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Assessing causal relationships between gut microbiota and abortion: evidence from two sample Mendelian randomization analysis

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Background: While some studies have suggested a link between gut microbiota (GM) and abortion, the causal relationship remains unclear.

Methods: To explore the causal relationship between GM and abortion, including spontaneous abortion (SA) and habitual abortion (HA), we performed a two-sample Mendelian randomization (MR) analysis. We used summary statistics data from MiBioGen and FinnGen for genome-wide association studies (GWAS), with GM data as the exposure variable and abortion data as the outcome variable.

Results: In the absence of heterogeneity and horizontal pleiotropy, the inversevariance weighted (IVW) method identified five genetically predicted GM genera linked to the risk of abortions. *Lactococcus* was negatively correlated with the risk of SA, whereas the *Eubacterium fissicatena* group was positively correlated with the risk of SA. Genetic predictions of *Coprococcus3* and *Odoribacter* were linked to a reduced risk of HA, while the *Eubacterium ruminantium* group was associated with an increased risk of HA.

Conclusion: Our study suggests a genetic causal relationship between specific GM and two types of abortions, improving our understanding of the pathological relationship between GM and abortion.

KEYWORDS

gut microbiota, Mendelian randomization analysis, spontaneous abortion, habitual abortion, causal relationship

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1 Introduction

Abortion, commonly referred to as miscarriage, is a frequent complication in early pregnancy, usually occurring before the 20th week of gestation. According to the American Society for Reproductive Medicine (ASRM), 15-25% of pregnant women experience miscarriages, although the actual rate may be higher in reality (1). The causes of abortion are varied and complex, with chromosomal abnormalities believed to account for about 50% of cases globally (2). Despite this, the mechanisms behind abortion remain largely unknown (3). In cases of threatened abortion, medical professionals often prescribe hormones like progesterone and dydrogesterone, but their prolonged use can result in emotional disturbances and other pregnancy complications (4, 5). A 2021 report by The Lancet emphasized that the consequences of abortion extend beyond personal and family distress, affecting national health systems and societal economics (1). Therefore, it is crucial to address the negative impacts of abortion and prevent potential risk factors.

The gut microbiota (GM), the most complex microbial community in the human body, plays a significant role in health and disease (6, 7). It has been a focal point of life sciences research for decades. The GM can influence female pregnancy through mechanisms such as immunity regulation, metabolism, inflammation, and the gut-uterine axis (8-10). The balance of microbial communities within the endometrium directly affects reproductive outcomes and may be a factor in recurrent miscarriages (11). Current evidence suggests that changes in certain GM components may support healthy pregnancies, while an imbalance in GM is associated with complications, including abortion (12). Notably, butyrate produced by GM supports intestinal health and normal immune function (13, 14). A reduction in butyrate has been observed in patients with recurrent abortions, drawing researchers' attention (15).

Mendelian randomization (MR) is an epidemiological technique that uses single nucleotide polymorphisms (SNPs) as instrumental variables (IVs) to estimate the causal effects of specific exposures on outcomes (16). This method is particularly valuable in medical research because it can minimize the influence of confounding factors, thus offering significant potential for exploring causal relationships in healthcare studies (17). To date, the relationship between the GM and abortion has been preliminarily investigated in observational studies, but the causal relationship between GM and abortion has not yet been explored (18).

Therefore, this study employs genome-wide association study (GWAS) data from the MiBioGen consortium and the FinnGen database to investigate the causal relationship between the GM and abortion through a two-sample MR analysis. We anticipate that this research will uncover potential pathogenic mechanisms of abortion and propose new strategies for improvement, thereby informing new directions in clinical treatment.

2 Materials and methods

2.1 Study overview

This study followed the framework outlined in Figure 1, treating each bacterial genus in the gut microbiota (GM) as an independent



(B) provides a flowchart of this Mendelian randomization study. Abbreviations used include MR for Mendelian randomization, SNP for single nucleotide polymorphism, GM for gut microbiota, SA for spontaneous abortion, and HA for habitual aborter

exposure factor and considering two types of abortion as outcome variables. The two-sample MR method was used to investigate specific microbial taxa in the GM that have a causal relationship with abortion. The MR method in this study was based on three assumptions: 1) Single-nucleotide polymorphisms (SNPs) used as instrumental variables (IVs) are associated with the GM; 2) IVs are independent of confounding factors; 3) IVs affect abortion risk solely through the GM, not through other pathways (19, 20).

2.2 Data sources

The MR analysis utilized two distinct genome-wide association study (GWAS) datasets. First, GM data were sourced from the MiBioGen Consortium, which conducted a large-scale population genetics study involving 18,340 individuals from 24 cohorts. The SNPs in this study were derived from human samples, initially including 14,587 SNPs ($p < 1 \times 10^{-5}$) related to the gut microbiome (21). Second, GWAS data for spontaneous abortion (SA) and habitual aborter (HA) were obtained from the FinnGen database. The SA study included 181,667 participants (18,680 cases and 162,987 controls) with a total of 21,292,180 SNPs. The HA study included 112,234 participants (651 cases and 111,583 controls) with a total of 21,266,295 SNPs. In this study, the GM was considered the exposure factor, while the two distinct types of abortion were regarded as outcome factors. SNPs were used as IVs in this study. Further details can be found in Table 1.

2.3 Selection of instrumental variables

In our dataset of GM, we classified the genera at the genus level, resulting in a total of 131 genera. We excluded 12 unknown genera, leaving 119 bacterial genera for the MR analysis (22). To ensure the accuracy of the causal relationship between GM and abortion, we implemented a series of quality control procedures to select SNPs related to microbial features. First, we selected SNPs associated with the GM using a significance threshold of $p < 1 \times 10^{-5}$, ensuring a significant correlation between the selected SNPs and the GM. Second, we assessed the independence of the selected SNPs by performing a clumping process ($r^2 < 0.001$, kb = 10,000) to evaluate linkage disequilibrium (LD) (23). Third, we extracted SNP information relevant to both exposure and outcome, aligning the effect alleles to ensure data accuracy. Subsequently, the F-statistic of the SNPs was employed to assess the strength and stability of the IVs in relation to the exposure factor. IVs with an F-value ≤ 10 were deemed to have a weak correlation with the exposure and were

therefore excluded. The calculation formula for the F-statistic is $F=\beta^2$ exposure/SE² exposure (24).

2.4 Statistical methods and sensitivity analysis

This study thoroughly investigated the potential causal relationship between GM and two types of abortion using five analytical methods: Inverse Variance Weighted (IVW), Weighted Median, Simple Mode, MR-Egger, and Weighted Mode, with IVW serving as the primary method (25). To guard against false positives in multiple testing, we applied a Bonferroni correction to establish a statistically adjusted significance threshold $[p = 4.20 \times 10^{-4} (0.05/$ 119)] (26). We assessed the heterogeneity of the results using the pvalue from Cochran's Q test. A p-value < 0.05 indicated the presence of heterogeneity, while a p-value > 0.05 suggested no significant heterogeneity. The reliability of the MR analysis results was validated through the intercept test using the MR-Egger method. An intercept p-value > 0.05 indicated the absence of horizontal pleiotropy, thereby improving the robustness of the study findings. Additionally, a leave-one-out sensitivity analysis was conducted to sequentially exclude individual SNPs and identify any SNPs with a strong influence on the MR estimates. The reliability of the results was further assessed using funnel plots and forest plots. All statistical analyses were performed using R-4.3.2 and RStudio software, utilizing the Two Sample MR package (version 0.5.7). Our rigorous methods and procedures aimed to improve the scientific quality and credibility of the research on the potential causal relationship between GM and the two types of abortion.

3 Results

3.1 Instrumental variable selection

Based on predefined criteria, we selected 1531 SNPs as IVs for 119 GM genera. The analysis showed that the F statistics for these SNPs were greater than 10 (Supplementary Table S1), indicating their robustness as IVs. This suggests that there is no evidence of weak instrument bias, further confirming the reliability of the results. We presented all MR analysis results for the 119 GM genera and the risk of the two types of abortion in Figure 2. Additional details of the analysis results for the 119 GM genera and the two types of abortion can be found in Supplementary Tables S2, S3.

TABLE 1 Details of the GWASs included in the Mendelian Randomization.

Trait	Data Type	N_cases	N_controls	Consortium/Dataset
Gut Microbiota	Exposure	18,340		MiBioGen
Spontaneous abortion	Outcome	18,680	162,987	FinnGen_R10
Habitual aborter	Outcome	651	111,583	FinnGen_R10



3.2 Effects of genetically predicted gut microbiota on two types of abortion

Using IVW analysis, we identified five specific GM genera associated with the risk of abortion. *Lactococcus* (OR = 0.924, 95% CI: 0.868-0.984) exhibited a protective effect on spontaneous abortion (SA), while the *Eubacterium fissicatena* group (OR = 1.074, 95% CI: 1-1.153) was associated with an increased risk of SA. *Coprococcus3* (OR = 0.467, 95% CI: 0.226-0.966) and *Odoribacter* (OR = 0.466, 95% CI: 0.23-0.944) showed a protective effect on habitual abortion (HA), whereas the *Eubacterium ruminantium* group (OR = 1.402, 95% CI: 1.025-1.918) was associated with an increased risk of HA (Figure 3). A scatter plot in Figure 4 illustrates the estimated effects of GM SNPs on abortion based on our MR analysis results.

However, despite the identified causal relationships, the observed outcomes did not meet the stringent threshold set by the Bonferroni correction and thus lost statistical significance after adjustment.

3.3 Sensitivity analysis

We assessed the heterogeneity of SNPs using Cochran's Q test, as presented in Table 2. Additionally, we evaluated the horizontal pleiotropy of SNPs using Egger's intercept and MR-PRESSO. The results indicated no significant heterogeneity or horizontal pleiotropy (p > 0.05). Further confirmation of data robustness was achieved through leave-one-out sensitivity analysis, funnel plots, and forest plots (Figure 5; Supplementary Figures S1, S2).

4 Discussion

This study represents the first investigation into the causal relationships between the GM and different subtypes of abortion. Abortions are categorized based on clinical presentation, such as habitual aborter, and whether they occur spontaneously, such as spontaneous abortion (4, 27–29). We sourced GM data from the

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Simple mode 9 0.14 ● 0.288(0.069 to 1.206) Weighted mode 9 0.14 ● 0.30[(0.070 to 1.266) Odoribacter MR Egger 8 0.19 ● 0.400(0.153 to 1.042) Weighted modian 8 0.061 ● 0.4600(0.153 to 1.042) Inverse variance veighted 8 0.14 ● 0.4600(0.153 to 1.042) Simple mode 8 0.14 ● 0.319(0.079 to 1.297) Weighted mode 8 0.145 ● 0.319(0.078 to 1.257) Eubacterium ruminantium group MR Egger 19 0.078 ● 2.609(0.931 to 7.311) Weighted median 19 0.078 ● ● 1.4420(0.956 to 2.317) Inverse variance weighted 19 0.031 ● ● 1.402(1.025 to 1.918)			Inverse variance weighted	9	0.040		÷		0.467(0.226 to 0.966)
Weighted mode 9 0.14			Simple mode	9	0.127		÷.		0.288(0.069 to 1.206)
Odoribacter MR Egger 8 0.90 0.090(0.021 to 1.730) Weighter dendina 0.061 0.090(0.135 to 1.042) 0.090(0.135 to 1.042) Inverse variance weighted 8 0.034 0.046(0.230 to 0.944) Simple mode 8 0.145 0.319(0.079 to 1.297) Weighted mode 8 0.145 0.312(0.078 to 1.255) Eubacterium ruminantium group MR Egger 19 0.078 0.084(0.956 to 2.317) Inverse variance weighted 19 0.078 0.078 1.462(0.257 to 1.918) Eubacterium ruminantium group MR Egger 19 0.078 1.462(0.257 to 1.918) Keigen conde 19 0.078 1.462(0.257 to 1.912) 1.217(0.277 to 1.127)			Weighted mode	9	0.146		<u></u>		0.301(0.070 to 1.296)
Weighted median 8 0.01 →→→ 0.400(0.135 to 1.042) Inverse variance weighted 8 0.034 →→→ 0.316(0.029 to 1.297) Simple mode 8 0.154 →→→ 0.319(0.079 to 1.297) Weighted mode 8 0.154 →→→ 0.312(0.078 to 1.297) Eubacterium ruminantium group MR Egger 19 0.048 →→→→ 2.669(0.931 to 7.311) Weighted median 19 0.078 →→→→→→ 1.448(0.956 to 2.317) Inverse variance weighted 19 0.073 →→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→→	Odoriba	pacter	MR Egger	8	0.193				0.190(0.021 to 1.750)
Inverse variance weighted 8 0.154 ●●●● 0.319(0.079 to 1.297) Simple mode 8 0.154 ●●●● 0.319(0.079 to 1.297) Weighted mode 8 0.154 ●●●● 0.312(0.078 to 1.255) Eubacterium ruminantium group MR Egger 19 0.078 ●●●●● 2.609(0.931 to 7.311) Weighted mode 19 0.078 ●●●●●● 1.438(0.056 to 2.317) Inverse variance weighted 19 0.073 ●●●●●● 1.730(0.737 to 1.123)			Weighted median	8	0.061		4		0.400(0.153 to 1.042)
Simple mode 8 0.154 ••••• 0.319(0.079 to 1.297) Weighted mode 8 0.145 ••••• 0.312(0.078 to 1.255) Eubacterium ruminantium group MR Egger 19 0.086 ••••• 0.312(0.078 to 1.255) Weighted median 19 0.078 ••••• 1.488(0.956 to 2.317) Inverse variance weighted 19 0.078 ••••• 1.402(1.025 to 1.918)			Inverse variance weighted	8	0.034		(0.466(0.230 to 0.944)
Weighted mode 8 0.145 ••••• 0.312(0.078 to 1.255) Eubacterium ruminantium group MR Egger 19 0.086 ••••• 2.669(0.931 to 7.311) Weighted median 19 0.078 ••••• 1.448(0.956 to 2.317) Inverse variance weighted 19 0.031 ••••• 1.402(1.025 to 1.918) Simple mode 0 0.731 •••••• 1.702(0.72 to 1.127)			Simple mode	8	0.154		<u> </u>		0.319(0.079 to 1.297)
Eubacterium ruminantium group MR Egger 19 0.086 - 2.609(0.931 to 7.311) Weighted median 19 0.078 - 1.488(0.095 to 2.317) Inverse variance weighted 19 0.031 - 1.402(1.025 to 1.918) Simult moves ariance weighted 19 0.031 - 1.101/027 to 1.172			Weighted mode	8	0.145				0.312(0.078 to 1.255)
Weighted median 19 0.078 1.488(0.956 to 2.317) Inverse variance weighted 19 0.034 1.402(1025 to 1.918) Simonb mode 19 0.324 1.402(1025 to 1.918)	Eubacte	terium ruminantium group	MR Egger	19	0.086	18		•	 2.609(0.931 to 7.311)
Inverse variance weighted 19 0.034 ↓ ↓ 1.402(1.025 to 1.918)			Weighted median	19	0.078		—		1.488(0.956 to 2.317)
Simple mode 19 0.221 1721(0.727 to 4.122)			Inverse variance weighted	19	0.034		—		1.402(1.025 to 1.918)
Simple mode 19 0.251 - 1.151(0.127 (0.4.122)			Simple mode	19	0.231	-	•		 1.731(0.727 to 4.122)
Weighted mode 19 0.192			Weighted mode	19	0.192	-	•		 1.830(0.763 to 4.391)
Inverse variance weighted was the primary analysis method, 0 1 2 3 4	Inverse variance weighted was the P value<0.05 was considered stati	e primary analysis method, tistically significant			(0	1 2	3	4
protective factor risk factor		initially significant			protect	tive factor	risk factor		→



MiBioGen database and data on spontaneous and habitual aborter from the FinnGen database. We conducted MR and sensitivity analyses on 119 bacterial genera and two abortion subtypes. Our research identified five bacterial genera with a causal relationship to abortion, with sensitivity analyses showing no evidence of heterogeneity or pleiotropy. These findings support our hypothesis of a causal link between GM and abortion.

The World Health Organization (WHO) defines spontaneous abortion as the natural death of an embryo or fetus before the 20th week of pregnancy without external intervention (30). Our MR analysis found two GM significantly associated with spontaneous abortion. *Lactococcus*, recognized as beneficial microbiota in healthy pregnancies, has been extensively researched for its positive effects on colitis and its ability to induce apoptosis in colorectal cancer cells (31, 32). Research by Antonio González-Sánchez indicates that *Lactococcus* is highly active in the vagina during childbirth (33). We speculate that *Lactococcus* plays a protective role during pregnancy. Additionally, our MR analysis identified the *Eubacterium fissicatena* group as a risk factor.

Numerous studies have shown that the *Eubacterium fissicatena* group affects the host's immune system and may cause pregnancy failure leading to abortion (34).

It is a common misconception that habitual abortion is simply a series of spontaneous abortions; however, this is not accurate. The American Society for Reproductive Medicine (ASRM) specifically defines Recurrent Pregnancy Loss (RPL) as experiencing two or more miscarriages before the 20th week of pregnancy. Approximately 2.5% of pregnant women experience this condition (35). The primary causes include genetic issues, uterine structural abnormalities, hormonal imbalances, and immune system problems (36), such as chromosomal abnormalities, endometritis, thyroid disorders, and Celiac disease (37–39).

In our MR study, we identified three specific microbiota associated with habitual abortion. The Eubacterium rectale group is associated with an increased risk, suggesting it may contribute to habitual miscarriages. Observational studies by Yongjie Liu have shown an increased abundance of the Eubacterium rectale group in the feces of women with habitual abortions, indicating its role in

	Exposure	Hetero	geneity	Directional	MR-PRESSO	
Outcome		Cochran's Q	<i>p</i> -value	Egger intercept	<i>p</i> -value	<i>p</i> -value
Spontaneous abortion	Lactococcus	3.674	0.932	-0.020	0.357	0.913
	Eubacterium fissicatena group	0.861	0.997	0.001	0.979	0.998
Habitual aborter	Coprococcus3	3.199	0.866	-0.036	0.783	0.917
	Odoribacter	3.750	0.710	0.071	0.436	0.766
	Eubacterium ruminantium group	12.712	0.755	-0.064	0.232	0.710

TABLE 2 Sensitivity analysis of the MR analysis results of the gut microbiota and abortions.



increasing this risk (18). Conversely, Odoribacter and Coprococcus3 appear to have protective roles against habitual abortion. Studies by Gomez-Arango et al. found that Odoribacter was negatively correlated with systolic blood pressure at 16 weeks of pregnancy in women with healthy pregnancies (40), suggesting a protective role in maintaining normal blood pressure levels during pregnancy. However, it is noteworthy that Coprococcus3 has been reported to be positively associated with certain diseases, such as reduced immunity (34, 41, 42), but its role in miscarriage has not yet been reported. In our MR study, we speculate that Coprococcus3's preventive role against miscarriage may stem from its ability to produce butyrate (43, 44). Nonetheless, whether Coprococcus3 and Odoribacter influence the occurrence of habitual miscarriages through specific pathways and their mechanisms of action has not been detailed in clinical studies yet. These microbiota undoubtedly warrant further research.

However, this study has certain limitations. First, our data were primarily drawn from European populations provided by the MiBioGen and FinnGen consortia, which limits the diversity of the population in our MR study. Second, our analysis only explored potential causal relationships between GM at the genus level and miscarriage. Third, our Mendelian Randomization (MR) analysis primarily relied on the significance (p < 0.05) of the Inverse Variance Weighted (IVW) method. It is prudent to interpret the significance derived from a single method cautiously. Therefore, future studies should aim to validate these findings with larger datasets and explore other robust MR methods to further strengthen causal inference. Fourth, the MR analysis results did not meet the Bonferroni correction threshold $[p = 4.20 \times 10^{-4} (0.05/$ 119)], meaning the associations in this study are not statistically significant. Hence, these findings are indicative of potential associations rather than definitive evidence. More research is needed to reveal the specific mechanisms involved.

5 Conclusion

Overall, this study utilized two-sample MR to explore the potential causal relationships between the GM and miscarriage, identifying both beneficial and harmful microbial groups that affect miscarriage. This research could potentially assist in the early prevention of miscarriage and provide new insights into its treatment.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material. Further inquiries can be directed to the corresponding author.

Ethics statement

The studies involving humans were approved by Finnish Genome Center Ethics Committee. The studies were conducted in accordance with the local legislation and institutional requirements. Written informed consent for participation was not required from the participants or the participants' legal guardians/next of kin in accordance with the national legislation and institutional requirements.

Author contributions

HY: Conceptualization, Data curation, Methodology, Software, Writing – original draft, Investigation, Validation, Writing – review & editing. JC: Conceptualization, Data curation, Methodology, Software, Writing – original draft, Formal analysis, Writing – review & editing. YW: Conceptualization, Data curation, Formal analysis, Investigation, Writing – original draft, Writing – review & editing. YL: Methodology, Writing – original draft, Writing – review & editing. QJ: Conceptualization, Funding acquisition, Resources, Supervision, Writing – original draft, Writing – review & editing.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fendo.2024. 1415730/full#supplementary-material

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