# XX 基因通过促进糖酵解促进巨噬细胞 M1 极化

# 2024-05-06

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# 1 摘要

### 1.1 生信需求

疾病: 类风湿性关节炎 RA 物种: 临床患者或者动物模型都可以细胞: 巨噬细胞

目标: 筛出 XX 基因, XX 基因满足, 1、是糖酵解相关基因 2、与巨噬细胞极化相关 (M1/M2)

设想: XX 基因在 RA 中上调, RA 中 M1 巨噬细胞上调, 其中 XX 基因主要分布在 M1 巨噬细胞而非 M2 巨噬细胞上, XX 基因可能通过促进糖酵解促进巨噬细胞 M1 极化

M1 标志: iNOS, CD11c, CD86 等 M2 标志: CD206, IL-10, TGF-beta 等

### 1.2 结果

- 首先通过分析 GEO 单细胞数据,鉴定出巨噬细胞不同表型 Fig. 5。
- 该数据集为小鼠来源,鉴定 M0、M1、M2 的小鼠基因 Marker 参考<sup>1</sup>,实际使用的 Marker 见 Tab. 1
- 进行差异分析 (Tab. 2):
  - XX 在 RA 中 M1 巨噬细胞上调: GPI-day25-RA\_Macrophage\_M1 vs Control\_Macrophage\_M1
  - 其中 XX 基因主要分布在 M1 巨噬细胞而非 M2 巨噬细胞上: GPI-day25-RA\_Macrophage\_M1 vs GPI-day25-RA\_Macrophage\_M2
- 以上两组差异基因交集见 Fig. 6
- 小鼠基因映射到人类 Tab. 3
- 其中糖酵解相关的基因见 Fig. 7
- 筛选到唯一的基因: PPARG (小鼠 Pparg)。其表达特征见 Fig. 8

### 1.3 进一步分析需求

利用开源数据库进行生物信息学分析,筛选并验证类风湿性关节炎临床患者和动物模型中与巨噬细胞极化和糖酵解相关的关键基因 XX 的表达情况

- XX (VWF) 表达水平与炎症因子、巨噬细胞浸润、巨噬细胞极化相关因子、糖酵解相关因子的相关性
- 与患者状态 (例如血清类风湿因子 (RF)、抗链球菌溶血素抗体 (ASO)、红细胞沉降率 (ESR) 和 C 反应蛋白 (CRP)) 的相关性

### 1.4 进一步分析结果

- 关联分析结果见 Fig. 9, Tab. 6。
- 未找到可用的 RA 表型数据集。

# 2 前言

### 3 材料和方法

### 3.1 材料

All used GEO expression data and their design:

- **GSE184609**: scRNA-Seq analysis of FACS-sorted live synovial cells isolated from naïve mice (two replicates) or from mice at day 6, 14, or 25 of GPI-induced arthritis (one replicate per time point).
- **GSE17755**: Peripheral blood was obtained from patients with RA (n=112), SLE (n=22), polyJIA (n=6), sJIA (n=51), HC (n=8), and HI (n=45). Blood samples from 8 HC and 45 HI are used as control.

### 3.2 方法

Mainly used method:

- The biomart was used for mapping genes between organism (e.g., mgi\_symbol to hgnc\_symbol)<sup>2</sup>.
- The Human Gene Database GeneCards used for disease related genes prediction<sup>3</sup>.
- GEO https://www.ncbi.nlm.nih.gov/geo/ used for expression dataset aquisition.
- R package Limma and edgeR used for differential expression analysis<sup>4,5</sup>.
- The data in published article of Jablonski et al used for distinguishing macrophage phenotypes  $(M0/M1/M2)^1$ .
- The R package Seurat used for scRNA-seq processing<sup>6,7</sup>.
- SCSA (python) used for cell type annotation<sup>8</sup>.
- R version 4.4.0 (2024-04-24); Other R packages (eg., dplyr and ggplot2) used for statistic analysis or data visualization.

# 4 分析结果

- 5 结论
- 6 附:分析流程
- 6.1 scRNA-seq
- 6.1.1 数据来源

这是一批小鼠的单细胞测序数据。

#### Data Source ID:

GSE184609

#### data\_processing:

10X Genomics Cell Ranger v3.1

### data\_processing.1:

Gene-cell UMI matrix was generated for downstream analyses. Low-quality cells were removed based on their unique feature counts and mitochondrial gene content. Data was normalized and log transformed using the default setting of Seurat (version 3.1.4).

### data\_processing.2:

Genome\_build: mm10

### data\_processing.3:

Supplementary\_files\_format\_and\_content: For each sample, there is one mtx file with filtered gene expressing UMI counts for each sample, one tsv file containing gene names, and one tsv file with cell barcodes.

### (上述信息框内容已保存至 Figure+Table/GSE184609-content)

### 6.1.2 细胞聚类与初步注释

使用 SCSA 对细胞类型注释。

Figure 1 (下方图) 为图 SCSA Cell type annotation 概览。

(对应文件为 Figure+Table/SCSA-Cell-type-annotation.pdf)

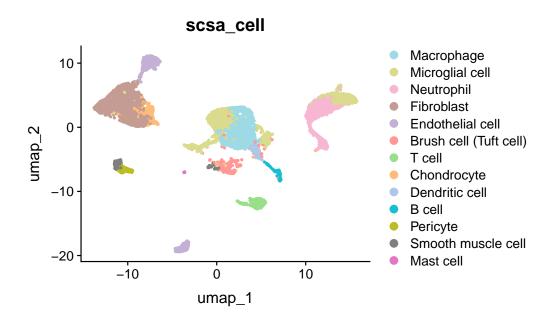


Figure 1: SCSA Cell type annotation

### 6.1.3 巨噬细胞重聚类

Figure 2 (下方图) 为图 Microphage UMAP Clustering 概览。

(对应文件为 Figure+Table/Microphage-UMAP-Clustering.pdf)

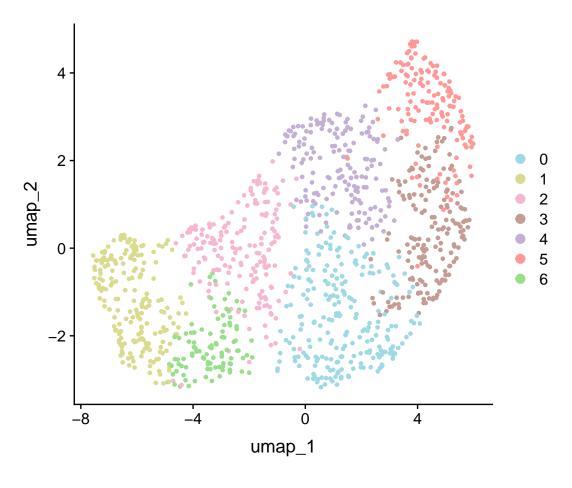


Figure 2: Microphage UMAP Clustering

### 6.1.4 巨噬细胞表型 M0、M1、M2 鉴定 Markers

Table 1 (下方表格) 为表格 The markers for Macrophage phenotypes annotation 概览。

### (对应文件为 Figure+Table/The-markers-for-Macrophage-phenotypes-annotation.csv)

注:表格共有 26 行 2 列,以下预览的表格可能省略部分数据;含有 3 个唯一'cell'。

Table 1: The markers for Macrophage phenotypes annotation

cell	markers
Macrophage_M0	Sh2d3c
Macrophage_M0	Slc13a3
Macrophage_M0	Rcan1
Macrophage_M0	Trp53inp1
Macrophage_M0	Slc40a1
Macrophage_M0	Il16
Macrophage_M1	Cfb

cell	markers
Macrophage_M1	Slfn4
Macrophage_M1	H2-Q6
$Macrophage\_M1$	Fpr1
Macrophage_M1	Slfn1
$Macrophage\_M1$	Ccrl2
$Macrophage\_M1$	Fpr2
$Macrophage\_M1$	Cxcl10
$Macrophage\_M1$	Oasl1

Figure 3 (下方图) 为图 Heatmap show the reference genes 概览。

### (对应文件为 Figure+Table/Heatmap-show-the-reference-genes.pdf)

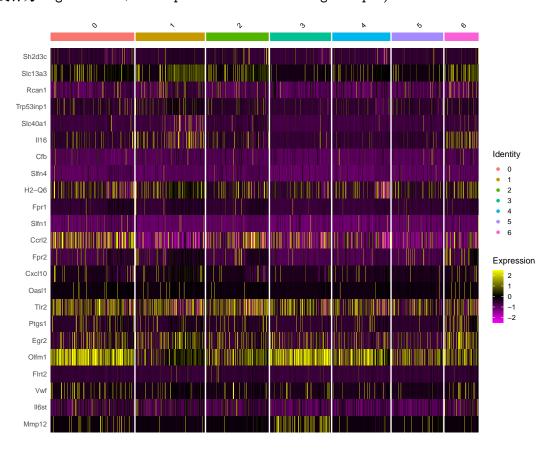


Figure 3: Heatmap show the reference genes

Figure 4 (下方图) 为图 Macrophage phenotypes type annotation 概览。

### (对应文件为 Figure+Table/Macrophage-phenotypes-type-annotation.pdf)

# macrophage\_phenotypes

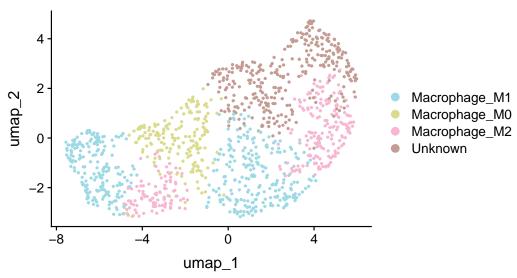


Figure 4: Macrophage phenotypes type annotation

### 6.1.5 RA 与 Control 的巨噬细胞表型

随后,根据数据集的来源(RA或Control,将巨噬细胞分类)

Figure 5 (下方图) 为图 The Phenotypes 概览。

(对应文件为 Figure+Table/The-Phenotypes.pdf)

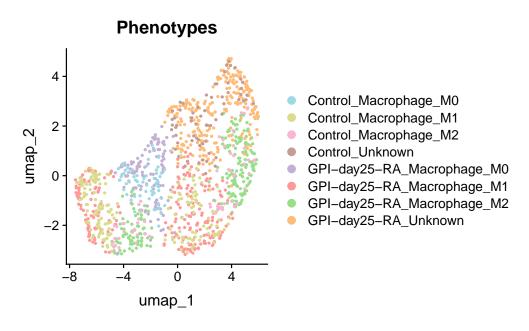


Figure 5: The Phenotypes

### 6.1.6 差异分析

- XX 在 RA 中 M1 巨噬细胞上调: GPI-day25-RA\_Macrophage\_M1 vs Control\_Macrophage\_M1
- 其中 XX 基因主要分布在 M1 巨噬细胞而非 M2 巨噬细胞上: GPI-day25-RA\_Macrophage\_M1 vs GPI-day25-RA\_Macrophage\_M2

Table 2 (下方表格) 为表格 DEGs of the contrasts 概览。

### (对应文件为 Figure+Table/DEGs-of-the-contrasts.csv)

注:表格共有 355 行 7 列,以下预览的表格可能省略部分数据;含有 2 个唯一 'contrast;含有 335 个唯一 'gene'。

Table 2: DEGs of the contrasts

contrast	p_val	avg_log2FC	pct.1	pct.2	p_val_adj	gene
GPI-day25	1.75831850	2.17770295	0.044	0.385	5.27495551	Adora3
$\label{eq:GPI-day25} \text{GPI-day25}$	2.23900708	9.52901476	0.069	0.427	6.71702126	F7
$\label{eq:GPI-day25} \text{GPI-day25}$	6.04375313	9.97788827	0.093	0.536	1.81312594	Hal
$\label{eq:GPI-day25} \text{GPI-day25}$	1.83819770	13.5930067	0.052	0.641	5.51459312	Cxcl13
$\label{eq:GPI-day25} \text{GPI-day25}$	1.00690808	4.79632080	0.153	0.583	3.02072424	Ifi44
$\label{eq:GPI-day25} \text{GPI-day25}$	2.16444867	8.41136649	0.153	0.87	6.49334601	Slc13a3
$\label{eq:GPI-day25} \text{GPI-day25}$	8.87142099	7.46793928	0.081	0.391	2.66142629	Cd4
$\label{eq:GPI-day25} \text{GPI-day25}$	1.52778766	0.95745400	0.141	0.307	4.58336298	Tnfsf14
$\label{eq:GPI-day25} \text{GPI-day25}$	4.80404528	3.93765968	0.169	0.484	1.44121358	Cd79b
$\label{eq:GPI-day25} \text{GPI-day25}$	8.42060311	3.12179649	0.06	0.651	2.52618093	Cd209e
$\label{eq:GPI-day25} \text{GPI-day25}$	8.42067724	1.96704094	0.145	0.786	2.52620317	Adgre4
$\label{eq:GPI-day25} \text{GPI-day25}$	8.24915617	8.83724798	0.161	0.766	2.47474685	Pparg
$\label{eq:GPI-day25} \text{GPI-day25}$	1.31157015	8.50505864	0.153	0.292	3.93471045	F10
$\label{eq:GPI-day25} \text{GPI-day25}$	2.28779328	2.51813054	0.024	0.411	6.86337986	Apoc4
$\label{eq:GPI-day25} \text{GPI-day25}$	2.75634208	4.11075823	0.073	0.755	8.26902626	Il10

Figure 6 (下方图) 为图 Intersection of RA M1 up with M1 not M2 概览。

(对应文件为 Figure+Table/Intersection-of-RA-M1-up-with-M1-not-M2.pdf)

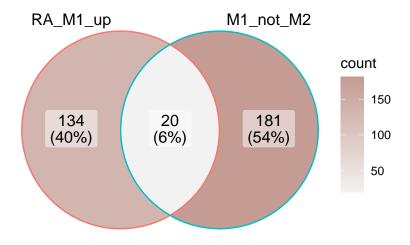


Figure 6: Intersection of RA M1 up with M1 not M2

### Intersection:

Ifi44, Adgre4, Pparg, Dppa3, Cadm1, P2ry14, Gm1673, Vwf, Ednrb, Fam43a, Bambi, Slc28a2, Plk2, Rcn3, Rrm1, Ifi204, Bmp2, Gfra2, Spon1, Gstm1

(上述信息框内容已保存至 Figure+Table/Intersection-of-RA-M1-up-with-M1-not-M2-content)

# 6.2 小鼠基因映射到人类基因

Table 3 (下方表格) 为表格 Mapped genes 概览。

(对应文件为 Figure+Table/Mapped-genes.csv)

注: 表格共有 19 行 2 列,以下预览的表格可能省略部分数据;含有 19 个唯一'mgi\_symbol;含有 19 个唯一'hgnc\_symbol'。

- 1. hgnc\_symbol: 基因名 (Human)
- 2. mgi\_symbol: 基因名 (Mice)

Table 3: Mapped genes

mgi_symbol	hgnc_symbol
Bmp2	BMP2
Ednrb	EDNRB
Dppa3	DPPA3
Spon1	SPON1
Gfra2	GFRA2
Bambi	BAMBI

mgi_symbol	hgnc_symbol
Cadm1	CADM1
Slc28a2	SLC28A2
Rrm1	RRM1
Ifi44	IFI44
Gm1673	C4orf48
Ifi204	MNDA
P2ry14	P2RY14
Rcn3	RCN3
Gstm1	GSTM1

### 6.3 糖酵解相关基因

Table 4 (下方表格) 为表格 Glycolysis related genes from GeneCards 概览。

(对应文件为 Figure+Table/Glycolysis-related-genes-from-GeneCards.xlsx)

注:表格共有 118 行 7 列,以下预览的表格可能省略部分数据;含有 118 个唯一 'Symbol'。

### The GeneCards data was obtained by querying :

Glycolysis

Restrict (with quotes):

FALSE

Filtering by Score: :

Score > 3

Table 4: Glycolysis related genes from GeneCards

Symbol	Description	Category	UniProt_ID	$\operatorname{GIFtS}$	$GC\_id$	Score
TIGAR	TP53 Induc	Protein Co	Q9NQ88	45	GC12P038924	22.4
PKM	Pyruvate K	Protein Co	P14618	58	GC15M072199	20.77
HK2	Hexokinase 2	Protein Co	P52789	55	GC02P074947	19.42
GAPDH	Glyceralde	Protein Co	P04406	59	GC12P038965	17.14
LDHA	Lactate De	Protein Co	P00338	59	GC11P018394	15.81
HIF1A	Hypoxia In	Protein Co	Q16665	57	GC14P061695	15.1
RRAD	RRAD, Ras	Protein Co	P55042	46	GC16M067483	15.1

Symbol	Description	Category	UniProt_ID	$\operatorname{GIFtS}$	$GC\_id$	Score
HK1	Hexokinase 1	Protein Co	P19367	59	GC10P069269	14.64
PKLR	Pyruvate K	Protein Co	P30613	55	GC01M155289	13.37
ENO1	Enolase 1	Protein Co	P06733	56	GC01M008861	13.36
ENO3	Enolase 3	Protein Co	P13929	54	GC17P004948	13.33
PFKP	Phosphofru	Protein Co	Q01813	53	GC10P003066	13.19
TPI1	Triosephos	Protein Co	P60174	55	GC12P006867	13.18
GLTC1	Glycolysis	RNA Gene (		2	GC11U909607	12.97
PGK1	Phosphogly	Protein Co	P00558	57	GC0XP078166	12.94

Figure 7 (下方图) 为图 Intersection of RA M1M2 related with Glycolysis related 概览。

### (对应文件为 Figure+Table/Intersection-of-RA-M1M2-related-with-Glycolysis-related.pdf)

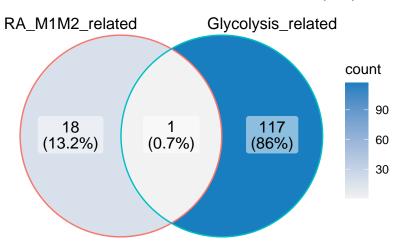


Figure 7: Intersection of RA M1M2 related with Glycolysis related  $\,$ 

Intersection:
PPARG

(上述信息框内容已保存至 Figure+Table/Intersection-of-RA-M1M2-related-with-Glycolysis-related-content)

# 6.4 交集基因的表达(小鼠单细胞数据)

Figure 8 (下方图) 为图 Violing plot of expression level of the Pparg 概览。

(对应文件为 Figure+Table/Violing-plot-of-expression-level-of-the-Pparg.pdf)

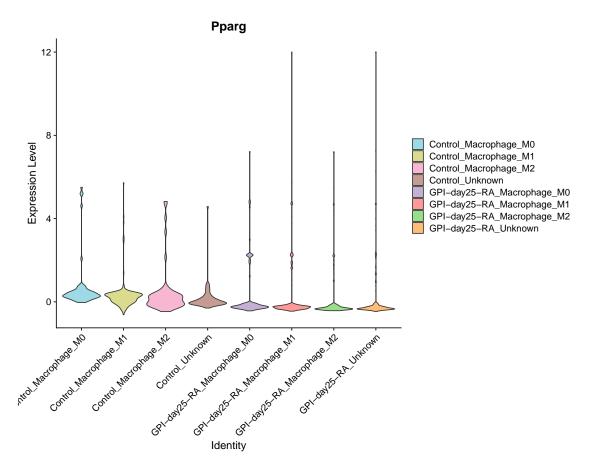


Figure 8: Violing plot of expression level of the Pparg

# 7 进一步分析

### 7.1 数据来源

### Data Source ID:

GSE17755

### $data\_processing:$

Log2 ratios of Cy3 to Cy5 were calculated and normalized by the method of global ratio median normalization.

(上述信息框内容已保存至 Figure+Table/HUMAN-GSE17755-content)

# 7.2 炎症因子、巨噬细胞浸润、巨噬细胞极化相关因子、糖酵解相关因子

使用 genecards 获取相关基因 (各取前 50 基因):

- IF: Inflammatory factors 炎症因子
- MI: Macrophage infiltration 巨噬细胞浸润
- MP: Macrophage polarization 巨噬细胞极化
- G: Glycolysis 糖酵解

Table 5 (下方表格) 为表格 All Factors 概览。

### (对应文件为 Figure+Table/All-Factors.csv)

注:表格共有 200 行 2 列,以下预览的表格可能省略部分数据;含有 4 个唯一'type'。

Table 5: All Factors

name
IL6
TNF
CRP
BDNF-AS
IL1B
LINC02605
TLR4
MIR146B
ADIPOQ
LINC01672
CXCL8
IL1A
NFKB1
CERNA3
IL18

对上述基因集去重复后,关联分析。

# 7.3 关联分析

Figure 9 (下方图) 为图 HUMAN correlation heatmap 概览。

(对应文件为 Figure+Table/HUMAN-correlation-heatmap.pdf)

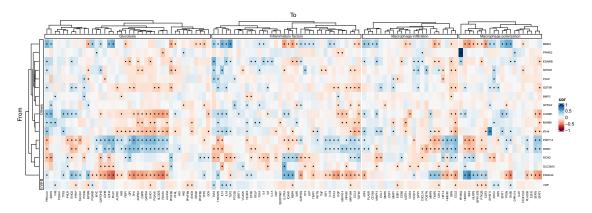


Figure 9: HUMAN correlation heatmap

Figure 10 (下方图) 为图 HUMAN correlation heatmap VWF significant 概览。

### (对应文件为 Figure+Table/HUMAN-correlation-heatmap-VWF-significant.pdf)

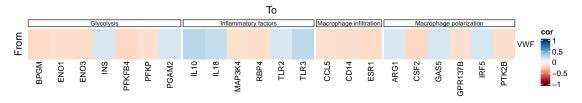


Figure 10: HUMAN correlation heatmap VWF significant

'HUMAN regression VWF significant'数据已全部提供。

### (对应文件为 Figure+Table/HUMAN-regression-VWF-significant)

注:文件夹 Figure+Table/HUMAN-regression-VWF-significant 共包含 4 个文件。

- 1. 1\_Glycolysis.pdf
- 2. 2\_Inflammatory factors.pdf
- 3. 3\_Macrophage infiltration.pdf
- 4. 4\_Macrophage polarization.pdf

Table 6 (下方表格) 为表格 HUMAN correlation 概览。

### (对应文件为 Figure+Table/HUMAN-correlation.csv)

注:表格共有 2023 行 9 列,以下预览的表格可能省略部分数据;含有 17 个唯一'From'。

- 1. cor: 皮尔逊关联系数, 正关联或负关联。
- 2. pvalue: 显著性 P。
- 3. -log2(P.value): P 的对数转化。
- 4. significant: 显著性。
- 5. sign: 人为赋予的符号,参考 significant。

Table 6: HUMAN correlation

From	То	cor	pvalue	-log2(	signif	sign	Factors	Type
EDNRB	GAPDH	-0.14	0.0931	3.4250	> 0.05	-	Glycol	Others
PPARG	GAPDH	-0.06	0.4812	1.0552	> 0.05	-	Glycol	Others
CADM1	GAPDH	-0.26	9e-04	10.117	< 0.001	**	Glycol	Others
BMP2	GAPDH	-0.1	0.225	2.1520	> 0.05	-	Glycol	Others
SLC28A2	GAPDH	0.06	0.4919	1.0235	> 0.05	-	Glycol	Others
RRM1	GAPDH	0.33	0	16.609	< 0.001	**	Glycol	Others
BAMBI	GAPDH	-0.16	0.0796	3.6510	> 0.05	-	Glycol	Others
PLK2	GAPDH	-0.08	0.3648	1.4548	> 0.05	-	Glycol	Others
P2RY14	GAPDH	0.43	0	16.609	< 0.001	**	Glycol	Others
MNDA	GAPDH	-0.02	0.7616	0.3928	> 0.05	-	Glycol	Others
GSTM1	GAPDH	-0.16	0.073	3.7759	> 0.05	-	Glycol	Others
IFI44	GAPDH	-0.33	0	16.609	< 0.001	**	Glycol	Others
RCN3	GAPDH	0.12	0.148	2.7563	> 0.05	-	Glycol	Others
SPON1	GAPDH	-0.14	0.0882	3.5030	> 0.05	-	Glycol	Others
GFRA2	GAPDH	-0.07	0.3662	1.4492	> 0.05	-	Glycol	Others

# Reference

- 1. Jablonski, K. A. et al. Novel markers to delineate murine m1 and m2 macrophages. PloS one 10, (2015).
- 2. Durinck, S., Spellman, P. T., Birney, E. & Huber, W. Mapping identifiers for the integration of genomic datasets with the r/bioconductor package biomaRt. *Nature protocols* 4, 1184–1191 (2009).
- 3. Stelzer, G. et al. The generards suite: From gene data mining to disease genome sequence analyses. Current protocols in bioinformatics **54**, 1.30.1–1.30.33 (2016).
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