ORIGINAL RESEARCH

Role of Serum Lipids, Blood Glucose and Blood Pressure in Breast Cancer Risk for Women with Type 2 Diabetes Mellitus

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Purpose: Women with type 2 diabetes mellitus (T2DM) have an increased risk of breast cancer. We aimed to determine the contribution of lipids, glucose and blood pressure to this risk based on the multifactorial nature of T2DM.

Patients and Methods: This population-based cohort study used data from a Dutch database (the Groningen Initiative to Analyse Type 2 Diabetes Treatment) for the period 2004–2013. The cohort included women diagnosed with T2DM, aged 30–80 years, with no history of breast cancer and with follow-up data for at least 1 year. We used Cox proportional hazards models to estimate the associations of exposures with breast cancer occurrence, reporting adjusted hazard ratios (aHR) with 95% confidence intervals (CI). Exposures of interest included total cholesterol, high-density lipoprotein cholesterol (HDL-C), low-density lipoprotein cholesterol (LDL-C), triglycerides, glycated hemoglobin A (HbA1c) and systolic blood pressure (SBP).

Results: During a median of 4.45 years' follow-up, 183 of 10,183 included women received a breast cancer diagnosis. We observed U-shaped associations with breast cancer incidence for total cholesterol and HDL-C at baseline. Compared with moderate elevations, women had significantly higher breast cancer risks associated with high total cholesterol (aHR, 95% CI: 1.72, 1.15–2.55) and HDL-C (aHR, 95% CI: 1.74, 1.18–2.58) levels, while low total cholesterol (aHR, 95% CI: 1.43, 0.94–2.19) and HDL-C (aHR, 95% CI: 1.44, 0.95–2.17) levels produced marginal effects without significance. Women with high LDL-C levels more often received a breast cancer diagnosis than those with medium levels (aHR, 95% CI: 1.56, 1.03–2.35).

Conclusion: This real-world dataset highlights the importance of balancing lipid profiles, particularly total cholesterol and HDL-C. Dysregulation of the lipid profile, not the glucose or blood pressure profiles, may increase the risk of breast cancer in women with T2DM.

Keywords: cholesterol, triglycerides, glycated hemoglobin A, systolic blood pressure, breast cancer

Plain Language Summary

Type 2 diabetes mellitus (T2DM) may increase the risk of breast cancer in women, but we know little about the causes. Although researchers initially thought that raised blood sugar levels offered a plausible explanation, the subsequent literature has not shown a clear association. Focus has since shifted to the roles of disordered fat levels in the blood (dyslipidemia) and raised blood pressure (hypertension), because we know that women with T2DM suffer from these more often than healthy women. Many observational studies have examined the relationships between these disorders and breast cancer risk, but these have not confirmed clear associations in women with T2DM. The present research examined the associations between different fat levels (total cholesterol, high- and low-density lipoprotein cholesterol, and triglycerides), blood pressure, and blood sugar control (glycated hemoglobin A) with the risk of breast cancer in women with T2DM. This study revealed U-shaped associations with breast cancer incidence for two fats (total cholesterol and high-density lipoprotein cholesterol) and a detrimental role for another (raised levels of low-density lipoprotein

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cholesterol). These findings improve our understanding of the roles of several modifiable risk factors for the occurrence of breast cancer, and they emphasize a need to closely monitor the lipid profiles of women with T2DM.

Introduction

The incidence of female breast cancer has increased globally over the last three decades, reaching 2.26 million cases in 2020, when it overtook lung cancer as the most commonly diagnosed malignancy. 1,2 Breast cancer has multifaceted genetic, lifestyle and disease-related risk factors, and shares many with type 2 diabetes mellitus (T2DM). Consistent with this, several reviews have even indicated that women with T2DM have a 14%-25% increased risk of breast cancer.³⁻⁵ Although obesity may intuitively contribute to this excess risk, research has revealed an increased obesity-related cancer incidence in women both after a diabetes diagnosis and as early as 5 years before this diagnosis. Thus, the mechanisms underlying the relationship between T2DM and breast cancer require further study.

Patients with T2DM often suffer from multiple metabolic disorders. Logically, we might assume that hyperglycaemia increases tumour growth in cancer cells. 7,8 However, the literature has not supported this hypothesis, 9-11 especially for glycated hemoglobin A (HbA1c), indicating the involvement of mechanisms other than hyperglycaemia alone. As a common comorbidity of diabetes, dyslipidaemia has been studied extensively in relation to breast cancer, albeit with inconsistent findings. Whereas biological experiments have implicated cholesterol as a risk factor for breast cancer, 12 research in general populations has uncovered an inverse association between total cholesterol, high-density lipoprotein cholesterol (HDL-C) and breast cancer risk. 13 Nevertheless, questions of reverse causality surround that finding, 14 and more recent Mendelian randomisation studies have consistently suggested that increased HDL-C levels have a positive association with increased breast cancer risk. 15-17 By contrast, research into low-density lipoprotein cholesterol (LDL-C) and triglycerides has revealed less conclusive findings. 13,15-20 Inconsistent results in the literature possibly result from the use of different age groups, producing inconsistent changes relative to age.^{21,22} Intriguingly, diabetes per se could also complicate this relationship by inducing quantitative and qualitative changes in serum lipids and lipoproteins. 23-25 Although dyslipidaemia is routinely measured in patients with T2DM, ²⁶ this has been the focus of only one study to our knowledge. This found no association between HDL-C and breast cancer, though the analysis relied on an unrepresentative sample of mostly low-income residents.²⁷ Finally, hypertension is a common comorbidity in adults with T2DM.²⁸ Although studies have considered the relationship between blood pressure and breast cancer in general populations, ^{29–31} this remains unclear in women with T2DM.

The contributions of total cholesterol, HDL-C, LDL-C, triglycerides, and systolic blood pressure (SBP) to the risk of breast cancer in women with T2DM warrant further research. Given that clinicians routinely collect data on these variables for patients with T2DM, we aimed to use a primary care database to clarify their contribution to the excess risk of breast cancer. The results could support the development of screening strategies for women with T2DM.

Materials and Methods

Study Design, Setting and Population

This retrospective population-based cohort study included women diagnosed with T2DM who received routine treatment in Dutch primary care and had at least one HbA1c or fasting plasma glucose (FPG) measurement between January 2004 and December 2013. The diagnosis of T2DM was established by general practitioners (GPs). Usually, this relied on elevated plasma glucose levels or HbA1c levels following recommendations of Dutch guideline for T2DM management.³² Patients were excluded if they met the following criteria at baseline: (1) age <30 or >80 years, (2) history of breast cancer, or (3) a follow-up period of <12 months after the inclusion date. We allowed a latency period of 12 months between patient inclusion and breast cancer diagnosis.

Data were obtained from the Groningen Initiative to Analyse Type 2 Diabetes Treatment (GIANTT) database, which comprises a cohort of approximately 60,000 patients with type 2 diabetes in primary care in Groningen, the Netherlands. More information about the database is available in the Supplementary Materials and online (https://www.giantt.nl/, accessed on 18th January 2022). This research required no ethics committee approval (see Supplementary Materials for explanations).

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Baseline Characteristics

Given the potential for delays in the diagnosis of T2DM, we defined the baseline index date as the first date of HbA1c or FPG measurement within a period from 1 year before the registered date of T2DM diagnosis to 1 year before the end of follow-up. Baseline data collection proceeded with different strategies to limit missing data, as follows: age (at the index date); body mass index (BMI) (first record within 1 year before the index date and the end of 2012); smoking status (most recent record up to the end of 2012); total cholesterol, HDL-C, LDL-C, triglycerides and SBP (first record within one year around the index date); medications (all records per drug within one year around the index date); and a history of diseases (all records before and within one year around the index date). Further details are included in Figure S1 and the Supplementary Materials.

Explanatory and Outcome Variables

The exposures of interest were the total cholesterol, HDL-C, LDL-C, triglyceride (mmol/L), HbA1c (%) and SBP (mmHg) at baseline. Data were collected from the electronic medical records of general practices. For laboratory measurements, certified labs performed the sample analyses. Blood pressure was measured in general practice according to a recommended procedure (Het Nederlands Huisartsen Genootschap [The Dutch College of General Practitioners]. Protocol Ambulante bloeddrukmeting [Protocol for measuring blood pressure], 30-03-2016, accessed on 22 October 2022. Url: https://www.nhg.org/downloads/protocol-ambulante-bloeddrukmeting). LDL-C was routinely calculated using the Friedewald equation.³³

The outcome of interest was time to breast cancer diagnosis, as assessed by patient follow-up to diagnosis, death, drop-out, or the end of data collection (December 2013), whichever came first. GPs recorded the breast cancer diagnosis in medical records according to the International Classification of Primary Care, 34 typically based on communication from medical specialists who have established the diagnosis. We defined drop-out as the absence of measurements or prescriptions (eg, antihyperglycemics, antihyperlipidemics, antihypertensives, antiplatelets, antidepressants, oral contraceptives, glucocorticoids and aspirin) to 31st December 2012.

Statistical Analyses

All data analyses were performed using R version 4.1.0. Descriptive statistics are presented as means and standard deviations or counts and percentages. The baseline characteristics of patients with and without a diagnosis of breast cancer were compared using independent t-tests (continuous variables) or chi-square tests (categorical variables; not considering missing values). Due to missing data, we used multiple imputation with a chained equations procedure, generating 25 imputed datasets with the "mice" R-package (see Supplementary Materials). 35

Cox proportional hazards models were conducted to test the associations of total cholesterol, HDL-C, LDL-C, triglycerides, SBP and HbA1c with breast cancer occurrence. We present the results as adjusted hazard ratios (aHRs) with 95% confidence intervals (CIs) and include crude cumulative incidence curves of breast cancer over time. The parameters were included as categorical variables in three levels (high, medium and low), with restricted cubic splines (RCS) used to establish the cut-off limits (see Supplementary Materials and Figure S2). For HbA1c, we used the low level as the reference because the RCS curves showed a nearly monotone decreasing trend; by contrast, we used the medium level as the reference for lipids and SBP. Age (cut-off values: 55 and 65 years) and BMI (cut-off values: 25 and 30 kg/m²) were categorised into three levels, using the low level as the reference. History of disease (yes/no) and medication usage (yes/no) were input as dichotomous variables.

Multivariate analyses were performed as follows. Model 1 was adjusted for age and BMI. Model 2 was adjusted for total cholesterol, HDL-C, LDL-C, triglycerides, SBP and HbA1c, other than the targeted explanatory variable, in addition to Model 1. Model 3 was adjusted for metformin, non-metformin glucose-lowering medications, angiotensin converting enzyme inhibitors/angiotensin-receptor blockers (ACE-i/ARB), non-ACE-i/ARB anti-hypertension medications, and statins, in addition to Model 1. Model 4 was adjusted for all covariates with a P-value <0.2 in the univariate analyses. See Supplementary Materials for more details. The proportional hazards assumption was examined by the score test,

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which revealed that only one variable (non-metformin anti-diabetic medications; P < 0.05) went against the assumption.³⁶ Therefore, we used this variable as a stratification factor in Model 3.

Two sensitivity analyses were conducted separately in both the univariate and multivariable models. First, we assessed the bias of reverse causality by extracting subsets of imputed datasets that only included patients with followup data for at least 2 years and fitted the Cox proportional hazards models. Second, we accounted for the competing risk of all-cause death by using the Fine-Gray model on the imputed datasets with a follow-up of at least 1 year.³⁷ To assess detection bias, we also conducted stratified analyses on the imputed datasets by age (≥65 and <65 years) and BMI (≥30 and <30 kg/m²), with the other included as a continuous variable (BMI or age, respectively). Finally, we performed an analysis stratified by statin use at baseline (yes/no).

Data Availability

The University Medical Centre Groningen maintains individual data, which the GIANTT steering committee can make available upon request. This manuscript reports aggregate data only.

Results

Cohort Characteristics

We identified 12,176 women with T2DM who had at least one blood glucose measurement recorded between 2004 and 2013; of these, 10,183 met all the eligibility criteria with a median follow-up of 4.45 years (Figure 1). Breast cancer was detected in 183 subjects, corresponding to a crude incidence rate of 3.69 per 1000 person-years. Compared to women without breast cancer, those who received a diagnosis were older (63.57 vs 61.79 years, P = 0.010), had lower BMIs (30.26 vs 31.37 kg/m², P = 0.007), and slightly higher HDL-C (mean: 1.38 vs 1.30 mmol/L, P = 0.029) at baseline (Table 1). Although total cholesterol did not differ as a continuous variable with breast cancer diagnosis, women without breast cancer more often had a total cholesterol level of 4.7-6.1 mmol/L, whereas those with breast cancer more often had a total cholesterol $\leq 4.7 \text{ mmol/L}$ (P = 0.003). We found no statistical significance for other exposures and covariates.

Association of Lipids with Breast Cancer

Univariate analyses of the imputed and complete datasets indicated that women with high total cholesterol, HDL-C or LDL-C levels at baseline had an increased risk of breast cancer compared to those with medium levels (Figure S3 and Table S1). The final multivariable-adjusted model based on imputed datasets for follow-up ≥1 year in Model 4 (Figure 2 and Table S2) revealed aHR values of 1.72 (95% CI, 1.15–2.55) for total cholesterol (\ge 6.1 versus \ge 4.7 to <6.1 mmol/L), 1.74 (95% CI, 1.18–2.58) for HDL-C (≥1.6 versus ≥1.1 to <1.6 mmol/L), and 1.56 (95% CI, 1.03–2.35) for LDL-C (≥4.1 versus ≥ 2.6 to < 4.1 mmol/L). Additionally, the risk of breast cancer increased with a low total cholesterol (< 4.7 vs ≥ 4.7 to <6.1 mmol/L) and low HDL-C ($<1.1 \text{ vs} \ge 1.1 \text{ to} <1.6 \text{ mmol/L}$) compared with the medium level, whereas the risk was less for low triglyceride levels (<1.3 versus ≥1.3 to <2.4 mmol/L); however, the 95% CIs in Model 4 did not reach significance for the respective aHR values of 1.43 (95% CI, 0.94–2.19), 1.44 (95% CI, 0.95–2.17), and 0.66 (95% CI, 0.42-1.05) (Figure 2 and Table S2). Considering each lipid measure as a continuous variable, Figure S2 depicts the estimates for the HRs and 95% CIs in the crude RCS, which shows a U-shaped pattern for both total cholesterol and HDL-C with nadir hazard ratios at approximately 4.9 mmol/L and 1.1 mmol/L, respectively. The crude cumulative incidence curves are shown in Figure S4.

Sensitivity analyses performed with the imputed datasets of women followed-up for at least 2 years produced similar findings, except for a marginal and insignificant association between a high LDL-C level and breast cancer risk in Model 4 (aHR, 1.56; 95% CI, 0.97-2.50; Table S3). The Fine-Gray models to account for the competing risk of death also produced results comparable to those of the primary analyses (Table S4).

Subgroup analyses (Table S5) indicated a positive association between higher total cholesterol levels and the risk of breast cancer, particularly for older (Model 3: aHR, 2.15; 95% CI, 1.21-3.80) and non-obese (Model 3: aHR: 2.13; 95%

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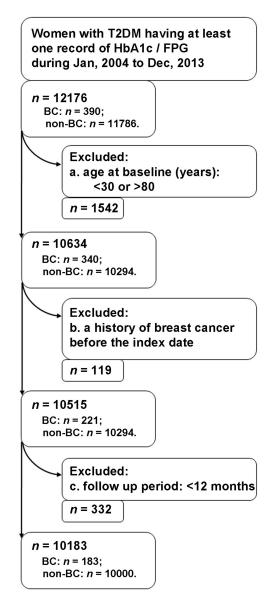


Figure I Flow chart of patient selection. Abbreviations: BC, breast cancer; FPG, fasting plasma glucose; HbA1c, glycated hemoglobin A; n, sample size.

CI: 1.26–3.62) women, but not for younger (Model 3: aHR, 1.36; 95% CI, 0.79–2.34) or obese (Model 3: aHR, 1.31; 95% CI, 0.73–2.37) peers. Statin prescriptions did not alter this relationship.

Association of Blood Glucose and Blood Pressure with Breast Cancer

Neither HbA1c nor SBP had a significant univariable or multivariable association with the risk of breast cancer in the imputed datasets for women who received at least 1 year of follow-up (Figures 2, S3 and Tables S1 and S2). Figures S2 and S4 show the corresponding crude RCS and cumulative incidence curves, while Tables S3-S4 show that the two sensitivity analyses produced similar results.

Subgroup analysis by age (Table S5) revealed a significant association between low SBP levels and the risk of breast cancer in Model 3 for women aged ≥65 years (HR, 2.06; 95% CI, 1.27–3.35), but not for those aged <65 years (HR, 1.03; 95% CI, 0.64–1.65). We also observed this association for women with a BMI ≥30 kg/m² in Model 3 (HR, 1.87; 95% CI, 1.13–3.10), but not for those with a BMI \leq 30 kg/m² (HR, 1.05; 95% CI, 0.65–1.69).

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Table I Baseline Characteristics by Breast Cancer Diagnosis in Women with Type 2 Diabetes Mellitus

Characteristics	Patients with Breast Cancer (n = 183)		Patients without Breast Cancer (n = 10,000)		P values ^a
	N (%)	Mean (SD)	N (%)	Mean (SD)	
Follow-up period (months)		47.50 (26.64)		59.46 (28.52)	<0.001
Duration of diabetes (years)					0.928
<2	176 (96.17)		9576 (95.76)		
≥2	7 (3.83)		424 (4.24)		
Age, groups (years)					0.011
<55	31 (16.94)		2682 (26.82)		
≥55, <65	64 (34.97)		3090 (30.90)		
≥65	88 (48.09)		4228 (42.28)		
Age continuous (years)		63.57 (9.13)		61.79 (10.96)	0.010
BMI groups (kg/m²)					0.152
<25	25 (13.66)		1115 (11.15)		
≥25, <30	65 (35.52)		2908 (29.08)		
≥30	77 (42.08)		4661 (46.61)		
Missing	16 (8.74)		1316 (13.16)		
BMI, continuous(kg/m²)		30.26 (5.21)		31.37 (6.12)	0.007
Smoking status					0.139
Non-smoker	132 (72.13)		6470 (64.70)		
Smoker	25 (13.66)		1730 (17.30)		
Missing	26 (14.21)		1800 (18.00)		
A history of cancers			, ,		0.637
No	170 (92.90)		9401 (94.01)		
Yes	13 (7.10)		599 (5.99)		
A history of cardiovascular diseases	, ,		, ,		0.238
No	150 (81.97)		8536 (85.36)		
Yes	33 (18.03)		1464 (14.64)		
A history of diabetes comorbidities	, ,		,		0.083
No	167 (91.26)		9450 (94.50)		
Yes	16 (8.74)		550 (5.50)		
A history of hypertension	, ,		, ,		0.181
No	97 (53.01)		5821 (58.21)		
Yes	86 (46.99)		4179 (41.79)		
Metformin			/		0.067
No	91 (49.73)		4268 (42.68)		
Yes	92 (50.27)		5732 (57.32)		
Non-Metformin			, ,		0.647
No	135 (73.77)		7552 (75.52)		
Yes	48 (26.23)		2448 (24.48)		
ACE-i or ARB			` '		0.558
No	91 (49.73)		5219 (52.19)		
Yes	92 (50.27)		4781 (47.81)		
Non-ACE-i or non-ARB			, ,		0.352
No	57 (31.15)		3473 (34.73)		
Yes	126 (68.85)		6527 (65.27)		
Statins			, ,		0.884
No	71 (38.80)		3799 (37.99)		
Yes	112 (61.20)		6201 (62.01)		

(Continued)

Table I (Continued).

Characteristics	Patients with Breast Cancer (n = 183)		Patients without Breast Cancer (n = 10,000)		P values ^a
	HDL-C, groups (mmol/L)				
<1.1	38 (20.77)		2071 (20.71)		
≥1.1, <1.6	58 (31.69)		4663 (46.63)		
≥1.6	45 (24.59)		1755 (17.55)		
Missing	42 (22.95)		1511 (15.11)		
HDL-C, continuous (mmol/L)		1.38 (0.44)		1.30 (0.35)	0.029
LDL-C, groups (mmol/L)					0.069
<2.6	51 (27.87)		2671 (26.71)		
≥2.6, <4.1	54 (29.51)		4098 (40.98)		
≥4.1	34 (18.58)		1661 (16.61)		
Missing	44 (24.04)		1570 (15.70)		
LDL-C, continuous (mmol/L)		3.06 (1.11)		3.14 (1.07)	0.396
Total cholesterol, groups (mmol/L)					0.003
<4.7	52 (28.42)		2717 (27.17)		
≥4.7, <6.1	43 (23.50)		3679 (36.79)		
≥6.1	49 (26.78)		2084 (20.84)		
Missing	39 (21.31)		1520 (15.20)		
Total cholesterol, continuous (mmol/L)		5.33 (1.37)		5.27 (1.21)	0.589
Triglycerides, groups (mmol/L)					0.056
<1.3	24 (13.11)		1993 (19.93)		
≥1.3, <2.4	85 (46.45)		4234 (42.34)		
≥2.4	34 (18.58)		2277 (22.77)		
Missing	40 (21.86)		1496 (14.96)		
Triglycerides, continuous (mmol/L)		2.02 (1.02)		2.00 (1.11)	0.807
HbAlc, groups (%)					0.616
<7.0	109 (59.56)		6119 (61.19)		
≥7.0, <8.0	36 (19.67)		1920 (19.20)		
≥8.0	21 (11.48)		1454 (14.54)		
Missing	17 (9.29)		507 (5.07)		
HbAlc, continuous (%)		6.98 (1.43)		7.05 (1.43)	0.525
SBP, groups (mmHg)					0.151
<140	63 (34.43)		3344 (33.44)		
≥140, <164	78 (42.62)		4634 (46.34)		
≥164	36 (19.67)		1442 (14.42)		
Missing	6 (3.28)		580 (5.80)		
SBP, continuous (mmHg)		145.44 (20.23)		145.10 (19.84)	0.827

Notes: ^aP values were estimated by chi-square tests for categorical variables and t-tests for continuous variables. Chi-square tests did not consider missing values. P-values <0.05 are shown in bold and italic. P-values 0.05–0.1 are shown in italic.

Abbreviations: ACE-i/ARB, angiotensin converting enzyme inhibitors/angiotensin-receptor blockers; BMI, body mass index; FPG, fasting plasma glucose; HbA1c, glycated hemoglobin A; HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; SBP, systolic blood pressure; SD, standard deviation.

Discussion

Summary

This study identified U-shaped associations between total cholesterol and HDL-C at baseline with the risk of breast cancer in women with T2DM, with increased risk at high (significant) and low (non-significant) levels compared with medium levels. Women with high LDL-C levels at baseline were also more likely to be diagnosed with breast cancer than those with medium levels. These associations appeared to be independent of statin therapy. Of note, we also observed no relationship between either glycemic control or blood pressure and the risk of breast cancer.

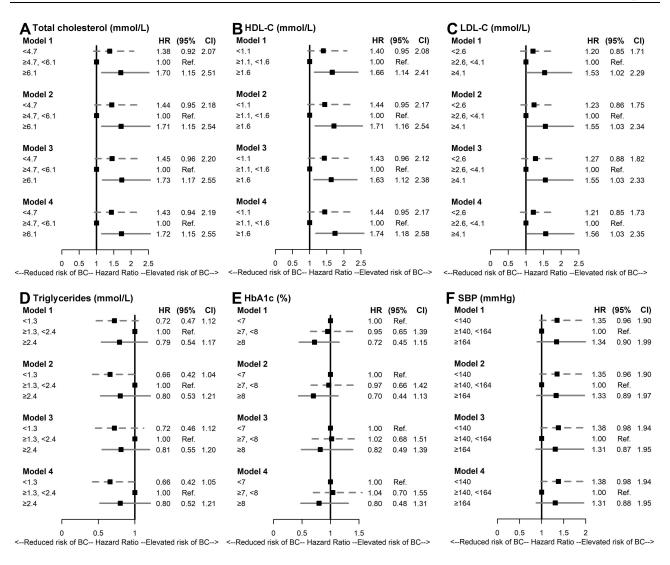


Figure 2 Forest plots of the multivariate analyses examining the associations of lipids, glycated hemoglobin A and systolic blood pressure with breast cancer risk. Associations of (A) total cholesterol, (B) HDL-C, (C) LDL-C, (D) triglycerides, (E) HbA1c and (F) SBP with breast cancer were indicated in HR and 95% CI. Model 1: adjusted for age and BMI; Model 2: adjusted for lipids, HbA1c and SBP (plus Model 1); Model 3: adjusted for metformin, non-metformin glucose-lowering medications, ACE-i or ARB, non-ACE-i or non-ARB antihypertensives, and statins (plus Model 1); Model 4: adjusted for any covariate with a P-value <0.2 in the univariate models. Abbreviations: ACE-i/ARB, angiotensin converting enzyme inhibitors/angiotensin-receptor blockers; BMI, body mass index; CI, confidence interval; HbA1c, glycated hemoglobin A; HDL-C, high-density lipoprotein cholesterol; HR, hazard ratio; LDL-C, low-density lipoprotein cholesterol; Ref., reference; SBP, systolic blood pressure.

Breast Cancer Risk Associated with Circulating Lipids Total Cholesterol, HDL-C and LDL-C

Cholesterol may affect breast cancer risk by altering the properties of membranes to facilitate lipid raft formation and increase signalling events. Moreover, levels affect the metabolite 27-hydroxycholesterol, which might act as an oestrogen receptor agonist that promotes associated cancer cell line proliferation. ^{12,38} Despite the lack of direct evidence for breastspecific cancer in women with diabetes, a low total cholesterol level appears to be associated with increased all-site cancer risk.³⁹ Intriguingly, we observed a U-shaped relationship between total cholesterol and breast cancer risk in both the primary and sensitivity analyses, notably for women aged ≥65 years, but unexpectedly for women with a BMI <30 kg/m². Given that total cholesterol increases around age 50 years in women, ^{21,22} we further adjusted the models with age as a continuous variable, and the observed associations remained.

Statins, which inhibit 3-hydroxy-3-methylglutaryl-coenzyme A (HMG-CoA) reductase, efficiently reduce plasma cholesterol. 40 Although in vitro and in vivo evidence suggests that statins have antiproliferative effects on various cancer, clinical evidence has failed to reveal an association with breast cancer risk. 41 To account for the possible effect of statins,

we nevertheless included their use as a covariate in Model 3 and as a stratification factor in the stratified analyses. The positive association between high total cholesterol levels and breast cancer risk persisted in patients not receiving statins.

Existing evidence suggests that HDL-C has concentration-dependent biphasic effects: at low concentrations it is associated with an increased risk of atherosclerotic cardiovascular disease⁴² and at high levels it is associated with increased risks of death from cardiovascular disease, cancer and other causes, 43 Although we observed a similar U-shaped association between HDL-C and breast cancer, we do not understand the underlying mechanism. It may be that patients with T2DM develop alterations in HDL metabolism, composition and structure, such as the replacement of cholesteryl esters with triglycerides or the glycation or oxidative modification of HDL apolipoproteins or enzymes. In turn, these changes may cause functional HDL deficiencies and even the acquisition of detrimental abilities that induce cancer cell proliferation. 44,45 The Louisiana State University Health Care Services Division (LSUHCSD) cohort found a non-linear association between HDL-C and all-site cancer risk (but not for breast-specific risk) under T2DM conditions.²⁷ This non-linearity was also observed in a prospective cohort of patients with T2DM from the Chinese Hong Kong Diabetes Registry.³⁹ LDL-C may also induce the proliferation of some breast cancer subtypes, ^{46,47} with a Hong Kong cohort showing elevated risk for high LDL-C levels compared to medium levels on a combined outcome of breast, bone, connective tissue and skin cancer. 48 In the GIANTT cohort, given the imbalanced sample sizes and overlapping CIs of HRs in subgroups, we cannot state that statins influence the association of HDL-C or LDL-C with breast cancer risk.

Triglycerides

A previous study showed a possible association between low triglyceride levels and increased cancer risk in patients with T2DM. 49 After adjusting for other lipid measures and covariates, we found a marginal and insignificant association between low triglyceride levels and a lower risk of breast cancer in women with T2DM. Although the underlying mechanisms are still unclear, lipid droplets may play a role in the resistance of cancer cells to lipotoxic stress. 50,51

Possible Explanations for Discrepancies

Compared with other population-based studies, three broad explanations may explain some of the discrepancies in the association between lipid measures and breast cancer risk in the current research.

First, our population had a different comorbidity status. Dyslipidemia induces both qualitative and quantitative changes in patients with diabetes. Glycated and oxidised HDLs cause a loss of physiological function, which results in a stronger ability to promote cell proliferation, thereby increasing the risk of breast cancer. 52,53 Furthermore, glycated LDL is more prone to oxidation than native LDL⁵⁴ and shows decreased LDL receptor-mediated catabolism.⁵⁵ Serum oxidised LDL levels are known to be associated with an increased risk of breast cancer. 46 Thus, diabetes might alter the association between lipid metabolism and breast cancer risk.

Second, we used different methodologies to analyse the data. Observations in both this and previous studies^{27,39} suggest the possibility of a non-linear association between lipids (especially total cholesterol and HDL-C) and breast cancer risk. Consequently, earlier research that has often used the lowest or highest level as the reference group may have obscured the true associations.^{27,56}

Third, our population used different medications, notably glucose-lowering and lipid-lowering drugs, which could have positively or negatively affected lipids and lipoproteins.⁵⁵ Thus, disease severity may have differed. For example, the LSUHCSD cohort relied on hospital-based data, 27 whereas our cohort relied on GP-based data for recently diagnosed women. Thus, we might expect different disease severity and medication use for women with T2DM in these cohorts, which may offset or obscure the impact of lipids on breast cancer risk.

Breast Cancer Risk Associated with Blood Glucose or Blood Pressure

In vitro biological evidence suggests that hyperglycemia creates a fertile ground for tumour growth by promoting reliance on aerobic glycolysis (known as the Warburg effect) and increasing glucose consumption in cancer cells. However, this outcome depends partly on the experimental conditions used, including the assay type, glucose concentration and incubation time. 57,58 Most women in the current study had a recent diagnosis of T2DM and relatively good control at

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baseline (HbA1c < 7.0%). The resulting failure to demonstrate a relationship between HbA1c levels and breast cancer risk is consistent with a recent review demonstrating that improved glycemic control through the use of glucose-lowering drugs or other treatment may not confer reduced cancer risk for patients with T2DM. ⁵⁹ Another review of observational studies found similar results. ¹⁰

Regarding SBP, shared dysregulation of the immune function may explain the biological link between hypertension and cancer. However, further investigation is needed because we only observed an increased risk of breast cancer with a very low SBP compared to a medium SBP in women aged \geq 65 years or with a BMI \geq 30 kg/m². This pattern may instead reflect the higher prevalence of hypertension and the longer history of antihypertensive treatment in older or obese patients. Although an association may exist between antihypertensive medications and the risk of breast cancer, research has not produced consistent findings. 62

Strengths and Limitations

This population-based cohort of women with T2DM (over 95% diagnosed recently) benefitted from a relatively large sample size with detailed information about laboratory tests, physical examinations, medications and disease history. The dataset extracted from electronic records also used real-world data from routine GP care. However, several limitations require careful consideration.

First, by using routinely collected data from electronic medical records, some variation may have existed between general practices in how the measurements and diagnoses were obtained. Nevertheless, we do not expect this to cause specific bias.

Second, we may not have used a sufficiently long latency period to allow the detection of breast cancer. ^{64,65} However, this should be considered in the context that abnormal lipid levels might have persisted for years in patients who develop diabetes. This study still uncovered significant associations between lipids and breast cancer risk, identifying groups at above-average risk of developing breast cancer.

Third, the results could suffer from reverse causality bias. To reduce this, we performed sensitivity analyses that excluded patients censored or diagnosed with breast cancer during the first 2 years of follow-up. Most of the associations persisted in these analyses.

Fourth, some patients not using statins may have received ezetimibe. We do not believe that this affected our findings because few women received this treatment (around 2%).

Fifth, we experienced a moderate number of missing values, especially for the lipid profile at baseline (15%–25%). Sticking to the complete dataset would have caused a substantial loss of patients and a reduction of power, leaving the multivariable analysis only feasible with a limited number of covariates. Thus, we applied multiple imputation by chained equation to create multiple predictions for each missing value based on all available information about a patient, which is considered superior to complete case analysis.³⁵ The broad comparability of the univariable analysis results with the imputed and complete datasets indicates that imputation caused trivial inference.

Sixth, residual confounding could not be excluded, even in the analyses adjusted for an extensive set of confounders, because of the many unmeasured factors (eg, physical activity, dietary factors and advanced glycation end products). Based on the four models with adjusted covariates, however, we observed no notable differences in the estimates. Additionally, we could not exclude selection bias. Among patients without a breast cancer diagnosis, a few baseline characteristics showed statistical differences between those with <12 and ≥12 months' follow-up (data not shown). The large numbers may explain this finding, though differences in the absolute numbers may not be clinically significant.

Conclusions

Dysregulation of the lipid profile, and not the glucose or blood pressure profiles, appears to be associated with the increased risk of breast cancer in women with T2DM in a real-world setting. This supports not only the importance of balancing total cholesterol and HDL-C levels but also the detrimental role of LDL-C on the occurrence of breast cancer in these women. Checking the lipid profile regularly may help to identify women with T2DM at an increased risk of breast cancer.

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Future Perspectives

The results of this study require validation in cohorts with longer follow-up periods. Other research should now investigate the roles of lipoproteins, their glycated and oxidated forms, and cholesterol metabolites in the molecular mechanism of breast cancer. Such research will provide a more complete understanding of the association between lipids and the risk of breast cancer.

Abbreviations

ACE-i/ARB, angiotensin converting enzyme inhibitors/angiotensin-receptor blockers; aHRs, Adjusted hazard ratios; BMI, Body mass index; CIs, Confidence intervals; FPG, Fasting plasma glucose; GPs, general practitioners; HbA1c, Glycated hemoglobin A; HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; LSUHCSD, Louisiana State University Health Care Services Division; Ref., reference; RCS, Restricted cubic splines; SBP, systolic blood pressure; T2DM, Type 2 diabetes mellitus.

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