Oncology and Cancer Screening

Research Article



A Mendelian Randomization Study of Insulin Therapy for Type 1 Diabetes Increasing the Potential Risk of Ovarian Cancer

Ling Zhang

Ling Zhang, Medicine, Oncology and Cancer Screening, China.

Article Info

Received: September 11, 2024 **Accepted:** September 16, 2024 **Published:** September 20, 2024

*Corresponding author: Ling Zhang, Medicine, Oncology and Cancer Screening, China.

Citation: Zhang L. (2024) "A Mendelian Randomization Study of Insulin Therapy for Type 1 Diabetes Increasing the Potential Risk of Ovarian Cancer", Oncology and Cancer Screening, 5(1); DOI: DOI: 10.61148/2994-8746/OCS/68

Copyright: © 2024 Ling Zhang. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Abstract:

Background Type 1 diabetes (T1D) has been associated with a higher risk of Ovarian cancer (OC), albeit the mechanisms underlying this association remain elusive. A better understanding of the relationship between T1D and OC may contribute to improved primary prevention of OC. We aimed to investigate the putative causal role of T1D on OC, and to identify the potentially mediatory effects of the usage of insulin product underlying this relationship.

Methods We performed a two-sample Mendelian randomization (MR) analysis , multivariable MR analysis to investigate whether T1DM has an independent effect on OC after adjusting for potential confounders. Then, the mediating role of insulin product was subsequently explored using mediation analysis via two-step MR.

Results the MR estimated based on IVW method indicated a causal association between genetically determined T1D and Ovarian cancer (OC) (OR: 1.0006, 95% CI 1.0001-1.0011; P=0.0164). After adjusting for body mass index , Smoking , physical activity , age at menopause and age at menarche, respectively ,we found that a causal relationship between T1DM and OC was still statistically significant (OR>1, P<0.05) .The two-step MR analysis revealed that insulin product acted as a mediating moderator between the T1D and OC (mediated proportion, 1.07%).

Conclusions Our findings suggest that T1D may confer a risk effect to OC, mediated in part by therapeutic insulin product. Therefore, precise dosage of insulin product or an alternative to insulin in T1D patients have a profound significance in terms of the prevention of OC.

Keywords: Type 1 diabetes (T1D) Insulin product Ovarian cancer (OC) Mendelian randomization (MR).

Introduction

Ovarian cancer (OC) is defined as an extremely threatening gynecological malignancy with crucial global impact on women's well-being and economic burden. It stands as a dominating cause of female mortality worldwide, with a miserable escalation in mortality rates. The absence of early clinical feature and the dearth of sensitive and specific diagnostic indicators system lead to the diagnosis of advanced-stage disease in over 70% of ovarian cancer patient. Furthermore, it is worth notice that the 5-year survival rate for individuals diagnosed with advanced-stage ovarian cancer stands at approximately 25%, an extraordinarily lower figure compared to the 5-year survival rate of individuals with early-stage ovarian cancer, which reaches 92% [1]. Despite the advancement of screening methods and treatment modalities, survival rates for patients with OC have not ameliorated. Consequently, exploring the.

Copy right © Ling Zhang

1 | Page

etiology and higher risk factors of OC are of great significance for prevention and treatment of disease actively and effectively.

The pathogenesis of OC is still a blue ocean of medical research. Reported nosogenesis include genetic factors, endocrine factors, fertility factors, immune factors, chronic inflammation, oxidative stress, and environmental factors[2]. Type 1 diabetes (T1D) is an autoimmune and metabolic disease that results in pancreatic β -cell damage, giving rise to a decrease in insulin synthesis or a complete lack of insulin secretion in patients, which contributes to elevated blood glucose[3]. Over the years, countless scientists have been devoting themselves to explore the relationship between T1D and OC. There is sufficient biological evidence that overexpression of IGF-I is demonstrated in OC [4]. Besides, low serum sex hormoneglobulin[5], inflammatory pathways[6], overproduction[7, 8], and microRNAs[9], may explain part of the mechanisms associating T1D to OC. Epidemiologic evidence also suggests that people with T1D are at significantly higher risk for Although there have been basic many forms of cancer. experimental researches and epidemiological investigation[10-12], regarding the association between T1D and OC, there are no randomized clinical trials. Meanwhile, many studies may cause deflection and even misjudgment of the research results due to various observable and unobservable confounding factors, reverse causality and bias.

Mendelian randomization (MR) is a new epidemiological method

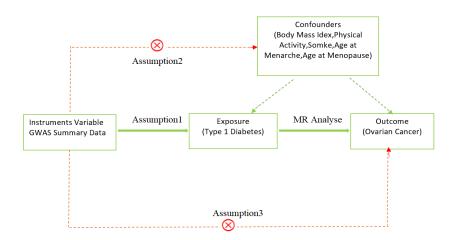
that imitates the design of randomized controlled studies[13]. It uses single nucleotide polymorphisms (SNPs) as instrumental variables (IVs) to infer causal relationship between the risk factors and interested outcomes. SNP is randomly assigned to individuals with gametes during meiosis[14], which is similar to the essential demands of randomized controlled trials. Meanwhile, genovariation precedes the occurrence of diseases, which avoids the potential influence of reverse causality. Consequently, MR is an ideal way to explore the causal relationship between T1D and OC.

This study used a two-sample MR and Mediated Mendelian randomization analyses design to investigate whether T1D has a causal association with OC and estimate the mediating roles between them to provide scientific evidence for OC prevention and treatment.

Materials and methods Experimental design

The current investigation selected T1D as the exposure factor and extracted IVs from the exposure dataset in the shape of SNPs that revealed conspicuous associations with T1D. The outcome variable of interest was OC. MR analysis was conducted to investigate the causal relationship between the exposure and outcome variables [15]. The reliability of MR analysis is predicated on the implementation of three conditional assumptions[16]. Refer to Figure 1A for a visual depiction of this assumptions.

Figure 1A: A diagram of our main MR study. Assumption1, there exists a firm correlation between instrumental variables and exposure factors; Assumption2, no confounding factors are existent that might affect the relationship between exposure and outcome variables; Assumption3, IVs do not exert a direct influence on the outcome, but exclusively effect the outcome via the exposure factors.



Data sources

All data were independently acquired from the IEU GWAS database (https://gwas.mrcieu.ac.uk/terms/), founded by the University of Bristol, UK. Detailed information is provided in Table 1.

Table 1: An overview of the GWAS datasets used in the MR analysis

Contribution	Traits	Sample Number of size SNPs		Author	Population	GWAS ID	
Exposure	Type 1 diabetes	457,695	24,182,422	Sakaue	European	ebi-a- GCST90018925	
Confounders	Body mass index	461,460	9,851,867	Ben Elsworth	European	ukb-b-19953	

Copy right © Ling Zhang

2 | Page

Oncology and	Cancer Screening			Aditum Publishing –www. <u>aditum.org</u>					
	smoked	461,066	9,851,867	Ben Elaworth	European	ukb-b-20261			
	Age at menopause	143,819	9,851,867	Ben Elsworth	European	ukb-b-17422			
	Age at menarche	182,416	2,441,816	Perry JR	European	ieu-a-1095			
	physical activity	377,234	11,808,007	Klimentidis	European	ebi-a- GCST006097			
Mediators	insulin product	462,933	9,851,867	Ben Elsworth	European	ukb-b-15445			
Outcome	Ovarian cancer	199,741	9,822,229	Burrows	European	ieu-b-4963			

Instrumental variables selection

Initially, SNPs revealing significant associations with exposures (selection criteria: P<5.0×10-8) were selected based on the 1,000 whole-genome European population. The linkage disequilibrium parameter (r2) was set at 0.001, and the genetic distance was founded in 10,000 kb to insure the independence of each instrumental variable (IV), excluding interference from other IVs [17]. And then, the instrumental strength of SNPs for each socioeconomic trait was calculated using the F-statistic. An F-statistic greater than 10 manifests that the combined SNPs serve as sufficiently strong instruments to illuminate phenotypic variation, while an F-statistic of 10 or less elucidates a weak instrument[18].

Mendelian randomization

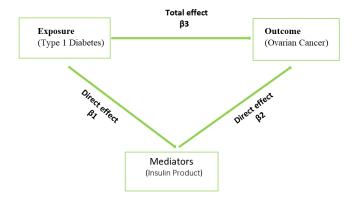
The study began with a two-sample Mendelian randomization employing T1D and confounders (including body mass index, Somke, physical activity, age at menopause, age at menarche) as exposures and OC as an outcome indicator severally. As T1D and

confounders are correlated in clinical studies, multivariate Mendelian randomization was performed to correct for the results. Inverse variance weighted (IVW), weighted median estimator (WME) and MR-Egger regression were the main methods in Mendelian randomization.

Mediated Mendelian randomization

To explore the role of insulin product in the relationship between T1D and OC, we used a two-step MR approach, as shown in Figure 1B.In MVMR, the total effect of each exposure is accounted for by direct effect and an indirect effect. There is a mediator if the following conditions were met:1) there was a correlation between T1D and the mediator (β 1); 2) the mediator was correlated with OC (β 2). 3) T1D was closely related to OC but not adjusted for the mediator (β 3'); The mediation ratio was calculated as (β 1 x β 2)/(β 3') . β 1 x β 2 is the indirect effect and the total effect(β 3) is β 3' + β 1 x β 2 .

Figure 1B: A diagram of mediator MR analysis. This diagram illustrates the potential causal pathway from T1D to OC, with Insulin Product serving as mediators.



Sensitivity analysis

This study implemented a 'leave-one-out' sensitivity analysis to assess whether the observed variation was dominant the association between the exposure and outcome variables. Secondly, to confirm the presence of horizontal pleiotropy in this MR analysis, the MR-Egger intercept test was absorbed [19]. Finally, Cochran's Q statistic was implemented in this study to ascertain heterogeneity.

Statistical analysis

In this study, the "TwoSampleMR" and "Mendelian randomization" [20]. Packages accessible within RStudio software were used to carry out the exposure and outcome analyses. The outcomes from the Mendelian randomization (MR) analysis were defined as beta (β) values, indicating the impact of T1D and insulin

product on OC. Furthermore, the associated 95% confidence intervals (CI) were recorded for all causal estimates. A significance threshold of p $<\!0.05$ was applied to discern statistical significance. Results Two-sample Mendelian randomization

In the IVW models, compelling evidence indicated a significant association between Type 1 Diabetes and the risk of ovarian cancer (OR: 1.0006, 95% CI 1.0001-1.0011; P=0.0164). Additionally, we assessed the association of the 17-SNP instrument with OC using MR-Egger, weighted median. MR-Egger regression showed an OR of 1.0009 (95% CI 1.0001-1.0011; P=0.0492), while weighted median (OR 1.0009, 95% CI 1.0002-1.0016; P=0.0087) techniques provided directionally consistent results. These findings suggest that the results remain robust even in the face of potential

Copy right © Ling Zhang

3 | Page

violations of MR assumptions. Furthermore, no evidence of pleiotropy or heterogeneity was detected in the MR-PRESSO global test, MR-Egger intercept test, and Cochran's Q test. The data is summarized in Supplementary Tables S1-2. To obtain MR estimates for each individual SNP, we conducted the analysis multiple times for each exposure-outcome combination. The scatterplots depicted in Figure 1C illustrate that SNPs exhibiting a larger effect on Type 1 Diabetes also exert a greater impact on the

risk of ovarian cancer. Each method is expressed by a different colored line, with the slope of the line manifesting the estimated causal effect. A leave-one-out analysis is feasible, wherein the MR analysis is repeated while excluding each SNP independently. This allows us to explicit that a majority of the associated signals were not primarily impacted by a single genetic marker, as is demonstrated in Figure 1D.

 Table 2: Two-sample MR analysis results under different methods

Exposure	Outcome	Method	nSNPs	Beta	SE	P-value	OR	OR_Ici95	OR_Uci95
Type 1 diabetes	Ovarian cancer	Inverse variance weighted	17	0.000621589	0.0002591	0.01643862	1.000622	1.000114	1.00113
		Weighted median	17	0.000900215	0.000343071	0.00869059	1.000901	1.000228	1.001574
		MR Egger	17	0.000915882	0.00042798	0.04919663	1.000916	1.000077	1.001756

SNP, single nucleotide polymorphism; SE, standard error; OR odds ratio; CI, confidence interval.

Figure 1C: Scatter plot of single SNP with T1D as the exposure and OC as the outcome.

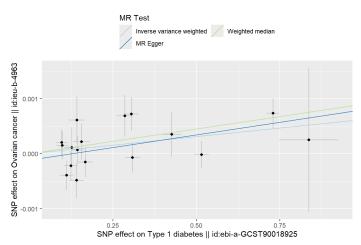
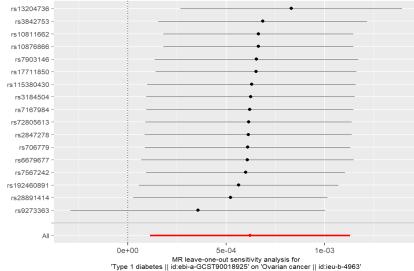


Figure 1D: Leave-one-out analysis of single SNP with T1D as the exposure and OC as the outcome.



Copy right © Ling Zhang

Mendelian randomization

To adjust for potential pleiotropic pathways that might confound the relationship between T1D and OC, an MVMR model was utilized. In this model, the combined effect of T1D on body mass index, Smoking, physical activity, age at menopause and age at

Multivariate

menarche respectively was used as an exposure for OC outcomes. The results of the IVW analysis revealed that the previously significant correlation between T1D and OC was not attenuated in the MVMR model and was still statistically significant , detailed results are presented in Table 3.

Table 3: Forest plot of the MVMR analysis. The direct causal effect of T1D on OC risk was determined by adjusting for traits such as Body mass index, Physical activity, Smoked, Age at menarche, Age at menopause.

exposure	MVMR	outcome	nsnp	method		OR(95%CI)	pval
T1D		OC	17	IVW		1.0006(1.0001 -1.0011)	0.016
adjusting B mas	ody ss index		7	IVW		1.0007(1.0002 -1.0013)	0.012
adjusting P	hysical ctivity		13	IVW		1.0007(1.0002 -1.0012)	0.011
adjusting S	•		14	IVW		1.0007(1.0002 -1.0012)	0.009
adjusting A n	ge at nenarche		9	IVW		1.0009(1.0003 -1.0014)	0.001
adjusting A menop	_		11	IVW		1.0007(1.0001 -1.0012)	0.024
					1 1.0005 1.0015		

Mediated Mendelian randomization

In order to investigate the mediating role of insulin product in the relationship between T1D and OC, a mediated MR analysis was conducted. The results of this study are detailed in

Table4. Our study revealed that insulin product serves as a mediator in the relationship between T1D and OC, with the mediating effect estimated to be ($\beta = 6.63 \times 10$ -4, 95% CI 2.10×10-4 to 1.12×10-3, P = 0.04).

Table 4: The mediation effect of mediators on the causal effect of T1D and OC

exposure	mediators	outcome		Direct	Direct	Mediation effect		Mediated
				effect A	effect B			proportion
			1	β1(95%CI)	β2(95%CI)	β2(95%CI)	P	β (
								95%CI)
T1D	insulin	OC	8.6	66e-	7.72e-	6.63e-	0.004	1.067(0.3
	product		03((7.52e-	02(2.55e-	04(2.10e-		38,1.794)
			03,	9.81e-03)	02,0.129)	04,1.12e-		
						03)		

Discussion

OC and T1D bring tremendous security challenges to public health. The relationship between T1D and OC remains a subject of debate, with the mechanism of this association not fully revealed. In this study, we utilized GWAS database and employed two-sample MR to examine the causal relationship between T1D and OC. The

findings indicate that T1D may potentially accelerate the progression of OC. The interconnectedness between T1D and OC was mediated by insulin product. Insulin product emerges as a risk factor for OC, underlining the precise control of insulin medicine to mitigate the risk of OC.

In recent decades, several observational studies have sought the

link between T1D and OC. Nonetheless, since there is inherent nature of recall and select bias in observational study, certain biases might contribute to inaccurate reporting of causal relationship. Several cohorts [21-24]and case—control [25]studies have been reported that a history of T1D is associated with an augmented risk of OC, however, other relevant studies found a negative significant association[26-29]. Besides, a subgroup meta-analysis based on DM types indicated that the risk of OC in T1DM group (44%) is higher than in T2DM group (17%)[30]. However, MR analysis overcomes the limitations of observational studies, which unambiguously interpret the causal relationship between T1D and OC. Moreover, the positive link was even more prominent in studies that adjusted for confounding factors than these for unadjusted analysis.

The underlying carcinogenesis effect of T1D to OC was not completely revealed at present, but several plausible mechanisms have been used to explain the links between them. Previous studies have shown that the neoplastic process has been considered to influenced by T1D through these mechanisms, including hyperglycemia, adipokine imbalances. chronic inflammation[31-33]. Because of a prolonged exposure to inflammation and hyperglycemic condition, the repair cycles which are associated with incessant ovulation process could be slowed down, thus, resulting in an underlying risk of OC[34]. In addition, hyperglycemia may lead to endothelial dysfunction, endothelial cell death and aberrant neoangiogenesis, ultimately, contributing to cancer progression[35, 36]. Studies have shown that adipose metabolic dysregulation as a hallmark of T1D, can result in high levels of inflammatory cytokines, including IL-6 and TNFa. These cytokines stimulate cell proliferation, invasion, and evasion of antitumor immunity pathways.

Mediated Mendelian randomization was employed to delve into the potential mechanism underlying this relationship. We find that therapeutic insulin product as a mediator could explain parts of this relationship. This is consistent with the results of some observational studies. Since 1972 in the animal experimental researches[37], it has been recognized that the insulin has promoted cancer growth due to the mitogenic effect of insulin by alteration in the PI3K/AKT pathway and mitotic kinase pathway[38]. T1D patients have widely used insulin replacement therapy to strictly control the metabolic status and prevent the progression of longterm complications associated with persistent hyperglycemia[39]. Although there are structural similarities between exogenous and endogenous insulin, the insulin replacement therapy is nonphysiologic, and thus, a supraphysiologic dose of insulin needs to be injected in order to achieve the Physiological needs. This contributes to an inevitable therapeutic increase in the dose of insulin therapy, leading exogenous systemic hyperinsulinemia[40]. Hyperinsulinemia increases the levels of estradiol and testosterone, which may result in cancer initiation and progression in patients.

We integrated data from GWAS of both T1D and OC. This robust methodology supports our causal inference by alleviating biases stemming from confounding and reverse causation. Furthermore, our findings provide a reliable theoretical framework for future investigations focused on improving therapeutic approaches for T1D, such as insulin preparation replacement reagent. This potential improvement may reduce the risk of OC, carrying an important directive function for clinical management. However,

there are several limitations in our research. On one hand, this investigation is the lack of comprehensive GWAS in non-European ancestries. Therefore, the assessment of these causal pathways in diverse ethnic groups should be paid to in the future studies. On the other hand, the utilization of pooled data from genome-wide association studies was short of individual-specific information, eliminating subgroup analyses for variables such as age, disease duration, disease classification, and son on. This limitation makes it impossible to compare potential differences in causal effects among these subgroups.

In conclusion, our study provides evidence supporting a causal relationship between T1D and OC. Furthermore, therapeutic insulin product mediated the effect between T1D and OC. Therefore, precise dosage of insulin product or an alternative to insulin in T1D patients have a profound significance in terms of the prevention of OC.

Acknowledgments

We would like to acknowledge the participants and investigators of the genome-wide association studies. We also want to thank the participants and investigators who developed a convenient software package for Mendelian randomization analysis.

Funding

This research received no specific grant from any funding agency in the public, commercial, or not-for-profit sectors

Disclosure statement

No potential conflict of interest was reported by the author(s).

Data availability statement

The processed datasets in this study are available in GWAS.

References:

- 1. Ha, H.I., et al., Incidence and treatment outcomes of ovarian sarcoma compared to epithelial ovarian cancer from the national cancer registry. Gynecologic Oncology, 2021. 163(3): p. 506-510.
- 2. Wentzensen, N., et al., Ovarian Cancer Risk Factors by Histologic Subtype: An Analysis From the Ovarian Cancer Cohort Consortium. Journal of Clinical Oncology, 2016. 34(24): p. 2888-+.
- 3. Weir, G.C., Islet Inflammation Can Be Linked to the Disruption of Proinsulin Processing in Type 1 Diabetes but not in Type 2 Diabetes. Journal of Clinical Endocrinology & Metabolism, 2023. 108(2): p. E21-E22.
- 4. Malaguarnera, R. and A. Belfiore, The emerging role of insulin and insulin-like growth factor signaling in cancer stem cells. Frontiers in Endocrinology, 2014. 5.
- 5. Osuna, J.A., et al., Relationship between BMI, total testosterone, sex hormone-binding-globulin, leptin, insulin and insulin resistance in obese men. Archives of Andrology, 2006. 52(5): p. 355-361.
- 6. Baig, M.S., et al., Editorial: Targeting signalling pathways in inflammatory diseases. Frontiers in Immunology, 2023. 14.
- 7. Ma, Y., et al., Serum leptin, adiponectin and endometrial cancer risk in Chinese women. Journal of Gynecologic Oncology, 2013. 24(4): p. 336-341.
- 8. Jin, J.H., et al., Association of plasma adiponectin and leptin levels with the development and progression of ovarian cancer. Obstetrics & gynecology science, 2016. 59(4): p. 279-85.

- Yu, T., et al., CircRNAs in cancer metabolism: a review. Journal of Hematology & Oncology, 2019. 12(1).
- 10. Giovannucci, E., et al., Diabetes and Cancer: A Consensus Report. Ca-a Cancer Journal for Clinicians, 2010. 60(4): p. 27. Chen, H.F., et al., A large scale population-based cohort study 207-221.
- 11. Inoue, M. and S. Tsugane, Insulin resistance and cancer: epidemiological evidence. Endocrine-Related Cancer, 2012. 19(5): p. F1-F8.
- 12. Pliszka, M. and L. Szablewski, Associations between Diabetes Mellitus and Selected Cancers. International Journal of Molecular Sciences, 2024. 25(13).
- 13. Thanassoulis, G. and C.J. O'Donnell, Mendelian Randomization Nature's Randomized Trial in the Post-Genome Era. Jama-Journal of the American Medical Association, 2009. 301(22): p. 2386-2388.
- 14. Davies, N.M., et al., Within family Mendelian randomization studies. Human Molecular Genetics, 2019. 28(R2): p. R170-R179.
- 15. Burgess, S., et al., Using published data in Mendelian randomization: a blueprint for efficient identification of causal risk factors. European Journal of Epidemiology, 2015. 30(7): p. 543-552.
- 16. Davies, N.M., M.V. Holmes, and G.D. Smith, Reading Mendelian randomisation studies: a guide, glossary, and checklist for clinicians. Bmj-British Medical Journal, 2018. 362.
- 17. Palmer, T.M., et al., Using multiple genetic variants as instrumental variables for modifiable risk factors. Statistical Methods in Medical Research, 2012. 21(3): p. 223-242.
- 18. Ma, W.W., et al., Causal relationship between body mass index, type 2 diabetes and bone mineral density: Mendelian randomization. Plos One, 2023. 18(10).
- 19. Teng, Y., et al., Association Between Telomere Length and Risk of Lung Cancer in an Asian Population: A Mendelian Randomization Study. World Journal of Oncology, 2023. 14(4): p. 277-284.
- 20. Yavorska, O.O. and S. Burgess, MendelianRandomization: an R package for performing Mendelian randomization analyses summarized data. International Journal Epidemiology, 2017. 46(6): p. 1734-1739.
- 21. Harding, J.L., et al., Cancer Risk Among People With Type 1 and Type 2 Diabetes: Disentangling True Associations, Detection Bias, and Reverse Causation (vol 38, pg 264, 2015). Diabetes Care, 2015. 38(4): p. 734-734.
- 22. Dankner, R., et al., Time-Dependent Risk of Cancer After a Diabetes Diagnosis in a Cohort of 2.3 Million Adults. American Journal of Epidemiology, 2016. 183(12): p. 1098-1106.
- 23. Fuchs, O., et al., The association between a history of gestational diabetes mellitus and future risk for female malignancies. Archives of Gynecology and Obstetrics, 2017. 295(3): p. 731-736.
- 24. Ballotari, P., et al., Diabetes and risk of cancer incidence: results from a population-based cohort study in northern Italy. Bmc Cancer, 2017. 17.
- 25. Reis, N. and N.K. Beji, Risk Factors for Ovarian Cancer: Results from a Hospital-Based Case-Control Study. Turkiye Klinikleri Tip Bilimleri Dergisi, 2010. 30(1): p. 79-87.
- 26. Gapstur, S.M., et al., Type II Diabetes Mellitus and the

- Incidence of Epithelial Ovarian Cancer in the Cancer Prevention Study-II Nutrition Cohort. Cancer Epidemiology Biomarkers & Prevention, 2012. 21(11): p. 2000-2005.
- on the risk of ovarian neoplasm in patients with type 2 diabetes mellitus. Gynecologic Oncology, 2014. 134(3): p. 576-580.
- 28. He, X.N., L.Z. Shi, and J. Wu, Retrospective database analysis of cancer risk in patients with type 2 diabetes mellitus in China. Current Medical Research and Opinion, 2018. 34(6): p. 1089-1098.
- 29. Han, K.T., G.J. Cho, and E.H. Kim, Evaluation of the Association between Gestational Diabetes Mellitus at First Pregnancy and Cancer within 10 Years Postpartum Using National Health Insurance Data in South Korea. International Journal of Environmental Research and Public Health, 2018.
- 30. Wang, L.H., et al., Diabetes mellitus and the risk of ovarian cancer: a systematic review and meta-analysis of cohort and case-control studies. Bmj Open, 2020. 10(12).
- 31. Kellenberger, L.D., et al., The role of dysregulated glucose metabolism in epithelial ovarian cancer. Journal of oncology, 2010. 2010: p. 514310.
- 32. Weinert, L.S., International Association of Diabetes and Pregnancy Study Groups recommendations on the diagnosis and classification of hyperglycemia in pregnancy: comment to the International Association of Diabetes and Pregnancy Study Groups Consensus Panel. Diabetes care, 2010. 33(7): p. e97; author reply e98.
- 33. Amato, M.C., et al., Visceral Adiposity Index (VAI) Is Predictive of an Altered Adipokine Profile in Patients with Type 2 Diabetes. Plos One, 2014. 9(3).
- 34. Piek, J.M.J., P. Kenemans, and R.H.M. Verheijen, Intraperitoneal serous adenocarcinoma: A critical appraisal of three hypotheses on its cause. American Journal of Obstetrics and Gynecology, 2004. 191(3): p. 718-732.
- 35. De Mattia, G., et al., Endothelial dysfunction and oxidative stress in type 1 and type 2 diabetic patients without clinical macrovascular complications. Diabetes Research and Clinical Practice, 2008. 79(2): p. 337-342.
- 36. Bhattacharyya, S., et al., Novel Tissue-Specific Mechanism of Regulation of Angiogenesis and Cancer Growth in Response to Hyperglycemia. Journal of the American Heart Association, 2012. 1(6).
- 37. Heuson, J.C. and N. Legros, Influence of insulin deprivation on growth of the 7,12-dimethylbenz(a)anthracene-induced mammary carcinoma in rats subjected to alloxan diabetes and food restriction. Cancer research, 1972. 32(2): p. 226-32.
- 38. Gryko, M., et al., EXPRESSION OF INSULIN-LIKE GROWTH FACTOR RECEPTOR TYPE 1 CORRELATES WITH LYMPHATIC METASTASES IN GASTRIC CANCER. Polish Journal of Pathology, 2014. 65(2): p. 135-140.
- 39. Cho, Y.H., M.E. Craig, and K.C. Donaghue, Puberty as an accelerator for diabetes complications. Pediatric Diabetes, 2014. 15(1): p. 18-26.
- 40. Chen, S.W., Z.X. Guo, and Q. Yu, Genetic evidence for the causal association between type 1 diabetes and the risk of polycystic ovary syndrome. Human Genomics, 2023. 17(1).