Association of microRNA gene polymorphisms with Type 2 diabetes mellitus: A systematic review and meta-analysis

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Background: Type 2 diabetes mellitus (T2DM) is a metabolic disorder with growing prevalence and increasing economic burden. Based on the role of genetics and epigenetic factors on T2DM, we aimed to carry a systematic review and meta-analysis for all miRNA gene polymorphisms and risk of T2DM. Materials and Methods: A computerized literature search was carried out on PubMed, Web of Science, Scopus, Embase, as well as references of relevant review/meta-analysis. Key search terms were "Diabetes Mellitus, Type 2," "MicroRNAs," and "Polymorphism, Single Nucleotide." All types of observational studies from January 1, 1992, to November 30, 2019, were included, without language restriction. Data analysis was performed using R programming language (3.5.2). Level of heterogeneity was obtained by Cochran's Q test (P 0.05), and subgroup analysis was performed based on ethnicity. Results: Thirty-two polymorphisms from fifteen articles were included. Meta-analysis was carried out based on minor allele frequencies. Seven studies with 2193 cases and 3963 controls were included for rs2910164 polymorphism. In subgroup analysis, there were significant results in Caucasian population in dominant model (odds ratio [OR] =1.12; 95% confidence interval [CI]: 0.83–1.51), homozygote model (OR = 1.78; 95% CI: 1.06-3.00), heterozygote model (OR = 1.77; 95% CI: 1.03-3.05), and recessive model (OR = 1.78; 95% CI: 1.07–2.96). Four studies with 2085 cases and 1933 controls were included for rs895819 polymorphism. Overall, there was no significant result for association with rs895819, but subgroup analysis revealed that minor allele significantly decreased the risk of T2DM in Caucasians by recessive model (OR = 0.34; 95% CI: 0.18–0.66), dominant model (OR = 0.70; 95% CI: 0.52–0.94), homozygote model (OR = 0.32; 95% CI: 0.16–0.62), heterozygote model (OR = 0.37; 95% CI: 0.19–0.74), allelic model (OR = 0.67; 95% CI: 0.52–0.85). Conclusion: The minor allele of rs2910164 may increase the risk of T2DM by leading to lower level of miR-146a. In contrast, minor allele of rs895819 may decrease the risk of T2DM by leading to higher level of miR-27a.

Key words: MicroRNAs, polymorphism, Type 2 diabetes

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INTRODUCTION

Type 2 diabetes mellitus (T2DM) as a metabolic disorder is caused by insulin resistance and beta-cell dysfunction.^[1,2] The prevalence of diabetes is rapidly growing, especially in middle- and low-income countries. In 2014, 422 million people had diabetes, which means 8.5/100 persons,^[3,4] which may increase

to 592 million till 2035. More than 90% of these patients develop type-2 diabetes.^[5,6] In recent years, diabetes leads to approximately 4 million deaths, as direct or indirect cause each year, and it is proclaimed as the seventh cause of death by the World Health Organization.^[7] T2DM micro- and macrovascular complications obligate increasing financial and social burden on communities. The world direct annual cost of T2DM is estimated over 800 billion dollars.^[3] Several

This is an open access journal, and articles are distributed under the terms of the Creative Commons Attribution-NonCommercial-ShareAlike 4.0 License, which allows others to remix, tweak, and build upon the work non-commercially, as long as appropriate credit is given and the new creations are licensed under the identical terms.

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Address for correspondence: Prof. Mahsa Mohammad Amoli, Metabolic Disorders Research Center, Endocrinology and Metabolism Research Institute, Shariati Hospital, 5th Floor, North Kargar Avenue, Tehran, Iran. E-mail: amolimm@tums.ac.ir Submitted: 04-Nov-2019; Revised: 22-Dec-2019; Accepted: 26-Feb-2020; Published: 30-Jun-2020 factors including genetic, family history, diet, low physical activity, smoking, high body mass index, and aging are responsible for the increasing prevalence of T2DM.^[3,4] Majority of T2DM patients suffer from obesity. High level of FFA in obese individuals correlates with beta-cell dysfunction and insulin resistance in T2DM.^[8] Both genetics and environmental factors are involved in the development of diabetes mellitus.^[9] In many studies, various candidate genes are proposed which are involved in the development of T2DM such as peroxisome proliferator-activated receptor gamma, glycogen synthase, insulin receptor substrate 1, insulin, potassium voltage-gated channel subfamily J member 1, ATP-binding cassette subfamily C member 8, and solute carrier family 2 member 1.^[10] These genes play roles in diabetes by several mechanisms.

MicroRNAs are small, conserved ncRNA molecules that are involved in most cellular processes through interaction with 3'-UTR region of the genes.^[11-13] miRNAs are important small noncoding RNAs that regulate RNAs and gene expression in various common diseases such as T2DM.^[14] miRNAs seem to play a role in the development of pancreatic islets and differentiation of insulin-producing cells.[15-18] For instance, mir-375 is involved in insulin secretion by interaction with myotrophin gene and development of islets.^[16,19,20] The interaction of FoxA2 and Rab27A genes with miR-124a has also been proven influential in the development of islet cells.^[15,21] On the other hand, polymorphisms in miRNAs may be functionally important by regulation of target genes or transcription of miRNAs.[22,23] These polymorphisms may affect the risk, development, progress, or morbidity associated with diabetes, by further effect on miRNA target genes.^[24,25] There are several studies on the role of miRNA polymorphisms in T2DM, but no systematic review or meta-analysis was performed yet. Only two reviews in 2017 and 2018 have been conducted on this subject. In 2017, Zhuang and Wang studied the polymorphisms related to diabetes mellitus and its complications in miRNA genes or their target sites.^[26] In 2018, Zhang et al. reviewed the miRNA polymorphisms associated with diabetes mellitus and its complications.^[27] However, their studies did not accomplish with all requirements. For instance, they included only few polymorphisms, from Ciccacci et al.,[28] while this study contains more miRNA polymorphism. Furthermore, some studies were inadequately added in their report including the study by Alipoor et al.[29] with MAF 0.25 which was designed to investigate the role of miRNA variant on miR-146a expression, or the study by Hara et al.[30] in Zhang review that was a GWA study and should be excluded in our systematic review.

The aim of the current study was to systematically search all miRNA polymorphism-related documents and further meta-analysis of results for their association with T2DM.

MATERIALS AND METHODS

Data sources (electronic and hand-searching)

The strategy has been developed based on the PRISMA-P 2015 checklist.^[31] Figure 1, the PRISMA 2009 flow diagram,^[32] shows the review process.

We systematically searched literatures in Web of Science, PubMed, Embase, and Scopus databases. Syntax definition was performed by combining Medical Subject Headings (MeSH) and free text. Key search terms were "Diabetes Mellitus, Type 2," "MicroRNAs," "Polymorphism, Single Nucleotide," and their equivalents. We also did a search on key journals and references of relevant review/meta-analysis. The second search was performed after first data collections; the search was based on included polymorphisms to find all relevant documents.

Types of studies and participants

Primary documents were screened according to this review strategy. We included case-control, cohort, and cross-sectional studies as three types of observational studies, with data being related to the association between miRNA polymorphisms and T2DM. All other studies were excluded. Studies that their controls were healthy people were included and others were excluded. Studies with insufficient data or with deviation from Hardy-Weinberg equilibrium^[33] were excluded. Insufficient data were resolved by contact with the corresponding author. If we had not received any response after three E-mails, the data were excluded from meta-analysis. Search was performed for documents restricted to publication dates from January 1, 1992, to November 30, 2019, to find the most recent documents which were more reliable. Documents in all languages were included. There was also no restriction about the age, gender, race, or genotyping methods. Documents with diabetes complications such as nephropathy, retinopathy, and other traits were excluded. The quality assessment has been applied by Newcastle-Ottawa Scale method.^[34]

Data collection

Data collection was carried out in three steps: first, removing duplications; second, three reviewers (M.Z., S.A., and A.V.) independently checked title/abstract to find included studies and studies with uncertainty. Disagreements and uncertainties were resolved by both consensus strategy and third-person strategies; at the final step, full texts were investigated by three reviewers (M.Z., S.A., and A.V.). Data extraction was performed by two reviewers (M.G. and M.M.A.) for the following data: first author name, country of study, publication year, study design, ethnicity, gender, polymorphisms, miRNAs, and genotyping methods.



Figure 1: The flow diagram of systematic review

Disagreements between reviewers were resolved by previously described strategies.

Statistical analysis

Data analysis was performed using R programming language 3.5.2 (R: A language and environment for statistical computing; R Foundation for Statistical Computing, Vienna, Austria; URL http://www.R-project. org/.)^[35] and STATA software V.13 (StataCorp. 2013. Stata Statistical Software: Release 13. College Station, TX, USA: StataCorp LP).^[36] The associations between included polymorphisms and T2DM were assessed by odds ratio (OR) and confidence interval (95% CI) for OR. The meta-analysis was performed based on allelic, homozygous, heterozygote, dominant, and recessive genetic models. All genetic models were analyzed in both two random and fixed models. Heterogeneity was assessed by the Cochran's Q test. We used random effects model if Q test was P < 0.05 and the I² statistic was more than 50% in forest plots. For each polymorphism, if sufficient studies were included, subgroup analysis was performed based on ethnicity. To find publication bias in primary studies, we assessed Egger's and Begg's tests. The results were depicted by funnel plots. Sensitivity analysis

was carried out by leave-one-out method, based on omit studies one by one.

RESULTS

Study selection

Based on our double systematic searching and after removing duplication, 1922 articles were included. The flow diagram and data related to screening are shown in Figure 1. Among 15 included studies^[28,37-50] (32 polymorphisms), we found two articles for two single-nucleotide polymorphisms (rs3820455 and rs11888095), three studies were included for rs3746444 and rs531564 polymorphisms, and also more than three studies were included for each of the following polymorphisms (rs895819, rs11614913, and rs2910164). These data are summarized in Table 1.

Meta-analysis

When we combined data related to polymorphisms with two included articles, it was observed that C and T alleles of rs3820455 and rs11888095, respectively, may increase the risk of T2DM, but the results were not significant (data are not shown). Furthermore, analysis of three included

Table 1: B	asic	charact	eristics o	f included s	studies						
First author	Year	Country	Ethnicity	Study design	Quality score	Gender	Source	Genotyping method	miRNAs	Single-nucleotide polymorphism	Sample size (case/control)
Alipoor	2016	Iran	Caucasian	Case/control	8	Male/female	PB	PCR-RFLP	miR-146a	rs2910164	183/192
et al.[38]									miR-149	rs2292832	183/192
Buraczynska et al. ^[39]	2014	Poland	Caucasian	Case/control	8	Male/female	РВ	PCR-RFLP	miR-196a2	rs11614913	920/834
Ciccacci	2017	Italy	Caucasian	Case/control	7	Male/female	PB	Sequencing	miR-375	rs6715345	100/75
et al.[28]		-							miR-27a	rs895819	148/147
									miR-27a	rs11671784	148/147
									miR-130b	rs861843	161/145
									miR-124a	rs531564	162/178
									miR-124a	rs73662598	163/178
									miR-128a	rs11888095	108/85
									miR-194a	rs3820455	144.92
									miR-146a	rs2910164	153/181
Ghaedi et al. ^[40]	2016	Iran	Caucasian	Case/control	8	Male/female	HB	PCR-RFLP	miR-27a	rs895819	204/209
Li <i>et al</i> .[41]	2015	China	ESA	Case/control	7	Male/female	PB	TaqMan	miR-27a	rs895819	738/610
									miR-124a	rs531564	738/610
									miR-128a	rs11888095	738/610
									miR-194a	rs3820455	738/610
									miR-146a	rs2910164	738/610
Sun <i>et al.</i> ^[42]	2018	China	ESA	Case/control	6	Male/female	PB	Sequencing	miR-34a	rs12128240	130/117
				,		,			miR-34a	rs2666433	130/117
									miR-34a	rs6577555	130/117
Wang	2015	China	ESA	Case/control	7	Male/female	PB	SNPscan Kit	miR-27a	rs895819	995/967
et al. ^[37]	20.0	onna	20/1			indio) ionidio			miR-146a	rs2910164	995/967
									miR-124a	rs531564	995/967
Zhou	2013	China	ESA	Case/control	7	Male /female	PB	PCR-RFLP	let-7a-2	rs1143770	104/62
et al. ^[43]	2010	onnia	LON		,	Maley Ternale	1 D		let-7a-2	rs547008	104/62
Chen	2014	China	FSΔ	Case/control	5	Male /female	PR		miR-14.9	rs71428439	33/856
et al.[44]	2014	Onna	LOA	003070011101	0	wate/ ternate	1 D	TOREDR	miR-146a	rs2010164	33/856
									miR-218	rs11134527	33/856
									miR_106a2	rs 1161/013	33/856
									mir_1002	rs3746444	33/856
Циара	2015	China	EGV	Casa (control	Б	Malo /fomalo	Цр	TagMan	miD 1460	rc2010164	16/495
et al. ^[45]	2015	Ghina	LOA	Case/ control	5	wale/ remale		laqiviali	miP_{10602}	ro 1161/012	40/405
7bi at al [46]	2012	China	EGV	Casa (control	Б	Malo /fomalo	Цр		miP 106o2	ro 11614012	40/403
ZIII et al.	2012	Ghina	LJA	Case/ control	5	wale/ lemale	IIF	FOR-KILF	miP 400	ro2746444	20/252
Huong	2015	China	ESA	Casa (control	7	Mala (famala	ЦП	TagMan	miD 1460	ro2010164	J6/2J2
et al ^[47]	2015	Ghina	LJA	Case/ control	/	wale/ lemale	IIF	Taqiviati	miD_{10602}	ro11614012	45/072
	2010	China		Casa (santral	7	Mala (famala	חח	МассАнкан	miD 125 a 2	1511014913	43/0/5
LI et al.	2019	Gillia	ESA	Case/ control	/	wale/remale	FD	Wassarray	IIIIK-155d-Z	rs 10002001	704/040
									let 7d	18 10993081	784/840
									Iet-70	157045890	784/840
									miR-107	rs2402050	784/840
									IIIIK-90	152402959	784/840
									MIK-90	rs0905043	784/840
									miR-29a	rs24168	784/846
									mik-23a	rs3/45453	/84/846
									miR-126	rs4636297	/84/846
									miR-133a-1	rs8089787	/84/846
									miR-133a-1	rs9948906	784/846
				. .					miR-106b	rs999885	784/846
Mohsen et al. ^[49]	2018	Iraq	Caucasian	Case/control	5	Male/female	PB	PCR-RFLP	miR-499	rs3746444	80/40
Moudi and Zohour ^[50]	2018	Iran	Caucasian	Case/control	6	Male/female	PB	PCR-RFLP	miR-605	rs2043556	300/300

articles for rs531564 and rs3746444 polymorphisms showed no significant results (data are not shown).

Seven studies were included for rs2910164.^[28,37,38,41,44,45,47] Seven studies with 2193 cases and 3963 controls were included for rs2910164 polymorphism. In overall analysis, the meta-analysis showed no significant association. Subgroup analysis for ethnicity showed that in Caucasians, there was a significant association in homozygote model (OR = 1.78; 95% CI: 1.06–3.00; I^2 = 35%, Q test P = 0.22), heterozygote model (OR = 1.77; 95% CI: 1.03–3.05; I^2 = 0%, Q test P = 0.57), and recessive model (OR = 1.78; 95% CI: 1.07–2.96; I^2 = 4%, Q test P = 0.31). The plots are shown in Figure 2.

Five studies were included in the meta-analysis^[39,4447] for rs11614913. Five studies with 1082 cases and 3102 controls were included for rs11614913 polymorphism. The results showed that T allele significantly decreased risk of T2DM in four genetic models: overdominant model (OR = 0.83; 95% CI: 0.70–0.97; I^2 = 32%, Q test P = 0.21), dominant model (OR = 0.73; 95% CI: 0.62–0.88; I^2 = 57%, Q test P = 0.05), homozygote model (OR = 0.76; 95% CI: 0.59–0.98; I^2 = 52%, Q test P = 0.08), and heterozygote model (OR = 0.75; 95% CI: 0.62–0.90; I^2 = 51%, Q test P = 0.09). However, subgroup analysis for rs11614913 in Chinese population showed no significant results [Figure S1].

Four studies with 2085 cases and 1933 controls were included for rs895819 polymorphism. Meta-analysis for

rs895819 showed no significant association with the C allele, but subgroup analysis revealed significantly decreased risk of T2DM in Caucasians with recessive model (OR = 0.34; 95% CI: 0.18–0.66; $I^2 = 0\%$, Q test P = 0.32), dominant model (OR = 0.70; 95% CI: 0.52–0.94; $I^2 = 0\%$, Q test P = 0.44, $I^2 = 13\%$), heterozygote model (OR = 0.37; 95% CI: 0.19–0.74; $I^2 = 0\%$, Q test P = 0.43), allelic model (OR = 0.67; 95% CI: 0.52–0.85; $I^2 = 0\%$, Q test P = 0.39). The plots are shown in Figure 3. The results for publication bias and sensitivity analysis are presented in Figures S2 and S3, receptively.

DISCUSSION

Main findings and interpretation

This study aimed to systematically investigate the role of miRNA gene polymorphisms on the risk of T2DM. For this purpose, we performed an overall meta-analysis and subgroup analysis for eligible studies.

Our results showed that C allele of rs2910164 polymorphism increased the risk of T2DM in subgroup analysis, as the risk effect was stronger in Caucasians. This polymorphism is located on miR-146a gene. The conversion of G to C in rs2910164 causes a mispairing in pre-microRNA structure and maturation process, which leads to decrease of both pre/mature miR-146a.^[51] Furthermore, the decreased expression of this miR-146a has been reported in the serum of patients with diabetes compared to the healthy control group.^[52] On the other hand, miRNA-mRNA interaction

Study Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)	Study Experimental Control Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)
$\begin{array}{c} \mbox{Chinase} \\ \mbox{Wang 2015} & 819 & 995 & 799 & 967 \\ \mbox{Li 2015} & 660 & 738 & 506 & 610 \\ \mbox{Huang 2015} & 44 & 46 & 432 & 485 \\ \mbox{Huang 2015} & 26 & 45 & 549 & 672 \\ \mbox{Chen 2015} & 22 & 33 & 566 & 856 \\ \mbox{Fixed effect model} & 1857 & 3590 \\ \mbox{Random effects model} \\ \mbox{Hetrogeneity} \ \vec{r} = 61\%, \ \tau^2 = 0.0858, \ p = 0.04 \\ \end{array}$	*	0.98 [0.78; 1.23] 42.1% 25.1% 1.74 [127; 2.38] 22.9% 21.3% 2.70 [0.64; 11.45] 1.1% 2.8% 0.90 [0.42; 1.91] 4.0% 8.3% 1.02 [0.49; 2.14] 4.2% 8.6% 1.18 [0.99; 1.41] 74.3% - 1.23 [0.85; 1.76] - 66.1%	$\begin{tabular}{ c c c c c c c } \hline Chinase & $$Wang 2015$ $$13$ $$19$ $$322$ 799 $$12015$ $$26$ $$66$ $$236$ $$506$ $$Huang 2015$ $$25$ $$49$ $$Chen 2015$ $$2$ $$49$ $$Chen 2014$ $$62$ $$148$ $$566$ $$Fixed effect model $$151$ $$2852$ $$Random effects model $$Hetrogenesity. $$\vec{r}$ = 0.554, $$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$$		0.92 [0.75; 1.12] 47.7% 27.8% 1.41 [1.12; 1.78] 35.2% 26.4% 1.61 [0.86; 3.02] 4.9% 11.8% 0.72 [0.35; 1.47] 3.7% 9.99% 1.06 [0.41; 2.76] 2.1% 6.4% 1.10 [0.96; 1.27] 93.6% 1.12 [0.83; 1.50] 82.3%
Caucasian 91 183 80 192 Ciccaci 2013 63 153 80 181 Fixed effect model 336 373 Random effects model Hetrogeneity: \hat{r} = 54%, r^2 = 0.0545, ρ = 0.14		1.38 [0.92; 2.08] 13.7% 17.5% 0.88 [0.57; 1.37] 12.0% 16.4% 1.12 [0.83; 1.51] 25.7% 1.11 [0.72; 1.73] 33.9%	$\begin{array}{c c} Catcasian & \\ \mbox{Alipor 2016} & 29 & 91 & 15 & 80 \\ \mbox{Ciccaccl 2013} & 14 & 63 & 13 & 80 \\ \mbox{Fixed effect model} & 154 & 160 \\ \mbox{Random effects model} \\ \mbox{Hetrogeneity: } \vec{\ell} = 0, \rho = 0.57 \end{array}$		2.03 [0.99: 4.14] 3.7% 9.9% - 1.47 [0.64: 3.41] 2.7% 7.8% 1.77 [1.03: 3.05] 6.4% 1.77 [1.03: 3.05] - 17.7%
Fixed effect model 2193 3963 Random effects model Heterogeneity: $\hat{f} = 52\%$, $\tau^2 = 0.0523$, $p = 0.05$ 2 0.1	0.5 1 2	1.17 [1.00; 1.36] 100.0% 1.19 [0.92; 1.53] 100.0% 10	Fixed effect model 1735 3012 Random effects model Heterogeneity: $\hat{f} = 55\%$, $\tau^2 = 0.0570$, $p = 0.04$	0.5 1 2	1.14 [0.99; 1.30] 100.0% 1.21 [0.93; 1.59] 100.0%
Experimental Control Study Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)	Experimental Control Study Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)
$\begin{tabular}{ c c c c c } \hline Experimental Control Events Total Events Total Events Total Chinase Wang 2015 131 489 322 490 Li 2015 364 442 236 340 Huang 2015 12 21 225 348 Chen 2014 6 17 148 438 Fixed effect model 996 1863 Random effects model Heterogenety: \vec{r} = 76%, \vec{r}^2 = 0.2217, \rho < 0.01$	Odds Ratio	OR Weight weight weight fixed 0.93 [0.71; 1.21] 48.8% 22.7% 2.06 [1.47; 2.88] 29.7% 21.5% 3.41 [0.76; 1.48] 1.6% 6.1% 0.73 [0.30; 1.78] 4.28% 10.7% 1.07 [0.39, 2.95] 3.3% 10.1% 1.24 [1.02; 1.50] 87.6% - 1.29 [0.76; 2.19] - 72.1%	Study Experimental Events Control Events Chinase Wang 2015 313 995 322 967 Huang 2015 364 738 236 617 Li 2015 25 46 194 485 Huang 2015 12 45 225 672 Chen 2014 6 33 148 856 Fixed effect model 1857 3590 Random effects model Heterogeneity: $t^2 = 75\%$, $t^2 = 0.0973$, $p < 0.01$ Vol	Odds Ratio	OR Weight 95%-Cl (fixed) (random) 0.92 [0.76; 1.11] 47.3% 23.5% 1.54 [1.24; 192] 35.7% 22.8% 0.79 [0.73; 2.8] 4.6% 12.9% 0.72 [0.37; 1.43] 3.7% 11.4% 0.60 [0.43; 2.62] 2.1% 8.0% 1.15 [1.01; 1.32] 93.4% 1.17 [0.82; 1.65] - 78.6%
$\begin{tabular}{ c c c c c c c } \hline Experimental Control Events Total Events Total$	Odds Ratio	OR Weigh (fixed) (random) 0.93 [0.71; 1.21] 48.8% 22.7% 0.06 [1.47; 2.88] 29.7% 21.5% 0.34 [0.78; 14.88] 1.6% 6.1% 0.73 [0.30; 1.78] 4.28 1.17% 0.74 [0.39; 2.95] 3.3% 1.0% 1.24 [1.02; 1.50] 87.6% 1.7% 1.29 [0.76; 2.19] - 72.1% 2.35 [1.19; 4.65] 7.2% 12.9% 1.76 [0.64; 2.71] 5.2% 12.9% 1.75 [0.91; 3.35] - 27.9%	Study Experimental Events Total Control Events Total Chinase Wang 2015 313 995 322 967 Li 2015 364 738 256 610 Huang 2015 25 46 194 485 Fixed effect model Heterogenety: /* r 75%, r ² = 0.973, p < 0.01	Odds Ratio	OR Weight 95%-CI (fixed) (random) 0.92 [0.76; 1.11] 47.3% 23.5% 1.54 [1.24] 35.7% 22.8% 1.76 [10.97; 3.28] 4.6% 12.2% 0.72 [0.37; 1.33] 3.7% 11.4% 1.76 [10.43; 2.62] 2.1% 8.0% 1.15 [1.01; 1.32] 93.4% - 1.17 [0.82; 1.65] - 7.8.6% 1.21 [0.59; 2.86] 2.7% 9.6% 1.30 [0.59; 2.86] 2.7% 9.6% 1.78 [1.06; 2.98] - 21.4%

Figure 2: Forest plot related to rs2910164 and risk of Type 2 diabetes mellitus. (a) Dominant model, (b) homozygote model, (c) heterozygote model, and (d) recessive model



Figure 3: Forest plot related to rs895819 and risk of Type 2 diabetes mellitus. (a) Recessive model, (b) dominant model, (c) homozygote model, (d) heterozygote model, and (e) allelic model

can play a role in disease-related gene expression or gene regulations. In T2DM, it is shown that this miRNA has interaction with inflammatory genes including *TRAF6*, *NF* κ *B TNF* α , and *IL-6*, which are involved in the pathogenesis of T2DM. Lower expression of miR-146a is associated with higher expression of inflammatory markers, insulin resistance, poor glycemic control, and the overall progression to diabetes.^[53] The reduced level of miR-146a, caused by C allele, may be an important factor for induction of inflammatory mechanism involved in T2DM. These causal relationships may justify the higher frequency of allele C and the role of rs2910164 polymorphism in T2DM patients.

The rs11614913 polymorphism in miR-196a is located on chromosome 17 between *HOX* genes and also has interaction with *HOX* genes.^[54,55] Hox genes play an important role in the initial development of fetal organs including pancreas.^[56] In addition, miR-196a can activate the *AKT* signaling pathway, which is involved in the development and treatment of type 2 diabetes,^[55,57,58] elucidating the importance of miR-196 in the pathology of diabetes. Converting C to T allele in rs11614913 polymorphism influences the processing of miR-196a2 and its function.^[59] Meta-analysis in this study showed that the T allele of this polymorphism has a protective effect on the risk of T2DM, which is similar to the study by Buraczynska et al. on Caucasians.^[39] No such association was observed in the subgroup analysis (Chinese studies), and even, it seems that in Chinese population, the T allele was tended to increase the risk of T2DM. The reason for the discrepancy in the total and subgroup analysis may be the result of the larger sample size in Caucasian study. Furthermore, previous studies such as gnomAD-Exomes and 1000 Genomes determined that the frequency of allele T was considerably different between East Asian and other parts of the world and C allele frequency was higher in East Asian population. Therefore, it may affect result for association between T allele and T2DM in this population. However, these results should be interpreted with cautious and nondecisively. Based on the important role of miR-196a and inconclusive results for the role of rs11614913 polymorphism in T2DM, additional observational studies on different ethnic populations seem to be essential.

We did not find a significant relation between rs895819 and T2DM. It may be related to few studies included and also ethnicity differences between groups. Based on subgroup analysis, we found significant associations in different genetic models, and variant allele seemed to decrease risk of T2DM in Caucasians, which suggests the protective effect for this allele. Most of the previous studies for this polymorphism were related to its role in cancers. Previously, a meta-analysis showed the ethnic effect on the role of this polymorphism in risk of cancers,^[60] which may approve the role of ethnic variation for association with this polymorphism. The miR-27a has high expression in high glycemic condition in adipose tissue.^[61] Higher level of this miRNA increases expression of pro-inflammatory cytokines and plays positive roles in inflammatory response.^[62] The rs895819 in pri-miR-27a may be involved in processing and function of this miRNA.^[63] In this study, the association with this polymorphism was merely affirmed in two included studies.

The rs531564 polymorphism in the pri-miRNA influences the expression of miR-124.^[64] However, we did not observe any correlation between rs531564 and T2DM, which can be as a result of an insufficient number of studies, small sample size, and various ethnicities.

Strengths and limitations

The strengths of our study are as follows: first, this is the first systematic review and meta-analysis, on the role of miRNA polymorphisms on T2DM, and has been performed without any language limitation. Second, we included four main databases as well as hand-searching. Third, three steps of title/abstract screening, full-text eligibility, and data extraction are done by two researchers independently. The limitations of our study are as follows: one of the main causes of heterogeneity in our study was different ethnic populations. Because of the small number of included studies in some cases, subgroup analysis was done by only two included articles. Furthermore, meta-analysis was performed for some included polymorphisms.

CONCLUSION

In this meta-analysis, we found that minor alleles of polymorphisms in miR-146a (rs2910164) and miR-27a (rs895819) inversely affect T2DM susceptibility, especially in Caucasian population. The minor allele of rs2910164 may increase the risk of T2DM by leading to lower level of miR-146a and consequently higher expression of inflammatory markers involved in T2DM. In contrast, minor allele of rs895819 may decrease the risk of T2DM by leading to higher level of miR-27a increasing the effect of inflammatory genes involved in T2DM.

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Conflicts of interest

There are no conflicts of interest.

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Study	Experimental Control Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)	Experimental Study Events Total E	Control events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)
European Buraczynska 2014	396 920 417 834		0.76 [0.63; 0.91] 74.7% 47.0%	European Buraczynska 2014 110 524	125 417		0.62 [0.46; 0.84] 74.1% 36.0%
Chinese Huang 2015 Chen 2014 Zhi 2012 Huang 2015 Fixed effect model Random effects mod Heterogeneity: $\hat{f} = 0\%$, 1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$		1.61 [0.87; 3.00] 6.9% 14.2% 0.99 [0.49; 1.99] 5.4% 11.8% 0.85 [0.43; 169] 5.7% 12.2% 0.95 [0.5]:1/4] 7.3% 14.8% 1.08 [0.78; 1.50] 25.3% - 1.08 [0.78; 1.50] - 53.0%	Chinese 11 18 Huang 2015 11 18 Chen 2014 12 18 Zhi 2012 14 19 Huang 2015 17 23 Eixed effect model 78 Random effects model	142 247 310 465 83 116 187 337 1165		1.16 [0.44; 3.10] 6.8% 16.5% 1.00 [0.37; 2.71] 6.5% 16.1% 1.11 [0.37; 3.34] 5.4% 14.3% — 227 [0.87; 5.91] 7.2% 17.0% 1.33 [0.81; 2.20] 5.9% - 1.33 [0.81; 2.20] - 64.0%
Fixed effect model Random effects mod Heterogeneity: $\hat{l} = 32\%$,	1082 3102 del i, τ ² = 0.0300, <i>ρ</i> = 0.21	0.5 1 2	0.83 [0.70; 0.97] 100.0% 0.91 [0.70; 1.19] 100.0%	Heterogeneity: $f' = 0\%$, $\tau^2 = 0$, $\rho = 0.64$ Fixed effect model 602 Random effects model Heterogeneity: $f^2 = 52\%$, $\tau^2 = 0.1689$, $\rho = 0.05$	1582 8 0.2	0.5 1 2	0.76 [0.59; 0.98] 100.0%
Study E	xperimental Control Events Total Events Total	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random)	Experimental Study Events Total Ev	Control vents Total	Odds Ratio	Weight Weight OR 95%-CI (fixed) (random)
Study E European Buraczvnska 2014	xperimental Control Events Total Events Total 506 920 542 834	Odds Ratio	Weight OR Weight 95%-CI Weight (frandom) 0.66 10.54: 0.801 84.5% 35.8%	Study Experimental Events Total Er European Buraczynska 2014 396 810	Control vents Total 417 709	Odds Ratio	Weight Weight OR 95%-Cl (fixed) (random) 0.67 [0.55; 0.82] 84.3% 38.8%
Example and the second	xperimental Control Events Total Events Total 506 920 542 834 39 46 380 485 27 33 701 856 33 38 219 252 39 45 525 675 162 2268 0, p = 0.71	Odds Ratio	Weight OR Weight 95%-CI Weight (fixed) Weight (random) 0.66 10.54: 0.801 84.5% 35.8% - 1.54 [0.67: 3.54] 4.5% 17.6% 1.00 [0.40; 2.45] 3.9% 16.1% 0.99 [0.36; 2.73] 3.1% 14.0% - 1.86 [0.77: 4.47] 4.1% 16.6% 1.33 [0.85; 2.09] - 64.2%	Experimental Events Total Events Buracynska 2014 396 810 Chinese - - - Huang 2015 28 35 - Zhi 2012 19 24 - - Huang 2015 28 35 - <td< td=""><td>Control vents Total 417 709 238 343 391 546 136 169 - 338 488 1546</td><td>Odds Ratio</td><td>Weight Weight 0R 95%-CI (fixed) 0.67 [0.55; 0.82] 84.3% 38.8% </td></td<>	Control vents Total 417 709 238 343 391 546 136 169 - 338 488 1546	Odds Ratio	Weight Weight 0R 95%-CI (fixed) 0.67 [0.55; 0.82] 84.3% 38.8%

Figure S1: Forest plot related to rs11614913 and risk of Type 2 diabetes mellitus. (a) Overdominant model, (b) dominant model, (c) homozygote model, and (d) heterozygote model



Figure S2: Funnel plots of publication bias for the relation between polymorphism and risk of Type 2 diabetes mellitus. rs2910164: (a1) dominant model, (b1) homozygote model, (c1) heterozygote model, and (d1) recessive model. rs11614913: (a2) overdominant model, (b2) dominant model, (c2) homozygote model, and (d2) heterozygote model. rs895819: (a3) recessive model, (b3) dominant model, (c3) homozygote model, (d3) heterozygote model, and (e3) allelic model



Figure S3: Sensitivity forest plots for the relation between polymorphism and risk of Type 2 diabetes mellitus. rs2910164: (a1) dominant model, (b1) homozygote model, (c1) heterozygote model, and (d1) recessive model. rs11614913: (a2) overdominant model, (b2) dominant model, (c2) homozygote model, and (d2) heterozygote model. rs895819: (a3) recessive model, (b3) dominant model, (c3) homozygote model, (d3) heterozygote model, and (e3) allelic model