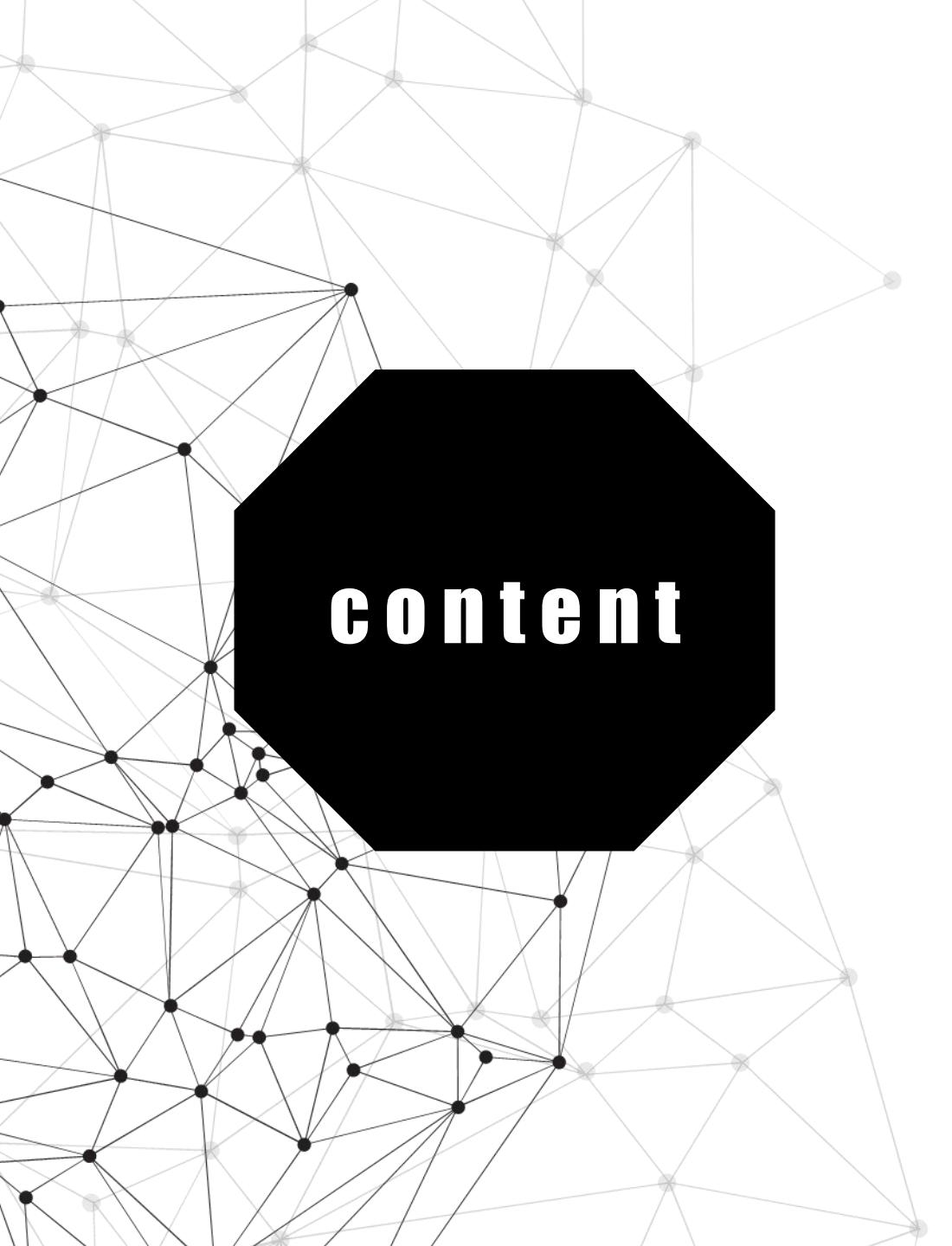


The background of the image features a complex network of interconnected nodes, resembling a molecular or neural network. It consists of numerous small, light-grey dots connected by thin grey lines, creating a sense of organic, flowing structure.

THE GALLERY OF CALLIGRAPHY

TEAM 6

GUO MINGHAO
NIU YIBO
YU LIN
WU YI



content

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CONSTRUCT DATASET

03

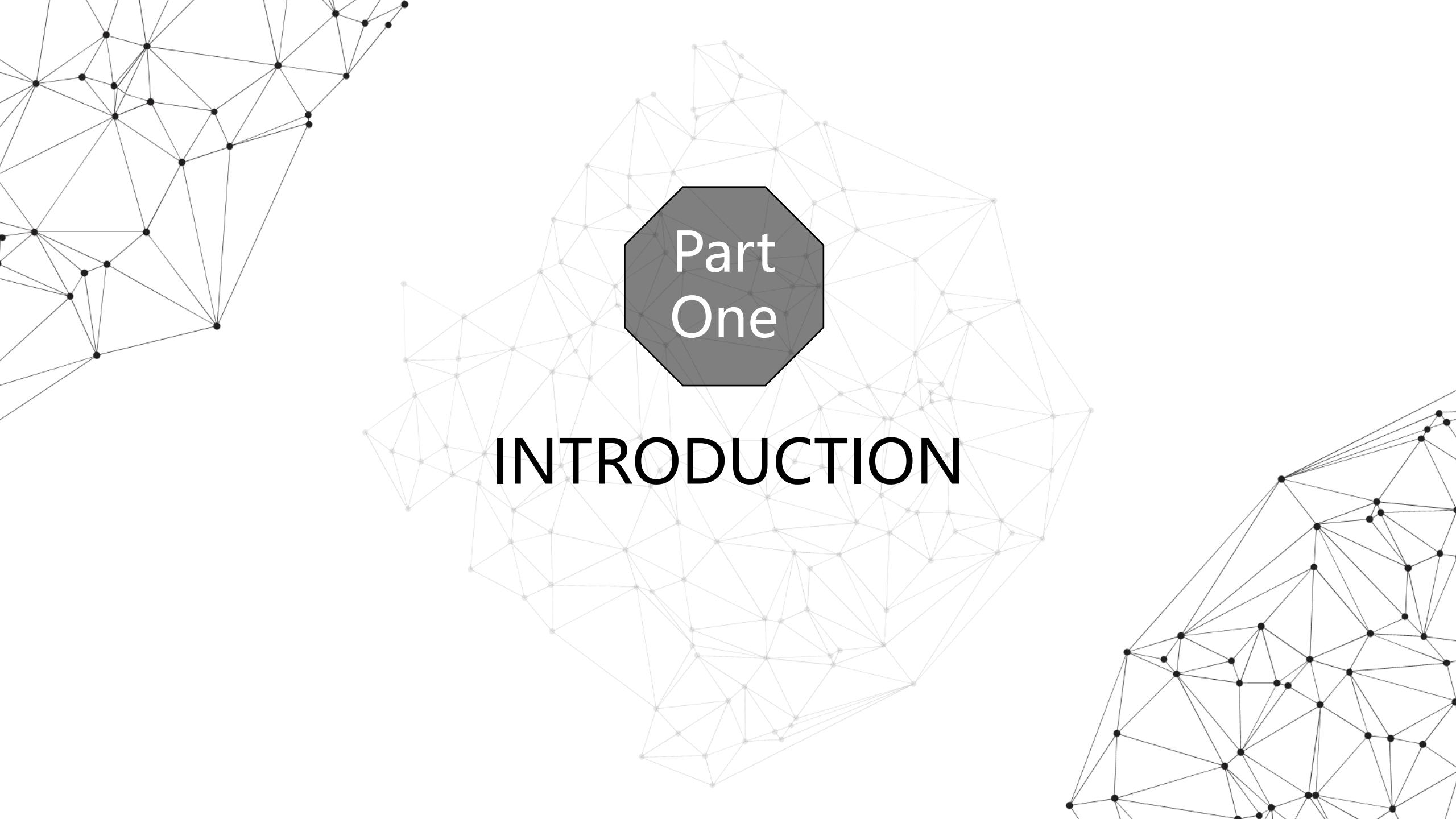
COMMUNITY DETECTION

04

ANALYSIS OF THE RESULTS

05

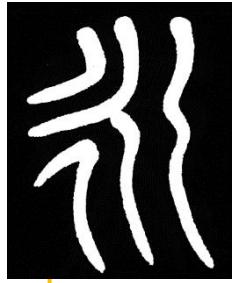
CONCLUSIONS

The background of the slide features a complex, abstract network structure composed of numerous small, light gray dots connected by thin white lines, creating a mesh-like pattern that covers the entire frame.

**Part
One**

INTRODUCTION

Five Styles of Chinese Calligraphy



Seal Script (篆书)

- Rectangular
- Curved brush.
- Beauty of balance



221-207 BCE



Clerical Script (隶书)

- Flat and square
- A silkworm head and a wild goose



Regular Script (楷书)

- Square and regular



Cursive Script (草书)

- Joined together in a flowing manner
- Writing faster

Han

202 BCE-220 CE

Jin

266 CE-420 CE

Semi-cursive Script (行书)

- Between regular script and cursive script.

Objectives

- Explore the connections between different styles of calligraphy.
- Analyze and validate the evolution of calligraphy.
- Analyze the relationships between calligraphers

Five Styles

Clerical 隶书 Cursive 草书
Regular 楷书 Seal 篆书
Semi-Cursive 行书

Different Calligraphers

Xizhi Wang Song Dynasty
Fu Mi Song Dynasty
Zhengqing Yan Tang Dynasty

Procedure

Collect data and form two datasets.

Use Siamese neural network to generate dissimilarity matrix.

Apply different community detection algorithms.

Do the visualization, analyze and compare the results of the two datasets.



Part Two

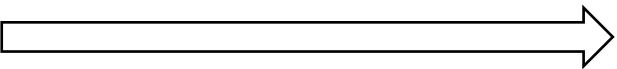
CONSTRUCT DATASET

- Obtain and process images

Obtain and Process Data



Using Java to get hundreds
of images from the Internet



www.shufami.com 书法迷



After image binarization

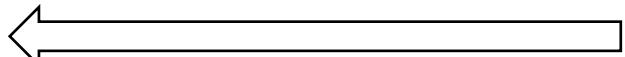
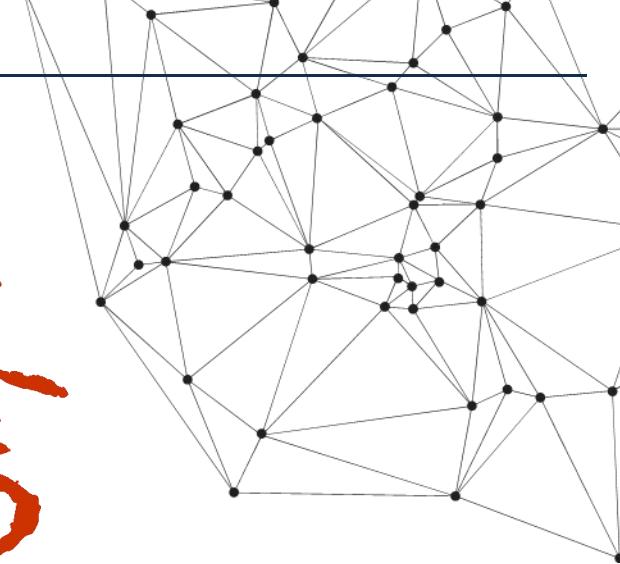
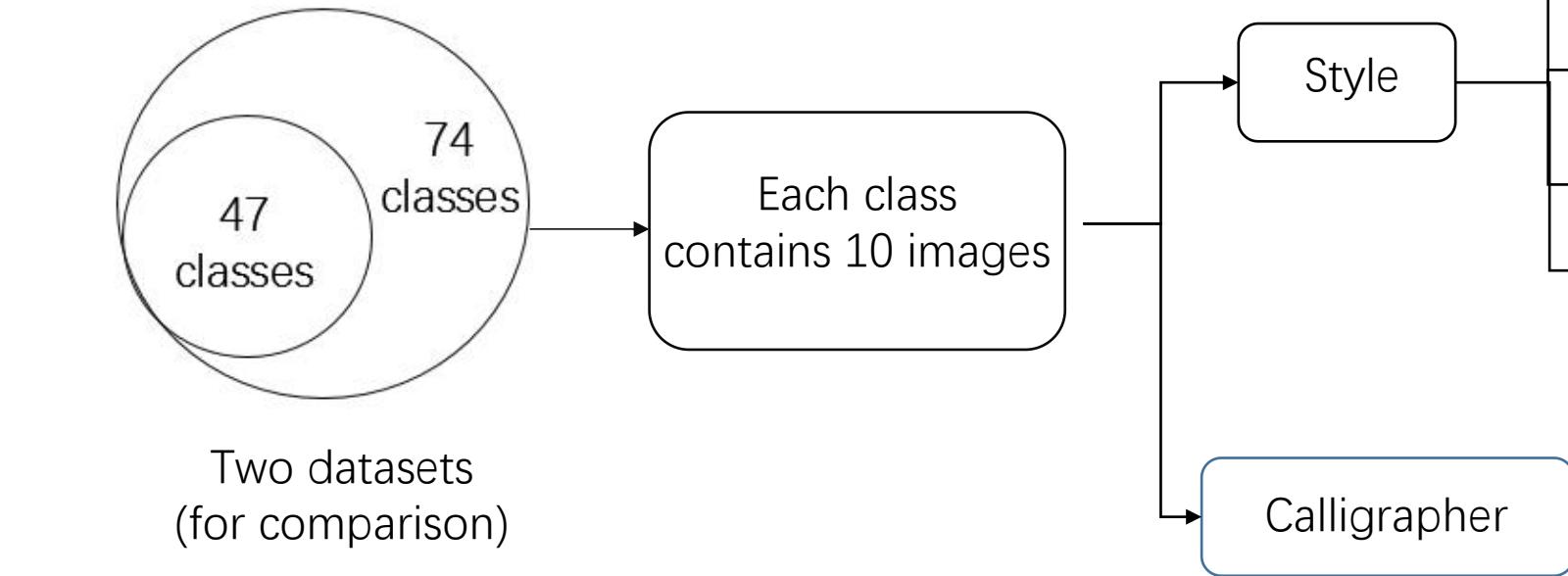


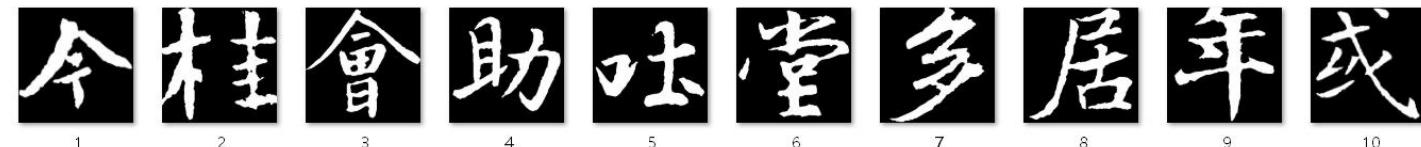
Image cropped



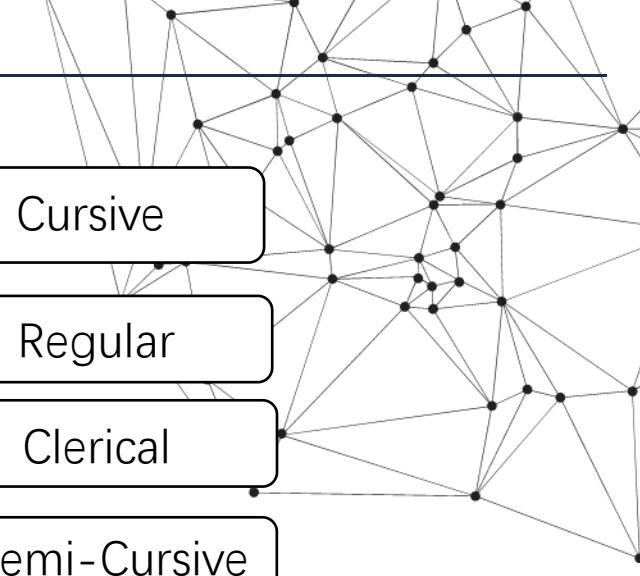
Structure Of Dataset

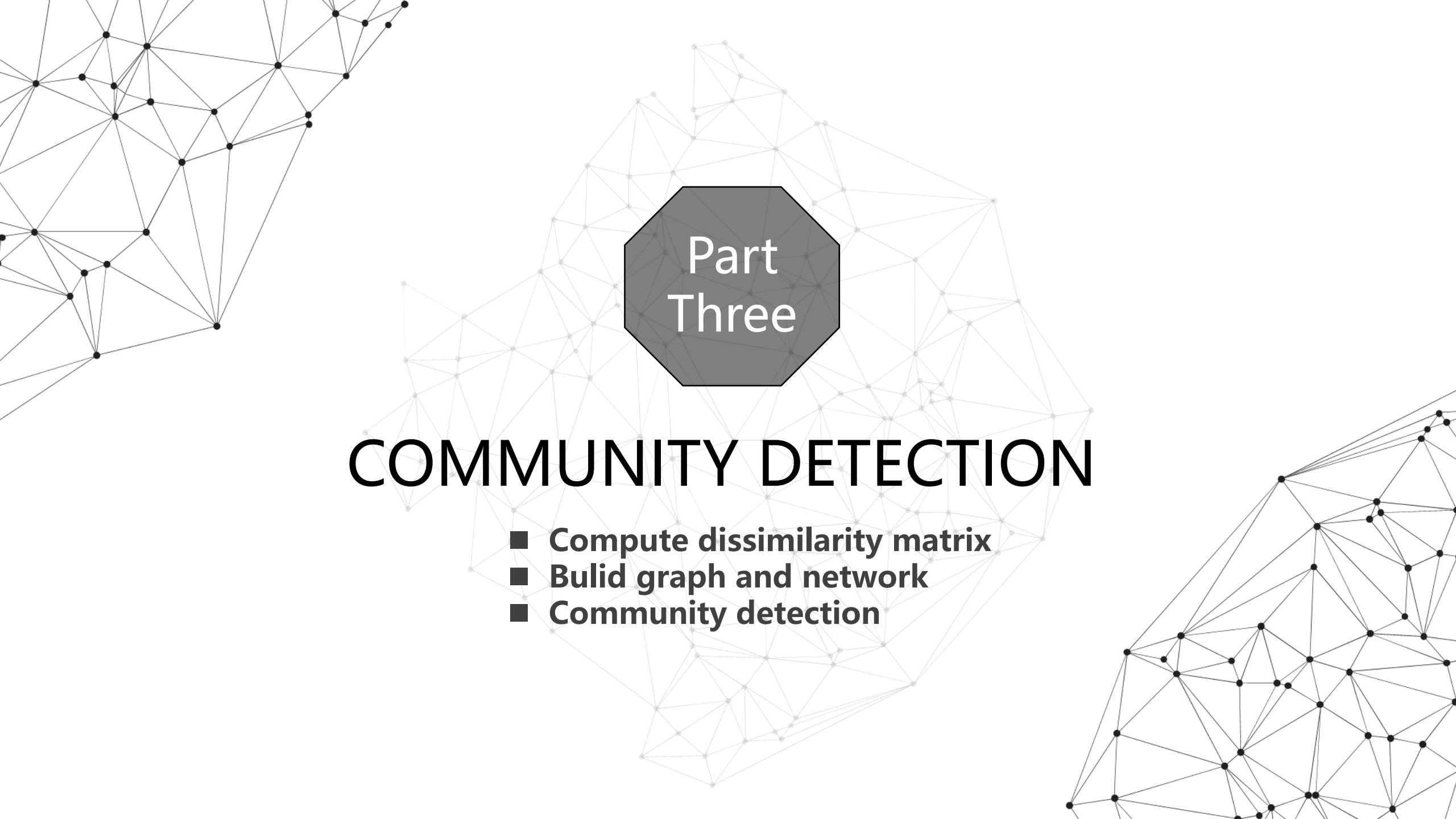


e.g.



Xizhi Wang's Regular Characters



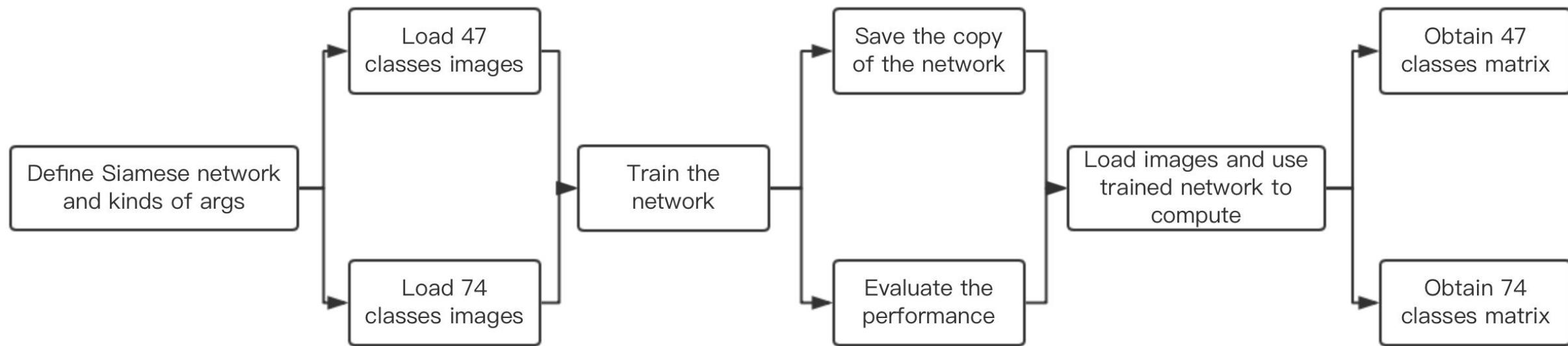
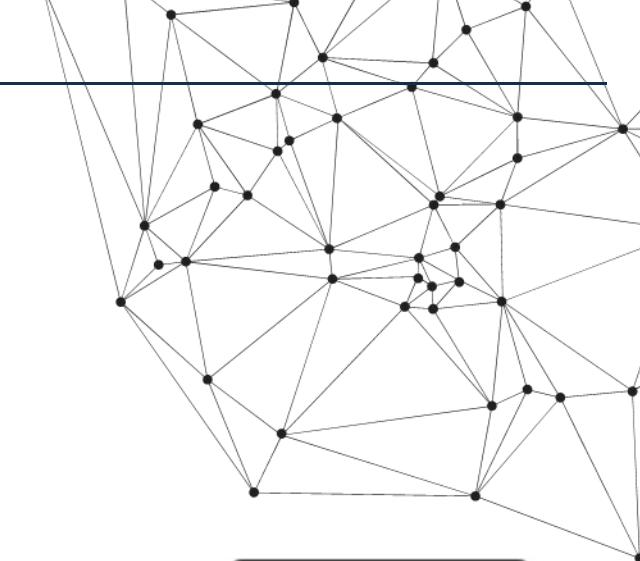


Part Three

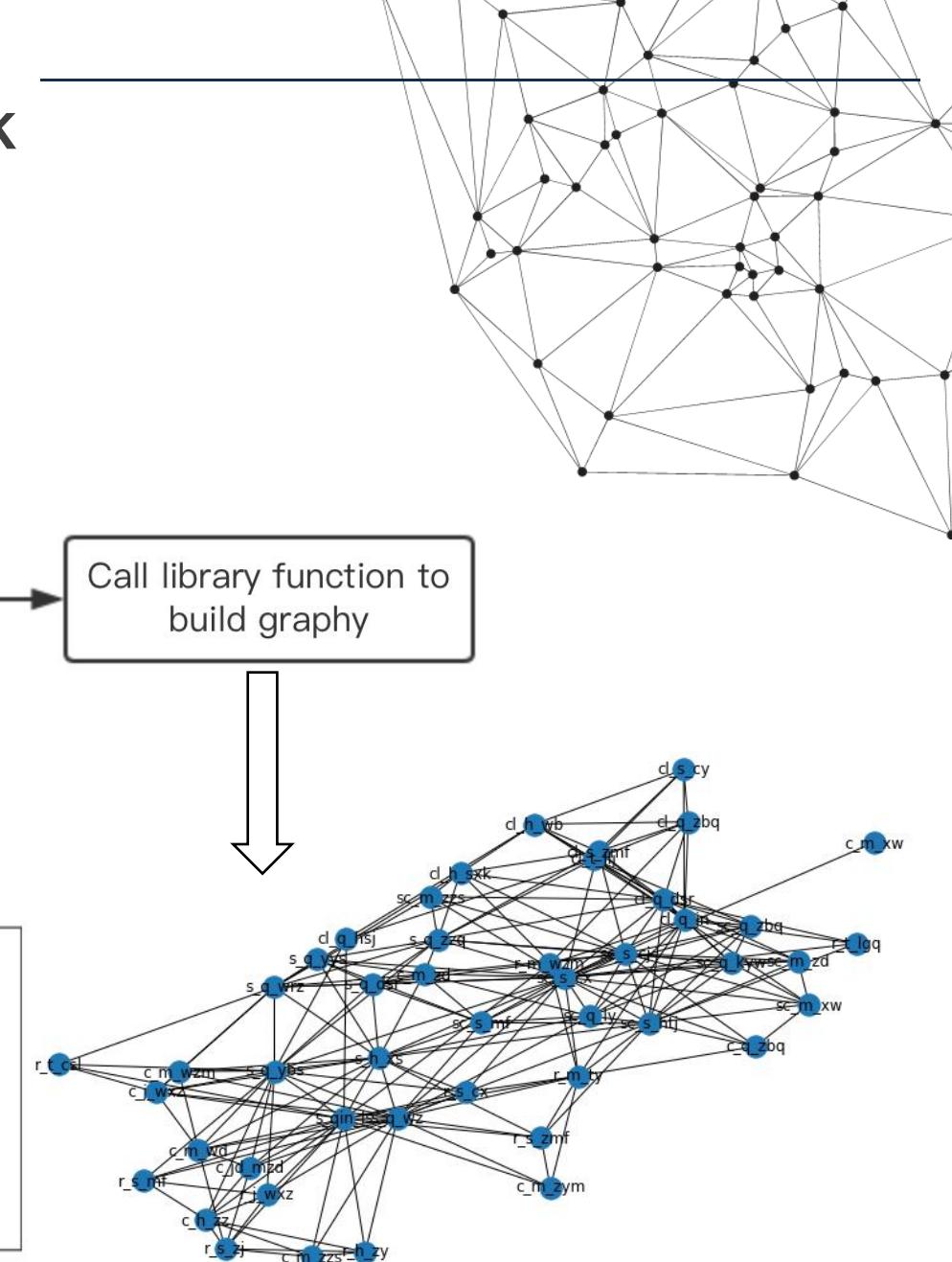
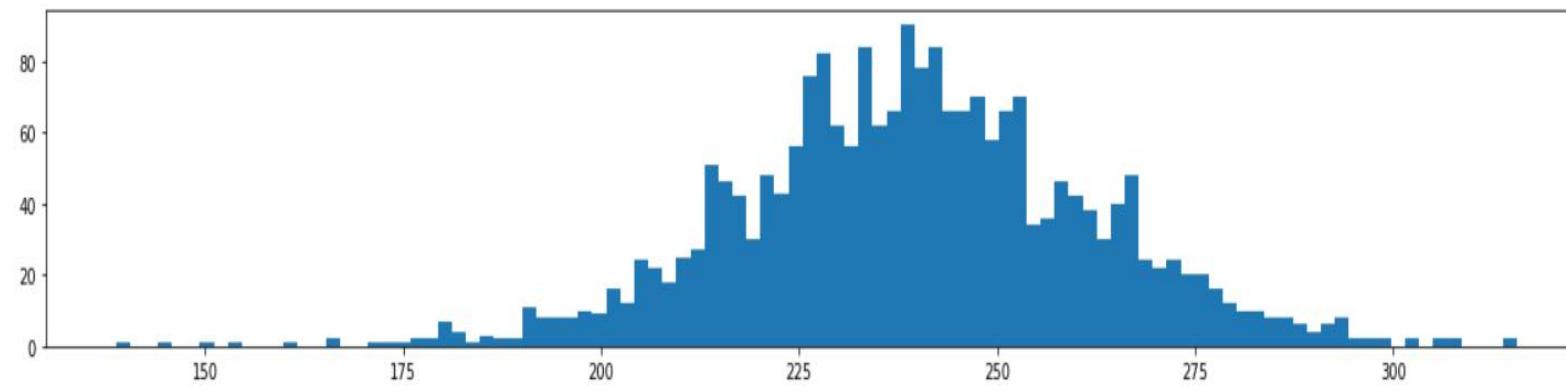
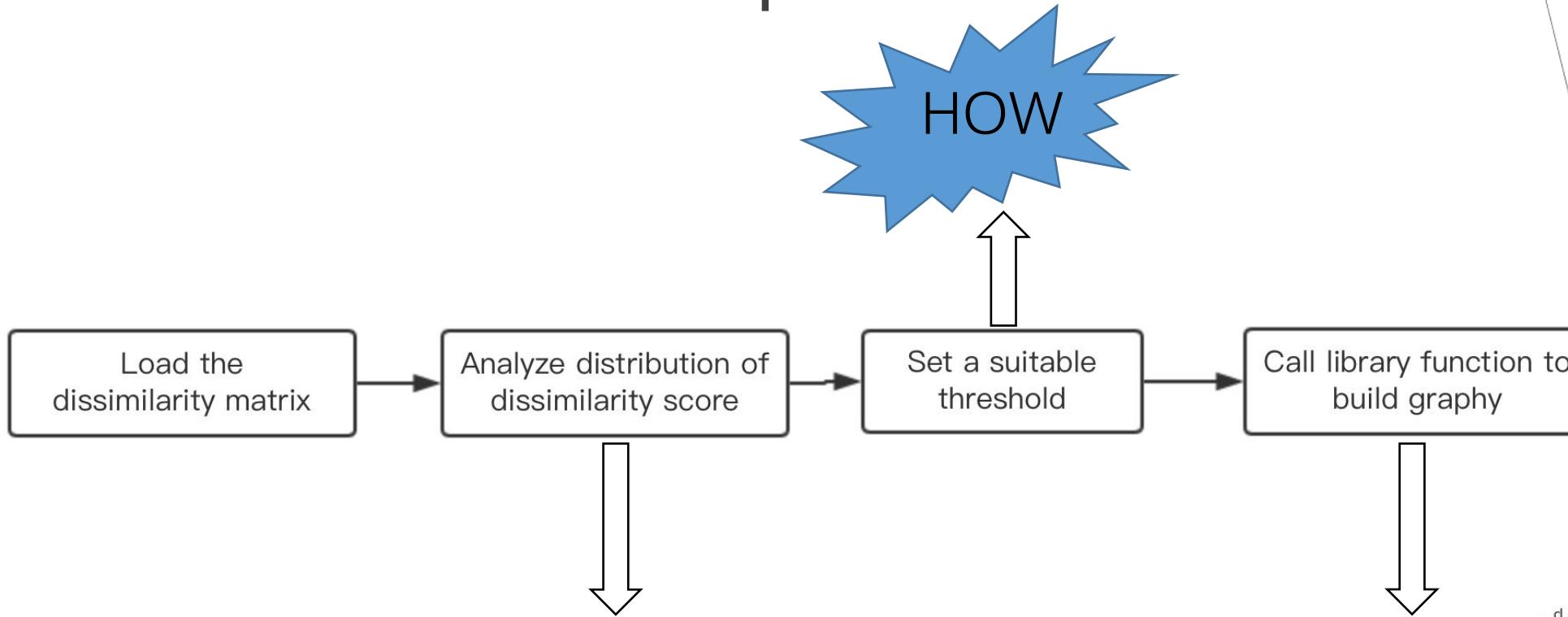
COMMUNITY DETECTION

- Compute dissimilarity matrix
- Build graph and network
- Community detection

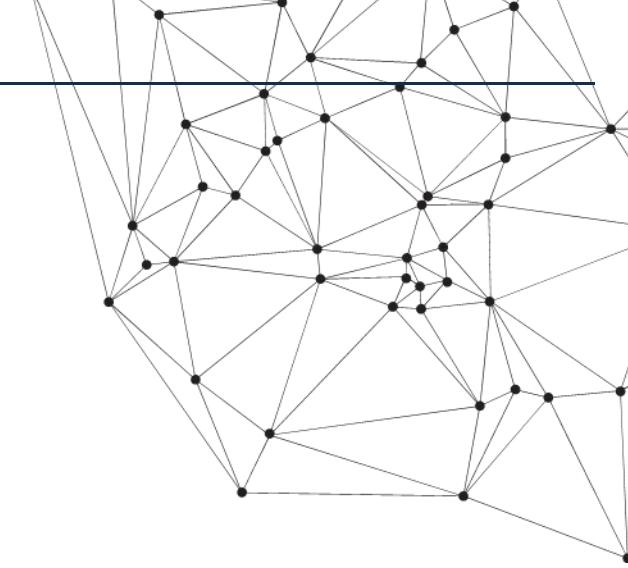
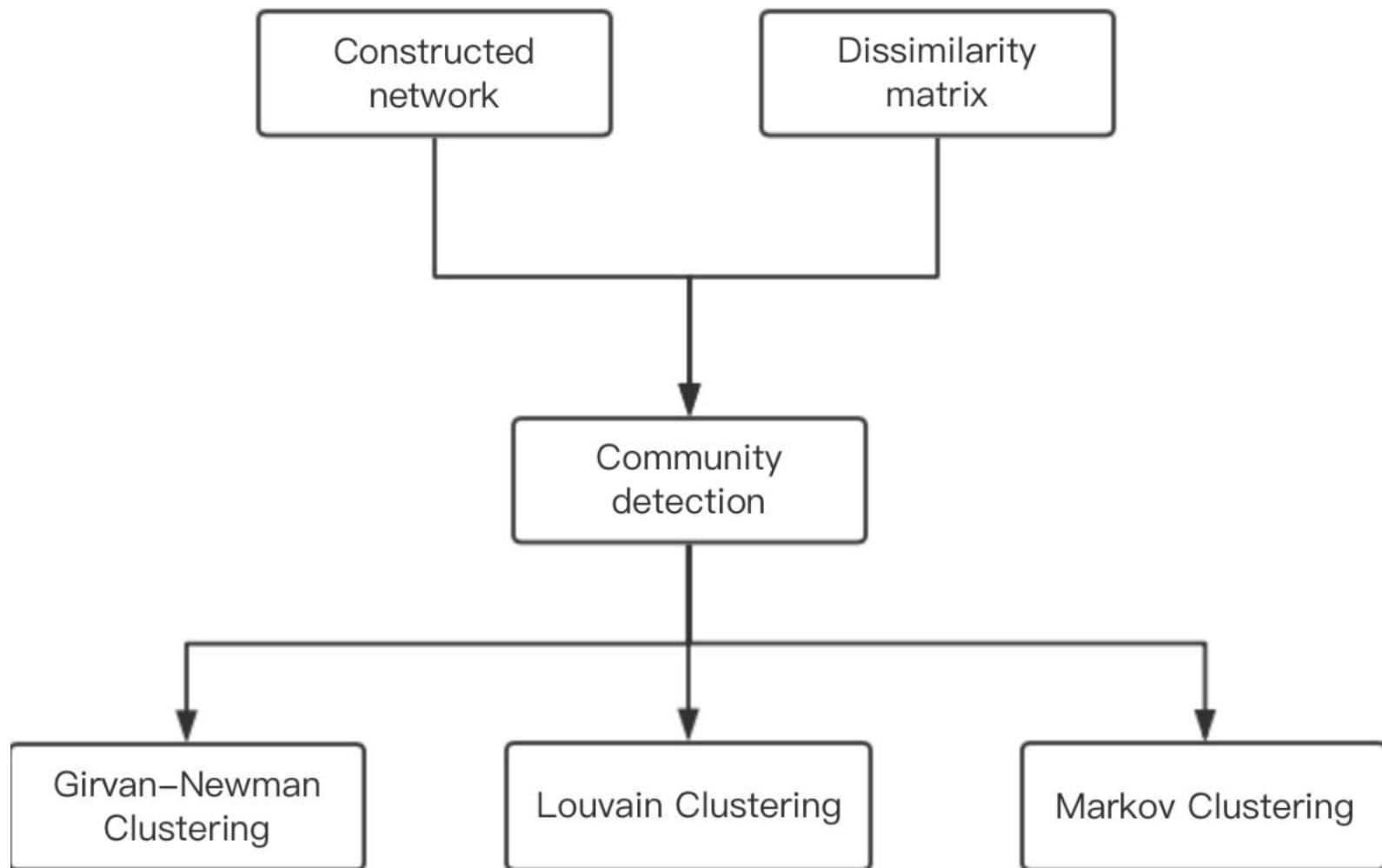
Compute Dissimilarity Matrix



Build Graph and Network

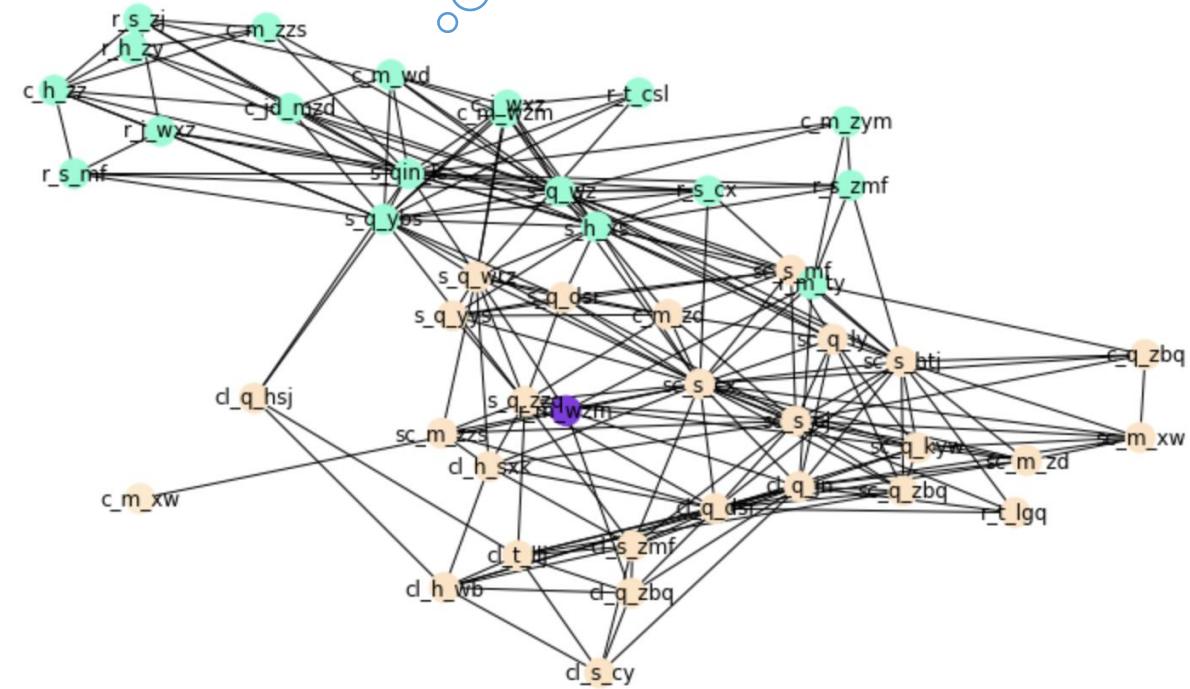


Community Detection



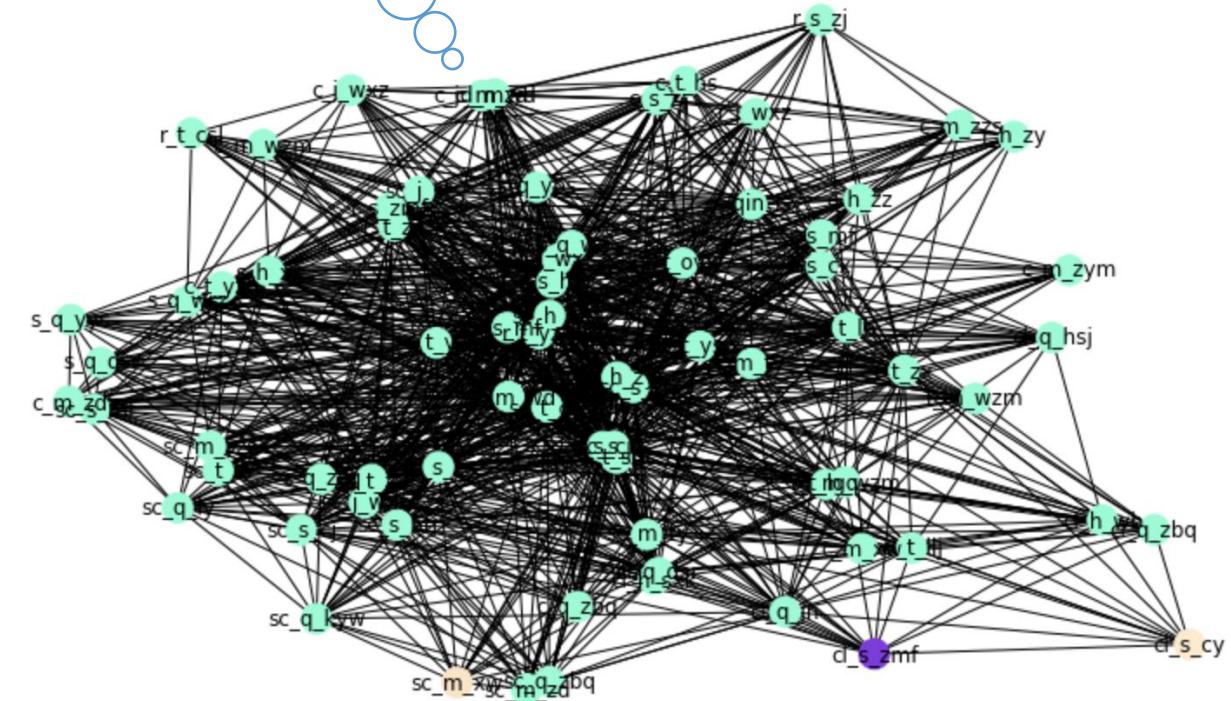
Girvan Newman

The number
of
community is
very few



47 Classes images

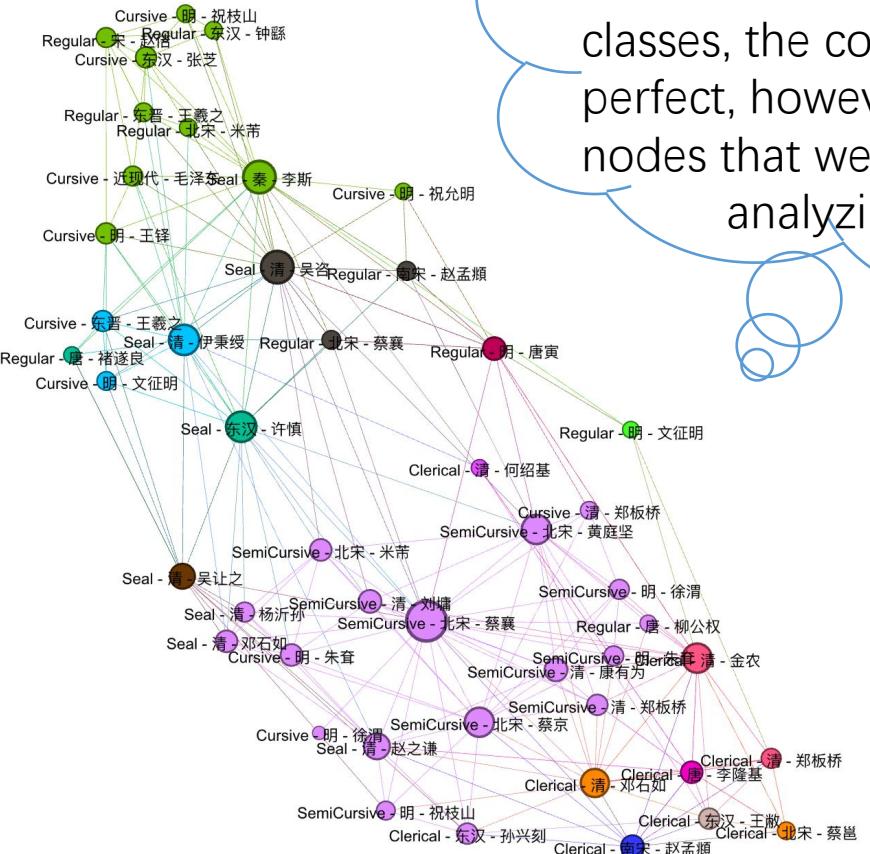
The core of the
community is
too dense



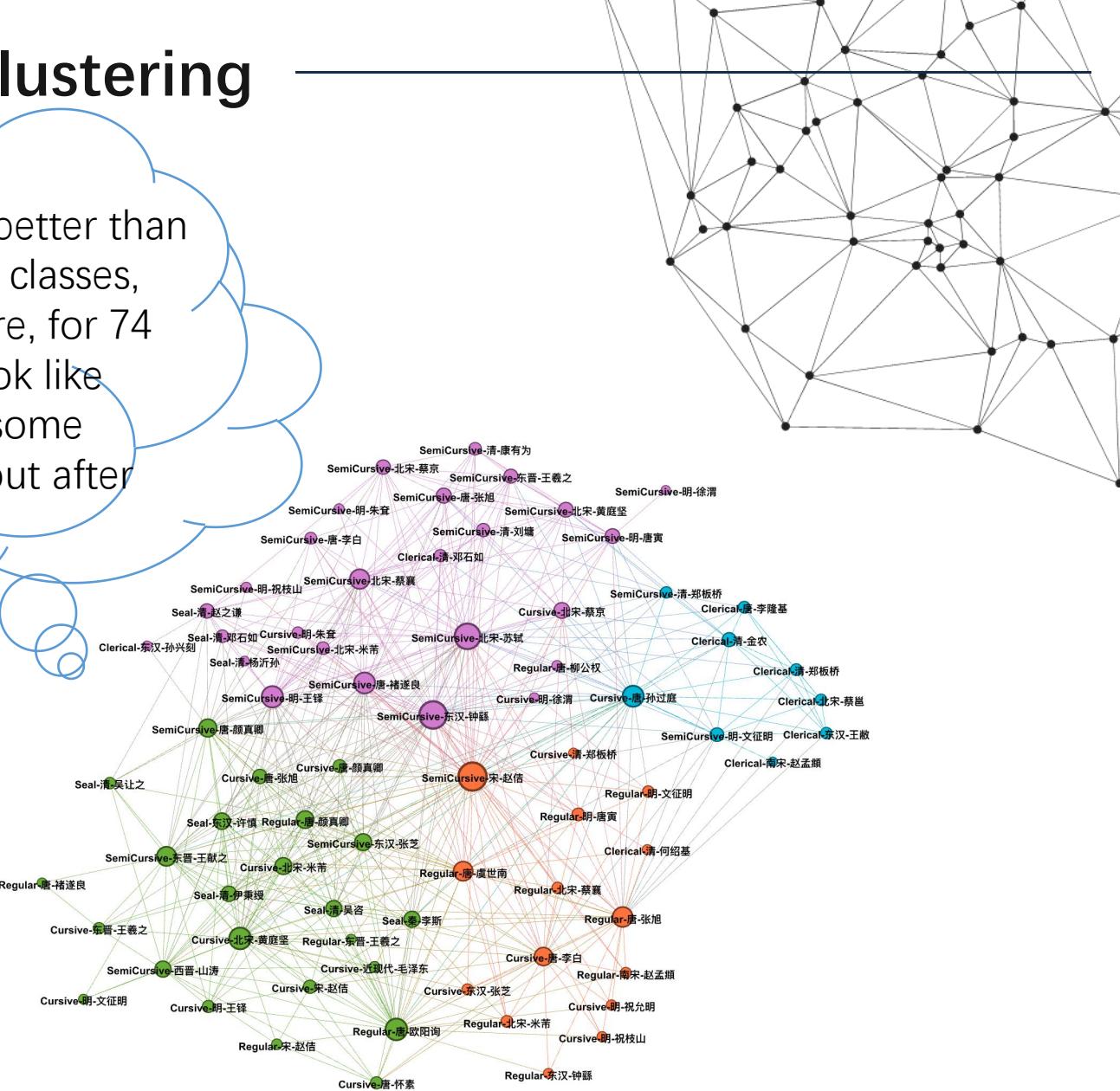
74 Classes images

Markov Clustering

Markov Clustering is much better than Girvan-Newman. but for 47 classes, communities are a little more, for 74 classes, the communities look like perfect, however there are some nodes that we can't figure out after analyzing



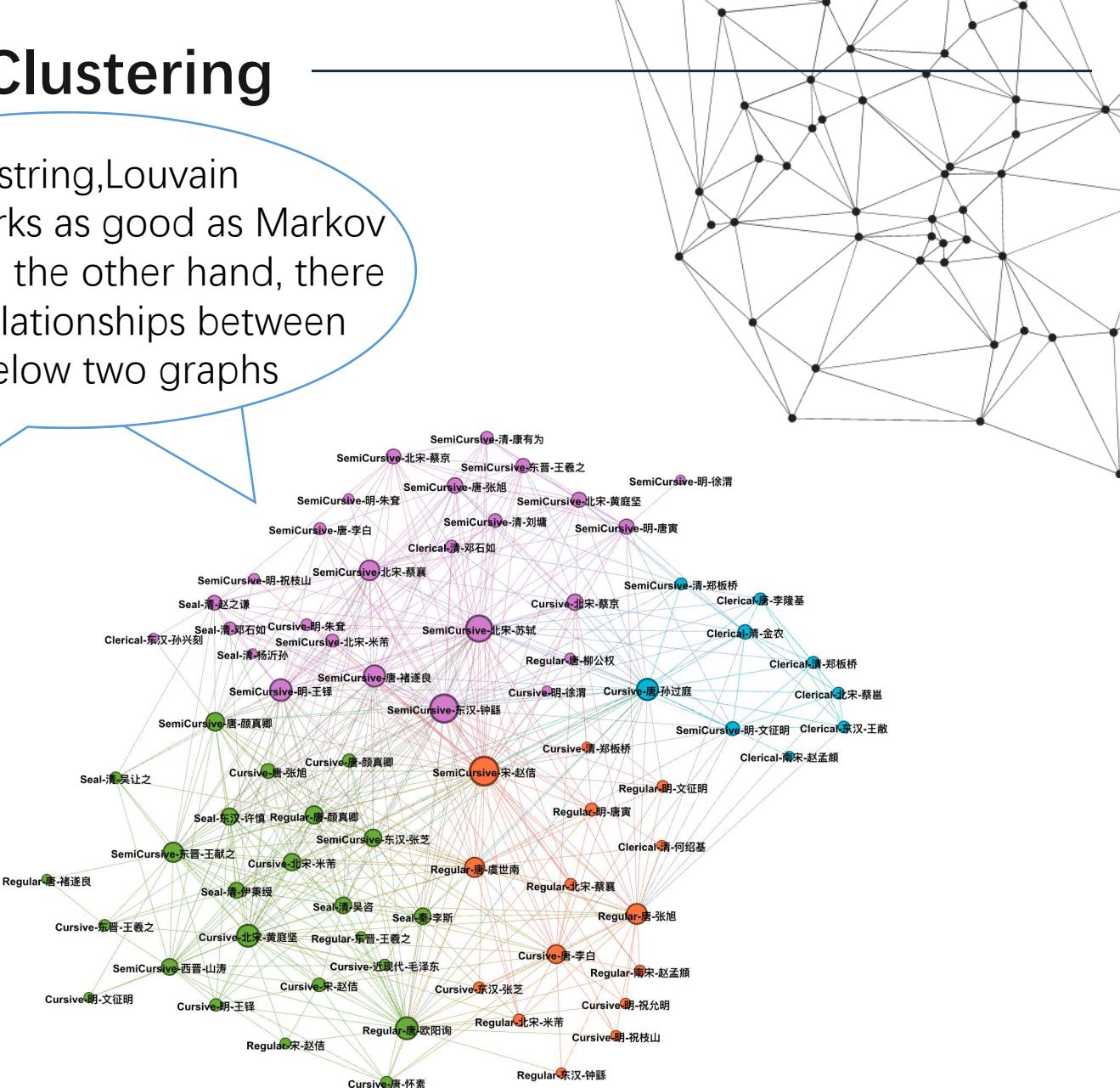
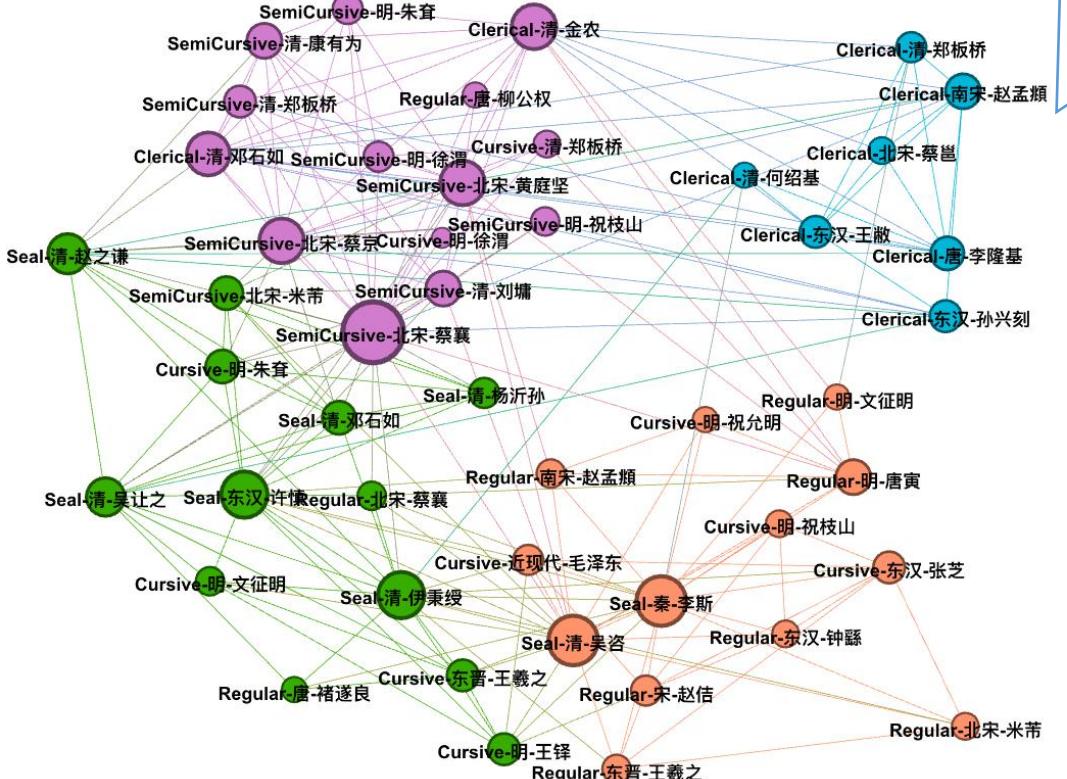
47 Classes images



74 Classes images

Louvain Clustering

About clustering, Louvain clustering works as good as Markov Clustering, on the other hand, there are better relationships between nodes in below two graphs

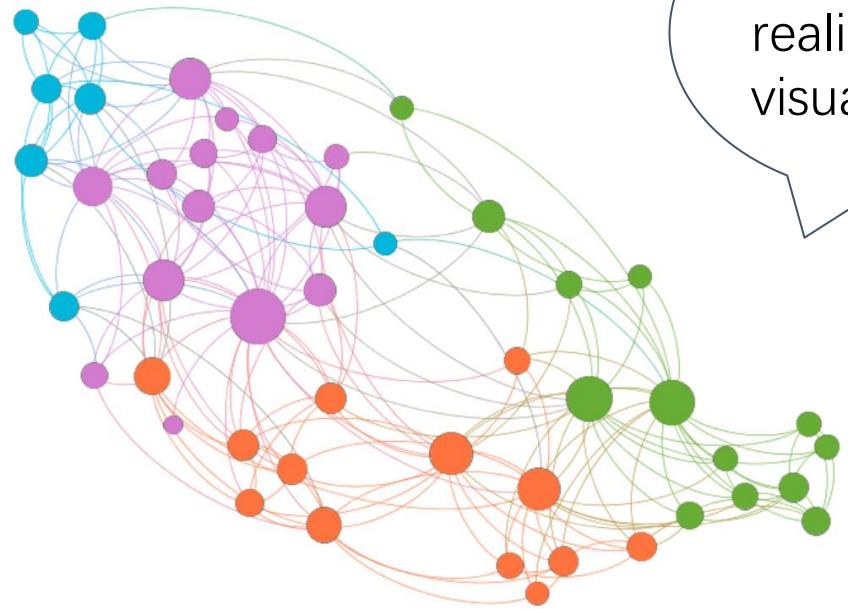


The background of the slide features a complex, abstract network graph composed of numerous small, light gray dots connected by thin white lines, creating a mesh-like pattern that covers the entire frame.

Part
Four

ANALYSIS OF THE RESULT

The visualization of the clustering results



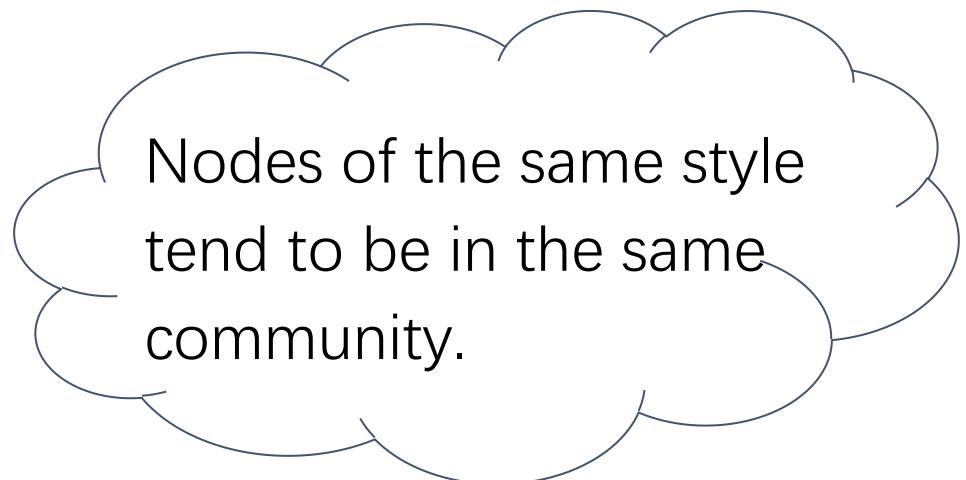
the result of 47 categories

Use gephi to
realize the
visualization.



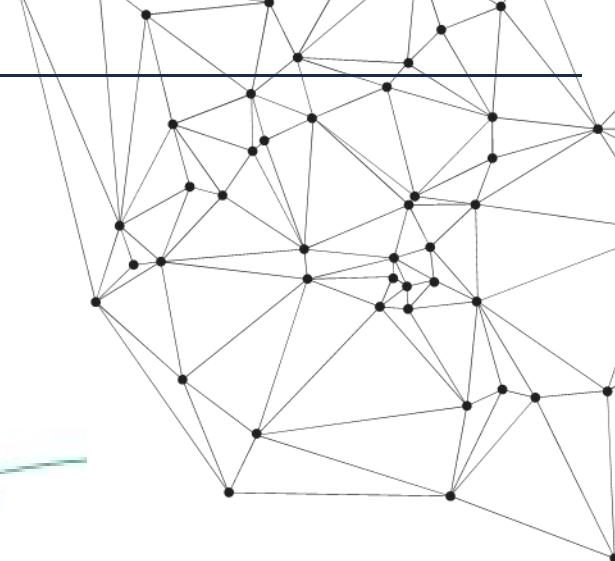
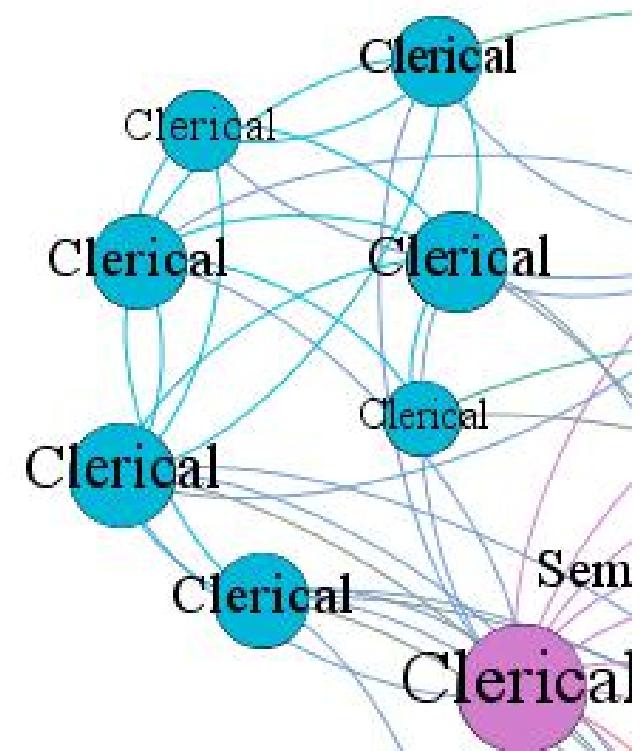
the result of 74 categories

What we expected from the old data set

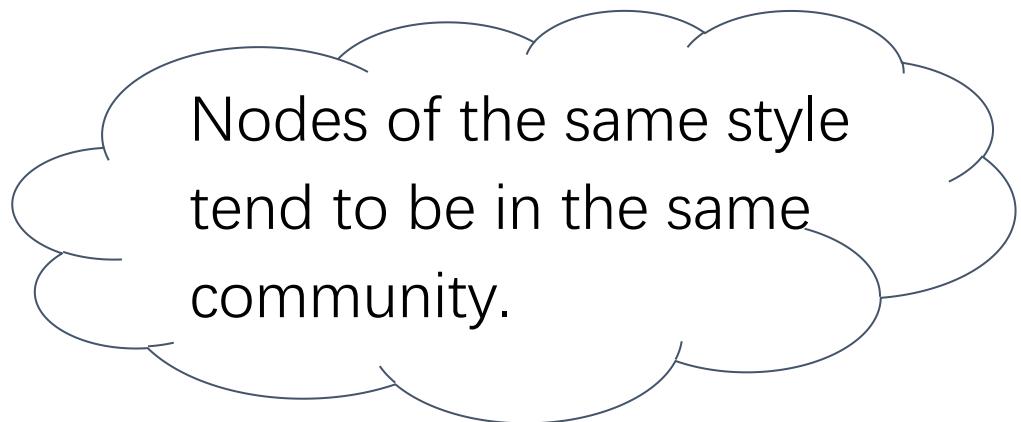


In this community:

- Only has clerical script
- Most clerical scripts are in it

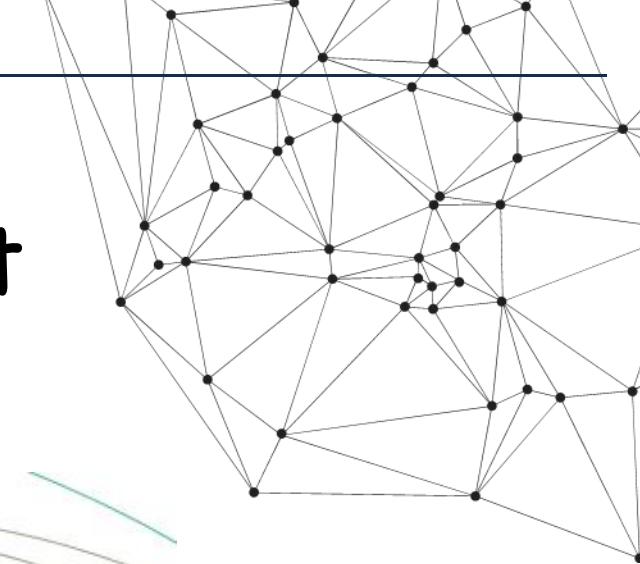
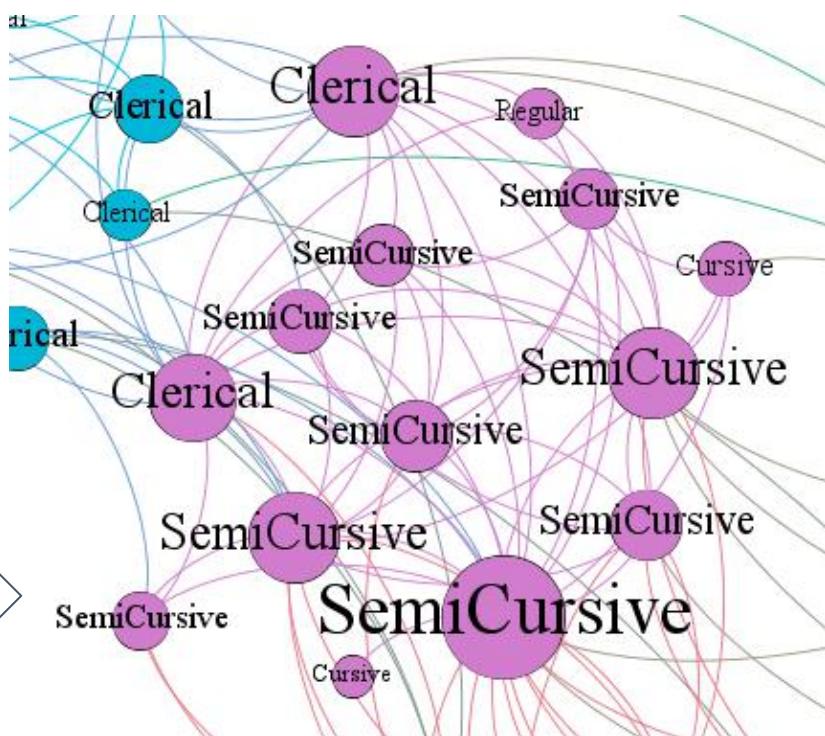


What we expected from the old data set



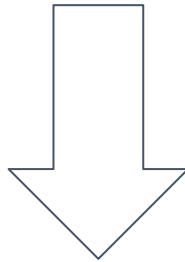
In this community:

- Mainly consist of semi-cursive script
- Most semi-cursive scripts are in it

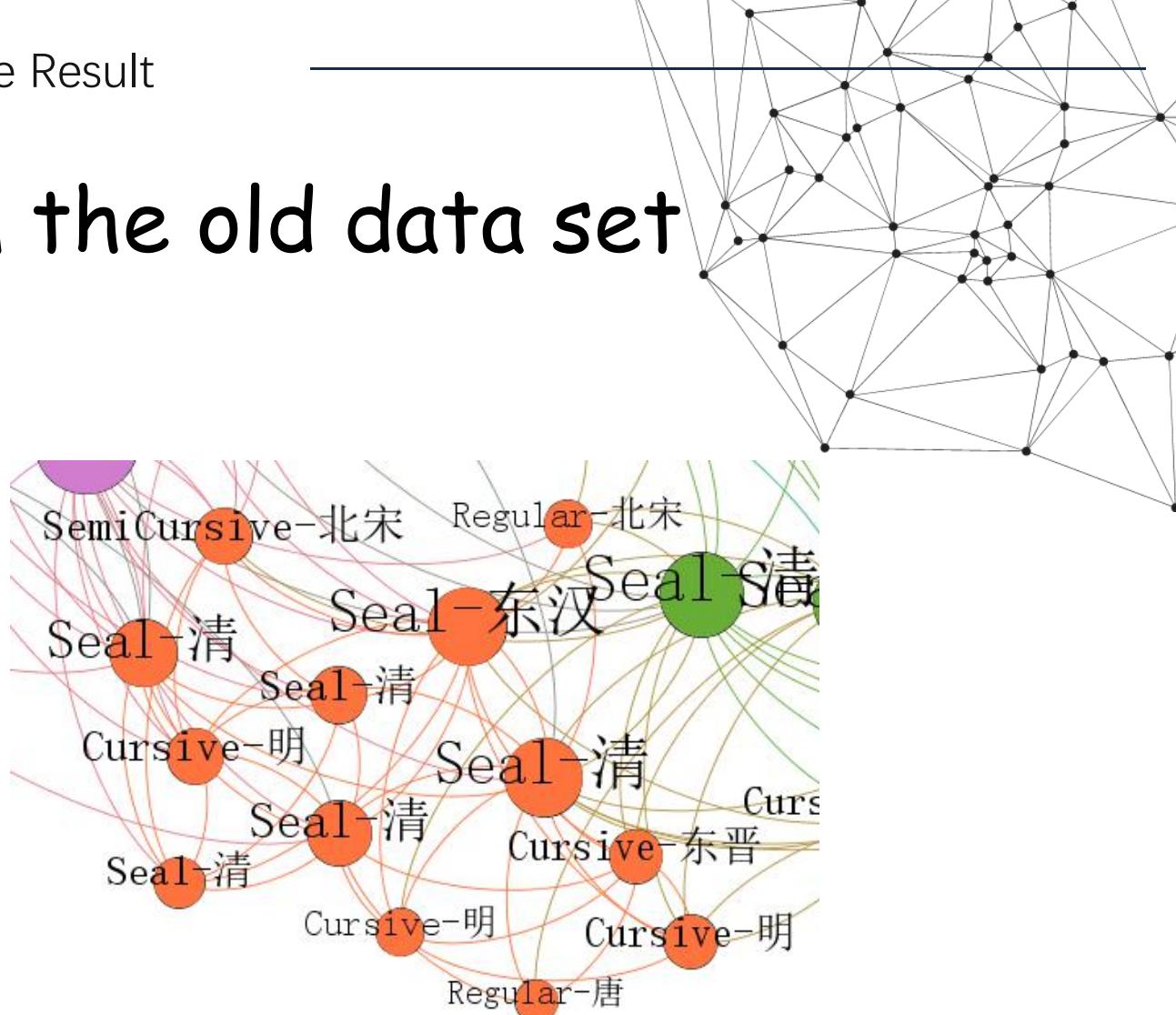


New insights we get from the old data set

Many seal script fonts in Qing dynasty are clustered with cursive script fonts



The evolution of seal script in Qing dynasty made it more similar to the characteristics of cursive script

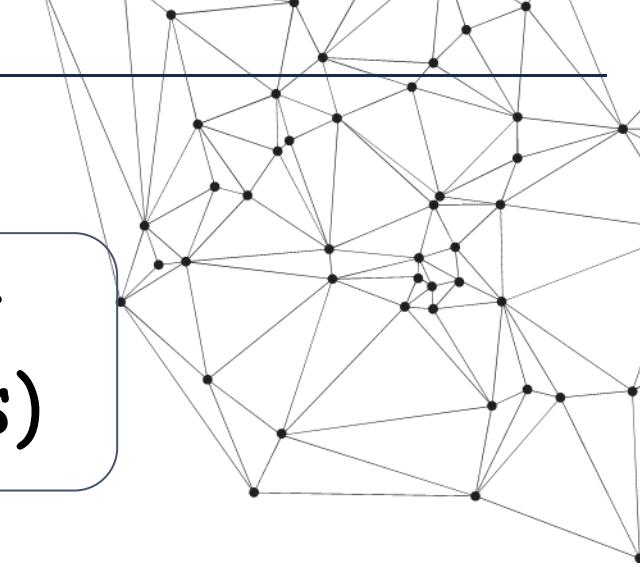


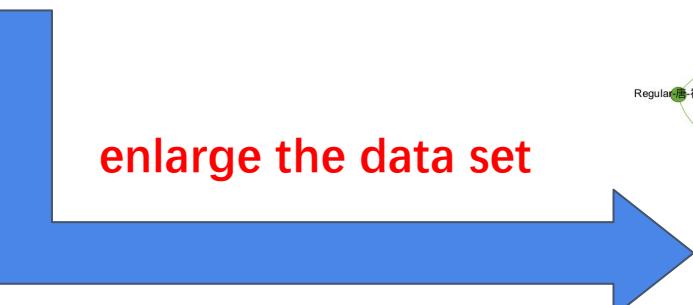
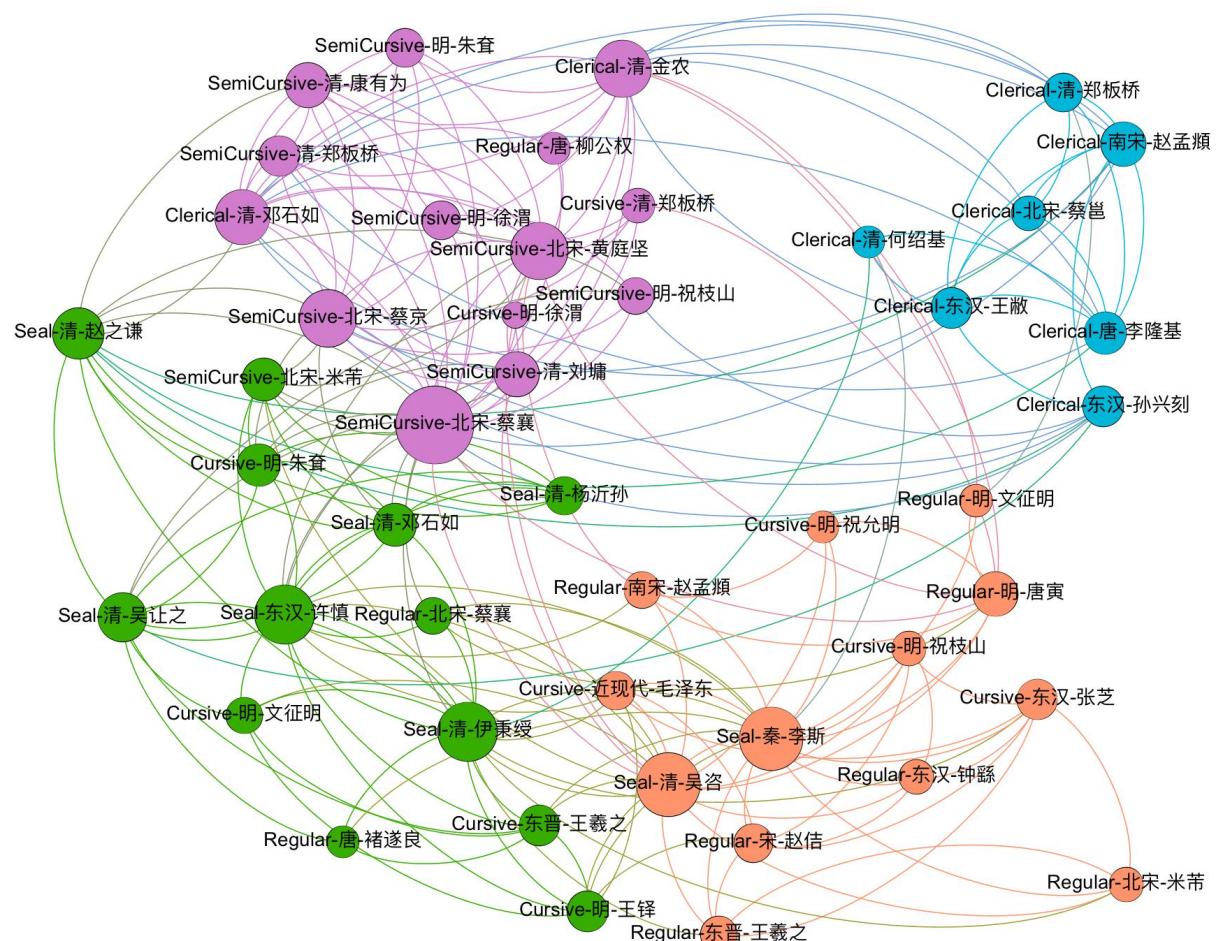
Old data set
(47 categories)

New data set
(74 categories)

Expand

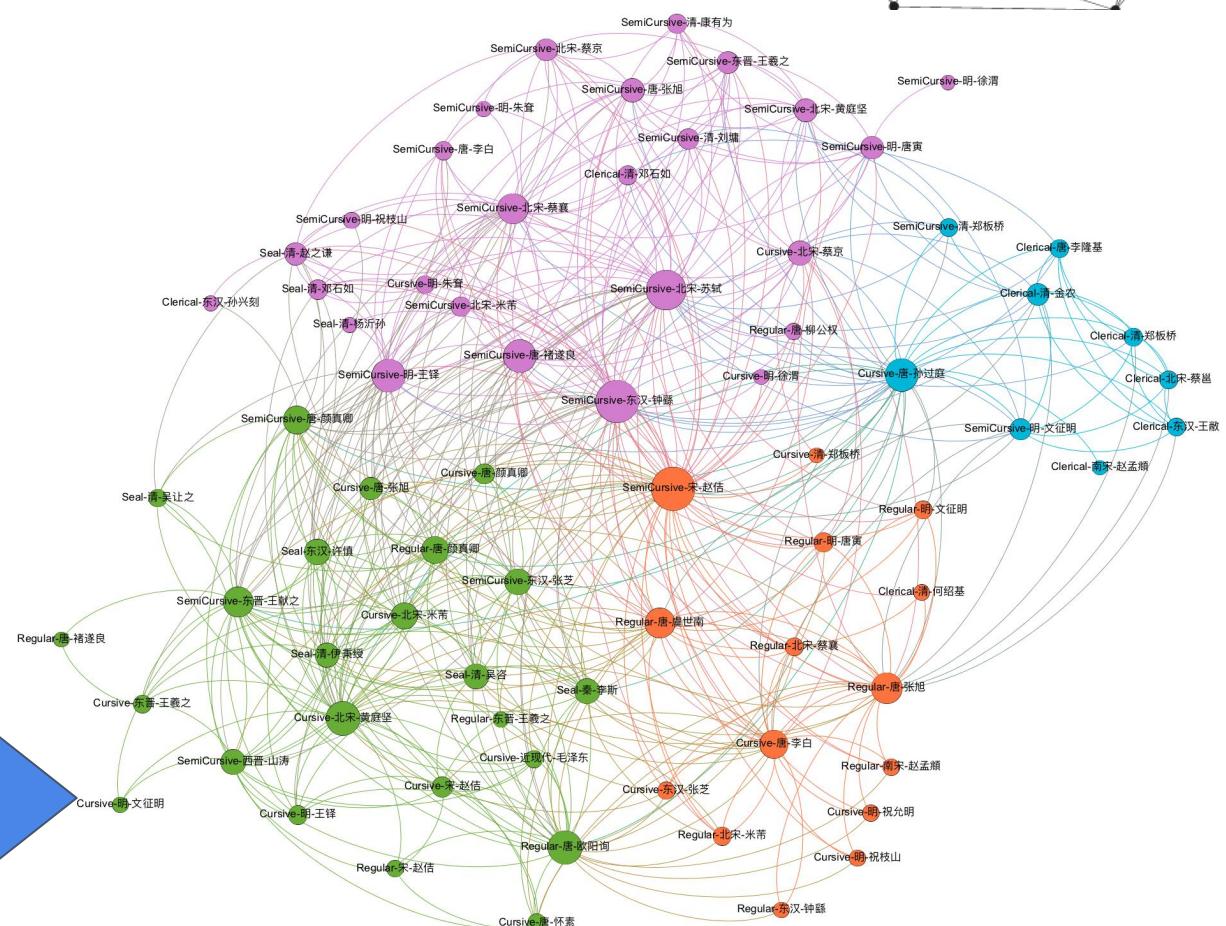
Connection and
Difference





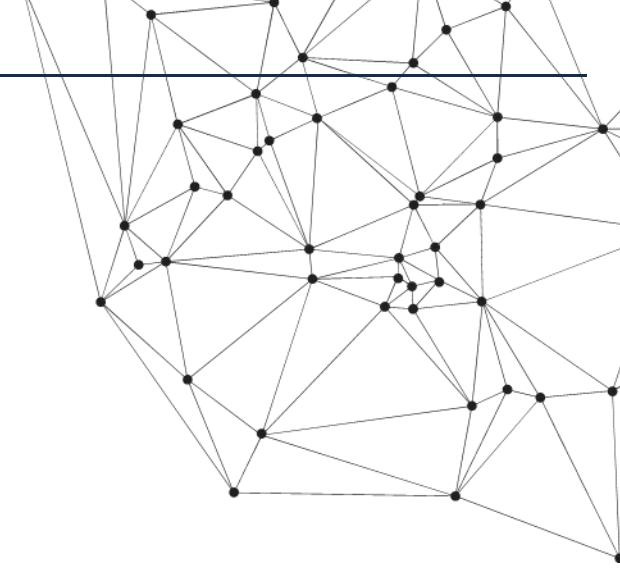
Communities of the same colour nearly have the same core.

Graph of 74 classes

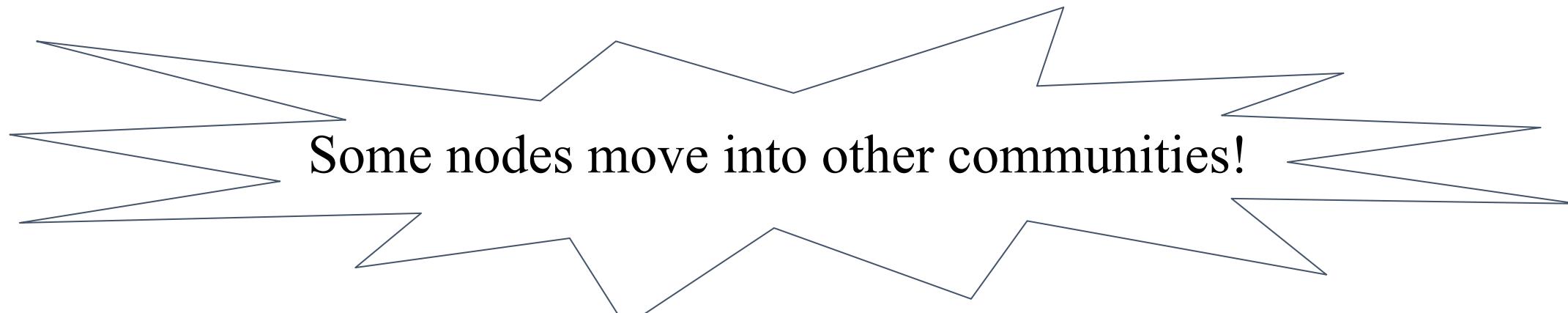


Connection

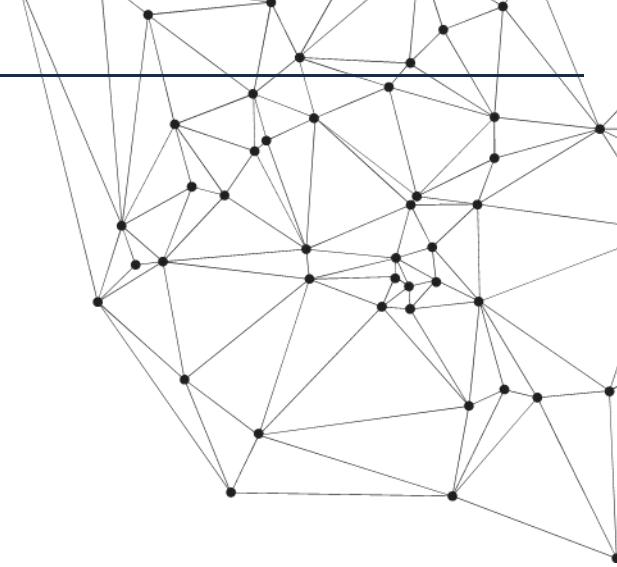
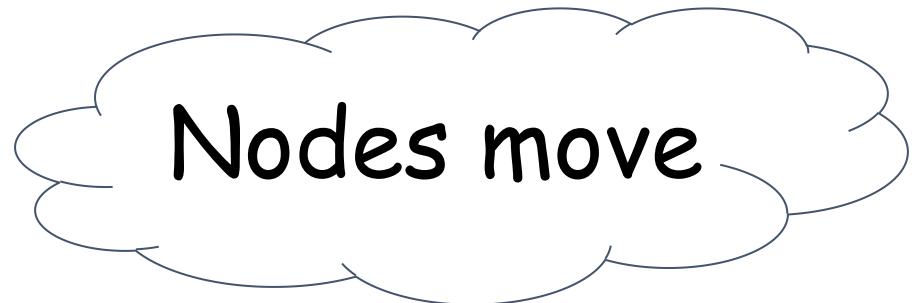
- ❖ Both data sets are clustered into four communities.
- ❖ The core of each community changes little.



Difference



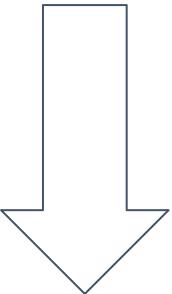
Some nodes move into other communities!



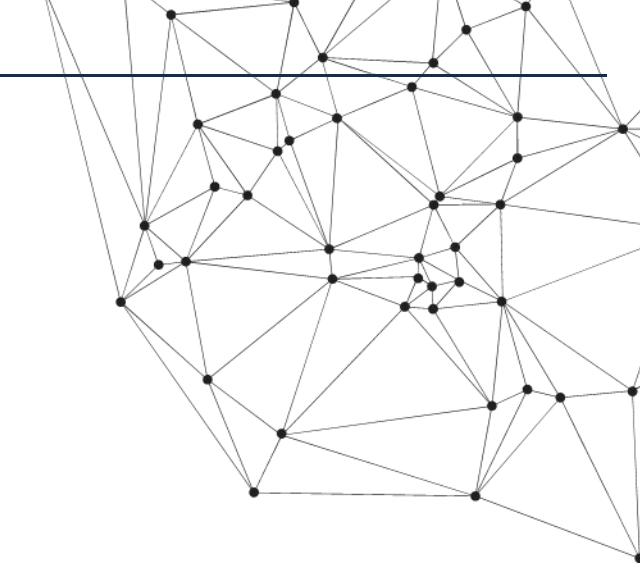
- ❖ Why they move
- ❖ Why some of them move together

Why they move

Some new nodes which have more connections with some old nodes are added



Such old nodes tend to move into a more corresponding community together with those new nodes

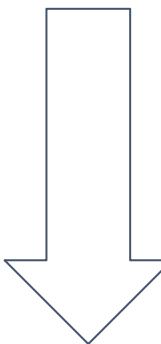


Analysis of the Result

Examples

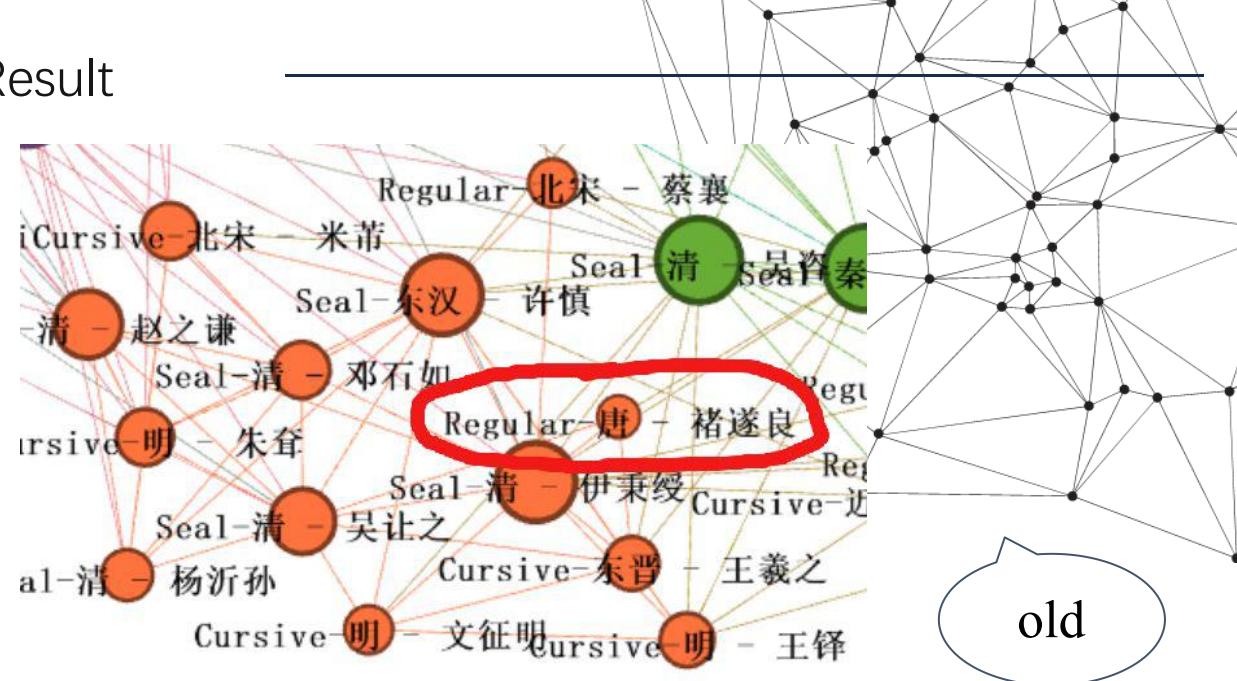
Each color means one community

Chu Suiliang studied the calligraphy
of Ouyang Xun and Wang Xizhi



the regular script of Ou
Yangxun was added

The same script of Chu Suiliang
moved to the same community with
those two



old

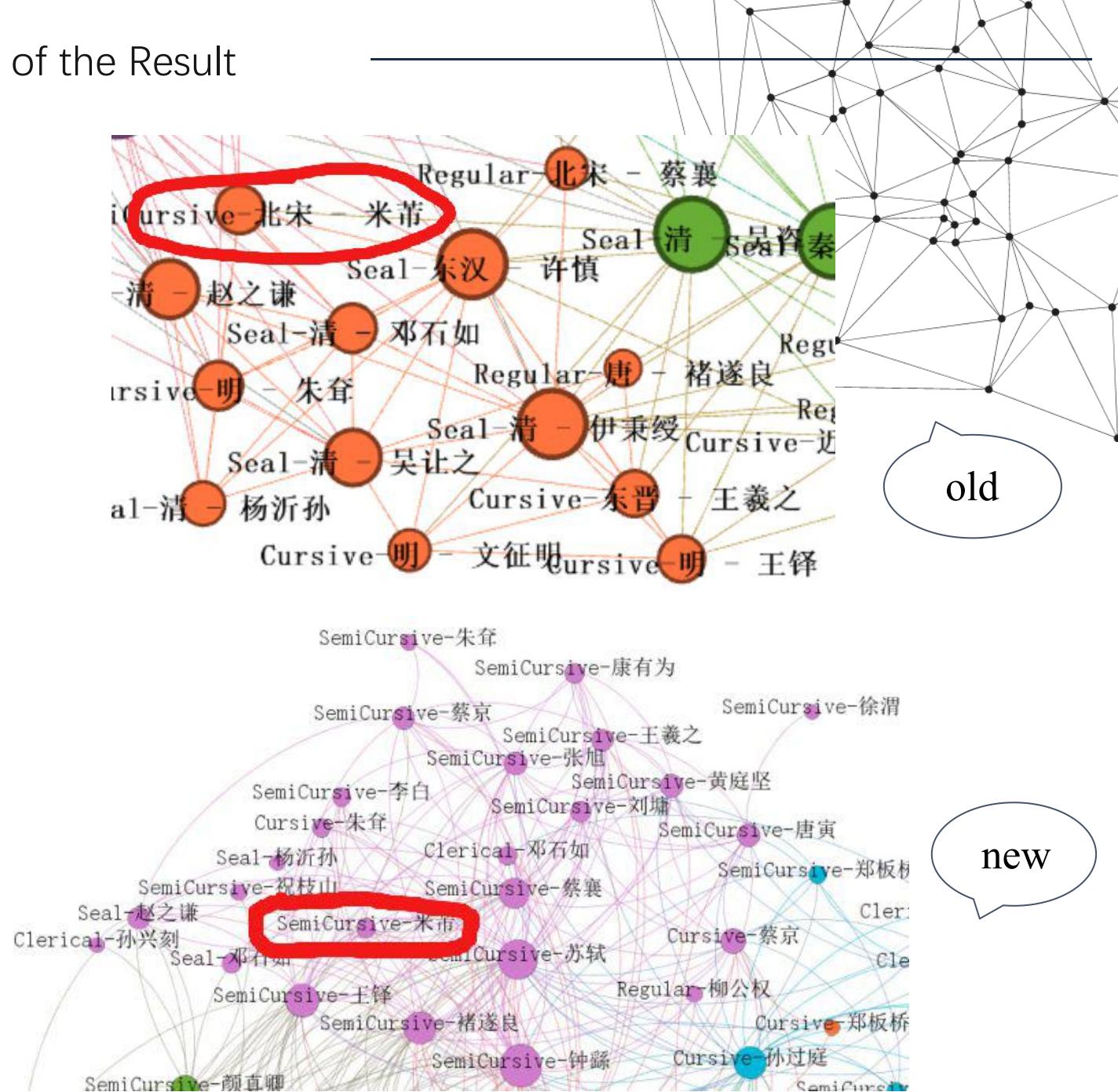
new

Examples

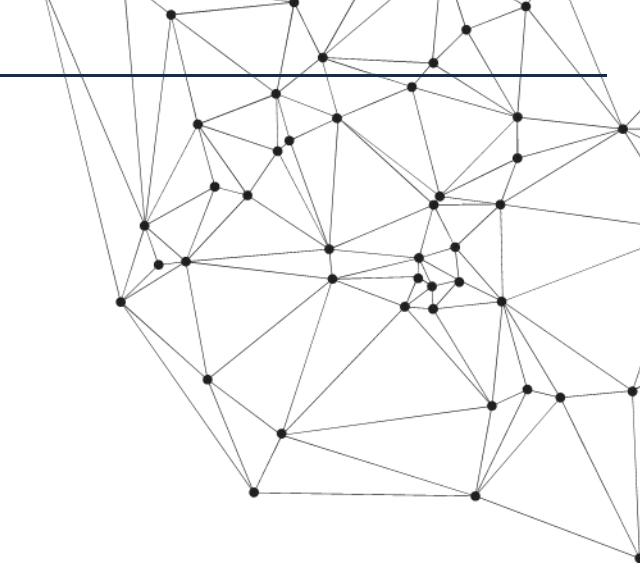
It's difficult to explain why Mi Fu's semi-cursive script were clustered into the orange community before

more semi-cursive scripts are added

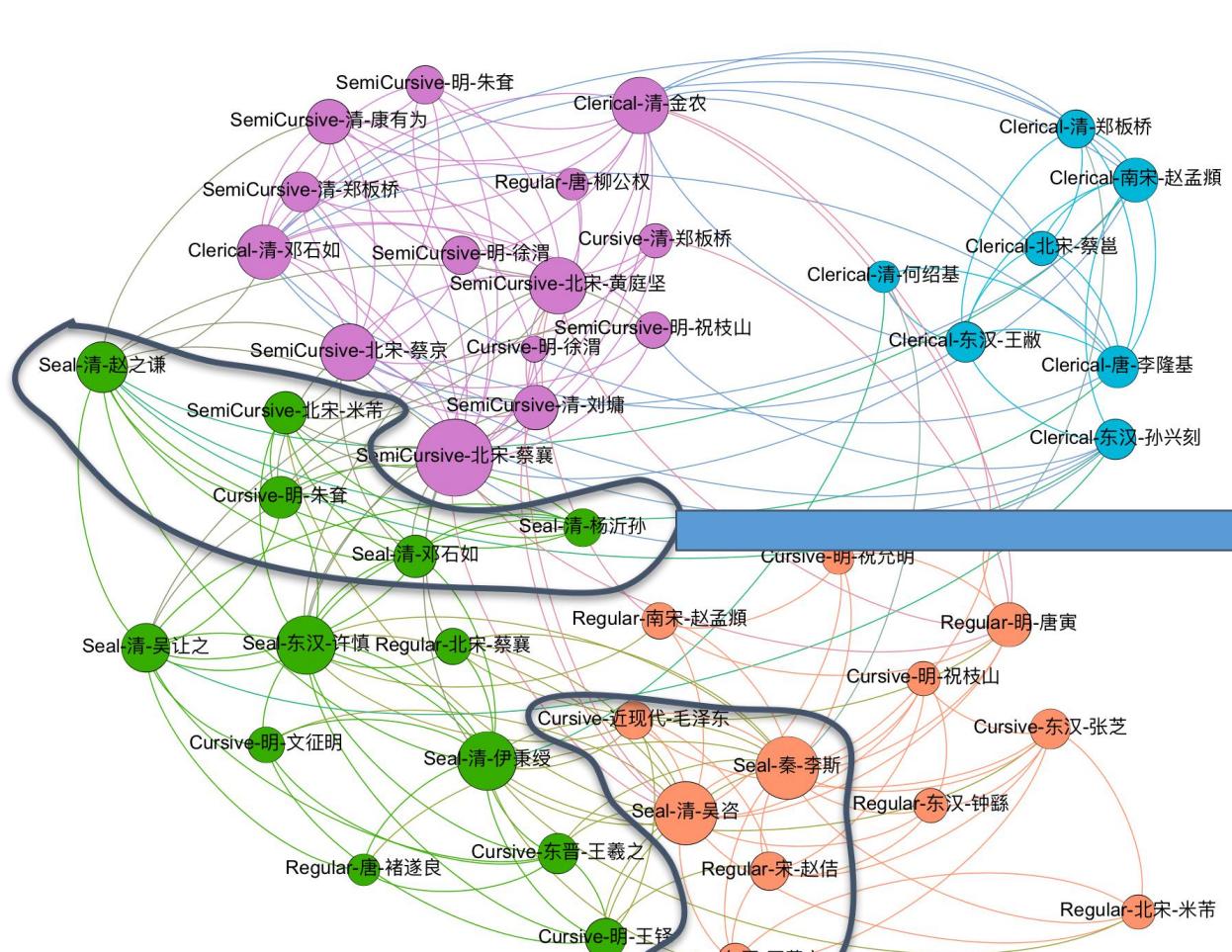
It moved into the purple community which contains most semi-cursive scripts



Why they move together



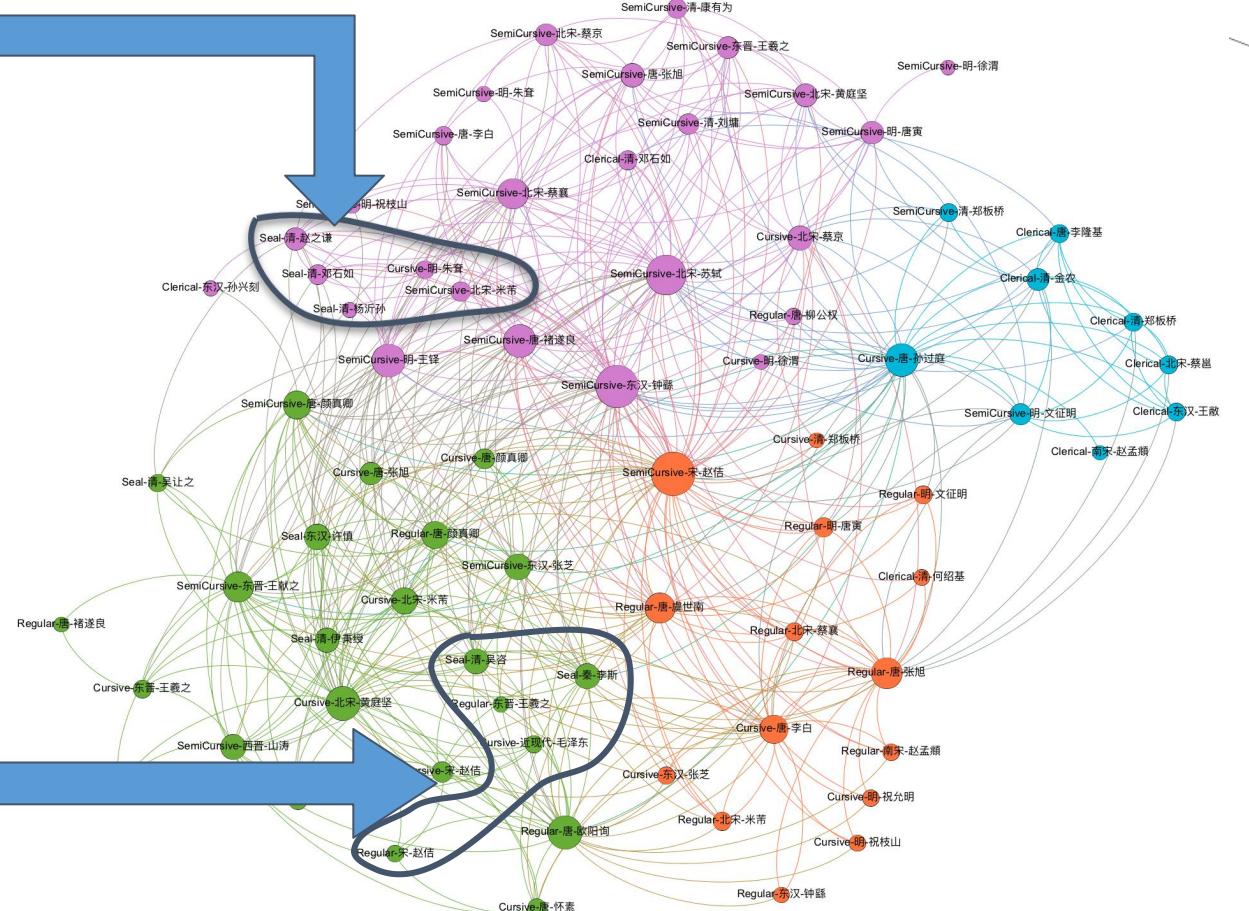
Most moving nodes in the same
community move to the same other
community



1.cursive-朱耷 seal-邓石如 seal-杨沂孙
seal-赵之谦 semicursive-米芾

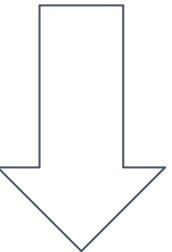
2.seal-吴咨 seal-李斯 regular-赵佶
cursive-毛泽东 regular-王羲之

Graph of 74 classes

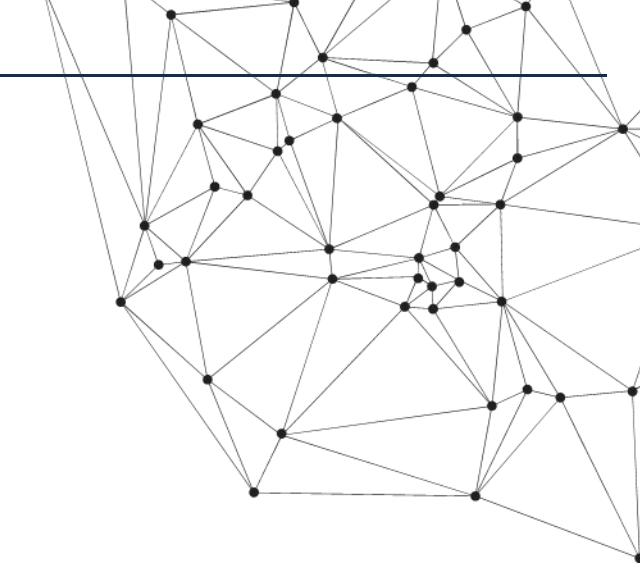


Why they move together

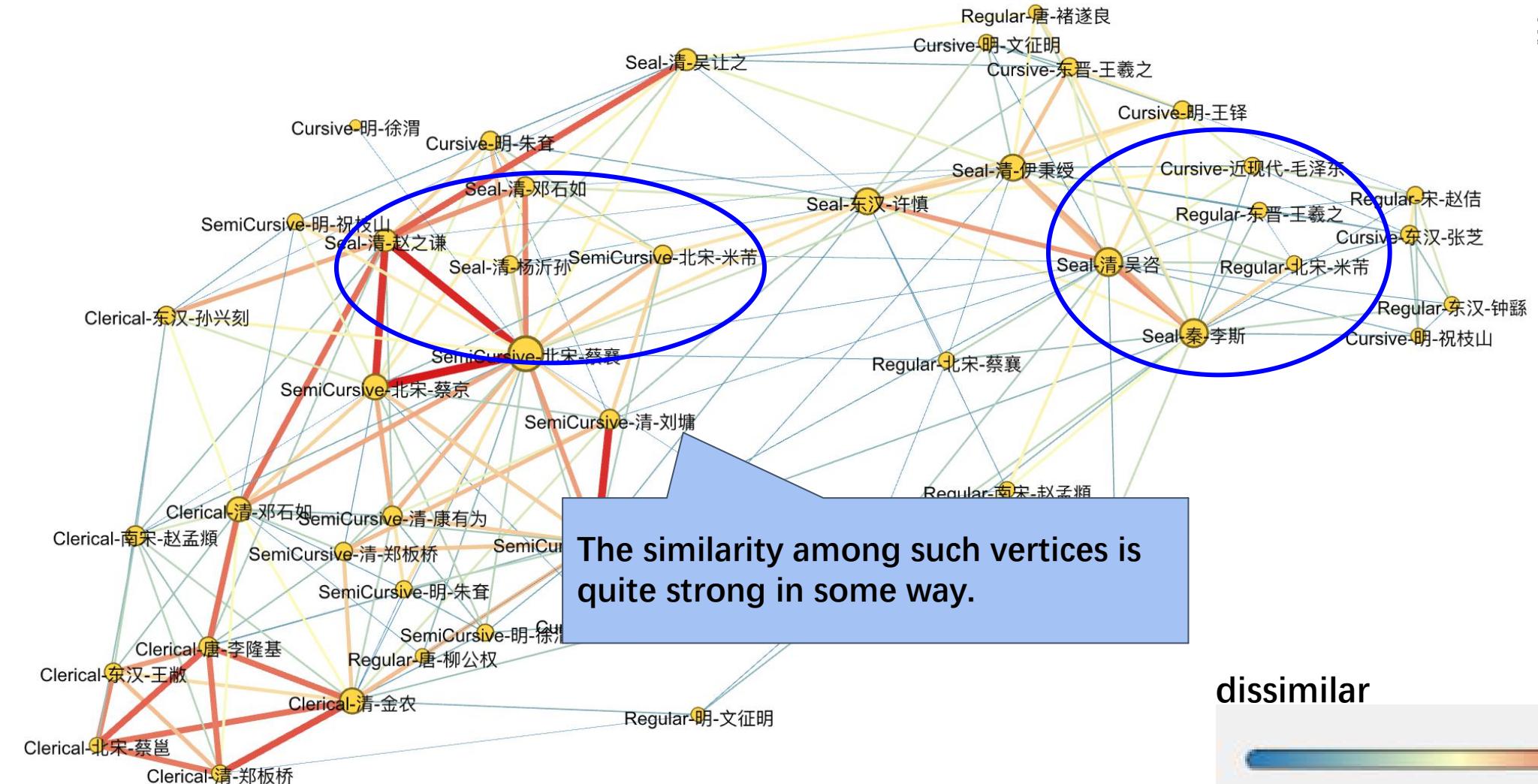
Most moving nodes in the same community move to the same other community



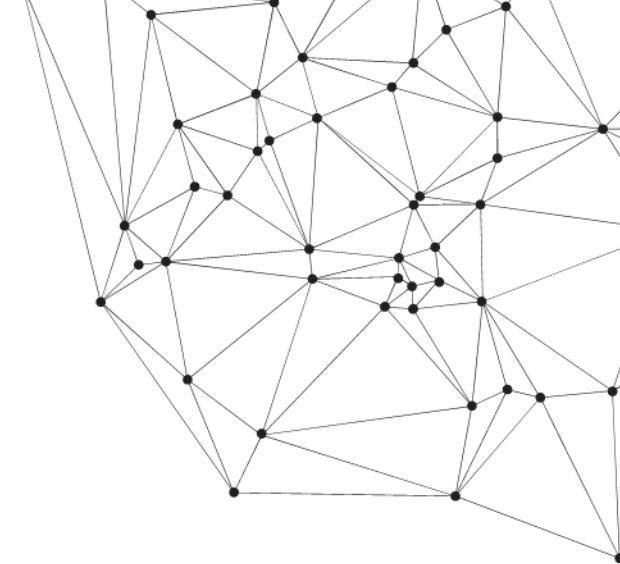
They connect with each other strongly



Weighted Graph



Possible reasons



- Similar characteristics

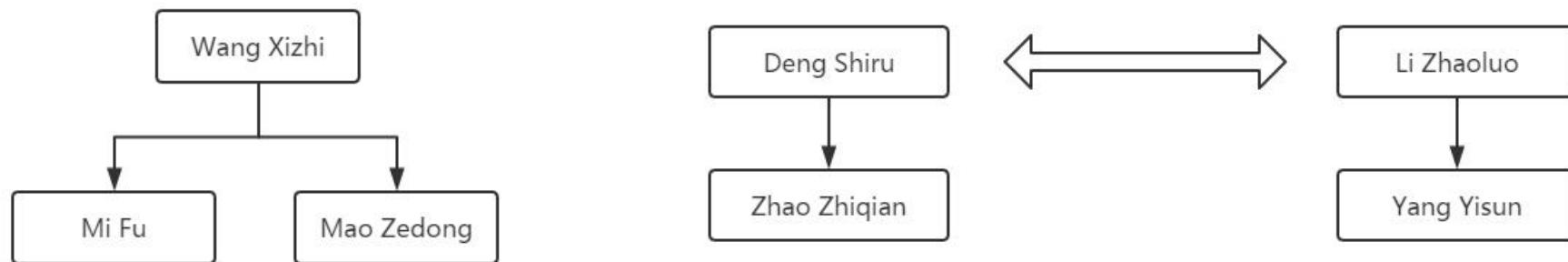
鵠 桂 離 魏 越 開 祖 穢 爾

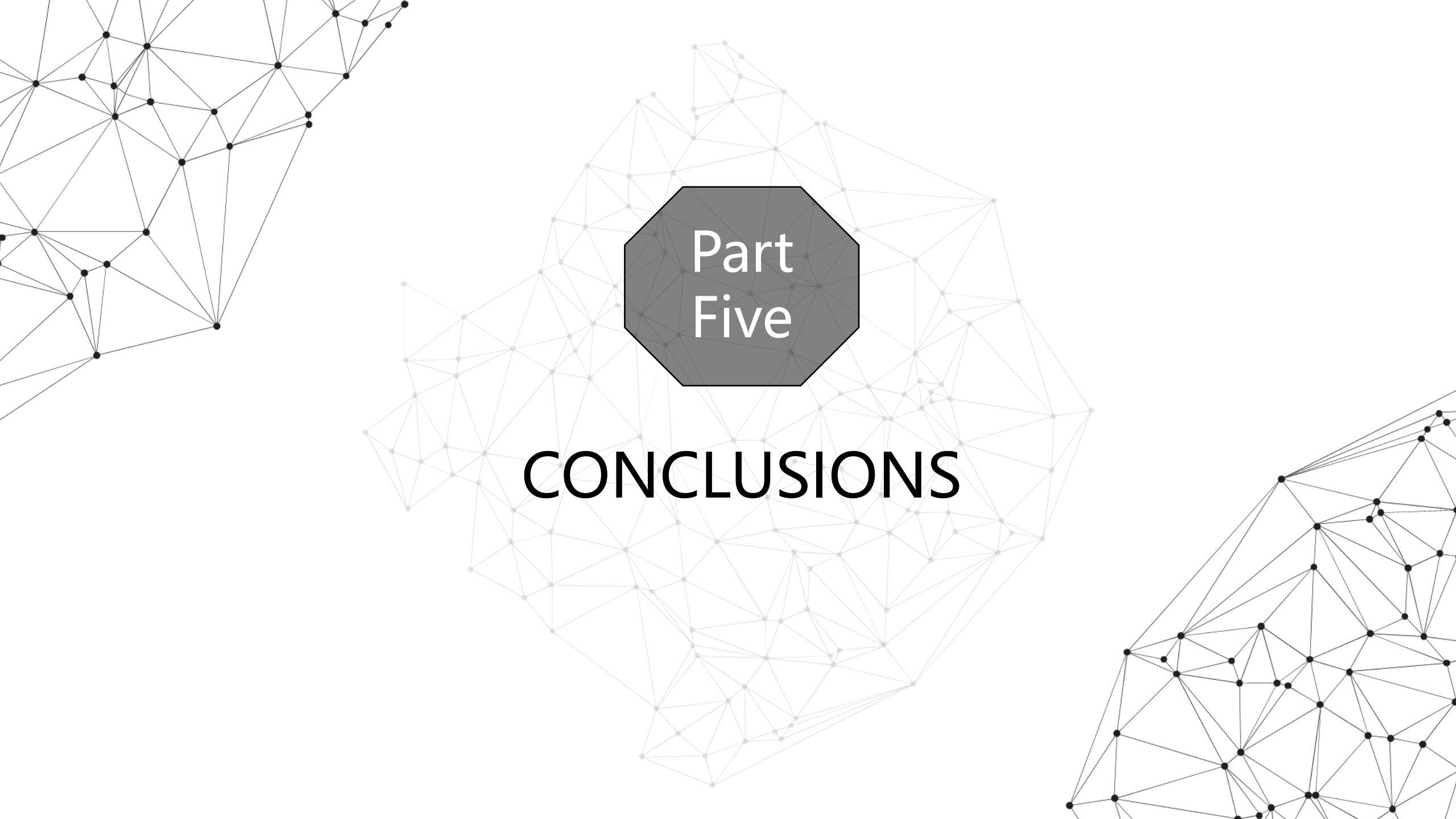
Seal of Li Si

懷 香 技 宿 淑 戶 乃 稹 穰 開

Seal of Wu Zi

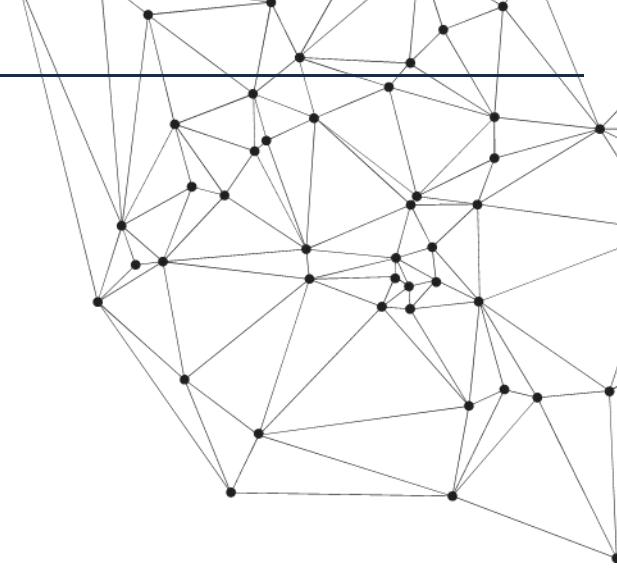
- The learning relationship between calligraphers



The background of the slide features a complex, abstract network structure composed of numerous small, light gray dots connected by thin white lines, creating a mesh-like pattern that covers the entire frame.

Part
Five

CONCLUSIONS



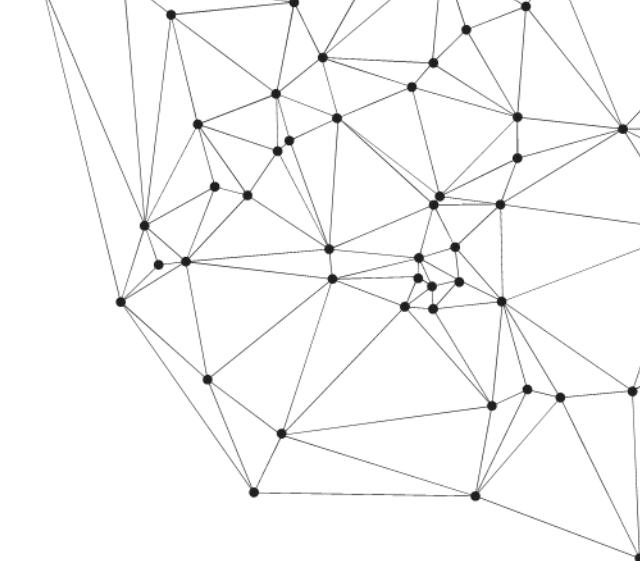
Conclusions

❖ What we expected

- Nodes of the same style tend to be in the same community.

❖ Interesting findings

- Even the same style of calligraphy works may be clustered into different communities because of historical factors, such as evolution or revolution of calligraphy, and so on.
- On the contrary, different styles may be clustered into the same community.
- As the expanding of data set, even the core of each community changes little, some vertices are clustered to other communities. Moving nodes in one community always move together.

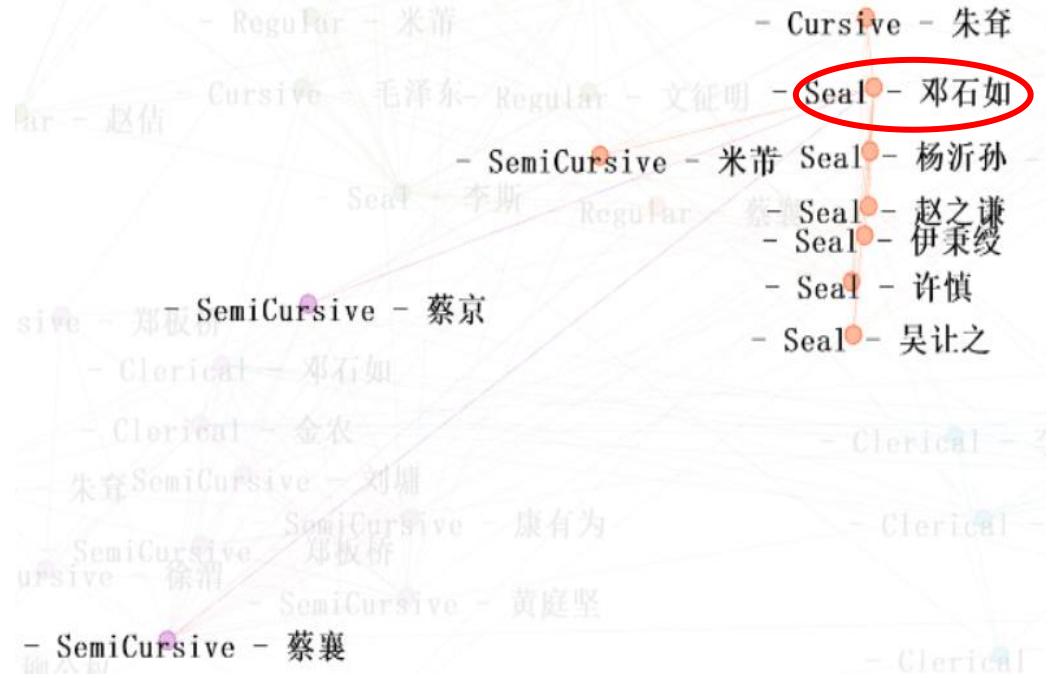


How can the findings be used?

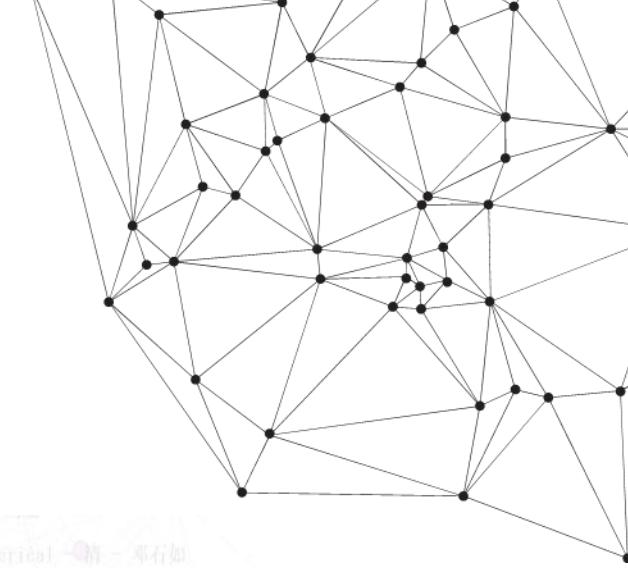
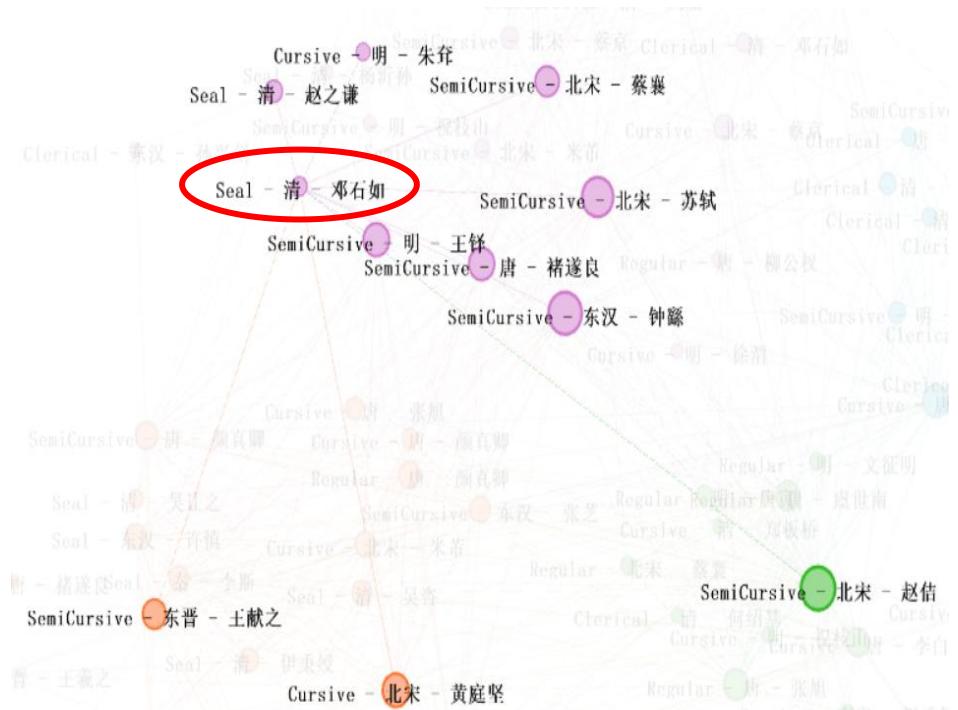
- Study calligraphy history or validate calligraphy knowledge.
- Find relationships among calligraphy or calligraphers which are rarely paid attention to.
- Gain new perspectives to classify calligraphy instead of traditional classification (types of style).



Seal-邓石如

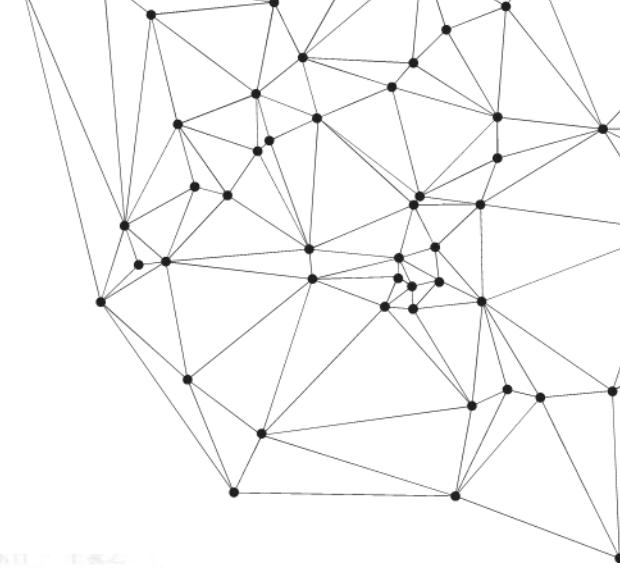


Communities of the same colour nearly have the same core.

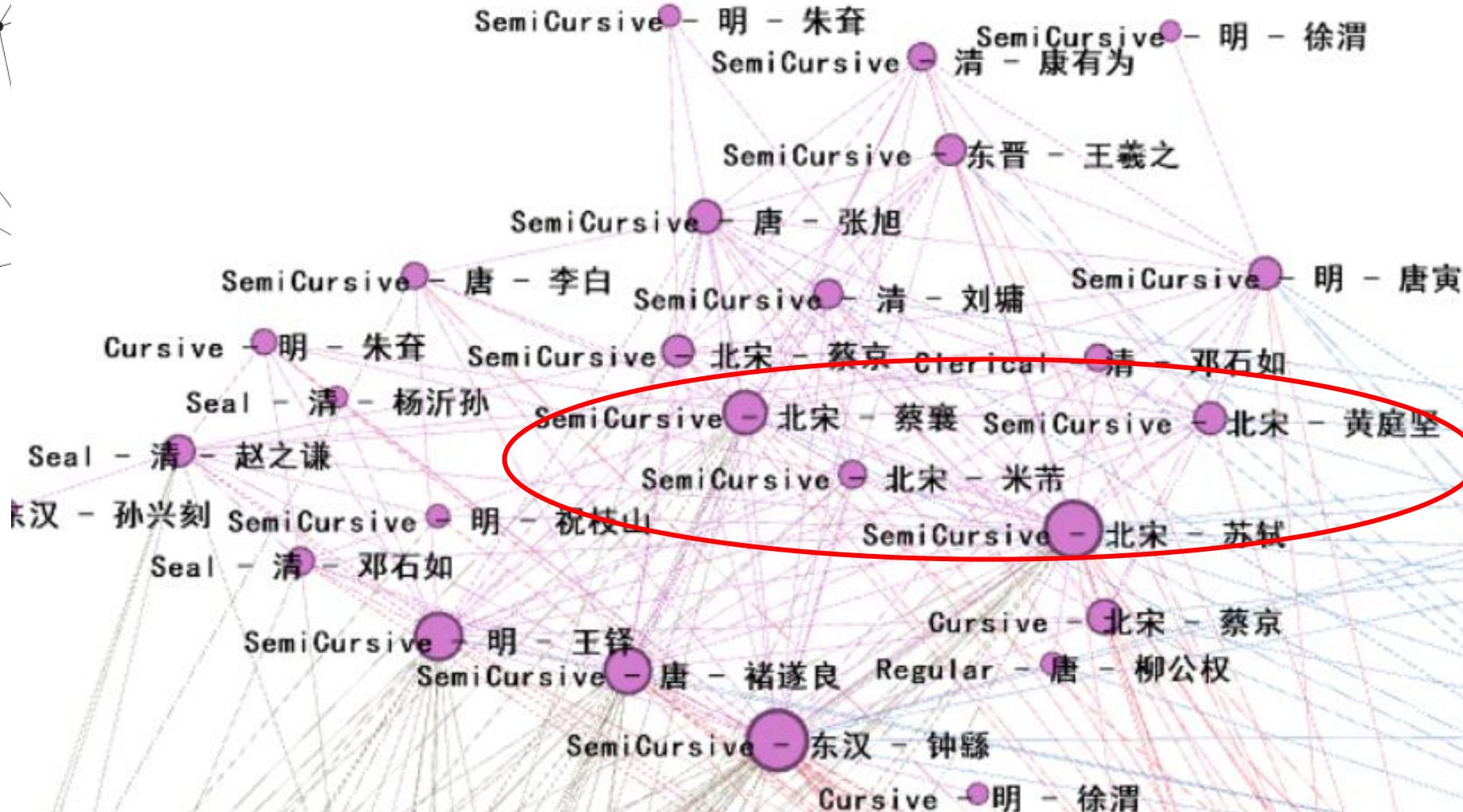


semicursive-米芾

Communities of the same colour nearly have the same core.



the Four Masters of the Song Dynasty (Chinese: 宋四家)



Four Talents of Wu (Suzhou) (Chinese: 吴中四才子)

