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Differently Expressed Genes (DEGs) Relevant to Type 2 Diabetes Mellitus Identification and **Pathway Analysis via Integrated Bioinformatics** Analysis

Idy Design A Collection B al Analysis C rpretation D reparation E ure Search F Collection G	DE DF AD AB ADFG	Ran Zhao Hua Xu Xue Liu Shumiao Zhao Hongwei Ma	P.K. China	
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Background: Material/Methods: Results:		The aim of this study was to evaluate the differently expressed genes (DEGs) relevant to type 2 diabetes mel- litus (T2DM) and pathway by performing integrated bioinformatics analysis. The gene expression datasets GSE7014 and GSE29221 were downloaded in GEO database, and DEGs from type 2 diabetes mellitus and normal skeletal muscle tissues were identified. Biological function analysis of the DEGs was enriched by GO and KEEG pathway. A PPI network for the identified DEGs was built using the STRING database. Thirty top DEGs were identified from 2 datasets: GSE7014 and GSE29221. Of the 30 top DEGs, 20 were up- regulated and 10 were down-regulated. The 20 up-regulated genes were enriched in regulation of mRNA, pro- tein biding, and phospholipase D signaling pathway. The 10 down-regulated genes were enriched in telomere maintenance via semi-conservative replication, AGE-RAGE signaling pathway in diabetic complications, and in- sulin resistance pathway. In the PPI network of 20 up-regulated DEGs, there were 40 nodes and 84 edges, with an average node degree of 4.2. For the 10 down-regulated DEGs, we found a total of 30 nodes and 105 edges, with an average node degree of 7.0 and local clustering coefficient of 0.812. Among the 30 DEGs, 10 hub genes (CNOT6L, CNOT6, CNOT1, CNOT7, RQCD1, RFC2, PRIM1, RFC4, RFC5, and RFC1) were also identified through		
Conclusions:		DEGs of T2DM may play an essential role in disease development and may be potential pathogeneses of T2DM.		
MeSH Keywords: Diabetes Mellitus, Type 2 • Genes, vif • Local Area Networks				
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Background

Skeletal muscle is key metabolic tissue of the human body and accounts for most glucose absorption and utilization [1]. Skeletal muscle is essential for the balance of glucose metabolism [2,3]. Insulin resistance of skeletal muscle is an important factor in T2DM. Therefore, identification of DEGs of skeletal muscle tissue expressed differently between patients with T2DM and healthy controls can provide essential information on the pathogenesis of T2DM [4]. In recent years, with the development and application of gene chip technology, a large amount of gene expression profile data has been generated [5,6]. How to extract valuable information from these data sets has become an important bioinformatics research topic. In the present work, the differently expressed genes (DEGs) relevant to T2DM and pathway were analyzed through integrated bioinformatics analysis to elucidate the pathogenesis of T2DM.

Material and Methods

DEGs identification form datasets GSE7014 and GSE29221

The gene expression profile datasets GSE7014 [7] and GSE29221 [8] were downloaded from the GEO database, and DEGs from T2DM and normal skeletal muscle samples were identified. The included 2 datasets were tissue samples from human skeletal muscle of T2MD patients versus healthy subjects. For the GSE7014 dataset, gene expression of a total of 26 skeletal muscle samples, including 20 samples from T2DM patients and 6 healthy subjects were analyzed using the GPL570 [HG-U133_Plus_2] Affymetrix Human Genome U133 Plus 2.0 Array. For the GSE29221 dataset, gene expression of 24 samples, including 12 skeletal muscle samples of T2MD and 12 normal skeletal muscle samples were analyzed using GPL6947 Illumina Human HT-12 V3.0 expression bead chips. The differently expressed genes of the 2 datasets (GSE7014 and GSE29221) were first screened independently with the restriction of FC>5 and p<0.01, then the differently expressed genes of GSE7014 and GSE29221 were merged to identify the overlapping DEGs.

Functional and pathway analysis of the DEGs

The biological function and pathway were analyzed by GO and KEGG enrichment. GO enrichment includes 3 aspects: biological process, cellular component, and molecular function. The enrichment of GO and KEGG was expressed with a bubble plot showing the gene ratio, gene count, gene function description, and p value.

PPI and hub genes identification

We used the STRING database in the construction of the PPI network for the 30 identified DEGs with the restriction of a minimum required interaction score of 0.4. The PPI network active interaction sources were identified from text mining, experiments, databases, co-expression, neighborhood, gene fusion, and co-recurrence. Hub genes of the 30 DEGs were identified by Cytohubba ranking method.

Results

DEGs identification

We used 2 data sets (GSE7014 and GSE29221) for bioinformatics analysis. In GSE7014, there were a total of 26 skeletal muscle samples, including 20 samples from T2DM patients and 6 normal samples. In GSE29221, there were 24 samples, including 12 skeletal muscle tissues of T2DM patients and 12 normal skeletal muscle samples. The DEGs were initially screened independently from GSE7014 and GSE29221 with the restriction of FC >5 and p<0.01. We identified 30 DEGs from 2 datasets (GSE 7014 and GSE29221) (Figure 1). Of the 30 top DEGs, 20 were up-regulated (Table 1) and 10 were down-regulated (Table 2). The 30 top DEGs identified from T2DM samples and normal skeletal muscle samples through GSE 7014 and GSE29221 are represented by a heat map in Figure 2.

GO and KEGG enrichment of the 30 DEGs

Genes ontology biological functional evaluation indicated that the 20 up-regulated genes were mainly enriched in regulation of mRNA process for biological process, protein biding related to molecular function, and cytosol for cellular component (Figure 3A–3C). For KEGG pathway evaluation, the 20 up-regulated genes were mainly enriched in RNA degradation and phospholipase D signaling pathway (Figure 3D). For the 10 down-regulated genes, GO was mainly enriched in telomere maintenance via semi-conservative replication, DNA binding, and DNA replication factor C complex (Figure 4A–4C). For KEGG evaluation, the 10 down-regulated genes were mainly enriched in AGE-RAGE signaling pathway in diabetic complications and insulin resistance (Figure 4D).

PPI hub genes identification

A PPI network for the DEGs was constructed using STRING (Figure 5A). In the PPI network of 20 up-regulated DEGs, there were 40 nodes and 84 edges, with an average node degree of 4.2 and local clustering coefficient of 0.684, which indicated statistically significant PPI enrichment (p<0.001). For the 10 down-regulated DEGs, there were 30 nodes and 105 edges,



Figure 1. (A–C) Identification of DEGs in GSE29221 and GSE7014 data series (A: Volcano plot GSE29221 in identification DEGs;
 B: Volcano plot GSE7014 in identification DEGs; C: Venn plot in identification overlap DEGs of GSE29221 and GSE7014 data series).

with an average node degree of 7.0 and local clustering coefficient of 0.812, which also demonstrated significant PPI enrichment (p<0.001) (Figure 5C). Ten hub genes (CNOT6L, CNOT6, CNOT1, CNOT7, RQCD1, RFC2, PRIM1, RFC4, RFC5, and RFC1) were also identified through Cytoscape (Figure 5B, 5D).

Discussion

Type 2 diabetes (T2DM) is a metabolic disorder characterized by impaired glucose uptake in muscle and fat, altered glucoseinduced insulin secretion, and increased hepatic glucose production, which lead to hyperglycemia [9–11]. There are now about 410 million people with T2DM globally, and this number is projected to increase to over 640 million by the year 2040 [12]. T2DM has become a heavy burden around the world [13,14]. Studies have demonstrated that the development of T2DM is a result of multi-gene interaction [15–17]. The extensive application of gene expression profiles in diabetes mellitus raises the possibility of evaluating the pathogenesis of T2DM using bioinformatics analysis.

In this work, we identified 30 DEGs (20 up-regulated and 10 down-regulated genes) from 2 datasets: GSE7014 and GSE29221. The 30 DEGs were generally enriched in the aspects of mRNA regulation, protein biding and phospholipase D signaling pathway, telomere maintenance via semi-conservative replication, AGE-RAGE signaling pathway in diabetic complications, and insulin resistance pathway. Ten hub genes (CNOT1, CNOT6, CNOT6L, CONT7, RQCD1, RFC1, RFC2, RFC2, RFC4, and RFC5) were screened according to the PPI network constructed in Cytoscape. CNOT has a scaffolding component essential for

 Table 1. Twenty up-regulated genes identified through datasets GSE7014 and GSE29221.

Gene symbol	Gene name	Fold change	P value
CNOT7	CCR4-NOT transcription complex subunit 7	3.771885	3.63E-03
IGFBPL1	Insulin-like growth factor binding protein-like 1	3.767067	2.50E-05
MAPT	Microtubule-associated protein tau	3.557061	2.33E-04
HILPDA	Hypoxia inducible lipid droplet-associated	3.50515	8.78E-04
TNKS2	Tankyrase 2	3.383897	9.31E-04
NIPSNAP3B	Nipsnap homolog 3B	3.046473	6.37E-03
CTU2	Cytosolic thiouridylase subunit 2	2.931942	7.95E-03
ANKH	Ankyrin repeat and KH domain containing 1	6.309422	2.13E-03
TPM3	Tropomyosin 3	2.873881	3.67E-03
ARF6	ADP ribosylation factor 6	2.81937	1.37E-03
TNRC6B	trinucleotide repeat containing 6B	2.791015	4.94E-05
ZNF451	Zinc finger protein 451	2.645071	1.04E-03
N4BP2L1	NEDD4 binding protein 2-like 1	2.604427	4.39E-03
MALAT1	Metastasis associated lung adenocarcinoma transcript 1 (non-protein coding)	2.535709	7.52E-03
HYAL2	Hyaluronoglucosaminidase 2	2.444833	3.36E-03
UACA	Uveal autoantigen with coiled-coil domains and ankyrin repeats	2.436325	8.59E-03
CCDC85B	Coiled-coil domain containing 85B	2.369664	7.11E-03
TEAD3	TEA domain transcription factor 3	2.292119	1.74E-03
CREB3	cAMP responsive element binding protein 3	2.243004	8.84E-04
EPHX1	Epoxide hydrolase 1	2.213641	4.26E-03

Table 2. Ten down-regulated genes identified through datasets GSE7014 and GSE29221.

Gene symbol	Gene name	Fold change	P value
LTF	Lactotransferrin	-2.2336	2.15E-03
ZDHHC9	Zinc finger DHHC-type containing 9	-2.36458	9.97E-03
PLXNC1	Plexin C1	-2.44666	1.10E-03
RFC2	Replication factor C subunit 2	-2.55141	2.48E-03
IL6R	Interleukin 6 receptor	-2.59581	2.74E-04
BRD4	Bromodomain containing 4	-2.83271	4.88E-03
WNK1	WNK lysine deficient protein kinase 1	-2.93786	6.86E-03
SCAI	Suppressor of cancer cell invasion	-3.95926	5.49E-03
NR4A2	Nuclear receptor subfamily 4 group A member 2	-4.06962	1.55E-03
FECH	Ferrochelatase	-5.19389	3.18E-04

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Figure 3. (A–D) Bubble plot of Go term and KEGG enrichment of the 20 up-regulated DEGs. (A: Biology process enrichment of the 20 up-regulated DEGs; B: Molecular function enrichment of the 20 up-regulated DEGs; C: Cellular component enrichment of the 20 up-regulated DEGs; D: KEGG pathway enrichment of the 20 up-regulated DEGs).



Figure 4. (A–D) Bubble plot of Go term and KEGG enrichment of the 10 down-regulated DEGs. (A: Biology process enrichment of the 10 down-regulated DEGs; B: Molecular function enrichment of the 10 down-regulated DEGs; C: Cellular component enrichment of the 10 down-regulated DEGs; D: KEGG pathway enrichment of the 10 down-regulated DEGs).





Figure 5. (A–D) PPI network and hub genes identification of the DEGs (A: PPI network of the 20 up-regulated DEGs; B: 5 hub genes identified for the 20 up-regulated DEGs by Cytoscape; C: PPI network of the 10 down-regulated DEGs; D: 5 hub genes identified of the 10 down-regulated DEGs by Cytoscape).

mRNA degradation [18], but its correlation with T2DM has not been previously reported. Only 1 study, by Franco [19], demonstrated that CNOT1 *de novo* missense mutation can result in a syndrome of pancreatic agenesis in mice. According to Franco [19] and our findings, CNOT1 may be a potential target for T2DM treatment. RFC is a replication factor C subunit that binds to the primer-template junction, PO-B transcription element, and other GA-rich DNA sequences [20]. RFC plays a role in DNA transcription regulation and in DNA replication and/or repair [21,22], but its correlation with T2DM had not been reported previously. The present study if the first to report that RFC is correlated with T2DM development, but our findings need to be confirmed by *in vitro* and *in vivo* experiments.

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Conclusions

DEGs in skeletal muscle between T2DM patients and healthy controls can be identified through integrated bioinformatics analysis. New genes that were not found to be correlated with T2DM can also be identified and may provide new information about T2MD development and help develop better treatments.

Conflict of interest

None.

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