

Integrated analysis of differentially expressed genes in esophageal squamous cell carcinoma using bioinformatics

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Esophageal squamous cell carcinoma (ESCC) is a deadly disease. To identify key genes in esophageal squamous cell carcinoma, we followed a strategy utilizing the larger microarray dataset (GSE38129) as the training set and another independent microarray dataset (GSE20347) as the validation set. Following quality control, differentially expressed genes (DEGs) were obtained using R software. Functional enrichment analysis was performed using DAVID database and the DEG co-expression network was established with Weighted Gene Co-Expression Network Analysis (WGCNA) and visualized by Cytoscape. The prognosis-related hub genes were then identified by Kaplan-Meier analysis based on the TCGA database. A total of 188 DEGs were obtained; 88 up-regulated genes and 100 down-regulated. The up-regulated DEGs were significantly associated with extracellular matrix organization and disassembly while down-regulated DEGs were significantly related to keratinocyte differentiation. Blue and turquoise co-expression modules were established and 18 hub genes were identified. The blue module was associated with mitotic nuclear division, cell division and mitotic cytokinesis and the turquoise module was associated with collagen catabolic process, extracellular matrix organization and keratinocyte differentiation. We established that the TPX2, CDK1 and CEP55 blue module hub genes were associated with relapse-free survival, and our overall results not only identify key genes but also provide potential novel biomarkers for ESCC diagnosis and treatment.

Key words: esophageal squamous cell carcinoma, differential expression genes, functional enrichment analysis, WGCNA, Kaplan-Meier analysis

Esophageal cancer is the 8th most common and 6th most fatal cancer worldwide [1]. It is a deadly disease with roughly 480,000 new patients every year [2]. The esophageal squamous cell carcinoma (ESCC) is histologically the most prevalent type of esophageal cancer with increasing morbidity [3]. Lacking effective early diagnosis, ESCC is usually detected at an advanced stage when the patients cannot swallow solid foods anymore with poor prognosis and clinical outcome. Thus, further research is needed to discover more effective diagnostic methods in early stage and the molecular mechanisms underlying ESCC in order to improve prevention and prognosis.

Recently, molecular markers are thought to be predictive and prognostic markers in ESCC [4]. As for immunohistochemical result, a systemic review demonstrated that the survival rate of ESCC patients with HER2-positive expression decreased because of radiation resistance possibly [5]. Additionally, ESCC patients with abnormal p53 expression show several times more rapid progression of the disease [6]. In terms of blood-based markers, Shimada et al found that the

combination of four antibodies (SURF1, HOOK2, LOC146223 and AGENCOURT_7565913) were highly specific for ESCC [7]. Moreover, tubulin beta chain, filamin A alpha isoform 1 and cytochrome b-c1 complex subunit 1 were identified as the differentially expressed proteins and biomarkers in ESCC patients serum [8]. According to epigenetic markers, the hypermethylation of CDKN2A, MGMT, APC, DAB2, CDH1 and DACT2 were found to be more frequent in ESCC tissues [9]. However, these biomarkers are still not effective for early diagnosis and prognosis prediction. Accordingly, it is worth exploring additional biomarkers for more effective diagnosis.

Microarray technology is a widely used tool to explore genetic alteration during tumorigenesis [10]. In our study, we followed a strategy utilizing the larger microarray dataset as training set and another independent microarray dataset as the validation set. The quality control was performed on both datasets in order to increase credibility of result. Functional enrichment and co-expression network analysis were applied for DEGs while survival analysis was used to identify key genes in ESCC.

Materials and methods

Microarray data. The GSE38129 [11] and GSE20347 [12] gene expression profiles were downloaded from the Gene Expression Omnibus [13] which is based on the Affymetrix Human Genome U133A 2.0 Array platform. The GSE38129 dataset included 30 ESCC tissues and 30 pairs of normal esophageal squamous epithelium tissue (N) and GSE20347 comprised 17 ESCC tissues and 17 pairs of N tissue. We utilized the larger dataset (GSE38129) as the training set and an independent dataset (GSE20347) as the validation set.

Quality control and DEG analysis. Quality control included relative logarithmic expression (RLE), RNA degradation and principal component analysis (PCA). Differentially expressed genes (DEGs) in ESCC tissue compared with N tissue were identified by Linear Models for Microarray Analysis (Limma) package in R software [14], and the Benjamini-Hochberg false discovery rate corrected P values. Adjusted p-value at <0.05 and fold change (FC) ≥ 4 formed DEG analysis cut-off criteria. Finally, hierarchical clustering analysis and DEG heatmap in GSE38129 and GSE20347 was conducted by heatmap package in R software.

Functional enrichment analysis of DEGs. Gene Ontology [15] and Kyoto Encyclopedia of Genes and Genomes [16] are widely used in bioinformatics to identify the most correlative biological process (BP) and relevant pathway informa-

tion. KEGG and GO BP analysis through the Database for Annotation Visualization and Integrated Discovery online tool (DAVID; david.ncifcrf.gov/) identified DEG biological significance [17]. P-value adjusted by Benjamini-Hochberg to <0.05 established the cut-off criteria.

Co-expression module detection. Weighted gene co-expression network analysis (WGCNA) [18] was restricted to training dataset DEGs by the R software WGCNA package. An unsupervised co-expression relationship was constructed by adjacency matrix of connection strengths with Pearson's correlation coefficients. The adjacency was defined by a soft threshold β selected to amplify strong connections between genes and penalize weak ones. Herein, the soft threshold was set at $\beta=18$ with scale-free topology criterion and the modules were identified as gene sets with a high topologic overlap. The top 10% of genes with the highest network connectivity were identified as module hub genes. Average linkage hierarchical clustering comprised topological overlap matrix based dissimilarity measure, and minimum gene module size of 30 cut the branches. Finally, each module's co-expression network was presented by Cytoscape 3.4.0 [19].

Survival analysis of hub genes. The mRNA transcript per million (TPM) of 96 ESCC tissue samples and corresponding patient follow-up information were downloaded from UCSC Xena (<https://xenabrowser.net/datapages/?host=https://tcga.xenahubs.net>). The survival R package [20] was subjected to survival analysis to explore the prognosis value of hub genes. Kaplan-Meier survival curves were plotted and relapse-free survival (RFS) provided survival endpoints. Patients with ESCC were divided into low and high expression groups according to the median of each hub gene expression, with logrank p-value <0.05 significant.

Table 1. The significant enriched GO BP terms and KEGG pathways.

	Description	No. enriched genes	Adjusted p-value
Upregulated			
	GO:0030574 Collagen catabolic process	12	1.20×10^{-11}
	GO:0030198 Extracellular matrix organization	14	6.08×10^{-9}
	GO:0001501 Skeletal system development	11	4.15×10^{-7}
	GO:0022617 Extracellular matrix disassembly	9	8.00×10^{-7}
	GO:0007067 Mitotic nuclear division	11	7.00×10^{-5}
	GO:0030199 Collagen fibril organization	6	1.87×10^{-4}
	GO:0051301 Cell division	10	7.12×10^{-3}
	GO:0035987 Endodermal cell differentiation	4	3.20×10^{-2}
	GO:0000281 Mitotic cytokinesis	4	3.52×10^{-2}
	GO:0008283 Cell proliferation	9	3.83×10^{-2}
	hsa04512 ECM-receptor interaction	8	2.86×10^{-5}
	hsa05146 Amoebiasis	8	5.52×10^{-5}
	hsa04510 Focal adhesion	9	3.25×10^{-4}
	hsa04974 Protein digestion and absorption	6	2.13×10^{-3}
	hsa04151 PI3K-Akt signaling pathway	9	7.34×10^{-3}
	hsa05202 Transcriptional misregulation in cancer	6	2.73×10^{-3}
Downregulated			
	GO:0030216 Keratinocyte differentiation	7	1.81×10^{-3}

GO BP, Gene Ontology Biologic Process. KEGG, Kyoto Encyclopedia of Genes and Genomes.

Results

Quality control and DEG analysis. Quality control included RLE, RNA degradation curve and PCA. The RLE in Figures 1A and 2A for both results showed high normalization level in the samples. The constantly rising RNA degradation curve in both results demonstrated the samples' undegraded RNA (Figures 1B, 2B) and PCA determined no absolute separation between different samples in either dataset, and thus imperfect quality (Figures 1C, 2V). Following integrated analysis of all of quality control processes, only 27 pairs of samples in GSE38129 and 16 in GSE20347 were retained and the following were excluded; GSM509791 and GSM509808 in GSE20347 and GSM935156, GSM935157, GSM935174, GSM935175, GSM935186 and GSM935187 in GSE38129. The PCA result of residual samples in GSE38129 and GSE20347 are presented in Figures 1D and 2D. The total of 188 DEGs obtained by Limma package comprised 88 up-regulated DEGs and 100 down-regulated and results are detailed in Supplementary Table 1. The DEGs in training and validation set heatmaps are listed in Figure 3.

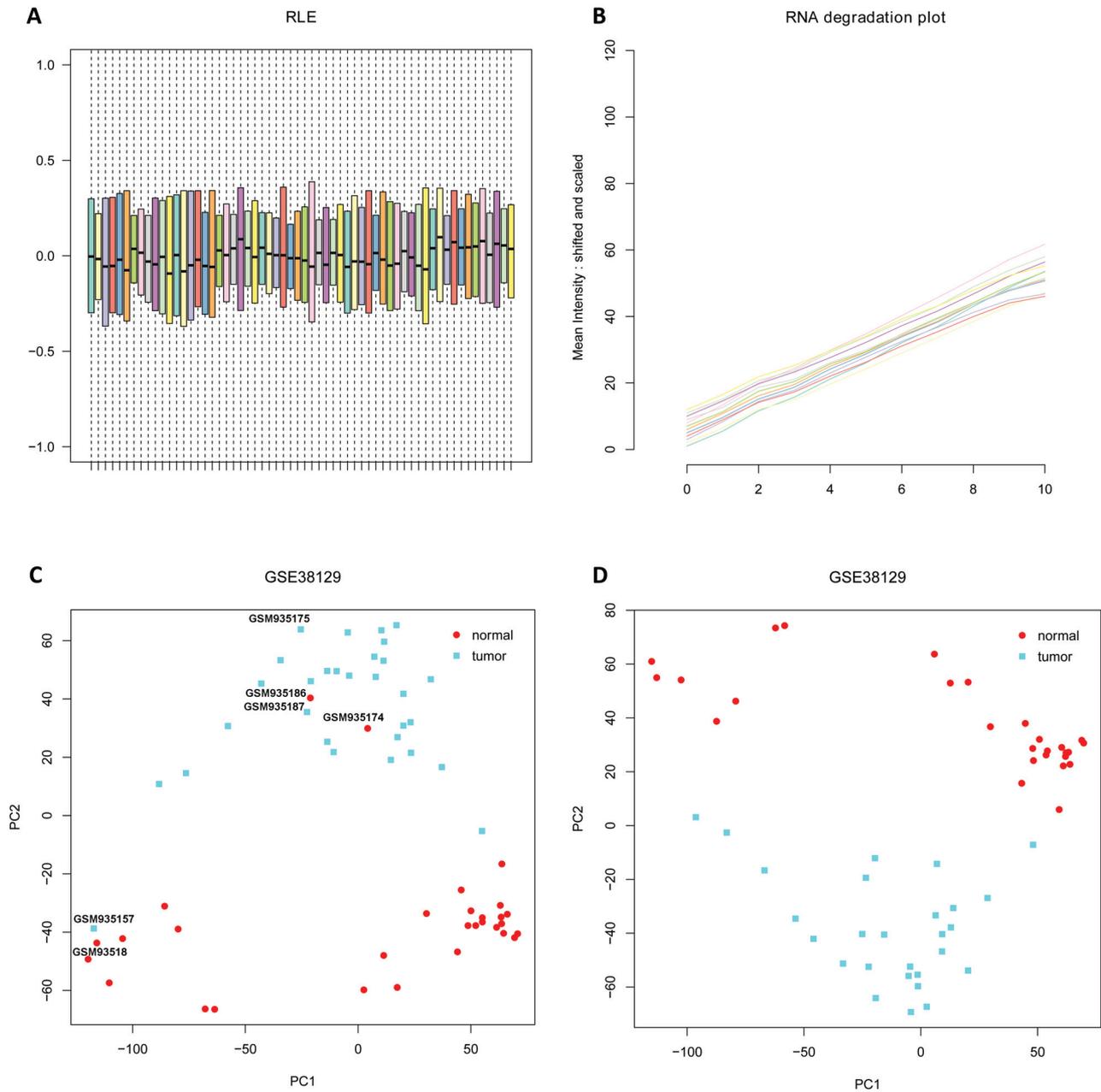


Figure 1. Quality control of the GSE38129 dataset. RLE and RNA degradation results performed on (A) and (B). PCA result before and after excluding GSM935156, GSM935157, GSM935174, GSM935175, GSM935186 and GSM935187 performed on (C) and (D), respectively. The higher level of normalization in RLE and separation of different samples in PCA present higher quality micro-array. The continuously rising RNA degradation curve reveals undegraded RNA obtained from samples. RLE = relative logarithmic expression. PCA = principal component analysis.

Functional enrichment analysis. GO BP analysis revealed that up-regulated DEGs were significantly associated with collagen catabolic process, extracellular matrix organization and extracellular matrix disassembly and down-regulated DEGs were significantly related to keratinocyte differentiation. In contrast, the KEGG pathway analysis demonstrated that the up-regulated genes were significantly enriched in ECM-receptor interaction, focal

adhesion and the PI3K-Akt signaling pathway but it established no significantly enriched pathways for down-regulated genes. The basic results are highlighted in Table 1 and more detailed analysis is contained in Supplementary Table 2.

Co-expression module detection and functional enrichment analysis. WGCNA analyzed the 188 DEGs expression profiles in the construct of gene co-expression modules.

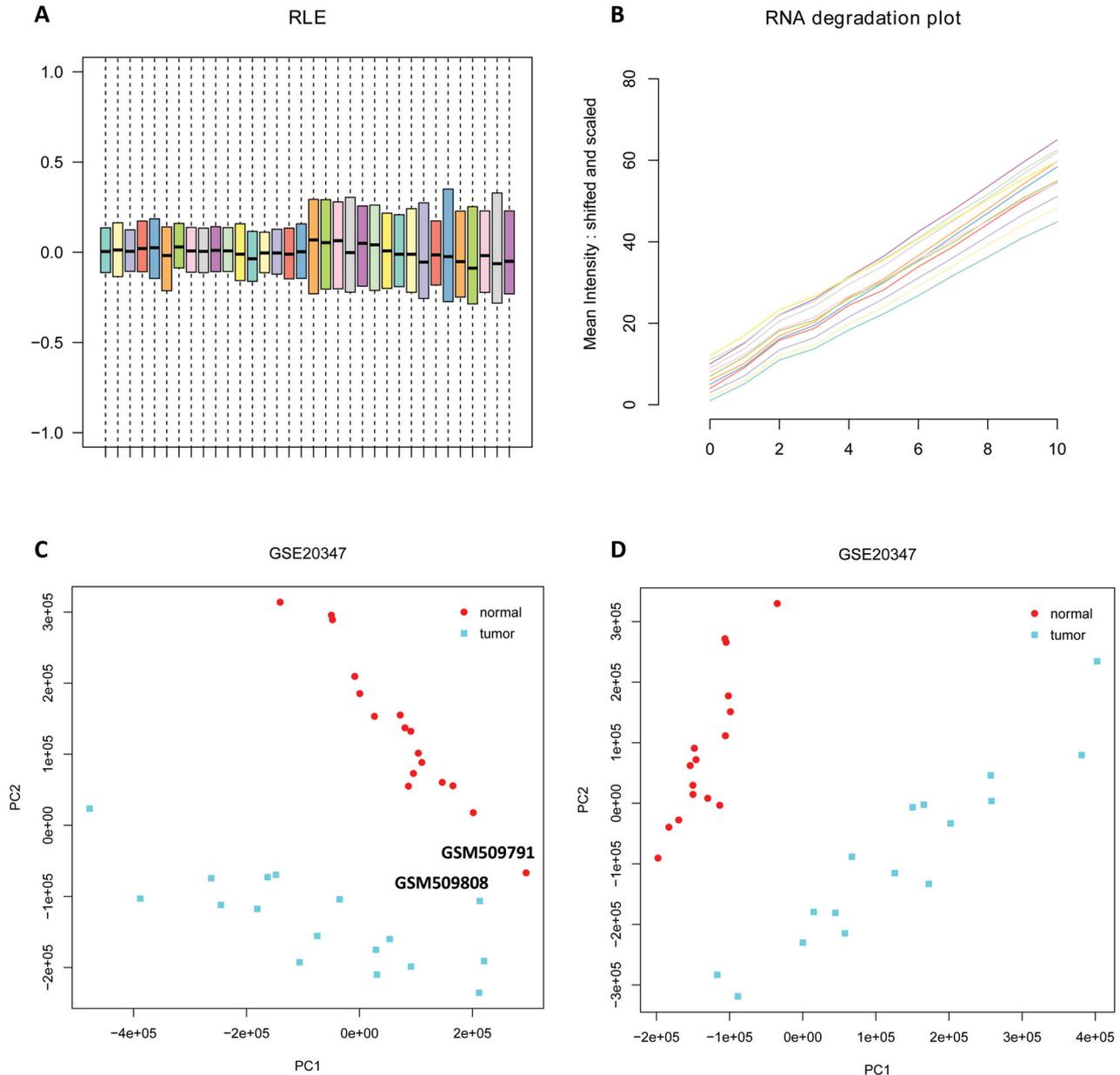


Figure 2. Quality control of the GSE20347 dataset. RLE and RNA degradation results performed on (A) and (B). PCA result before and after excluding GSM509791 and GSM509808 performed on (C) and (D), respectively. The higher level of normalization in RLE and separation of different samples in PCA present higher quality micro-array. The constantly rising RNA degradation curve rising constantly reveals undegraded RNA obtained from samples. RLE = relative logarithmic expression. PCA = principal component analysis.

'Blue' and 'Turquoise' modules were identified, with 61 and 106 genes respectively, and 18 hub genes were identified. The co-expression networks of these modules and the hub genes of each module are pictured in Figures 4 and 5. Genes in blue modules are associated with mitotic nuclear division, cell division and mitotic cytokinesis in GO BP analysis and those in the turquoise modules denote collagen catabolic processes, extracellular matrix organization and keratinocyte differentiation in GO BP analysis, and protein digestion,

amoebiasis and ECM-receptor interaction in KEGG analysis. Basic results are in Table 2 and greater detail is provided in Supplementary Table 3.

Survival analysis of hub genes. Follow-up information revealed that only 82 Relapse-Free Survival (RFS) patient notifications were obtained from the TCGA database. The prognostic value of the 18 hub genes was assessed by the R software survival package and Kaplan-Meier analysis established that the low expression of TPX2 ($p < 0.001$), CEP55

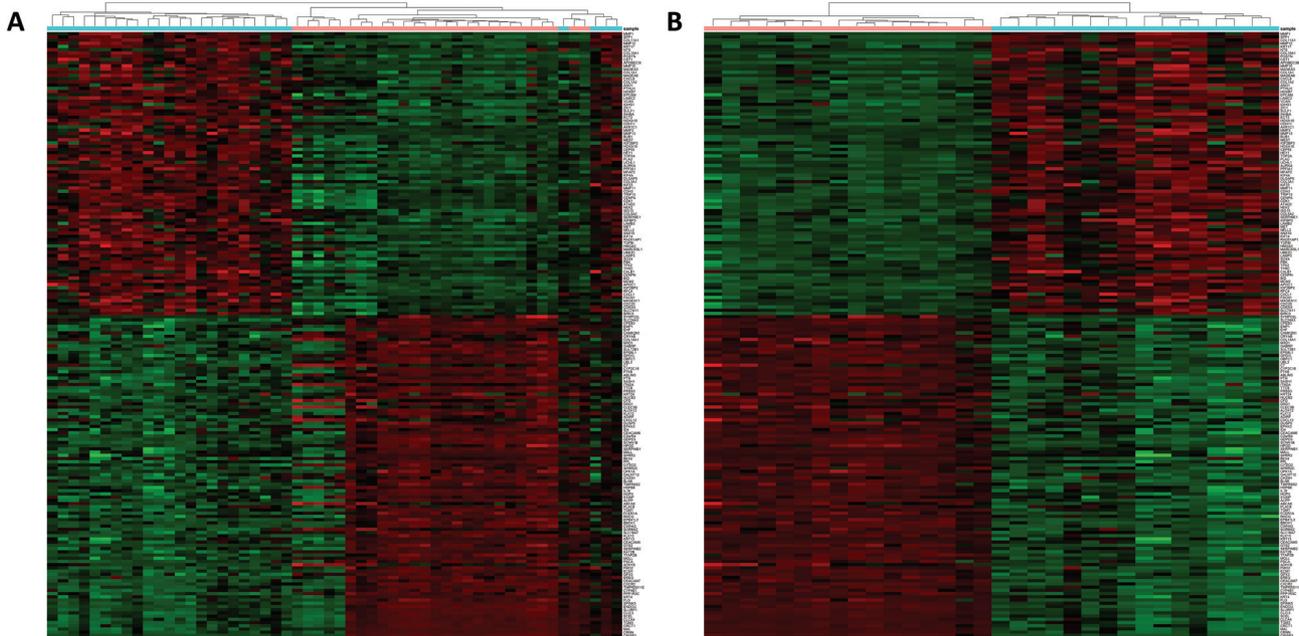


Figure 3. Heatmap and hierarchical clustering analysis of DEGs in training set (A) and validation set (B). Color depth shows the expression of DEGs; with red illustrating over-expression and green low expression. The side color on the top shows sample classification; with turquoise illustrating ESCC samples and pink highlighting normal esophageal squamous epithelium. DEGs = differentially expressed genes. ESCC = esophageal squamous cell carcinoma.

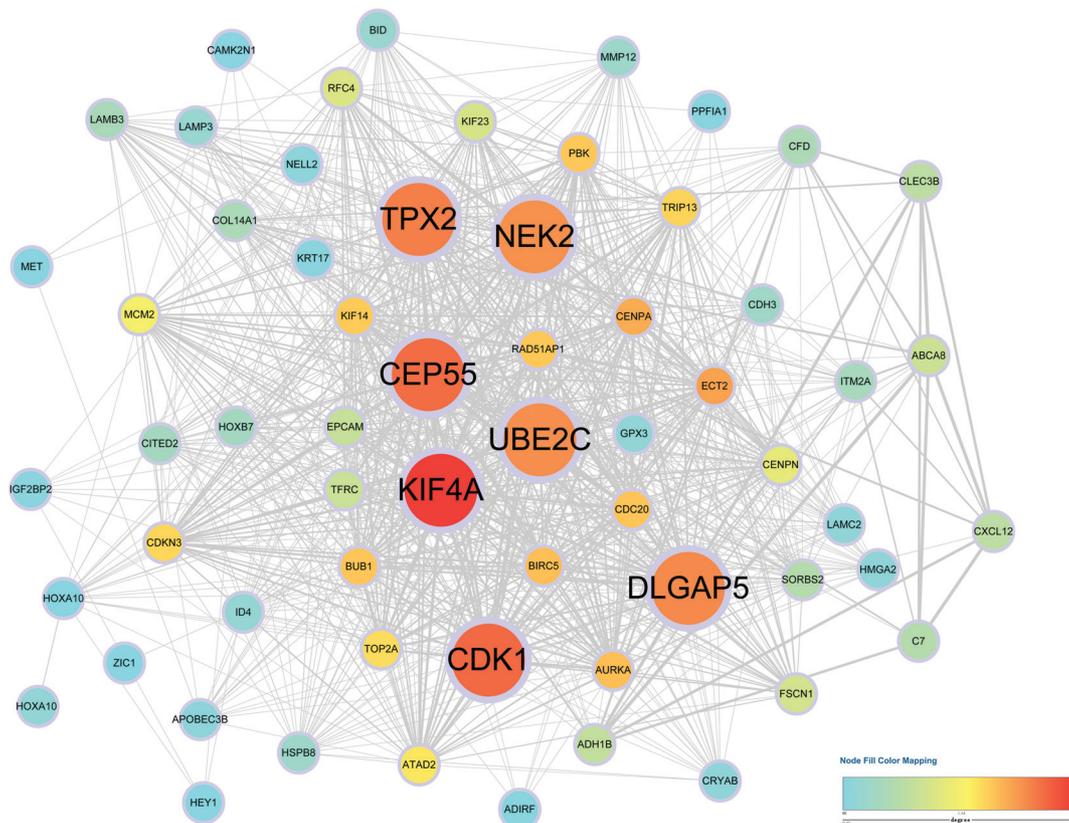


Figure 4. The co-expression network of blue modules identified by WGCNA. Color depth presents node connectivity degree, with red illustrating higher degree and turquoise denoting lower degree. Larger size nodes identify hub genes with the top 10% highest network connectivity degree. WGCNA = Weighted Gene Co-Expression Network Analysis.

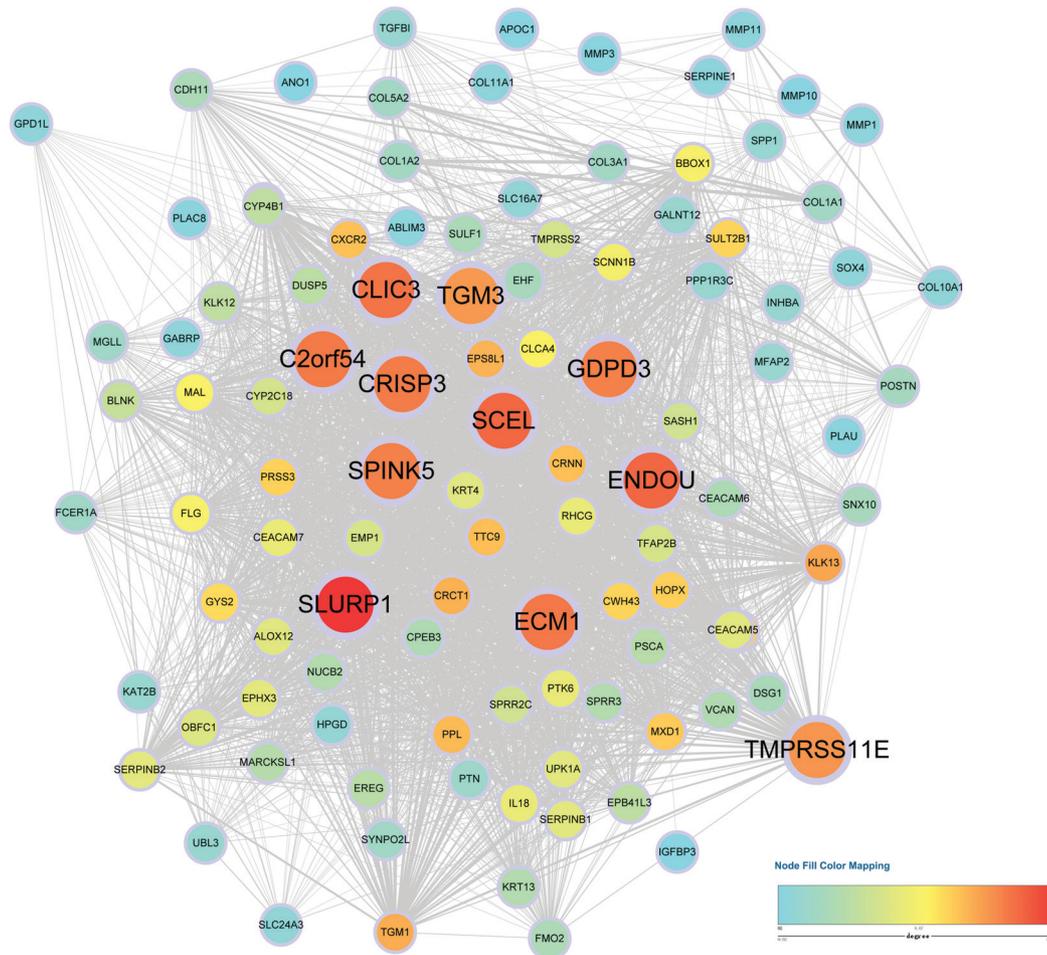


Figure 5. The co-expression network of turquoise modules identified by WGCNA. Color depth identifies node connectivity degree; with red illustrating higher degree of connectivity and turquoise is for lower connectivity. The larger size nodes denote hub genes with the top 10% highest network connectivity degree. WGCNA = Weighted Gene Co-Expression Network Analysis.

Table 2. The significantly enriched GO BP and KEGG pathways for co-expression modules.

	Description	No. enriched genes	Adjusted p-value
Module Blue			
GO:0007067	Mitotic nuclear division	11	8.01×10^{-6}
GO:0051301	Cell division	10	9.81×10^{-4}
GO:0000281	Mitotic cytokinesis	4	2.71×10^{-2}
Module Turquoise			
GO:0030574	Collagen catabolic process	10	7.46×10^{-8}
GO:0030198	Extracellular matrix organization	13	4.91×10^{-7}
GO:0001501	Skeletal system development	9	3.00×10^{-4}
GO:0030199	Collagen fibril organization	6	5.13×10^{-4}
GO:0030216	Keratinocyte differentiation	7	7.20×10^{-4}
hsa04974	Protein digestion and absorption	7	1.26×10^{-3}
hsa05146	Amoebiasis	7	1.82×10^{-3}
hsa04512	ECM-receptor interaction	6	5.20×10^{-3}

GO BP, Gene Ontology Biologic Process. KEGG, Kyoto Encyclopedia of Genes and Genomes.

($p=0.017$) and CDK1 ($p=0.039$) genes was associated with worse RFS in ESCC patients (Figure 6).

Discussion

A total of 188 genes were identified herein. These comprised 88 up-regulated genes and 100 down-regulated genes and almost all DEGs obtained in the training set were verified in the validation set. The up-regulated genes were associated with extracellular matrix organization and disassembly while down-regulated genes were associated with keratinocyte differentiation. Further, WGCNA identified blue and turquoise co-expression modules consisting of 61 and 106 genes, respectively, and also 18 hub genes. The blue module was associated with mitosis and cell division and the turquoise module involved extracellular matrix organization and catabolic processes. Kaplan-Meier analysis of hub genes then revealed that low TPX2, CDK1 and CEP55 expression was significantly related to worse relapse-free

survival. Interestingly, these 3 prognosis-related hub genes were all part of the blue module and GO BP mitotic nuclear division. These are considered key genes and their related biologic processes are crucial to the mechanisms involved in ESCC progression. Moreover all entities in the enriched GO BP and KEGG pathways most likely participate in mechanisms underlying ESCC progression and these therefore require urgent attention.

The TPX2 target protein for *Xenopus* kinesin-like protein 2 is a microtubule-associated protein-coding gene located on chromosome 20q11.21. TPX2 is upregulated in multiple tumor types such as cervical and gastric cancer [21, 22] and its expression is associated with growth and metastasis in hepatocellular carcinoma [23]. Researchers report that TPX2 regulates ESCC cells proliferation and invasiveness [24] and Hsu et al demonstrated that high TPX2 expression was associated with worse overall survival and shorter disease-free survival [25]. While Hsu's work supports and confirms our determination of the crucial role of TPX2 in ESCC progression, we did not establish its same effect in prognosis.

The Cyclin-dependent kinase 1, protein coding gene on chromosome 10q21.2 has a crucial role in the G2/M cell cycle. CDK1 activation and the formation of the cyclin B1-CDK1 complex is controlled by inhibitory phosphorylation of tyrosine and threonine in the early cell cycle phases, and CDK1 is activated by CDC25C phosphatase in the late G2 phase which is an obligatory step in G2/M transition [26]. Moreover, over-expression of CDK1 has been reported in many tumors, including breast cancer and hepatocellular carcinoma [27, 28].

Although some research has demonstrated that aberrant CDK1 expression is associated with poor patient prognosis [29], to the best of our knowledge no previous research has investigated CDK1 in ESCC. Our experimental results strongly suggest that CDK1 is an esophageal squamous cell carcinoma G2/M pathway regulator and that the combination of CDK1 with other regulators such as CDC25 should enhance the prediction of patient prognosis.

Centrosomal protein 55 (CEP55) is the protein coding gene on chromosome 10q23.33, and research has elucidated that this protein has an important role in regulating the PI3K/AKT pathway and promoting tumorigenesis [30]. CEP55 has been reported to bind PIK3CA and regulate the PI3K/AKT pathway; thus promoting phosphorylation and stimulating activation of AKT which has diverse roles in the cell cycle, cell survival and protein synthesis. CEP55 has also been identified in prognostic signatures for multiple cancer cell lines [31] and its over-expression has been related to poor clinical parameters including tumor stage, margin status and plasma tumor marker levels. Moreover, Jiang et al illustrated that CEP55 over-expression is significantly associated with reduced overall patient survival after surgery and, importantly, that the 5-year survival rate of ESCC patients with CEP55 over-expression was lower

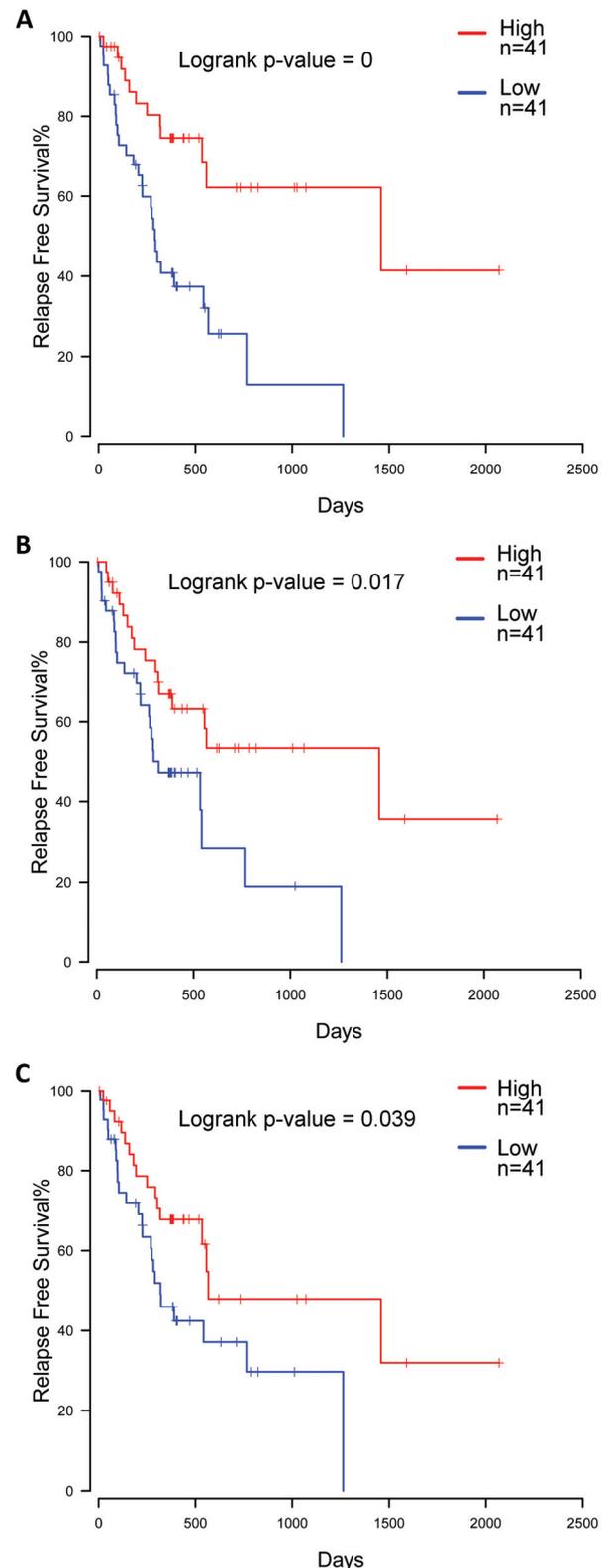


Figure 6. Prognostic value of TPX2(A), CDK1(B) and CEP55(C) for relapse-free survival in ESCC patients. The patients were divided into high and low expression groups according to the median of each DEG expression. ESCC = esophageal squamous cell carcinoma.

than those with under-expression [32]. However, we did not establish this finding, and we consider that our lack of support for Jiang's conclusion is due to the different populations, different expression measurements and different tumor tissue location in our individual research. Nevertheless, further assessment of CEP55 and ESCC relationships certainly appears warranted.

In conclusion, the results of our study proceed from screening 188 DEGs, and 2 co-expression modules and 18 hub genes were identified in these DEGs. These encode the enriched pathway, mitotic nuclear division, cell division, extracellular matrix organization and catabolic processes closely related to ESCC progression. Further, TPX2, CDK1 and CEP55 are considered key genes in esophageal squamous cell carcinoma prognosis. Since additional experiments and clinical research were not performed to verify resultant proteins ultimate significance, further study is advised. Focus on these key genes, their pathways, molecular mechanisms and clinical applications should enhance esophageal squamous cell carcinoma diagnosis and treatment.

Supplementary information is available in the online version of the paper.

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Supplemental Material

Supplementary Table 1. Differentially expressed genes of ESCC.

Genes	log FC	Adjusted p-value	Genes	log FC	Adjusted p-value	Genes	log FC	Adjusted p-value	Genes	log FC	Adjusted p-value
Up-regulated			Up-regulated			Down-regulated			Down-regulated		
MMP1	-5.248	1.36E-11	KIF23	-2.279	2.46E-12	CAMK2N1	2.039	8.71E-09	TMPRSS2	2.413	2.41E-07
SPP1	-4.696	4.27E-12	MMP11	-2.267	1.44E-11	CRYAB	2.048	5.63E-08	HSPB8	2.439	8.11E-11
COL11A1	-4.334	4.18E-13	CDH3	-2.266	7.97E-08	COL14A1	2.05	2.70E-07	IL18	2.466	3.74E-05
MMP12	-3.772	5.07E-13	TRIP13	-2.247	1.75E-14	MXD1	2.051	4.75E-07	HOPX	2.473	2.40E-05
KRT17	-3.667	3.94E-08	CENPA	-2.239	5.19E-13	GABRP	2.053	2.83E-03	S100P	2.491	1.30E-04
NTS	-3.459	1.94E-04	CDK1	-2.229	1.18E-12	SULT2B1	2.056	1.29E-05	ACPP	2.494	2.35E-05
COL10A1	-3.398	4.51E-14	ATAD2	-2.229	1.75E-14	EPS8L1	2.08	5.55E-09	ABCA8	2.512	7.12E-06
POSTN	-3.35	2.53E-07	NEK2	-2.219	1.78E-14	GPD1L	2.083	4.49E-14	PLAC8	2.532	7.69E-05
CST1	-3.092	2.86E-09	ISG15	-2.216	3.34E-07	OBFC1	2.096	4.56E-10	TGM1	2.559	2.90E-05
APOBEC3B	-3.085	7.90E-11	COL5A2	-2.216	7.80E-07	UBL3	2.103	1.06E-18	FCER1A	2.569	3.98E-14
MMP10	-3.065	6.94E-07	SERPINE1	-2.213	1.11E-08	C7	2.103	5.15E-04	RHCG	2.585	3.08E-03
MAGEA3	-3.053	7.94E-05	IGFBP3	-2.207	3.43E-06	CYP2C18	2.105	5.75E-06	EPB41L3	2.614	8.80E-13
COL1A1	-3.046	3.68E-08	LAMB3	-2.2	4.66E-09	PTK6	2.112	1.58E-04	BBOX1	2.645	5.20E-07
MAGEA6	-2.984	1.57E-04	MET	-2.184	1.41E-11	ABLIM3	2.113	3.04E-13	CWH43	2.664	3.02E-08
CXCL8	-2.929	9.00E-08	NELL2	-2.184	2.41E-09	PTN	2.126	1.84E-07	SORBS2	2.672	6.30E-10
COL1A2	-2.901	1.76E-07	SNX10	-2.181	8.44E-09	SASH1	2.133	2.51E-11	SLC16A7	2.713	2.38E-09
ANO1	-2.881	2.63E-06	KIF14	-2.179	1.75E-14	ITM2A	2.136	1.85E-06	KLK13	2.718	3.27E-08
PTHLH	-2.88	1.83E-05	RAD51AP1	-2.167	1.78E-13	TTC9	2.15	3.28E-07	KRT13	2.722	4.66E-03
HOXB7	-2.774	2.54E-15	TGFBI	-2.153	2.30E-06	PRSS3	2.15	2.39E-06	CEACAM5	2.723	1.20E-04
EPCAM	-2.725	1.81E-10	HMGA2	-2.152	3.96E-07	KRT24	2.164	1.40E-03	GYS2	2.74	5.42E-10
LAMC2	-2.714	1.56E-09	MARCKSL1	-2.151	2.80E-08	NUCB2	2.171	3.65E-12	SERPINB2	2.743	8.22E-04
VCAN	-2.713	1.88E-08	UBE2C	-2.144	8.45E-13	CFD	2.177	4.65E-06	KAT2B	2.772	1.65E-16
IGHG1	-2.635	6.95E-05	LAMP3	-2.142	7.74E-09	DSG1	2.196	4.64E-04	TFAP2B	2.796	3.76E-08
ZIC1	-2.63	6.11E-08	SOX4	-2.134	8.33E-10	CLEC3B	2.197	2.00E-04	MGLL	2.828	5.91E-14
SULF1	-2.627	3.49E-08	PBK	-2.127	1.72E-11	ALOX12	2.201	3.12E-04	PSCA	2.841	5.33E-07
INHBA	-2.574	9.49E-07	TPX2	-2.123	6.68E-12	KLK12	2.211	9.40E-05	ADH1B	2.87	2.52E-06
ECT2	-2.561	4.59E-16	TFR3	-2.114	1.70E-11	ADIRF	2.214	1.19E-09	FMO2	2.966	1.70E-06
HOXA10	-2.559	3.79E-06	CALB1	-2.101	4.14E-03	CXCL12	2.233	1.31E-04	ECM1	2.982	1.78E-07
CDH11	-2.548	7.13E-08	CENPN	-2.081	1.96E-10	DUSP5	2.248	7.88E-08	GPX3	3.014	9.95E-13
AKR1C1	-2.507	3.34E-04	BID	-2.072	9.24E-14	EPHX3	2.252	9.47E-05	EREG	3.118	7.05E-08
MMP3	-2.507	2.56E-06	MCM2	-2.071	8.07E-16	ID4	2.254	2.84E-11	CEACAM7	3.264	4.36E-06
MMP13	-2.482	1.37E-07	APOC1	-2.067	5.37E-10	CEACAM6	2.267	3.24E-03	CXCR2	3.317	1.17E-09
BUB1	-2.48	2.70E-14	IGF2BP2	-2.065	1.61E-09	C2orf54	2.267	5.70E-08	TMPRSS11E	3.322	5.09E-05
MEST	-2.417	8.70E-10	RFC4	-2.06	2.70E-14	GDPD3	2.276	1.75E-09	CYP4B1	3.345	1.29E-08
IGF2BP3	-2.409	5.32E-06	CXCL1	-2.052	2.53E-06	SCNN1B	2.284	6.00E-08	PPP1R3C	3.476	1.33E-12
HOXA10	-2.398	4.80E-12	FSCN1	-2.051	1.78E-13	HPGD	2.287	1.51E-09	KRT4	3.632	1.12E-04
CEP55	-2.397	3.53E-13	MAGEA11	-2.051	2.41E-04	SERPINB1	2.306	4.21E-06	FLG	3.667	1.79E-07
HEY1	-2.387	3.68E-10	CDC20	-2.025	1.90E-10	MALL	2.31	3.31E-05	SPINK5	3.684	1.84E-05
TOP2A	-2.378	1.87E-12	CDKN3	-2.018	1.68E-12	SPRR3	2.32	1.01E-02	ENDOU	3.703	9.76E-10
PLAU	-2.377	2.22E-13	SLC7A11	-2.015	4.29E-04	BEX4	2.327	6.38E-07	SLURP1	3.789	4.27E-07
UCHL1	-2.358	2.35E-05	BIRC5	-2.009	7.08E-11	PPL	2.34	4.13E-05	CLIC3	3.798	9.27E-07
AURKA	-2.357	6.41E-17	Down-regulated			CITED2	2.346	1.78E-14	SCEL	3.827	2.80E-05
PPFIA1	-2.353	1.89E-06	SYNPO2L	2	1.55E-05	SPRR2C	2.36	7.29E-04	CLCA4	3.966	5.96E-05
MFAP2	-2.34	3.34E-11	SLC24A3	2.006	3.66E-10	UPK1A	2.365	2.54E-06	TGM3	4.206	3.04E-06
KIF4A	-2.339	4.93E-16	CPEB3	2.022	5.23E-14	GALNT12	2.38	9.48E-11	CRCT1	4.228	5.01E-06
DLGAP5	-2.322	3.82E-12	EMP1	2.037	3.49E-06	CH25H	2.403	2.37E-08	MAL	4.754	1.02E-08
COL3A1	-2.281	1.09E-05	EHF	2.037	4.10E-03	BLNK	2.412	5.56E-07	CRNN	4.917	1.62E-06
									CRISP3	5.781	4.91E-09

ESCC, esophageal squamous cell carcinoma. FC, fold change.

Supplementary Table 2. The significant enriched GO BP terms and KEGG pathways.

	Description	Enriched genes
Upregulated		
GO:0030574	Collagen catabolic process	MMP10, COL3A1, COL1A2, COL1A1, MMP3, COL11A1, MMP13, COL5A2, MMP12, MMP1, COL10A1, MMP11
GO:0030198	Extracellular matrix organization	COL3A1, POSTN, COL5A2, LAMB3, SERPINE1, TGFB1, COL1A2, LAMC2, MFAP2, VCAN, COL1A1, COL11A1, SPP1, COL10A1
GO:0001501	Skeletal system development	PTHLH, COL3A1, COL1A2, HOXA10, SOX4, POSTN, VCAN, COL1A1, COL5A2, COL10A1, CDH11
GO:0022617	Extracellular matrix disassembly	MMP10, LAMB3, LAMC2, MMP3, MMP13, MMP12, MMP1, SPP1, MMP11
GO:0007067	Mitotic nuclear division	CDK1, CENPN, NEK2, BUB1, TPX2, AURKA, BIRC5, CDC20, PBK, CEP55, HMGA2
GO:0030199	Collagen fibril organization	COL3A1, COL1A2, COL1A1, COL11A1, COL5A2, MMP11
GO:0051301	Cell division	KIF14, CDK1, NEK2, BUB1, TPX2, AURKA, BIRC5, CDC20, UBE2C, HMGA2
GO:0035987	Endodermal cell differentiation	INHBA, LAMB3, HMGA2, COL11A1
GO:0000281	Mitotic cytokinesis	KIF23, KIF4A, CENPA, CEP55
GO:0008283	Cell proliferation	CXCL1, CDK1, DLGAP5, FSCN1, TGFB1, UCHL1, MET, BUB1, TPX2
hsa04512	ECM-receptor interaction	LAMB3, COL3A1, COL1A2, LAMC2, COL1A1, COL11A1, COL5A2, SPP1
hsa05146	Amoebiasis	LAMB3, COL3A1, COL1A2, CXCL8, LAMC2, COL1A1, COL11A1, COL5A2
hsa04510	Focal adhesion	LAMB3, COL3A1, MET, COL1A2, LAMC2, COL1A1, COL11A1, COL5A2, SPP1
hsa04974	Protein digestion and absorption	COL3A1, COL1A2, COL1A1, COL11A1, COL5A2, COL10A1
hsa04151	PI3K-Akt signaling pathway	LAMB3, COL3A1, MET, COL1A2, LAMC2, COL1A1, COL11A1, COL5A2, SPP1
hsa05202	Transcriptional misregulation in cancer	MET, CXCL8, HMGA2, MMP3, IGFBP3, PLAU
Downregulated		
GO:0030216	Keratinocyte differentiation	CRCT1, EREG, FLG, TGM1, SPRR3, TGM3, SCEL

GO BP, Gene Ontology Biologic Process. KEGG, Kyoto Encyclopedia of Genes and Genomes.

Supplementary Table 3. The significantly enriched GO BP and KEGG pathways for co-expression modules.

	Description	Enriched genes
Module Blue		
GO:0007067	Mitotic nuclear division	CDK1, CENPN, NEK2, BUB1, TPX2, AURKA, BIRC5, CDC20, PBK, CEP55, HMGA2
GO:0051301	Cell division	KIF14, CDK1, NEK2, BUB1, TPX2, AURKA, BIRC5, CDC20, UBE2C, HMGA2
GO:0000281	Mitotic cytokinesis	KIF23, KIF4A, CENPA, CEP55
Module Turquoise		
GO:0030574	Collagen catabolic process	MMP10, COL3A1, COL1A2, COL1A1, MMP3, COL11A1, COL5A2, MMP1, COL10A1, MMP11
GO:0030198	Extracellular matrix organization	COL3A1, POSTN, COL5A2, SPINK5, SERPINE1, TGFB1, COL1A2, MFAP2, VCAN, COL1A1, COL11A1, COL10A1, SPP1
GO:0001501	Skeletal system development	COL3A1, COL1A2, SOX4, POSTN, VCAN, COL1A1, COL5A2, COL10A1, CDH11
GO:0030199	Collagen fibril organization	COL3A1, COL1A2, COL1A1, COL11A1, COL5A2, MMP11
GO:0030216	Keratinocyte differentiation	CRCT1, EREG, FLG, TGM1, SPRR3, TGM3, SCEL
hsa04974	Protein digestion and absorption	COL3A1, PRSS3, COL1A2, COL1A1, COL11A1, COL5A2, COL10A1
hsa05146	Amoebiasis	COL3A1, COL1A2, SERPINB2, SERPINB1, COL1A1, COL11A1, COL5A2
hsa04512	ECM-receptor interaction	COL3A1, COL1A2, COL1A1, COL11A1, COL5A2, SPP1

GO BP, Gene Ontology Biologic Process. KEGG, Kyoto Encyclopedia of Genes and Genomes.