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# Revealing the role of metformin in gastric intestinal metaplasia treatment

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**Objective:** Gastric intestinal metaplasia (IM) is a precancerous stage associated with gastric cancer. Despite the observed beneficial effects of metformin on IM, its molecular mechanism remains not fully elucidated. This study aims to reveal the effects and potential mechanisms of metformin in treating IM based on both bioinformatics and *in vivo* investigations.

**Methods:** The seven public databases (GeneCards, DisGeNET, OMIM, SuperPred, Pharm Mapper, Swiss Target Prediction, TargetNet) were used in this work to identify targeted genes related to intestinal metaplasia (IM) and metformin. The shared targeted genes between metformin and IM were further analyzed by network pharmacology, while the interactions in-between were investigated by molecular docking. In parallel, the therapeutic effect of metformin was evaluated in IM mice model, while the core targets and pathways effected by metformin were verified *in vivo*.

**Results:** We screened out 1,751 IM-related genes and 318 metformin-targeted genes, 99 common genes identified in between were visualized by constructing the protein-protein interaction (PPI) network. The top ten core targeted genes were *EGFR*, *MMP9*, *HIF1A*, *HSP90AA1*, *SIRT1*, *IL2*, *MAPK8*, *STAT1*, *PIK3CA*, and *ICAM1*. The functional enrichment analysis confirmed that carcinogenesis and HIF-1 signaling pathways were primarily involved in the metformin treatment of IM. Based on molecular docking and dynamics, we found metformin affected the function of its targets by inhibiting receptor binding. Furthermore, metformin administration reduced the progression of IM lesions in Atp4a<sup>-/-</sup> mice model significantly. Notably, metformin enhanced the expression level of *MUC5AC*, while inhibited the expression level of *CDX2*. Our results also showed that metformin modulated the PI3K/AKT/mTOR/HIF-1a signaling pathway.

**Conclusion:** This study confirms that metformin improves the efficacy of IM treatment by regulating a complex molecular network. Metformin plays a

functional role in inhibiting inflammation/apoptosis-related pathways of further IM progression. Our work provides a molecular foundation for understanding metformin and other guanidine medicines in IM treatment.

KEYWORDS

network pharmacology, molecular docking, intestinal metaplasia, precancerous diseases, metformin

### **1** Introduction

Gastric cancer, a prevalent malignant neoplasm affecting the gastrointestinal tract, ranks as the fifth in incidence and third in mortality globally (Sung et al., 2021). Gastric intestinal metaplasia (IM), a type of precancerous condition, is highly associated with gastric cancer (Correa and Piazuelo, 2012). A previous study has found that stem cells play a crucial role in sustaining the equilibrium of the gastric mucosa, and its abnormality can be the primary cause related to IM and gastric cancer (Hayakawa et al., 2021). Due to the monoclonal proliferation of IM cells in human stomach, aberrant stem cells that underwent a specific differentiation may generate IM (Tatematsu et al., 2003). For instance, Mist1 gastric stem cells can result in metaplasia which Kras activation initiates (Hayakawa et al., 2015). In Mist1-Kras mice, increased Ras expression in the chief cells can lead to a wide range of metaplastic lineage changes, including IM and spasmolytic polypeptide-expressing metaplasia (SPEM) (Hayakawa et al., 2015). Moreover, blocking the Ras signaling pathway by MEK inhibitor has been reported to reverse gastric precancerous lesions (Choi et al., 2016). A latest evidence suggests that the origin of SPEM may attributed to a subset of chief cells through trans-differentiation, which are developed from dedifferentiation from chief cells, and shows more easily induced and expanded by acute parietal cell loss under inflammation (Nam et al., 2010). Chronic inflammation is exacerbated by the loss of parietal cells, resulting in foveolar hyperplasia and SPEM, which may eventually form IM (Giroux and Rustgi, 2017).

Metformin, a commonly prescribed biguanide medication, is commonly used in the treatment of type-2 diabetes (LaMoia and Shulman, 2021). It increases the parietal cell differentiation and inhibits the proliferation of the progenitor cell by regulating AMPK, KLF4, and PGC1α, which induces gastric acid secretion and prevents gastric cancer development (Miao et al., 2020). Metformin activates AMPK and inhibits mTORC1 pathway to trigger apoptosis and impede cell proliferation/cycle progression in gastric carcinogenesis (Cheung et al., 2022). It significantly decreases the incidence of gastric cancer in Helicobacter pylori-eradicated diabetes in a timea potential and dose-dependent manner, suggesting chemopreventive effect on gastric cancer (Cheung et al., 2019). Although metformin reduces the invasive phenotype and intestinal marker expression in CDX1-overexpressing HFE 145 cells (Choi et al., 2019), the role of metformin in IM treatment has not been well studied.

This study herein tried to obtain metformin-targeted genes and IM-related genes from seven public databases (GeneCards, DisGeNET, OMIM, SuperPred, Pharm Mapper, Swiss Target Prediction, TargetNet). Network pharmacology and molecular docking were performed to screen core genes, and then the interaction between metformin and selected genes was studied. The effects and targets-related mechanisms of metformin were investigated in IM mice model to confirm *in vivo*. Our work aims to preliminarily reveal the mechanism of metformin and its potential in IM therapeutics.

### 2 Materials and methods

### 2.1 Experimental design

The work flow of this study was shown in Figure 1. The databases and their websites were showed in the Supplementary Table S1.

### 2.2 Identification of IM-related genes

The genes related to IM were obtained from GeneCards (Wang et al., 2022), DisGeNET (He et al., 2022) and Online Mendelian Inheritance in Man (OMIM) (Li et al., 2022) databases. The search term in this work was "intestinal metaplasia." The Uniprot database (Zhang et al., 2021b) was used for gene correction.

### 2.3 Identification of metformintargeted genes

Genes targeted by metformin were collected from online databases, including SuperPred (Ma et al., 2022), Pharm Mapper (Jiao et al., 2022), Swiss Target Prediction (Shang et al., 2023), and TargetNet (Zhang et al., 2022c), with "metformin" as the keyword. Uniprot database was used to identify the genes.

### 2.4 Construction and analysis of proteinprotein interaction network

With shared targeted genes supplied from IM and metformin collections sourced from the Venny 2.1 tool, the protein-protein interaction (PPI) network was built by the STRING database framework. The visualization of the PPI network was performed using Cytoscape software version 3.9.1 (Zhang et al., 2021a).

### 2.5 Function enrichment analysis

The R clusterProfiler package (Li et al., 2021; Zhou et al., 2022) was conducted for Gene Ontology (GO) and Kyoto Encyclopaedia of Genes and Genomes (KEGG) pathway enrichment analysis. The



Bioinformatics website (http://www.bioinformatics.com.cn/) was employed to visualize the enrichment analysis.

### 2.6 Molecular docking

The chemical structure of metformin was obtained from the PubChem database (Zhang et al., 2022b). The PDB database was used to collect the structures of targeted proteins and ligands based on the genes identified above (Nag et al., 2023). The molecular docking study was conducted using AutoDock Vina 1.5.7, and the resulting data was shown using PyMol 2.4.0 (Zhang et al., 2022a; Zhu et al., 2022).

### 2.7 Experimental animals

The Atp4a<sup>-/-</sup> mice (certificate number: 20170010010146) were constructed utilizing CRISPR/Cas9 system provided by the Shanghai Model Organisms Center. The animals in a pathogen-free room were maintained in standard settings (relative humidity: 50%–60%, temperature:  $22^{\circ}C \pm 3^{\circ}C$ , 12-h/ 12-h light/dark cycle). The animal experiments conducted in this study were approved by the Animal Ethics Committee of the Fifth Affiliated Hospital of Zhengzhou University (No. KY2023017). The six wild-type mice (mice-C57bl/6) were purchased from

Beijing Weitong Lihua Experimental Animal Technology Company, which were set as the control group, while 12 Atp4a<sup>-/-</sup> mice were set to the model group and metformintreated group randomly. Six Atp4a<sup>-/-</sup> mice in metformin-treated group were given a diet containing 1,000 ppm (0.1%) metformin (Martin-Montalvo et al., 2013; Strong et al., 2016) (Macklin, China), starting at 12 weeks of age. After 12 weeks, gastric tissues were harvested for further experimental analysis.

### 2.8 Pathological examination

The gastric tissue was fixed with 4% paraformaldehyde. The samples were dehydrated by 75%–95% alcohols gradually, immersed in xylene, embedded in paraffin, and then cut into approximately 3- $\mu$ m sections using the microtome. Haematoxylin and eosin (H&E) staining kit (Celnovte, China) and alcian blue-periodic acid-schiff (AB-PAS) staining kit (pH 2.5) (Celnovte, China) staining were used to observe the degree of gastric lesions.

### 2.9 Immunohistochemistry (IHC) staining

The gastric tissue slices were further dewaxed and hydrated. The gastric tissue sections were incubated with MUC5AC (1:200 dilution; GeneTex, GTX11335, United States) and CDX2 (1:100 dilution; Abcam,

ab101532, United Kingdom). The incubation steps were done by the universal two-step kit (mouse/rabbit enhanced polymer test system) (ZSGBBIO, China). The cellular nuclei were stained with haematoxylin (Celnovte, China). The analysis of the integrated optical density (IOD) and area of the images was conducted using the ImageJ software (Schindelin et al., 2012). The protein expression was evaluated using the mean density (IOD/area).

# 2.10 RNA extraction and real-time PCR analysis

Total RNA was extracted from mouse gastric tissues with RNAiso Plus (TAKARA, Japan) following the manufacturer's procedure. The obtained RNA was reverse-transcribed into cDNA using the ReverTra Ace qPCR RT Kit (Toyobo, Japan). Quantitative real-time polymerase chain reaction (qRT-PCR) was performed with the Roche Lightcycler480II system (Roche, Switzerland) using ChamQ SYBR qPCR Master Mix (Vazyme, China). GAPDH was used as internal control and the qRT-PCR analysis for each sample was performed in triplicate with  $2^{-\Delta\Delta CT}$ . The primer sequences were provided in the Supplementary Table S2.

### 2.11 Western blot

The total proteins were extracted from gastric tissues with radio immunoprecipitation assay (RIPA) buffer (Solarbio, Beijing, China) and quantified by Bicinchoninic Acid Assay (Solarbio, Beijing, China). Before sodium dodecyl sulfate polyacrylamide gel electrophoresis (SDS-PAGE), the protein lysis buffer was mixed with  $5 \times$  SDS-PAGE loading buffer and boiled for 5 min at 100°C with the thermostat metal bath. The proteins were separated using a 5%-10% SDS-PAGE gel, followed by transferring into a polyvinylidene fluoride (PVDF) membrane. The membranes were blocked using a 5% non-fat milk solution for 2 h at ambient temperature. The membranes were then incubated with the primary antibody at 4°C overnight. After washing three times with Tris-buffered saline with Tween 20 (TBST), the membranes were incubated with the secondary antibodies for 1 h at room temperature. Enhanced chemical luminescence (ECL) reagent (Meilunbio, Shanghai, China) was used to detect the protein bands. The analysis was conducted using the ChemiDocTM XRS + equipment (Bio-Rad, United States).

The information of the primary antibodies are as below: rabbit anti-NF-kB (1:1,000, Cell Signaling Technology, 8,242, United States), rabbit anti-phospho-NF-KB (1:1,000, Cell Signaling Technology, 3,033, United States), rabbit anti-PI3K (1: 1,000, Cell Signaling Technology, 4,257, United States), rabbit antiphospho-PI3K (1:1,000, Affinity Biosciences, AF3241, United States), mouse anti-AKT (1:10,000, Proteintech, 60,203, United States), rabbit anti-phospho-AKT (1:10,000, Proteintech, 80,455, United States), rabbit anti-mTOR (1:5,000, Proteintech, 28,273, United States), rabbit anti-phospho-mTOR (1:1,000, Proteintech, 28,879, United States), rabbit anti-HIF-1a (1:1,000, Cell Signaling Technology, 36,169, United States), and mouse anti- $\beta$ -Actin (1:10,000, Proteintech, 66,009, United States) (Liu et al., 2020; Ba et al., 2021).

### 2.12 Statistical analysis

GraphPad Prism 9.0 (Berkman et al., 2019) was used to perform all statistical analyses. The Shapiro-Wilk test was used to examine the normal distribution. The independent samples *t*-test was used to compare the two groups of measurement data which met the requirements of normal distribution and homogeneity of variance. One-way analysis of variance (ANOVA) was performed to compare more than two groups of data conforming to the normal distribution and homogeneity of variance requirements. Kruskal–Wallis tests were conducted to compare two groups and multiple groups that did not conform to the normal distribution and homogeneity of variance requirements. The data was shown using the mean  $\pm$  standard deviation (SD). A *p*-value of <0.05 is indicative of a statistically significant difference.

### **3** Results

# 3.1 Screened IM-related genes and metformin-targeted genes

1,751 IM-related genes were obtained from the DisGeNET, GeneCards, and OMIM databases. In the SuperPred, Pharm Mapper, Swiss Target Prediction, and TargetNet databases, 318 metformin-targeted genes were identified.

# 3.2 GO and pathway enrichment analyses of IM-related genes

GO enrichment analysis indicated that cellular components of IM-related genes included collagen–containing extracellular matrix, membrane raft, membrane microdomain, etc (Supplementary Figure S1A). These genes were involved in the activity of receptor ligand, cytokine receptor binding, transcription factor, and growth factor (Supplementary Figure S1B). They were also implicated in several biological processes, such as epithelial cell proliferation, gland development, and tissue migration (Supplementary Figure S1C). We found related signaling pathways, such as microRNAs, proteoglycans, and PI3K-AKT signaling pathway (Figure 2A). Notably, the Enriched ReactomePA Pathways were mainly involved in interleukins related pathways, PI3K/AKT signaling in carcinogenesis and apoptosis (Figure 2B).

# 3.3 GO and pathway enrichment analyses of metformin-targeted genes

The GO enrichment results of metformin-targeted genes showed that the cellular components were mainly related to the vesicle lumen, membrane raft and membrane microdomain (Supplementary Figure S1D), molecular



#### FIGURE 2

Pathway enrichment analyses of IM-related genes and metformin-targeted genes. (A) Top ten enriched KEGG pathways of IM-related genes. (B) Top ten enriched ReactomePA pathways of IM-related genes. (C) Top ten enriched KEGG pathways of metformin-targeted genes. (D) Top ten enriched ReactomePA pathways of metformin-targeted genes.



functions were mostly related to the activity of peptide receptor, nuclear receptor and transcription factor (Supplementary Figure S1E), and biological process enrichment items were mainly related to the cellular response to peptide, response to peptide hormone and small molecule catabolic process (Supplementary Figure S1F). The main KEGG pathways were found to be

involved in neuroactive ligand-receptor interaction, chemical carcinogenesis-receptor activation, and cancer metabolism (Figure 2C). The ReactomePA pathways enriched analysis primarily included rhodopsin-like receptors, GPCR ligand binding, the nuclear Receptor transcription route, and signaling by Interleukins (Figure 2D).



### 3.4 The analysis of PPI network

A total of 99 targets were screened out as the shared targeted genes between IM and metformin clusters (Figure 3A). The PPI

network was composed of 94 nodes and 510 edges (Figure 3B). The core targets of metformin in IM treatment were determined by selecting the top ten targets based on their degree value. The PPI network of core targets consists of a total of ten nodes and 41 edges.



The top ten targets are *EGFR*, *MMP9*, *HIF1A*, *HSP90AA1*, *SIRT1*, *IL2*, *MAPK8*, *STAT1*, *PIK3CA*, and *ICAM1* (Figure 3C). The shade of the circle represents the importance of the target.

# 3.5 Annotation of targeted genes of metformin in IM treatment

GO and KEGG enrichment analyses were performed on 99 shared targeted genes of metformin for IM treatment. The top ten from each category of GO enrichment analysis were made for visual inspection. Biological processes (Figure 4A) included cellular response to peptide, collagen metabolic process, collagen catabolic process, reactive oxygen species metabolic process, and regulation of reactive oxygen species metabolic process. Cellular components included the vesicle lumen, secretory granule lumen, and specific granule (Figure 4B). The main molecular functions (Figure 4C) were nuclear receptor activity, ligand-activated transcription factor activity, and endopeptidase activity.

The KEGG enrichment analysis indicated that the metformin in IM treatment primarily influenced 80 signaling pathways. The top 20 pathways were charted in Figure 5. Common targeted genes were mostly involved in pathways in tumorigenesis, HIF-1 signaling pathway, PD-L1 expression and PD-1 checkpoint pathway in cancer, FoxO signaling pathway, prolactin signaling pathway, proteoglycans in cancer, and relaxin signaling pathway.

### 3.6 Molecular docking verification

We employed a docking approach to analyse ten hub targets, including EGFR, MMP9, HIF1A, HSP90AA1, SIRT1, IL2, MAPK8, STAT1, PIK3CA, and ICAM1 (Figure 6; Supplementary Figure S2). The information of core target proteins was shown in Supplementary Table S3. The molecular docking analysis revealed that the binding energies between metformin and the target molecules were all below zero (Table 1). Notably, the docking energies between EGFR, MMP9, HIF1A, IL2, PI3KCA and metformin were all less than –5 kcal/Mol.

Metformin showed a strong association with amino acid residues via the formation of hydrogen bonds. Furthermore, the binding affinities of metformin to EGFR, MMP9, HIF1A, SIRT1, IL2, MAPK8, and ICAM1 were found to be lower than those of the positive control. This observation suggested that metformin performed a strong binding capacity to these target proteins compared. The binding affinities of metformin with STAT1, HSP90AA1, and PIK3CA showed no significant difference from those of the positive control, indicating that metformin's capacity to



bind with these target proteins was nearly the same as that of the positive control.

# 3.7 Metformin inhibited the progression of IM

To examine the impact of metformin, histological slices of gastric tissues were prepared and stained with H&E once the  $Atp4a^{-/-}$  mice reached the age of 24 weeks. In the control group, the gastric mucosal epithelial cells were neatly distributed, the glandular structure was normal and tightly arranged, and no

pathological changes were observed. Atp4a<sup>-/-</sup> mice showed disordered distribution of gastric mucosal epithelial cells, loss of parietal cells, the appearance of goblet cells, inflammatory cell infiltration, and disordered glandular structures. In the metformin-treated group, the inflammatory cell infiltration was reduced, glandular structures tended to be regular, and the pathological changes were less severe than those in the Atp4a<sup>-/-</sup> model group (Figure 7A). The level of IM lesions in gastric tissues was assessed using AB-PAS staining, a technique that involves the staining of neutral mucins in red and sialomucins in blue. The presence of IM lesions in the gastric mucosa of Atp4a<sup>-/-</sup> mice was observed to be extensive compared with normal mucosa. In

Molecular name	Targets	Residues	Hydrogen bond length	Binding energy (kcal/Mol)
Metformin	EGFR	ASP-1006; GLU-1005	2.1; 1.8; 2.1	-5.95
Metformin	MMP9	ASP-131; ASP-206; ASP-207	1.8; 1.9; 2.5; 2.0; 2.5	-5.91
Metformin	HIF1A	LYS-75; ASP-77; GLU-79	2.1; 1.7; 1.9	-5.71
Metformin	HSP90AA1	GLU-62; ASP-102	2.1; 1.9	-4.67
Metformin	SIRT1	ASP-289	2.5; 2.3; 2.0	-2.53
Metformin	IL2	GLU-136; GLU-130	1.8; 1.8; 2.0; 2.3; 2.3	-6.18
Metformin	MAPK8	GLU-329; TRP-324	1.8; 2.3; 1.9	-4.82
Metformin	STAT1	GLU-657	2.1	-3.57
Metformin	PIK3CA	ASP-1017; ASP-454	2.1; 2.1; 1.9	-5.51
Metformin	ICAM1	GLU-90; ALA-178	1.8; 2.6; 1.9	-4.45
ANP	EGFR	GLN-791; MET-793; CYS-797; ALA-722; LYS-745; ASP-837; ASN-842; ARG-841	3.0; 2.2; 2.6; 2.0; 2.4; 2.8; 2.3; 2.7; 2.0	-4.54
PZE	MMP9	_	_	-2.71
SO4	HIF1A	_	_	-2.13
AB4	HSP90AA1	ASP-93; THR-104; LYS-58	2.9; 3.0; 2.5	-7.53
GOL	SIRT1	ILE-316; ILE-347	2.2; 1.8	-0.87
SO4	IL2	LYS-74	2.2	-4.14
EDO	MAPK8	GLU-272; ASP-277	3.0; 2.0	-2.29
SO4	STAT1	SER-606; LYS-587; ARG-602	2.9; 2.6; 2.3; 2.1	-5.41
799	PIK3CA	SER-854; GLN-859; VAL-851	2.8; 2.8; 2.8; 1.9	-8.93
NAG	ICAM1	ARG-13; TYR-83; THR-85	2.1; 2.1; 3.1; 3.4	-3.71

### TABLE 1 Summary of metformin and target proteins' molecular docking.

metformin-treated Atp4a<sup>-/-</sup> mice, IM lesions were regressed apparently (Figure 7B), suggesting that metformin reduced the lesion of gastric mucosa.

# 3.8 Metformin regulates the expression pattern of IM markers

To further understand the characters of IM tissues in Atp4a<sup>-/-</sup> mice and investigate the effect of metformin on IM, IHC staining was used to determine the expression of MUC5AC and CDX2 (Figure 8A). In the control group, its gastric mucosal epithelium showed strong positive expression of MUC5AC, with deep staining and wide distribution in the whole epithelium. The expression of MUC5AC in gastric mucosal epithelium of Atp4a<sup>-/-</sup> mice was significantly lower than that of the control group. After metformin treatment, the positive expression of MUC5AC increased significantly compared with the model group. The negative expression of CDX2 was observed in gastric mucosa in the control group. The expression of CDX2 in gastric mucosa of Atp4a<sup>-/-</sup> model mice was significantly increased compared with that of controls. Positive expression of CDX2 protein was significantly reduced in the metformin-treated group (Figure 8B). Thus, the

metformin treatment can effectively increase the expression of MUC5AC and reduce the expression of CDX2.

# 3.9 Metformin treatment regulated the core- and related-targets

The qRT-PCR results indicated that metformin increased the expression of MUC5AC and decreased the expression of CDX2 significantly (Figure 9A), which further verified the IHC staining results in Figure 8B. It was further proved that metformin could effectively inhibit the degree of intestinal metaplasia in the gastric mucosa of Atp4a<sup>-/-</sup> mice. Compared with the wide type mice, the mRNA expression of EGFR, MMP9, HIF1A, HSP90AA1, MAPK8, ICAM1, IL2, NF-κB, IL6, IL-1β, TNF-α, PI3KCA, AKT1, mTOR, and LDHA increased significantly, while the mRNA expression of STAT1 and FoxO3a decreased significantly in the gastric tissues of Atp4a-/- mice. Furthermore, the treatment of metformin in Atp4a<sup>-/-</sup> mice drive the expressions of these targets the other way around significantly. The expression of SIRT1 was increased significantly following the metformin intervention, although the difference between the normal and model groups was not significant (Figure 9B).



# 3.10 The effect of metformin in NF- $\kappa$ B and PI3K/AKT/mTOR/HIF-1 $\alpha$ signaling pathway

The Western blot analysis was conducted to evaluate the protein levels of NF- $\kappa$ B and PI3K/AKT/mTOR/HIF-1 $\alpha$  signaling pathway in gastric tissues of Atp4a<sup>-/-</sup> mice (Figure 10A). Compared with the control group, the expression of p-NF- $\kappa$ B, p-PI3K, p-Akt, p-mTOR, and HIF-1 $\alpha$  in IM mice increased significantly, indicating the activation of NF- $\kappa$ B and the PI3K/AKT/mTOR/HIF-1 $\alpha$  signaling pathway in the progression of IM. The results of metformin-treated group demonstrated that metformin inhibited the expression of p-NF- $\kappa$ B, p-PI3K, p-AKT, p-mTOR, and HIF-1 $\alpha$  significantly

(Figures 10B–E). These findings suggested that metformin could lower the activation of NF- $\kappa$ B and inhibit the PI3K/AKT/mTOR/ HIF-1 $\alpha$  signaling pathway in Atp4a<sup>-/-</sup> mice.

### 4 Discussion

IM-related genes were mainly concentrated in interleukins related pathways and apoptosis, while targeted genes of metformin were involved in transduction pathways and interleukins signaling pathways. Using network pharmacological analysis, 1,751 IM-related genes and 318 metformin-targeted genes



ImageJ software. (\*p < 0.05, \*\*p < 0.01, \*\*\*\*p < 0.0001).

were identified, and 99 common genes were identified. Common genes were mainly involved in the pathways in cancer, HIF-1 signaling pathway, PD-L1 expression and PD-1 checkpoint pathway in cancer, FoxO signaling pathway. Ten core targets (EGFR, MMP9, HIF1A, HSP90AA1, SIRT1, IL2, MAPK8,

STAT1, PIK3CA and ICAM1) identified in PPI analysis exhibited a strong correlation with the therapeutic effects of metformin in IM treatment. The molecular docking results demonstrated that metformin exhibited spontaneous binding to core target proteins, indicating its potential role in modulating the biological activity of



these core targets. Besides, metformin treatment regulated the expressions of core targets and related factors in the IM mice model.

Metformin can increase the radiosensitivity of cancer cells by down-regulating EGFR/PI3K/AKT (Zhang et al., 2014; Song et al., 2015). The PI3K/AKT/mTOR signaling pathway is known to facilitate the proliferation of cancer cells, and it has been observed that metformin possesses the ability to impede the activity of this pathway (Ala and Ala, 2021). The activation of the PI3K/AKT pathway has been found to induce the activation of NF- $\kappa$ B and the inactivation of FoxO3a, eventually facilitating the advancement of gastric cancer (Huang et al., 2017; Rena et al., 2017). FoxO3a can directly increase its transcriptional activity by phosphorylating AMPK (Greer et al., 2007). Furthermore, the decrease in FoxO3a expression results in a notable elevation in NF- $\kappa$ B activity and the production of inflammatory factors (Lin et al., 2004). In parallel, metformin improves the immune response to cancer cells by reducing NF- $\kappa$ B (Kheirandish et al., 2018). It inhibits the phosphorylation and nuclear translocation of the NFκB subunit p65 and suppresses the degradation of its inhibitory protein I kaB. This process leads to NF-kB being sequestered in the cytoplasm and unable to translocate to the nucleus to participate in inducing an inflammatory response, which reduces the upregulation of IL-1, IL-6, IL-2 and TNF-a (Feng et al., 2023). Metformin has been found to possess the capability to diminish the production of HIF-1a via AMPK activation, leading to a decrease in tumor metastasis (Meireles et al., 2022). Metformin can enhance the AMPK/mTOR pathway-dependent expression of STAT1, thus promoting the immunoregulatory effect. In intestinal inflammation, the drug also negatively regulates heat shock protein 90 (HSP90) (Saber and El-Kader, 2021). The previous study showed that metformin can exert inhibitory effects on cellular migration and invasion by regulating MMP9 activity (Jang et al., 2014; Cheng et al., 2022). Metformin also reduces the expression of ICAM1, thus inhibiting inflammatory signaling



and metastatic progression (Liu et al., 2014; Schexnayder et al., 2018).

Loss of H + -K + -ATPase a subunit leads to parietal cell loss resulting in loss of gastric acid, increased number of glands, and vesicle formation in Atp4a<sup>-/-</sup> mice at 12 weeks of age, while chronic inflammation leads to atrophy of the basal mucosa and development of IM (Judd et al., 2005; Liu et al., 2020). In the present study, we selected Atp4a<sup>-/-</sup> mice as the animal model and revealed metformin may have a therapeutic effect on IM in vivo, which provides firm evidence for our prediction and analysis. Studies have confirmed that insufficient expression of MUC5AC and ectopic expression of CDX2 could be observed in IM (Battista et al., 2021; Koide et al., 2022). The IHC staining results of MUC5AC and CDX2 provided further validation of the therapeutic effect of metformin in IM. After metformin treatment, the expressions of inflammatory factors such as IL-1 $\beta$ , TNF- $\alpha$ , IL-6 and IL-2 were reduced and the activation of NFκB was decreased, suggesting that metformin reduced the inflammation in Atp4a<sup>-/-</sup> mice. Additionally, Western blot results demonstrated that metformin can stop the further progression of IM by inhibiting PI3K/Akt/mTOR/HIF-1a signaling pathway. Combined with the enrichment analysis results, animal experiments indicated that metformin may participate in apoptosis and inflammation to slow down the IM process.

Although Atp $4a^{-/-}$  mice can mimic the pathological changes and molecular characteristics of IM, in extreme cases, H + -K + -ATPase  $\alpha$  subunit deletion can also cause dysfunction in other organs, which

is different from human gastric IM. While molecular docking is a valuable tool, other additional validation methods are needed to further confirm the interactions between metformin and its targets. The downstream factors of the selected signaling pathway need to be further explored. In parallel, this study provides an explanation and theoretical basis for the potential chemopreventive effect of metformin on gastric cancer in clinical trials. Clinical trials of metformin in human gastric IM patients should be performed and investigated to further validate the experimental results and lay a foundation for the clinical application of metformin.

# 5 Conclusion

This study demonstrates that metformin has a therapeutic effect on IM in Atp4a<sup>-/-</sup> mice. Combining the results of network pharmacology and molecular docking, it regulates the expressions of predicted core targets and related factors. Besides, metformin reduces inflammation and inhibits the PI3K/AKT/mTOR/HIF-1 $\alpha$ signaling pathway, which eventually slows down the IM process in inflammation and apoptosis.

### 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 authors.

### Ethics statement

The animal study was approved by the Animal Ethics Committee of the Fifth Affiliated Hospital of Zhengzhou University. The study was conducted in accordance with the local legislation and institutional requirements.

### Author contributions

RH: Conceptualization, Validation, Writing-original draft, Data curation. XX: Data curation, Writing-original draft. XS: Methodology, Writing-review and editing. YM: Writing-review and editing. HW: Validation, Writing-review and editing. HX: Validation, Writing-review and editing. FL: Writing-review and editing. PZ: Conceptualization, Funding acquisition, Supervision, Writing-review and editing. SL: Conceptualization, Funding acquisition, Supervision, Writing-review and 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/fphar.2024.1340309/ full#supplementary-material

#### SUPPLEMENTARY FIGURE S1

GO enrichment analyses of IM-related genes and metformin-targeted genes. (A) Top ten cellular components of IM-related genes. (B) Top ten molecular functions of IM-related genes. (C) Top ten biological processes of IM-related genes. (D) Top ten cellular components of metformin-targeted genes. (E) Top ten molecular functions of metformin-targeted genes. (F) Top ten biological processes of metformin-targeted gene.

#### SUPPLEMENTARY FIGURE S2

Patterns of molecular docking between the primary target proteins and the original ligands. (A) ANP- EGFR. (B) PZE-MMP9. (C) SO4- HIF1A. (D) AB4-HSP90AA1. (E) GOL- SIRT1. (F) SO4- IL2. (G) EDO- MAPK8. (H) SO4- STAT1. (I) 799- PIK3CA. (J) NAG-ICAM1.

#### SUPPLEMENTARY TABLE S1

The databases information and their websites.

#### SUPPLEMENTARY TABLE S2

The primer sequences of targets genes and related genes.

### SUPPLEMENTARY TABLE S3

The core targets proteins information.

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