Microflora Disturbance during Progression of Glucose Intolerance and Effect of Sitagliptin: An Animal Study

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Background. Emerging evidences have shown a close interplay between obesity, diabetes, and intestinal flora disturbance. Dipeptidyl peptidase-4 inhibitor, exemplified by sitagliptin, is highly efficacious in treating type 2 diabetes (T2DM), yet little is known if sitagliptin exerts beneficial effects on microbiota associated with obesity and T2DM. We evaluated changes of gut microbiota following the induction of obesity and T2DM in a streptozotocin treated high fat/high carbohydrate fed (HF/HC-STZ) rat model and explored the effect of sitagliptin on gut microbiota for HF/HC-STZ rats.

Methods. Sitagliptin was administered via oral gavage to diabetic rats. Fecal DNA extraction and 454 pyrosequencing based on analysis of 16S rRNA genes was utilized to determine the overall structure of microbiota in fecal DNA samples. Results. Results showed that, at the level of phylum, there was higher abundance of Firmicutes and Tenericutes and less abundance of Bacteroidetes in obese rats compared to their lean counterparts. At the level of genus, short-chain fatty acid- (SCFA-) producing bacteria, Blautia, Roseburia, and Clostridium, and probiotics Lactobacillus, Bifidobacterium, and so forth were identified significantly different from each other among conditions. Conclusion. Marked shifts of the gut microbiota structure were observed in the rats during development of glucose intolerance. Intestinal flora changed in the process of glucose intolerance, and treatment of sitagliptin moderately corrected the dysbiosis of microbiota in T2DM.

1. Introduction

Type 2 diabetes mellitus (T2DM) is exacting a huge level of patient suffering and social cost worldwide. Recent studies suggested that an altered composition and diversity of gut microbiota could play an important role in the development of metabolic disorders. As the second genome of human, intestinal flora has commensal relationship of mutual benefit with the host. Although not fully understood yet, the gut microbiota is implicated in various aspects of intestinal function integrity including epithelial cell turnover, immune modulation, and gastrointestinal motility. The gut microbiota also regulate energy metabolism such as breaking down dietary toxins and carcinogens, manufacturing micronutrients, fermenting indigestible food substances, facilitating the absorption of certain electrolytes and trace minerals, and governing the growth and differentiation of enterocytes through the production of short-chain fatty acids (SCFA) [1–4]. Studies have shown that obesity promotes the growth of the Firmicutes phylum and reduces the proportion of Bacteroidetes phylum in the gut [5–9]. Implantation of gut flora from obese mice to normal and germfree mice resulted in increased body weight and insulin resistance [9, 10] supporting the notion that the bacterial species from obese gut have metabolically unfavorable properties. The associations between microbiota and obesity, insulin resistance, and diabetes are presumably due to the impaired ability of the microbes to extract energy from the diet [9], altered fatty acid metabolism [11], changes in secretion of gut hormones such as peptide YY (PYY) [12], activation of lipopolysaccharide toll-like receptor-2 [13], and changes in the intestinal barrier integrity [14]. There seems to be an obvious growing interest toward the relationship between gut microbes and T2DM.

Sitagliptin, a dipeptidyl peptidase-4 inhibitor, received approval from the US FDA in 2006 for treatment of type 2 diabetes mellitus. It prevents the enzymatic degradation of glucagon-like peptide 1 (GLP-1) and glucagon-like peptide 2 (GLP-2). GLP-1 appears to increase insulin secretion,
decrease glucagon secretion, decrease hepatic gluconeogenesis, improve insulin sensitivity, and delay gastric emptying [15]. GLP-2 appears to be an intestinal specific growth factor, promoting the growth of intestinal mucosa, repairing damaged intestinal epithelium, and improving the intestinal mucosa barrier integrity [16, 17]. Since sitagliptin affects the metabolism of GLP-1 and GLP-2 secreted from enteroneodcrine, we wonder whether sitagliptin regulates intestinal flora.

Among available methodologies to study the microbial ecology of complex bacterial communities, the so-called metagenomics approach is considered to be the “golden standard” [18, 19]. In the current study, we analyzed the structure of intestinal flora in a HF/HC progressive glucose intolerance rat model with or without sitagliptin treatment by 454 pyrosequencing; we tried to explore whether and to what extent sitagliptin regulated microbiota.

2. Materials and Methods

2.1. Drug and Diets. The HF/HC diet (containing 19.8 g fat, 44.6 g carbohydrate, and 22.3 g protein per 100 g, and 40 kcal% fat, 40 kcal% carbohydrate, and 20 kcal% protein by energy) was purchased from Shanghai Laboratory Animal Center, Chinese Academy of Science (SLACCAS), Laboratory Animal Co. Ltd. (Shanghai, China). The streptozotocin (STZ) was purchased from Sigma, USA. Sitagliptin was a kind gift from MSD China.

2.2. Animal Experiments. Fifteen four-week-old male Sprague-Dawley (SD) rats (pathogen-free grade, average body weight 110 g) were purchased from SLACCAS Laboratory Animal (Shanghai, China). All rats were acclimatized in our laboratory for 7 days before the initiation of the experiment. After fasting for 12 hours, fresh stool samples were collected by stimulating the anus and immediately stored at −80°C for subsequent analysis. Then the rats were fed with HF/HC diet for 4 weeks, and stool samples were collected for the second time. After fasting for 12 hours, an OGTT test was performed to assess the insulin resistance. The rats were receiving 50% D-glucose solution by gavage at 2 g/kg body weight, and blood glucose levels were measured at 0, 30, 60, 90, and 120 minutes following the glucose challenge using blood samples collected by tail snipping. The day after OGTT test, the HF/HC fed (4 weeks on the diet) rats were injected intraperitoneally with STZ (30 mg/kg body weight) to induce diabetes. The animals were fed continuously on the HF/HC diet throughout the rest of the study. Ten rats developed diabetes with fasting blood glucose (FBG) >11.1 mmol/L at 2 weeks after STZ treatment [20, 21]. Two rats died and 3 rats failed to develop diabetes. Stool samples were collected from the rest of the 10 rats for the third time for 4 weeks after the induction of T2DM. Thereafter, sitagliptin (10 mg/kg body weight, oral gavage, once a day) was administered to all 10 diabetic rats for 12 weeks, followed by another stool sample collection (Figure S1 in Supplementary Material available online at http://dx.doi.org/10.1155/2016/2093171). Body weight was weighed and blood glucose was checked weekly with Bayer glucometer during the study.

All animal experimental protocols were approved by the Animal Ethics Committee of Animal Center of East Hospital, Tongji University, and all the animal experiments were carried out in strict accordance with the Guidelines for Care and Use of Laboratory Animals of the Animal Ethics Committee of Animal Center of East Hospital, Tongji University. All efforts were made to minimize animal suffering.

2.3. Fecal DNA Extraction and 454 Pyrophosphate Sequencing. Microbial DNA was extracted from stool samples using the E.Z.N.A. Stool DNA Kit (Omega Bio-tek, Norcross, GA, USA) according to manufacturer’s instructions. The V1–V3 regions of the bacteria 16S ribosomal DNA gene were amplified by polymerase chain reaction (95°C for 2 min, followed by 25 cycles at 95°C for 30 s, 55°C for 30 s, and 72°C for 30 s and a final extension at 72°C for 5 min) using primers 27F 5′-(CTATCCCGTGTGCGCAGACCT-3′) and 533R 5′-(CCTATCCGGGCTGCTGATC)-3′. Sequences were performed in a 20 μL mixture containing 4 μL of 5x FastPlfu Buffer, 2 μL of 2.5 mM dNTPs, 0.8 μL of each primer (5 μM), 0.4 μL of FastPlfu Polymerase, and 10 ng of template DNA. After purification with the AxyPrep DNA Gel Extraction Kit (Axygen Biosciences, Union City, CA, USA) and quantification using QuantiFluor-ST (Promega, US), the amplified mixture was used for pyrosequencing on a Roche 454 GS FLX+ Titanium platform (Roche 454 Life Sciences, Branford, CT, USA) according to standard protocols at Majorbio BioPharm Technology Co., Ltd., Shanghai, China. The raw reads were deposited into the NCBI Sequence Read Archive (SRA) database (Accession Number: SRP056522).

In total, 323,039 valid sequences were obtained from all 40 samples, with an average length of 419 bp per sequence. The resulting sequences were processed using QIIME (version 1.17). After removing sequences with average quality score <20 over a 50 bp sliding window, sequences shorter than 200 bp, sequences with homopolymers longer than six nucleotides, and sequences with ambiguous base calls or incorrect primer sequences, a total of 241,879 high-quality sequences were produced with an average length of 430 bp per sequence. Operational Taxonomic Units (OTUs) were clustered with 97% similarity cutoff using UPARSE (version 71 http://drive5.com/uparse/) and chimeric sequences were identified and removed using UCHIME. The phylogenetic affiliation of each 16S rDNA gene sequence was analyzed by RDP Classifier (Michigan State University, http://rdp.cme.msu.edu/) against the SILVA (SSU115) 16S rDNA database using confidence threshold of 70% [22].

2.4. Statistical Analysis. Comparisons of diversity estimators and the metabolic indices between conditions were analyzed by the one-way repeated measures ANOVA using SPSS version 13.0 for Windows. OTUs that reached 97% similarity level were used for diversity (Shannon), rarefaction curve and Shannon-Wiener curve analysis by using Mothur (version 1.30.1) [23]. Heatmap figure was generated using R packages gplots [24] at genus level. A differentially abundance feature was analyzed based on multiple hypotheses testing rare
Table I: Body weight and blood glucose changes during progression of glucose intolerance and after Sitagliptin treatment. Data are expressed as mean ± SD (compared with group obesity: △P < 0.01; compared with group diabetes: △△P < 0.01, *P < 0.05).

<table>
<thead>
<tr>
<th>Conditions</th>
<th>Body weight (g)</th>
<th>Blood glucose (mmol/L)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Normal</td>
<td>155.06 ± 6.4*</td>
<td>3.83 ± 0.25*</td>
</tr>
<tr>
<td>Obesity</td>
<td>398.92 ± 23.06</td>
<td>4.30 ± 0.37±</td>
</tr>
<tr>
<td>Diabetes</td>
<td>416.83 ± 35.64</td>
<td>18.81 ± 2.55*</td>
</tr>
<tr>
<td>Sitagliptin</td>
<td>432.67 ± 58.10</td>
<td>15.54 ± 1.39*</td>
</tr>
</tbody>
</table>

In this study, with an average of 6047 sequences per sample. A brief summary and the estimators of OTUs and diversity (Shannon) are shown in Figure 1, and detailed characteristics of each sample are listed in Table S1. There were statistically significant differences of Shannon indexes between obese condition and sitagliptin-treated condition (4.32±0.36 versus 4.65±0.20, P < 0.05) and between diabetic condition and sitagliptin-treated condition (3.99±0.25 versus 4.65±0.20, P < 0.01), suggesting significant higher diversity found in sitagliptin condition compared to obese or diabetic condition.

The rarefaction curves of all four conditions did not level off at the sequencing depth of 6,000 (Figure S2). Therefore, this sequencing depth was not sufficient to cover the whole bacterial diversity. Thus, it deserves further sequencing to detect more bacterial species in the rat feces. However, when the Shannon curves tended to be smooth, the sequencing data quantity became dramatically bigger, which reflected the vast majority of microbes in the sample (Figure S3).

3.4. Microbial Structures Differed Significantly among Conditions. There were significant differences at the phylum level in rat fecal microbiota during progression of glucose intolerance and after treatment with sitagliptin. Figure 2 demonstrated that Bacteroidetes and Firmicutes accounted for the majority of microflora (>90%) in each condition. The relative abundance of Firmicutes in obesity condition was significantly increased compared to that in the normal condition (86.28% versus 66.24%, P < 0.01), while the relative abundance of Bacteroidetes decreased substantially (12.16% versus 31.64%, P < 0.01). There was no significant difference between obesity and diabetes at phylum level. The relative abundance of Firmicutes in the sitagliptin condition was significantly less than that in the diabetic condition (63.19% versus 83.56%, P < 0.01). In contrast, the relative abundance of Bacteroidetes increased dramatically (32.46% versus 16.06%, P < 0.01). It is noteworthy that the relative abundance of Tenericutes in the obese condition significantly increased compared to the normal condition (1.15% versus 0.61%, P < 0.05) but decreased dramatically after inducing diabetes (1.15% versus 0.09%, P < 0.01) and then increased largely after sitagliptin treatment as compared to the diabetic condition.
Table 2: Statistical analysis was performed by Wilcoxon Signed-Rank test. Data of normal, obesity, diabetes, and sitagliptin are relative abundance (percentage) of all sequences in each group ($P^\#$ group obesity versus normal; $P^*$ group diabetes versus obesity; $P^{\Delta}$ group diabetes versus sitagliptin).

<table>
<thead>
<tr>
<th>Taxonomy</th>
<th>Normal (%)</th>
<th>Obesity (%)</th>
<th>Diabetes (%)</th>
<th>Sitagliptin (%)</th>
<th>$P^#$</th>
<th>$P^*$</th>
<th>$P^{\Delta}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bacteroides (genus)</td>
<td>1.4</td>
<td>0.2</td>
<td>0.6</td>
<td>5.4</td>
<td>0.013</td>
<td>0.022</td>
<td>0.005</td>
</tr>
<tr>
<td>Prevotellaceae (family)</td>
<td>9.9</td>
<td>1.8</td>
<td>6.8</td>
<td>9.2</td>
<td>0.005</td>
<td>0.001</td>
<td></td>
</tr>
<tr>
<td>Blautia (genus)</td>
<td>0.2</td>
<td>0.8</td>
<td>7.1</td>
<td>0.3</td>
<td>0.012</td>
<td>0.005</td>
<td>0.005</td>
</tr>
<tr>
<td>Roseburia (genus)</td>
<td>0.7</td>
<td>0.9</td>
<td>0.3</td>
<td>0.5</td>
<td>0.042</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ruminococcus (genus)</td>
<td>2.5</td>
<td>1.4</td>
<td>0.6</td>
<td>1.9</td>
<td>0.025</td>
<td>0.007</td>
<td></td>
</tr>
<tr>
<td>Clostridium (genus)</td>
<td>1.3</td>
<td>4.6</td>
<td>0.8</td>
<td>0.7</td>
<td>0.014</td>
<td>0.013</td>
<td></td>
</tr>
<tr>
<td>Lactobacillus (genus)</td>
<td>22.3</td>
<td>22.5</td>
<td>13.0</td>
<td>6.7</td>
<td>0.028</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bacillus (genus)</td>
<td>4.6</td>
<td>4.4</td>
<td>6.0</td>
<td>0</td>
<td>0.005</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bifidobacterium (genus)</td>
<td>0.024</td>
<td>0.005</td>
<td>0.002</td>
<td>0.085</td>
<td>0.007</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Parabacteroides (genus)</td>
<td>0.2</td>
<td>0.02</td>
<td>0.1</td>
<td>0.6</td>
<td>0.008</td>
<td>0.024</td>
<td>0.005</td>
</tr>
</tbody>
</table>

Figure 2: Relative abundance of different bacterial phyla in microbiota of four conditions.

Figure S4) which showed the significant differences of Blautia, Roseburia, Clostridium, and so on, among four conditions. There were significant variations in the composition of intestinal bacteria species among four study conditions. A cladogram representation of the microbiota structure of four conditions and their predominant bacteria performed by LEfSe and the greatest differences in taxa among four communities are shown in Figure 4 and Figure S5.

To compare the overall microbiota structures in all conditions, the unweighted Unifrac distance matrix was calculated based on the OTUs of each sample. The PCA result revealed a significant difference in gut bacterial structure among all conditions. The three principal component scores accounted for 30.57%, 20.06%, and 8.6% of total variations, respectively (Figure 5).

4. Discussion

In the study, we explored the changes of rat intestinal microbiota during progression of glucose intolerance and the effect condition (0.96% versus 0.09%, $P < 0.01$). Similar changes were observed at the phylum of Proteobacteria.

Furthermore, we explored the similarities and distinctions of species distribution in all conditions. As shown in Figure 3, we found that there were 441 species shared in the four conditions, accounting for around half of the OTUs in each condition. It is noteworthy that about one-third of the species only found in normal (64 OTUs) or obesity (22 OTUs) condition belonged to Lachnospiraceae, which is capable of fermenting complex carbohydrates to SCFAs playing an important role in maintaining intestinal homeostasis [27, 28]. Interestingly, one-third of the total species only found in sitagliptin condition (94 OTUs) belonged to Ruminococaceae, which is a major utilizer of plant polysaccharides [29, 30].

In addition, Metastat analysis showed significant differences of relative abundance at the level of genus or family among conditions (Table 2 and Table S2). And the result at level of genus was marked in the Heatmap plot (Figure
Figure 4: Taxonomic representation of statistically and biologically consistent differences among condition normal, obesity, diabetes, and sitagliptin. Differences are represented by the color of the most abundant class (red indicating normal condition, green obesity condition, blue diabetes condition, purple sitagliptin condition, and yellow nonsignificance). The diameter of each circle is proportional to the taxon's abundance.
of sitagliptin on microbiota. We found significantly more Firmicutes and less Bacteroidetes in obese rats compared to their lean counterparts, similar to previous studies [31] despite conflicting data showing no differences of Bacteroidetes and Firmicutes between obese and lean subjects [32–34]. The increased Firmicutes in obese mice may be related to the genes encoding enzymes that break down polysaccharides which cannot be digested by the host, increasing the production of monosaccharides and short-chain fatty acids (SCFA) and the conversion of these SCFA to triglycerides in the liver. The binding of SCFA to two G-protein-coupled receptors (GPR41 and GPR43) induces peptide YY secretion, which suppresses gut motility, retards intestinal transit, and subsequently results in the increase of nutrient uptake and deposition [35]. Interestingly, we also found that hydrogen-producing Prevotellaceae decreased significantly in the obese
condition. Given the fact that hydrogen inhibits digestion, the obese rats seemed likely to absorb more calories from similar energy intake.

Another interesting finding was the higher abundance of the phylum Tenericutes (class Mollicutes) in obese rats compared to their lean counterparts. Certain species of Mollicutes bloom have been proved to evolve the capacity to import certain types of carbohydrates common in westernized diet for both mice and human being (e.g., glucose, fructose, and sucrose) and to metabolize these imported sugars to SCFA which could be readily absorbed by the host [36]. Sitagliptin did restore the structure of gut microbe at the level of phylum, similar to the lean control condition without significant effects on body weight.

Unweighted Unifrac PCA analysis confirmed above results in the respect of overall microbial structures. The intestinal microbiota of the four conditions of rats were structurally separated from each other in the three principal components.

Various mechanisms have been proposed to explain the influence of microbiota on insulin resistance and T2DM including metabolic endotoxemia, modifications in the secretion of the incretins, and butyrate production. Decrease of SCFA-producing bacteria has been commonly observed in metabolic diseases including T2DM [37] and even colorectal cancer [38]. SCFA-producing bacteria have been previously shown to benefit the host through protecting the mucosa from damage induced by pathogens, supplying colonocyte nutrients, mitigating inflammation, and so forth [39, 40]. Recent studies in mice have shown that an increase in colonic production of short-chain fatty acids triggers intestinal gluconeogenesis (IGN) via complementary mechanisms. Butyrate activates IGN gene expression through a cAMP-dependent mechanism in enterocytes, whereas propionate, itself a substrate of IGN, activates IGN gene expression via the portal nervous system and the fatty acid receptor FPR3. In rodents, the result of increased IGN is beneficial to glucose and energy homeostasis with reductions in hepatic glucose production, appetite, and body weight [41]. Belonging to either Clostridium, Eubacterium, or Fusobacterium, the butyrate-producing bacteria mainly exist in cecum and colon. Fusobacterium prausnitzii and Roseburia intestinalis are predominant species in butyric acid producing bacteria in human and animal intestine [42]. At the level of genus, our data suggested that the relative abundance of butyrate-producing Clostridium and Roseburia in the diabetic rats decreased significantly compared to the lean and obese conditions. Moreover, we also observed that the SCFA-producing bacteria Blautia increased significantly in the diabetic condition compared to the obese condition. Becker et al. demonstrated that the addition of Clostridium butyricum as a second butyrate-producing bacterium to SIHUMI (a simplified human intestinal microbiota: Anaerostipes cacaee, Bacteroides thetaiotaomicron, Bifidobacterium longum, Blautia producta, Clostridium ramosum, Escherichia coli, and Lactobacillus plantarum) led to an increase of butyrate by 56% but decrease of Bifidobacterium longum, Blautia producta and E. coli, indicating that either of these organisms competes with Clostridium butyricum for the same substrates or Clostridium butyricum forms inhibitory substances [43]. Thus it seems reasonable to suppose that changes of Clostridium, Roseburia, and Blautia show an opposite trend. In other words, the increase of SCFA-producing bacteria like Blautia leads to decrease of butyrate-producing bacteria like Clostridium and Roseburia. In the current study, the structure of SCFA-producing bacteria seemed to be abnormal in diabetic rats. After sitagliptin treatment, Roseburia increased and Blautia decreased while Clostridium showed no change.

Probiotics such as Lactobacillus and Bifidobacterium normally reside in the intestinal tract and regulate gut microflora and mucosal immunity. Recent data also suggest that microflora, especially probiotics, exert beneficial effects on gastrointestinal discomforts such as diarrhea, abdominal pain, and bloating [44]. Lactobacillus produces lactic acid, CO₂, acetic acid, and/or ethanol which may contribute to a more acidic environment through homo- or heterofermentative metabolism [45, 46]. Our data showed that Lactobacillus and Bifidobacterium decreased in the diabetic condition. However, sitagliptin made Bifidobacterium increased with no significant effect on Lactobacillus, which seemed to be consistent with the notion we discussed above.

Marked changes in microbiota composition after metformin treatment were observed in high-fat diet- (HFD-) fed conditions, especially Akkermansia belonging to the Verrucomicrobia phylum showed the most conspicuous changes [47], suggesting a possible interaction between HFD, metformin, and intestinal microbiota. Forslund et al. found significant increase of Escherichia spp. and decrease of Intestinibacter spp. in human gut microbiome in metformin-treated T2DM; especially the latter is resistant to oxidative stress and able to degrade fucose, indicative of an indirect involvement in mucus degradation [41]. Interestingly, there are consistent data from numerous studies showing that sitagliptin results in reduced gastrointestinal discomfort when combined with metformin, a classic antihyperglycemia agent also known for its gastrointestinal side effects. For instance, Reasner et al. showed that sitagliptin and metformin fixed dose combination resulted in a significant reduction on abdominal pain and diarrhea compared to metformin monotherapy [48]. In addition, a 104-week clinical study (Harmony) also showed that diarrhea incidence was reduced in sitagliptin plus metformin group as compared to metformin monotherapy group [49]. Though the underlying mechanism of effect of metformin and sitagliptin on intestinal flora remains unclear, our results seem to provide a possible explanation arguing that this effect could be related to favorable changes of microflora after sitagliptin treatment.

We supposed that the effect of sitagliptin on microbiota was related to GLP-2, which helps improve the intestinal mucosa barrier integrity. Besides, recent studies found that DPP-4 may act as the protease activated receptor 2 (PAR2) agonists involved in the pathophysiology of PAR2, leading to proliferation and inflammation in smooth muscle cells. Thus DPP-4 inhibitor can relieve edema of intestinal wall and alleviate the intestinal inflammation [50, 51]. Integrated intestinal mucosa barrier and improved intestinal environment may be not fit for bacteria related with obesity and diabetes or opportunistic pathogen to colonize.
In summary, our data highlights the potential beneficial effect of sitagliptin on gut microbe of diabetic or obese animals, which constituted a novel, exciting observation calling for further investigation on numerous involved microbiota species. These data, when combined with the results of future clinical studies, may facilitate the development or optimization of novel microbiota based T2DM interventional strategies.

5. Conclusions

Taken together, our data suggest that the structure of gut flora, especially SCFA-producing bacteria, seems to be abnormal in a progressive glucose intolerance rat model. Treatment with sitagliptin improved gut dysbiosis moderately. To the best of our knowledge, our study seems to be the first report showing that a DPP-4 inhibitor may exert positive regulation on gut flora dysregulated during progression of glucose intolerance and obesity in an animal model. Future research will be directed to elucidate whether the favorable changes of microflora contribute to better metabolic control and reduced gastrointestinal side effects of DPP4 inhibitor when combined with metformin.

Disclosure

The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the paper.

Competing Interests

The authors declare no competing financial interests.

Authors’ Contributions

Bo Feng designed the study; Xinfeng Yan, Peicheng Li, and Zhaosheng Tang performed the experiments; Bo Feng, Xinfeng Yan, Zhaosheng Tang, and Lin Wang analyzed the data and contributed to the discussion for interpreting the data. Bo Feng and Xinfeng Yan wrote and revised the paper. Xinfeng Yan prepared all the figures and tables. All authors reviewed the paper.

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