (1.16)	25,025	0.42 (0.57)	< 0.0001
Alkaline phosphatase ( $\mu$ kat/L)	494,381	2.73 (1.22)	951	3.05 (1.97)	23,775	2.73 (1.26)	< 0.0001
C-reactive protein (mg/L)	343,978	5.96 (16.91)	582	7.96 (30.56)	14,550	5.69 (15.03)	ns
Leukocytes (10 <sup>9</sup> cells/L)	188,057	6.58 (2.38)	323	7.03 (2.47)	8,075	6.49 (1.96)	0.0001
ns, not significant at $\alpha = 0.05$ . <sup>a</sup> Age is biomarkers, e.g., 987 type 1 diabetes	at recorded diagno case subjects had	ssis of type 1 diabetes or so I glucose levels assessed at	election as control s baseline. <sup>c</sup> Category	subjects. <sup>b</sup> Due to the availabl "unclassified" also includes ii	e measurements, th Idividuals age ≤16	ie number of case subjects years and pensioners.	studied varied between

Table 1-Participant characteristics at the baseline investigation

			HR (9	5% CI)
Biomarker	Person-years	Case subjects, n (%)	Model 1	Model 2
Biomarkers related to glycemia				
Glucose (mmol/L)	9,887,018	987 (0.19)	1.66 (1.59–1.73)	1.72 (1.65–1.80)
Fructosamine (mmol/L)	7,956,801	813 (0.19)	1.50 (1.40–1.61)	1.45 (1.35–1.56)
Biomarkers of lipid metabolism				
Triglycerides (mmol/L)	10,196,681	1,040 (0.19)	1.49 (1.41-1.58)	1.49 (1.40–1.58)
Total cholesterol (mmol/L)	10,227,990	1,040 (0.19)	1.05 (0.98-1.12)	1.06 (0.99–1.13)
ApoB-to-apoA-I ratio	2,810,460	305 (0.19)	1.36 (1.25–1.49)	1.35 (1.23–1.48)
ApoA-I (g/L)	3,306,276	348 (0.19)	0.65 (0.58–0.73)	0.67 (0.59–0.76)
ApoB (g/L)	3,035,459	332 (0.19)	1.20 (1.08–1.34)	1.17 (1.05–1.31)
Biomarkers related to inflammation				
Haptoglobin (g/L)	8,145,669	823 (0.20)	1.16 (1.09-1.24)	1.16 (1.09–1.24)
Uric acid (μmol/L)	9,537,617	964 (0.19)	1.43 (1.35–1.52)	1.46 (1.36–1.57)
Albumin (g/L)	9,202,623	930 (0.19)	1.02 (0.95-1.09)	0.97 (0.90–1.04)
C-reactive protein (mg/L)	5,964,550	582 (0.17)	1.05 (1.01-1.09)	1.04 (1.01-1.08)
Leukocytes (10 <sup>9</sup> cells/L)	3,138,493	323 (0.17)	1.08 (1.05–1.12)	1.08 (1.05–1.12)
Other biomarkers				
Urea (mmol/L)	7,276,004	745 (0.19)	1.10 (1.03-1.18)	1.05 (0.97-1.14)
γ-glutamyl transferase (μkat/L)	9,844,951	1,001 (0.19)	1.09 (1.07-1.11)	1.09 (1.07-1.11)
Alkaline phosphatase (µkat/L)	9,238,047	952 (0.19)	1.16 (1.14–1.19)	1.16 (1.13–1.19)
BMI (kg/m <sup>2</sup> )	2,387,455	230 (0.17)	1.31 (1.15-1.48)	1.28 (1.12–1.46)

# Table 2–HRs and 95% CIs of type 1 diabetes for continuous biomarkers

HRs were estimated using adjusted Cox models with attained age as the underlying timescale and date of birth as the time origin. We used standardized, continuous variable values to allow that a 1-unit change in HR corresponds with the change of 1 SD in the standardized variable. Model 1 was not adjusted for any additional confounders; model 2 was adjusted for sex, age at first blood sampling, fasting status (overnight fasting vs. nonfasting), occupation, and country of birth.

more pronounced associations in individuals with older age (Supplementary Table 3 and Supplementary Table 4).

# Nested Case-Control Study Analysis

The case-control analyses yielded biomarker-type 1 diabetes associations similar to those obtained with the full cohort (Supplementary Table 5). By plotting mean biomarker levels by time to index year, we found that type 1 diabetes case subjects had consistently higher levels of glucose, fructosamine, triglycerides, and uric acid up to 15 years before diagnosis compared with control subjects (Fig. 1). The mean levels of other biomarkers did not differ significantly between case and control subjects prior to diagnosis (Supplementary Fig. 2); spikes occurring in some of the plotted biomarker levels are likely due to small numbers of case subjects at certain time points.

Glucose levels were higher in type 1 diabetes case subjects, with an average of 5.42 mmol/L (95% CI 5.16–5.68) at 15 years before diagnosis and an average difference of 0.57 mmol/L in comparison with control subjects. Two years prior to diagnosis, the average glucose levels reached 5.67 mmol/L (95% CI 4.94–6.39), indicating prediabetes. Fructosamine levels were elevated in type 1 diabetes cases 15 years prior to diagnosis (average 2.17 mmol/L [95% CI 2.11–2.22]), and the difference between case and control subjects increased inversely to time of diagnosis (average difference 5 years prior to diagnosis 0.08 mmol/L [95% CI 0.03–0.14]).

Triglyceride levels were elevated in type 1 diabetes case subjects up to two decades prior to diagnosis, with average levels of 2.09 mmol/L (95% Cl 1.70–2.49) and an average difference of 0.81 mmol/L (95% Cl 0.46–1.15) compared with control subjects 16 years prior to diagnosis. Similarly, levels of uric acid were elevated in type 1 diabetes cases 15 years prior to diagnosis compared with control subjects (average difference 42.34 µmol/L [95% Cl 18.83–65.84]), but these differences between case and control subjects decreased again and leveled off in the years preceding type 1 diabetes diagnosis.

To account for the influence of BMI on biomarkers of lipid metabolism, we repeated the analysis for the biomarkers triglyceride, apoA-I, and the apoB-toapoA-I ratio with additional adjustment for BMI; the HRs were comparable, but the CIs were wider (Supplementary Table 5*B*).

# CONCLUSIONS

In our population-based, longitudinal study, we found that biomarkers related to glycemia, lipid metabolism, and inflammation were associated with future type 1 diabetes risk. Furthermore, the levels of glucose, fructosamine, triglycerides, and uric acid were elevated in case subjects up to 15–20 years prior to diagnosis compared with randomly selected population control subjects. These results indicate that metabolic conditions influencing the development of type 1 diabetes in adults may operate decades before disease manifestation.

We found increased incidence of type 1 diabetes associated with higher levels of triglycerides, apoB, and especially elevated apoB-to-apoA-I ratio. Studies in childhood type 1 diabetes also found dysregulation of lipid metabolism prior to diagnosis (11,12,23); consistent with our findings, data from the German BABYDIAB study showed increased levels of triglycerides and other lipids in autoantibody-positive children

		Men		Women	Age of onse	tt ≤30 years	Age of onse	et ≤40 years
Biomarker	N	HR (95% CI)	N	HR (95% CI)	N case subjects	HR (95% CI)	N case subjects	HR (95% CI)
Biomarkers related to glycemia Glucose (mmol/L) Erurtosamine (mmol/I)	276,274 276 584	1.65 (1.56–1.75) 1.46 (1.33–1.60)	247,828 204 967	1.85 (1.73–1.98) 1.44 (1.78–1.62)	83	1.67 (1.40–1.99) 1.56 (1.21–2.00)	222 30	1.54 (1.37–1.74) 1 49 (1 27–1 74)
	100,044		100/101	(70.1 07.1)	4	(00:3 13:1) 00:1	2	(+
Biomarkers of lipid metabolism Triglycerides (mmol/L)	286,762	1.46 (1.35–1.57)	256,072	1.53 (1.37–1.71)	86	1.04 (0.82–1.31)	228	1.12 (0.98–1.29)
Total cholesterol (mmol/L)	287,889	1.04 (0.95–1.13)	256,938	1.06 (0.94–1.19)	86	1.05 (0.79–1.41)	228	1.02 (0.86–1.21)
ApoB-to-apoA-I ratio	89,616	1.34 (1.19–1.51)	70,828	1.34 (1.14–1.58)	13	1.49 (1.05–2.12)	47	1.15 (0.85–1.56)
ApoA-I (g/L)	103,016	0.62 (0.53–0.73)	81,147	0.74 (0.61–0.90)	16	0.59 (0.33–1.05)	54	0.72 (0.53–0.99)
ApoB (g/L)	95,262	1.13 (0.98–1.31)	76,209	1.19 (0.99–1.43)	18	1.45 (0.84–2.51)	54	0.98 (0.69–1.39)
Biomarkers related to inflammation								
Haptoglobin (g/L)	217,344	1.11 (1.03–1.21)	193,812	1.25 (1.13–1.39)	65	1.01 (0.79–1.30)	173	1.10 (0.95–1.28)
Uric acid (p.mol/L)	264,102	1.43 (1.31–1.56)	239,831	1.50 (1.33–1.69)	82	1.06 (0.79–1.42)	213	1.12 (0.94–1.33)
Albumin (g/L)	256,437	0.91 (0.83–1.00)	241,455	1.03 (0.92–1.16)	81	0.82 (0.66–1.02)	203	0.91 (0.79–1.05)
C-reactive protein (mg/L)	176,509	1.05 (1.01–1.09)	167,469	1.02 (0.94–1.12)	51	0.96 (0.64–1.43)	131	0.92 (0.64–1.31)
Leukocytes (10 <sup>9</sup> cells/L)	80,434	1.08 (1.03–1.13)	107,623	1.08 (1.03–1.14)	22	1.02 (0.63–1.65)	55	1.11 (0.83–1.49)
Other								
Urea (mmol/L)	198,705	1.01 (0.91–1.12)	184,443	1.11 (0.98–1.27)	53	1.07 (0.85–1.34)	144	0.99 (0.80–1.23)
$\gamma$ -glutamyl transferase ( $\mu$ kat/L)	276,793	1.09 (1.07–1.11)	251,036	1.10 (1.07–1.13)	86	0.97 (0.55–1.69)	226	1.06 (0.94–1.19)
Alkaline phosphatase (µkat/L)	257,418	1.15 (1.11–1.19)	236,962	1.18 (1.13–1.23)	78	1.13 (1.00–1.27)	208	1.11 (1.01–1.21)
BMI (kg/m <sup>2</sup> )	58,584	1.41 (1.17–1.70)	73,623	1.12 (0.92–1.37)	15	0.99 (0.56–1.74)	48	1.19 (0.89–1.60)
Restricted analysis by sex and restrictutimescale and date of birth as the time	ed analysis by a e origin. Model	age of onset; values de s were adjusted for sex	ttermined in ov ., age at first bl	vernight fasting sample ood sampling, fasting s	s. HRs were estimated tatus (overnight fasting	using adjusted Cox mo g vs. nonfasting), occup	odels with attained ag ation, and country of	e as the underlying birth.

(11), whereas downregulated lipid levels, including triglycerides, were found in children progressing to type 1 diabetes in the Finnish Type 1 Diabetes Prevention and Prediction Study (DIPP) (12,23). Notably, these differences were primarily detected in early infancy prior to the appearance of autoantibodies (11,12). High triglycerides and apoB-toapoA-I ratio and low apoA-I levels have been described to be associated with insulin resistance and adiposity (24,25), and our findings could thus reflect that elevated lipid levels contribute to development of type 1 diabetes in adults via adverse effects on insulin sensitivity. In addition, alkaline phosphatase and uric acid levels were elevated in individuals who later developed type 1 diabetes, and these biomarkers are also associated with reduced insulin sensitivity, adiposity, and the metabolic syndrome (26,27). Finally, we observed a positive association between BMI and the risk of type 1 diabetes, which is in line with previous findings regarding type 1 diabetes in children (28,29), including a recent Mendelian randomization study (30). Taken together, these results provide support for the contribution of reduced insulin sensitivity, adiposity, and other elements of the metabolic syndrome to the development of type 1 diabetes in adults. Hypothetically, this implies that the onset of type 1 diabetes may be delayed by efforts to contain insulin sensitivity, e.g., weight loss or increased physical activity. Our findings differ somewhat from observations in children, which could reflect that insulin resistance plays a less crucial role in the development of childhood type 1 diabetes compared with type 1 diabetes with adult onset. To our knowledge, this is the first prospective study investigating glycemic, lipid, and other metabolic markers in relation to type 1 diabetes with adult onset, and validation studies are clearly warranted.

In theory, the moderate elevations in biomarkers observed in individuals who subsequently develop diabetes may reflect genetic susceptibility. It is indeed possible that individuals with HLA risk genotypes or family history of diabetes have a phenotype that is inherently more diabetogenic than that of others (31,32), including higher glucose or lipid levels. On the contrary, most previous studies indicate that sporadic type 1



Figure 1—Trajectories of biomarker values during the 20 years before the diagnosis of type 1 diabetes. Weighted mean biomarkers values and 95% confidence bands are plotted by time to index date for type 1 diabetes case subjects (red) and control subjects (blue). Graphs for additional biomarkers can be found in Supplementary Fig. 1.

diabetes is associated with more severe metabolic disturbances compared with familial type 1 diabetes (33,34), but these characteristics have primarily been assessed at time of diagnosis; prospective data, on the other hand, are scarce. Notably, the downregulation of lipids observed before type 1 diabetes diagnosis in children was found to be independent of HLA genotype (23). Furthermore, the influence of environmental factors may also be exerted through alterations of lipid metabolism or inflammation, and such factors could potentially include dietary factors or virus infections in addition to overweight and obesity. Since we did not have information on either genetic or lifestyle factors, these hypotheses will need to be explored in future studies.

We previously reported on positive associations between these biomarkers (e.g., glucose, fructosamine, triglycerides, apoB-to-apoA-I ratio, and uric acid) and incidence of type 2 diabetes in the AMORIS cohort (6). Our findings regarding type 2 diabetes were strikingly similar to those seen with type 1 diabetes, but biomarker associations with type 2 diabetes were generally stronger and with indications of an even longer preclinical phase. This provides support for the hypothesis of a partly shared pathogenesis of diabetes with adult onset (35), with and without an autoimmune component. In terms of genetic background, it should, however, be noted that the overlap between type 1 and type 2 diabetes appears to be minor (36).

The strengths of our study include the population-based design and the size of the study, including >500,000 individuals and up to 1,000 type 1 diabetes cases and the long follow-up period of up to 25 years. Furthermore, the biomarker measurements were performed on fresh blood samples in the same laboratory with a well-docu-

mented and consistently implemented methodology, assuring high quality of the biomarkers. The use of the PR in combination with the NDR and the PDR allowed a high coverage of all patients with type 1 diabetes (37) and, together with the Swedish Cause of Death Register and population registers, ensured minimal loss to follow-up. Our approach to defining type 1 diabetes cases was rigorous and included several steps to ensure exclusion of type 2 diabetes. Notably, we could verify a type 1 diabetes diagnosis using multiple sources and exclude all individuals who at any time, in any register, received a type 2 diabetes diagnosis. In addition to being exclusively registered with type 1 diabetes, the additional criteria of insulin treatment or age  $\leq$  30 years at diagnosis was used. It should be noted that a diagnosis of type 1 diabetes based on information in NDR has been shown to be accurate in 97% of cases with age of onset <30 years (16). We additionally performed sensitivity analyses restricted to 1) samples obtained after overnight fasting, 2) case subjects both age <30 years and age <40 years at end of follow-up, and 3) case subjects without any record of oral glucose-lowering drugs; most associations were comparable between the main and the restricted analysis, although there was less precision due to smaller numbers in the sensitivity analyses. Furthermore, in separate analyses including only case subjects with a record of insulin dispensed during the year of diagnosis, the associations with glucose, triglycerides, and uric acid remained. In support of our method, incidence of type 1 diabetes with onset at ages between 15 and 29 years ranged between 5 and 13 per 100,00 person-years (2), which aligns well with our cohort (10 per 100,000 person-years). It is possible that diagnosis was timed incorrectly; this would most likely lead to an overestimation of the time between biomarker assessment and diagnosis. To circumvent this, we used three different national registers for identification of cases, using the first recording as time of diagnosis.

The limitations of this study include the lack of repeated measurements for the biomarkers. We did not have information on family history of diabetes, genetic susceptibility toward HLA, measurements of autoantibodies, and insulin and C-peptide levels with which to estimate insulin resistance and  $\beta$ -cell function. We did not have complete data on BMI; thus, our additional analyses with adjustment for BMI are based on a smaller number of individuals. Due to the nature of the available data, it is possible that our findings could at least partly be explained by the presence of individuals with misclassified type 2 diabetes within our population with type 1 diabetes. These could, for instance, include the insulin-deficient subgroup of adult-onset diabetes as defined by Ahlqvist et al. (38), who were autoantibody negative but whose clinical features were similar to those of patients with autoimmune-deficient diabetes. Furthermore, we were unable to separate type 1 diabetes case subjects with adult onset from case subjects with latent autoimmune diabetes in adults. We acknowledge that the moderate effect sizes and the similarity with previous findings in a cohort with type 2 diabetes suggest that potential misclassification of subjects may have contributed to the observed associations. Future studies with an even stricter case definition, e.g., including autoantibody testing, are clearly needed to confirm our findings.

In conclusion, elevated levels of various biomarkers related to glycemia, lipid metabolism, and inflammation are associated with an increased risk of developing type 1 diabetes in adulthood, and these biomarkers may be elevated already 15–20 years preceding diagnosis.

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