



# Selecting Most Influential Topics In The Future

**Zhao Jinghao**

**5140309102**

# CONTENTS

## Abstract

PART ONE

## Topic Map

PART TWO

## Features Extraction

PART THREE

## Prediction

PART FOUR

## Conclusion

PART FIVE

## References

PART SIX



# **Abstract**

## **PART ONE**

## Abstract

### **K-core Analysis of Scholarly Network**

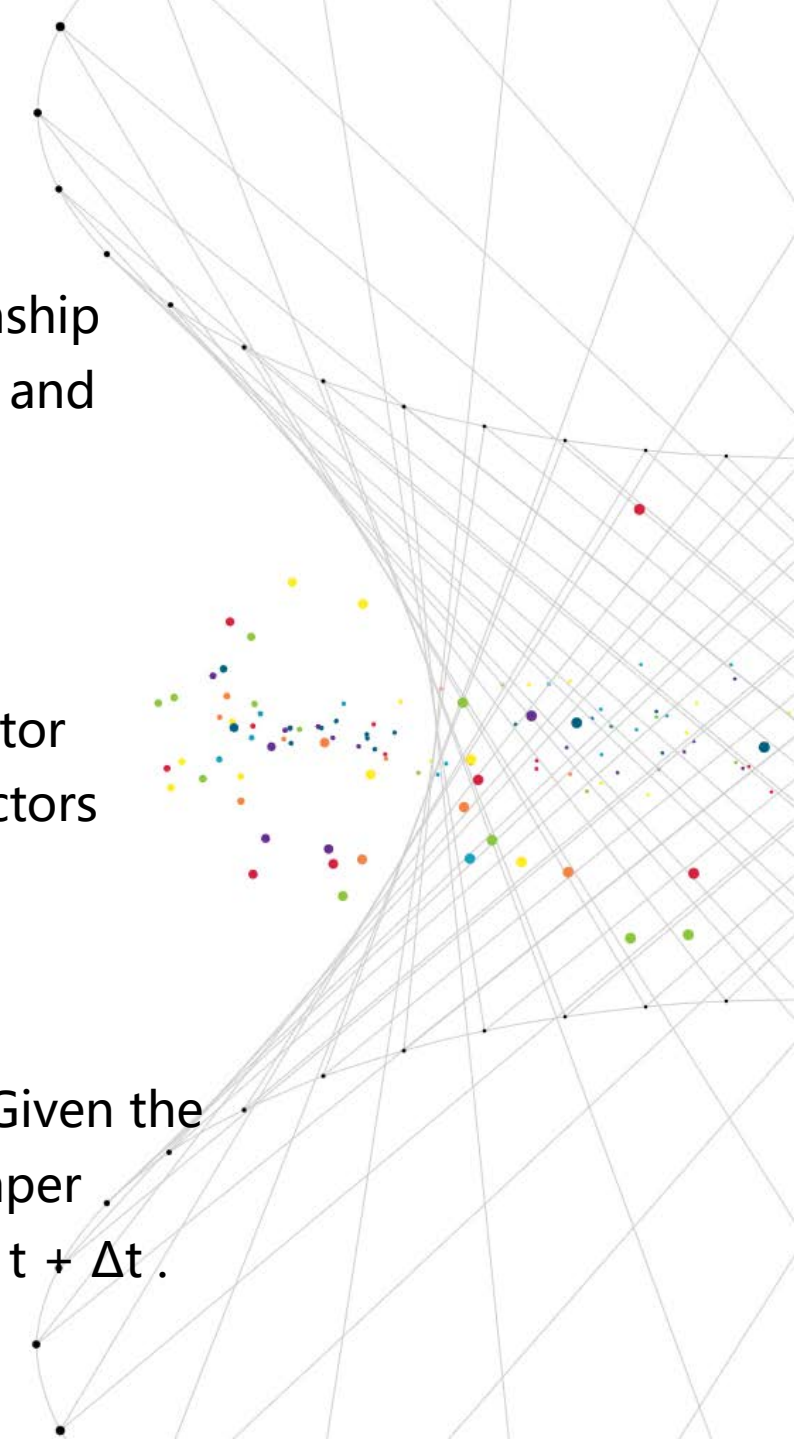
Visualization of basic structure of the CS field and show the relationship between different topics. Including topic clustering, k-core analysis, and heat representing.

### **Topic Factor Extraction**

Extract factors that can influence the future development or the factor that can show the present state of a topic. Determine how these factors influence the growth rate of the topic.

### **Topic Scale Prediction**

The goal is to regard the scale prediction as a regression problem. Given the factor matrix  $M$  of topic  $T$  at time  $t$ , the problem is to predict the paper number  $N$ , which means the size and scale of this topic, at the time  $t + \Delta t$ .





# Topic Map

## PART TWO

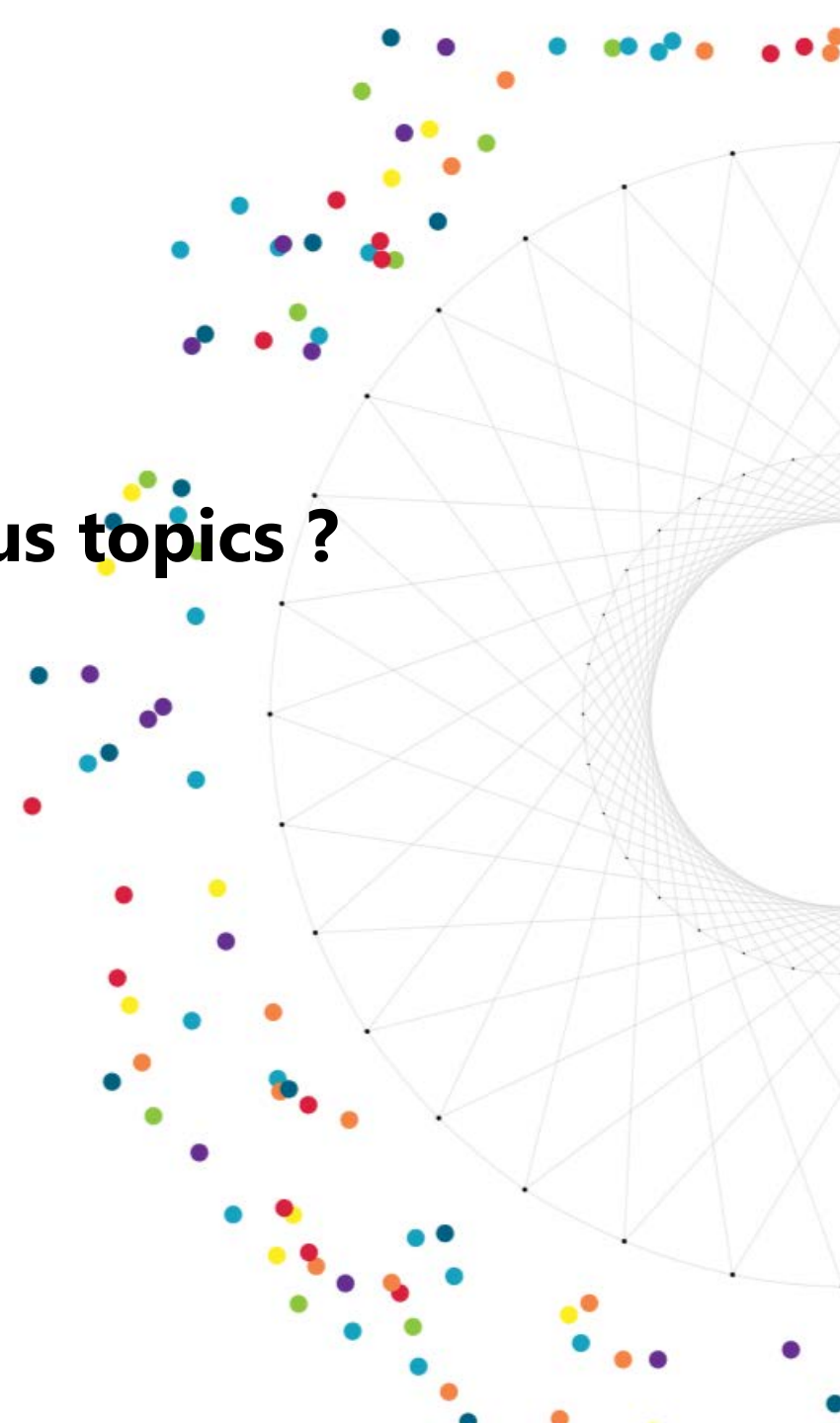


**What does the computer domain contain ?**

**What is the relationship between the various topics ?**

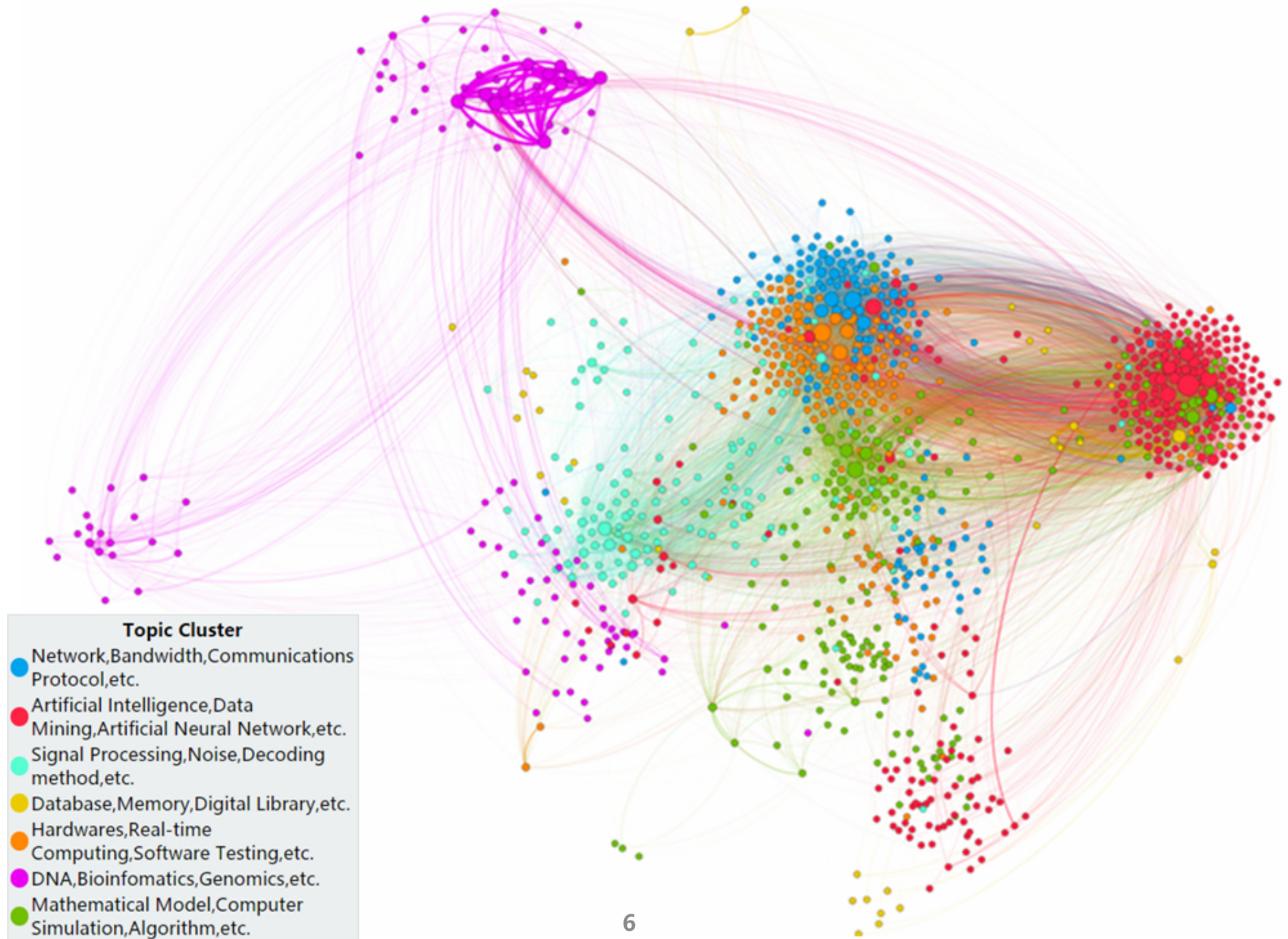
**What is the basis of the computer field ?**

**How various topics influence each other ?**

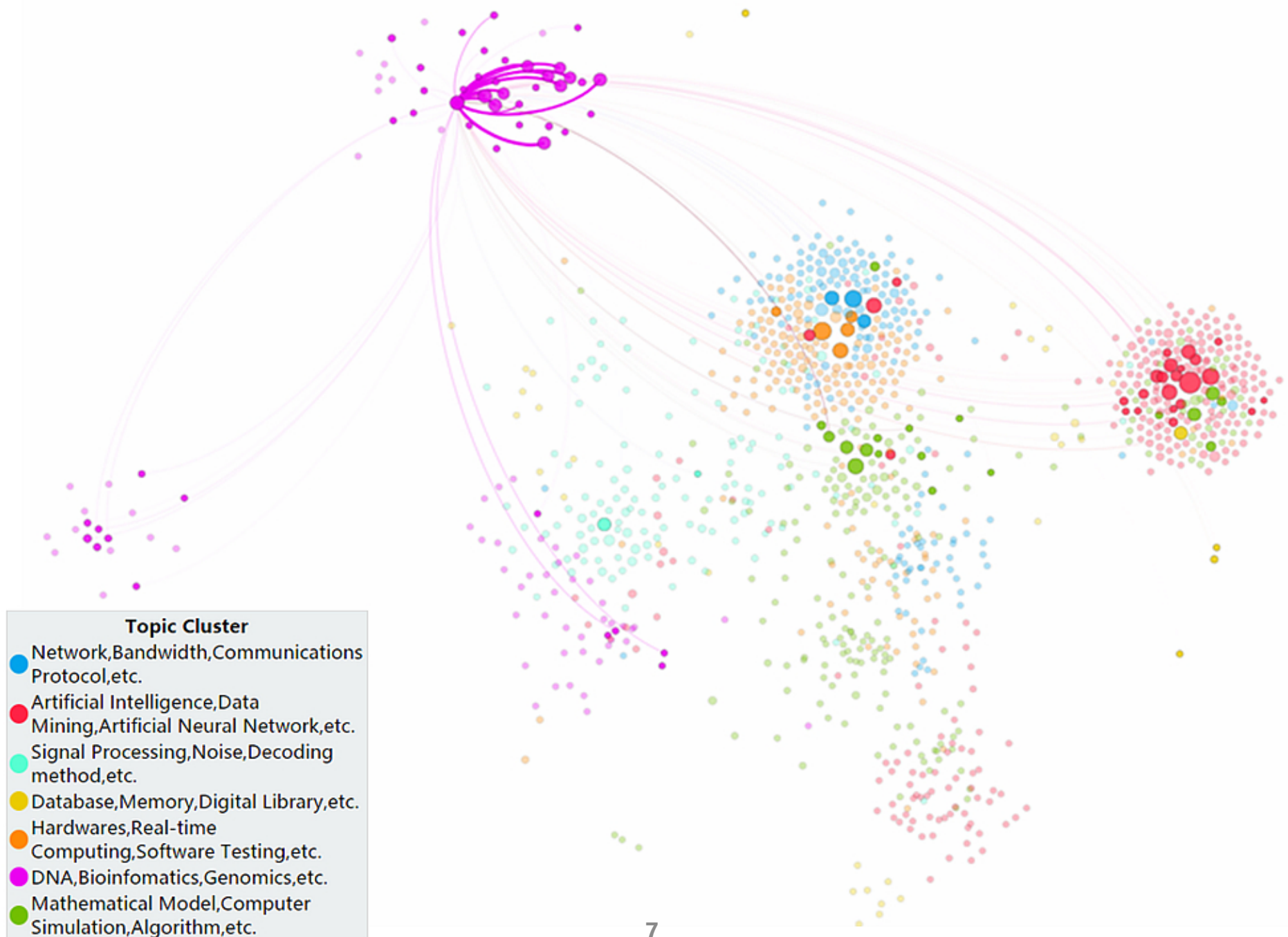




## Topic Map



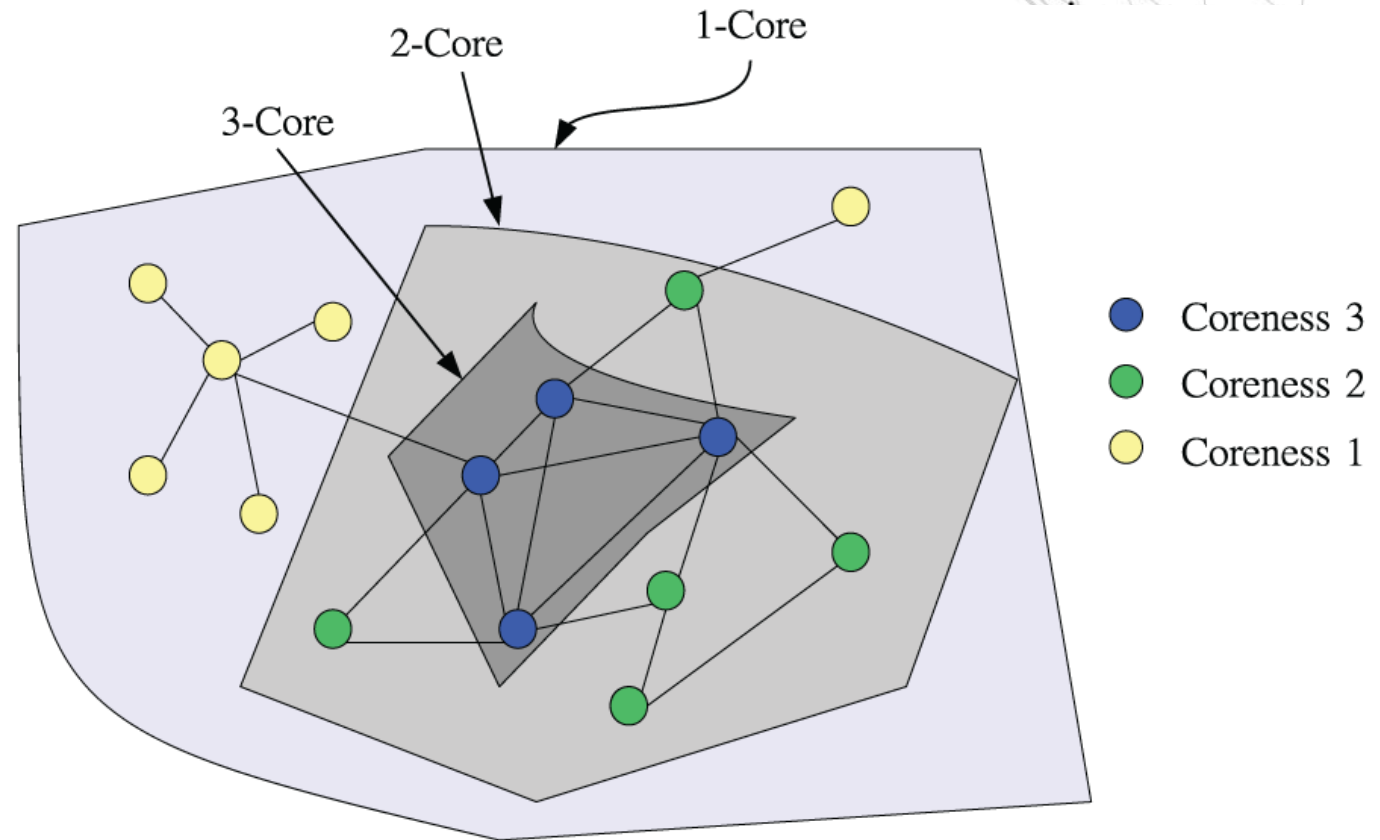
## Topic Map



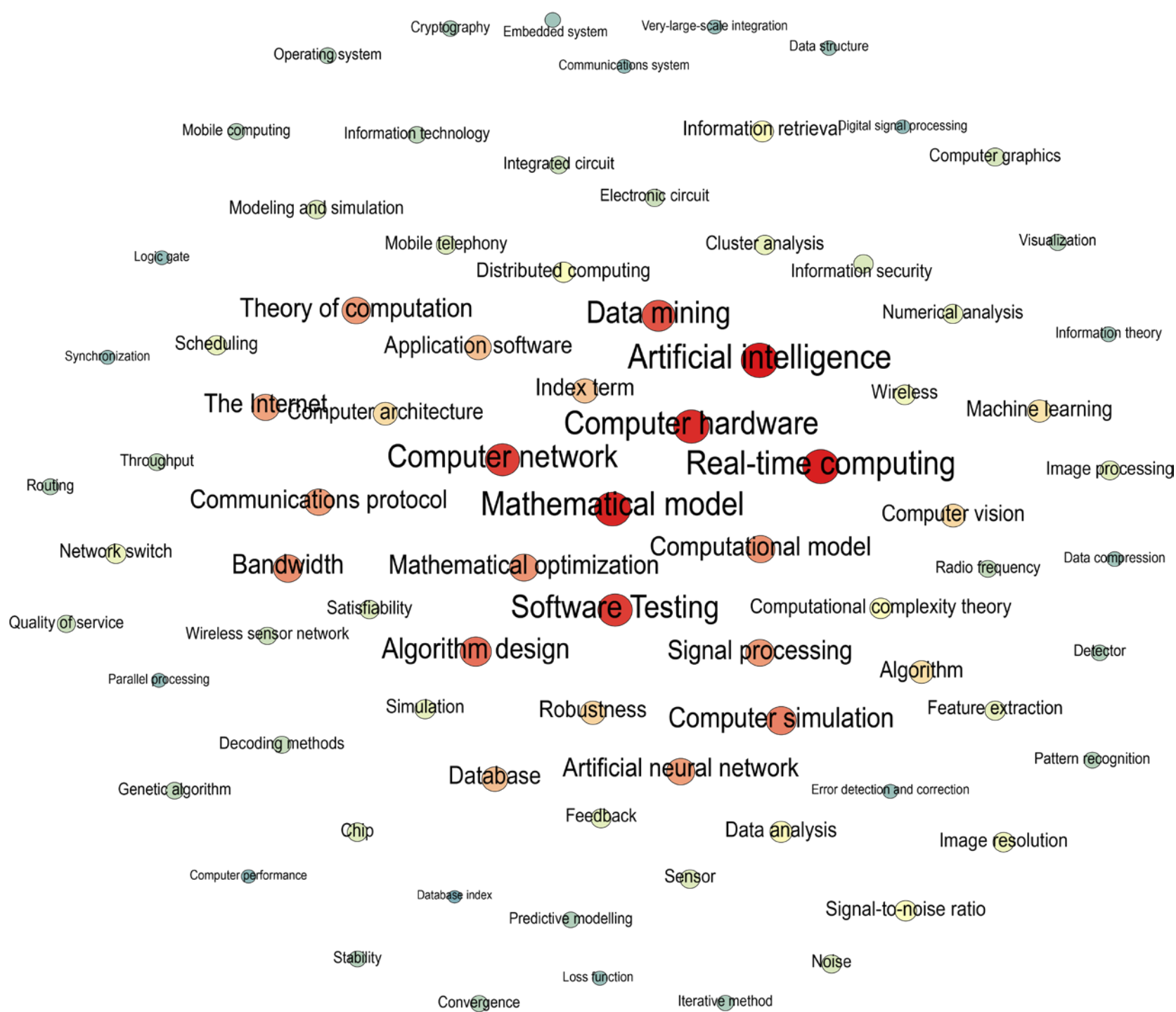


# K-core Analysis of Scholarly Network

A **k-core** is the maximal subgraph where all vertices have degree at least  $k$ .



# Topic Map





# Features Extraction

## PART THREE

---

## Features Extraction

### Paper Factor

Paper-num  
Citation-ave  
Citation-max

### Author Factor

Author-num  
Author-hindex-ave  
Author-hindex-max  
Author-hindex-var

### Growth Factor

Increase-num  
Increase-num-ave  
Increase-num-max

### Venue Factor

Venue-num  
Venue-distinct-num  
Venue-index-ave

### Interaction Factor

interaction-growthnum-ave  
interaction-growthnum-max

TABLE I  
TOPIC FACTORS AND CORRELATION COEFFICIENTS BETWEEN THIS ELEMENT AND TOPIC SCALE AFTER  $t$  YEARS

Factor	Element	Definition	$cc_1$	$cc_5$	$cc_{10}$
Paper	<i>paper-num</i>	The number of papers in this topic	0.9927	0.9861	0.9570
	<i>citation-ave</i>	The average value of papers' citations in this topic	-0.0103	-0.0007	-0.0029
	<i>citation-max</i>	The max value of papers' citations in this topic	0.3368	0.3413	0.3373
Author	<i>author-num</i>	The number of authors in this topic	0.9618	0.9532	0.9371
	<i>author-hindex-ave</i>	The average value of authors' h-index in this topic	0.0688	0.0629	0.0637
	<i>author-hindex-max</i>	The max value of authors' h-index in this topic	0.3580	0.3691	0.3811
	<i>author-hindex-var</i>	The variance of authors' h-index in this topic	0.0542	0.0486	0.0500
Growth	<i>increase-num</i>	The growth of paper number between current year and last year	0.8885	0.9432	0.9438
	<i>increase-num-ave</i>	The average value of growth number in the past five years	0.9487	0.9586	0.9558
	<i>increase-num-max</i>	The max value of growth number in the past five years	0.9381	0.9385	0.9294
Venue	<i>venue-num</i>	The total number of venues in this topic	0.7054	0.6767	0.6511
	<i>venue-distinct-num</i>	The number of distinctive venues in this topic	0.5669	0.5616	0.5550
	<i>venue-index-ave</i>	The weighted average of the <i>venueIndex</i> of venues appeared in this topic.	0.0123	0.0280	0.0528
Interaction	<i>interaction-growthnum-ave</i>	The average value of <i>increase-num</i> of neighboring topics	0.0291	0.0356	0.0290
	<i>interaction-growthnum-max</i>	The max value of <i>increase-num</i> of neighboring topics	0.0331	0.0381	0.0290





## Time Serialization

Time serialization to each factor of **12243 topics**  
From **1950 to 2015**

*Containing*

more than **14.4 million** authors  
more than **30 million** papers



# Prediction

## PART FOUR

---

# Models

**Linear regression(LR)**



**Decision Tree Regression(DT)**



**Random Forest Regression(RF)**



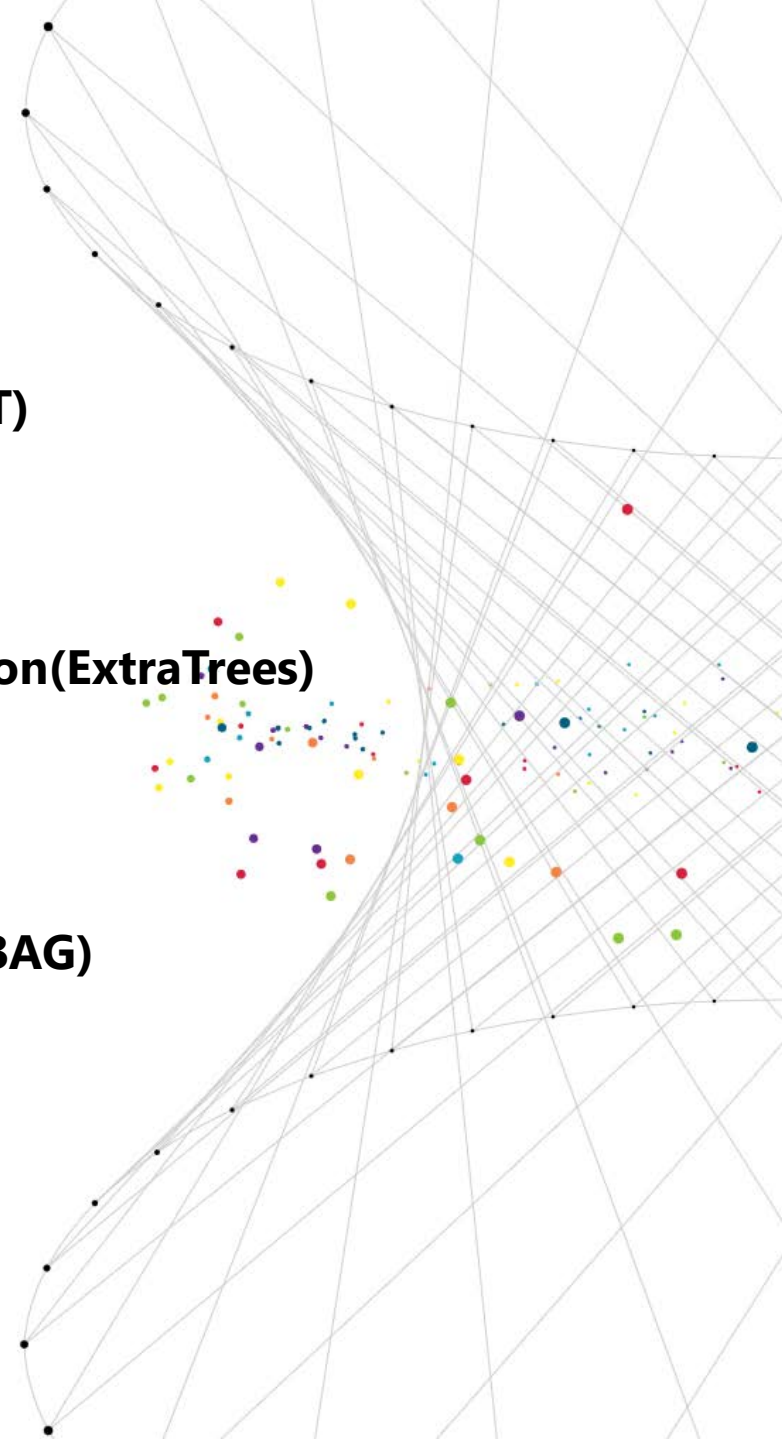
**Extremely Randomized Trees Regression(ExtraTrees)**



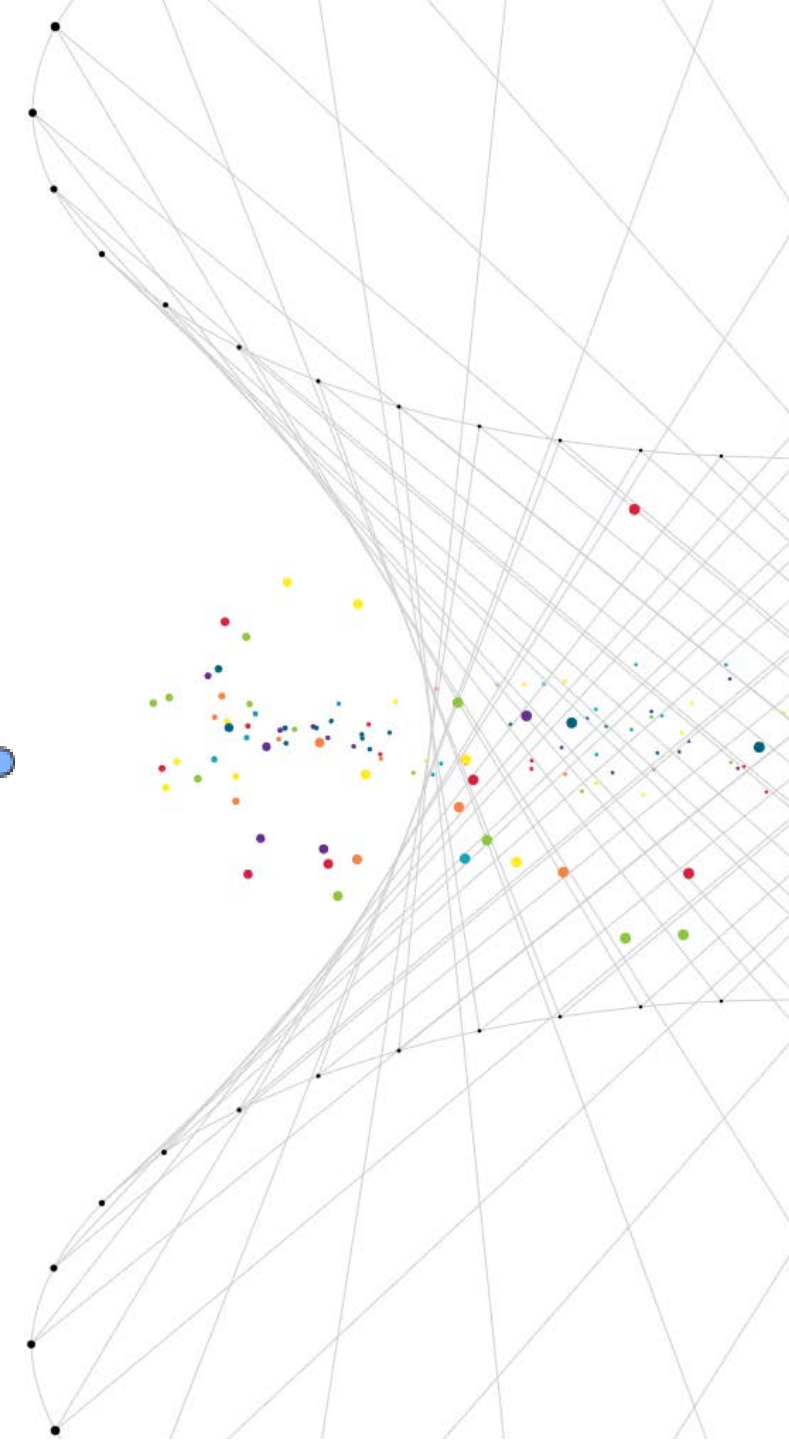
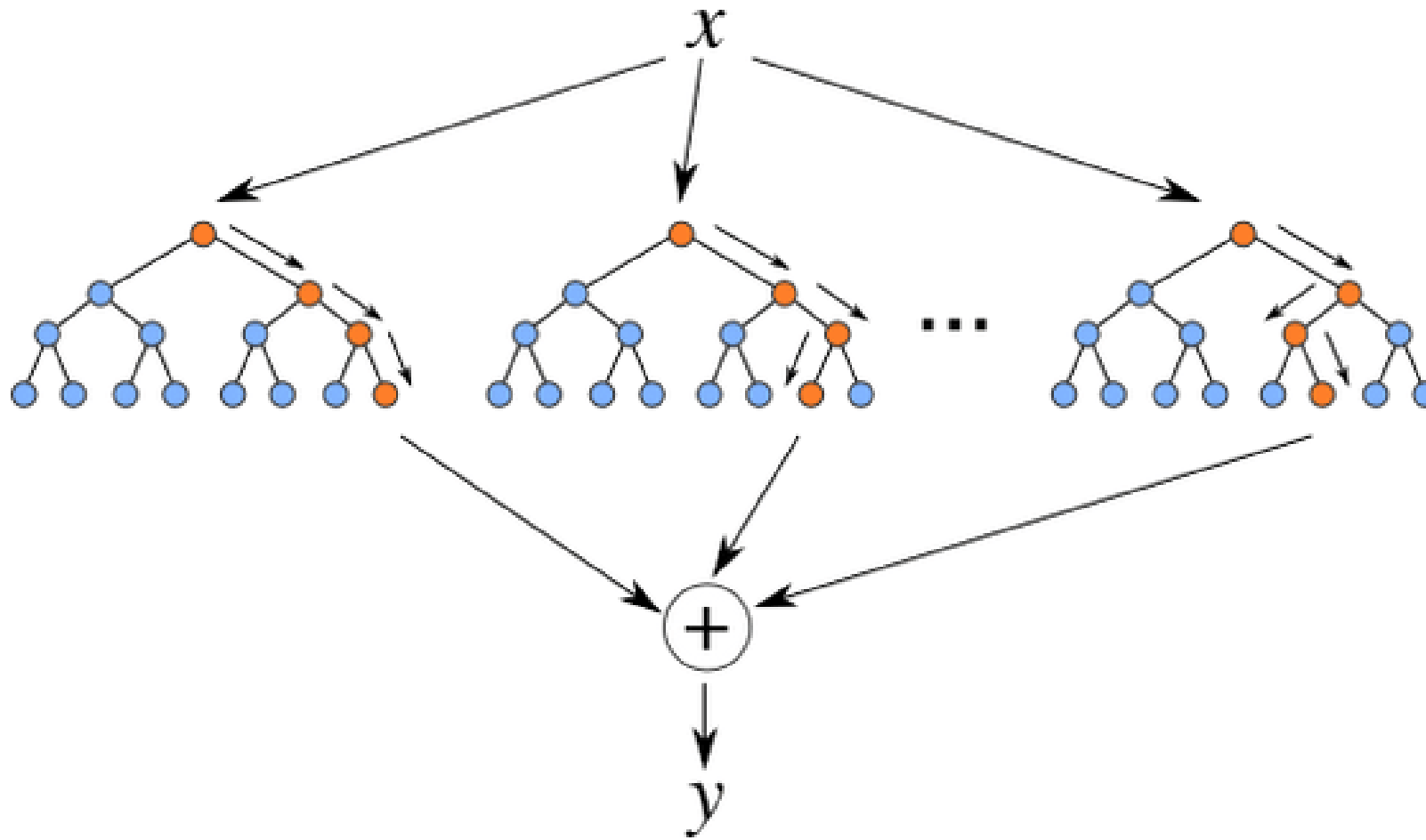
**Gradient Boosting Regression(GBDT)**



**bagged decision trees(BAG)**



# Random Forest



## Predictive Performance

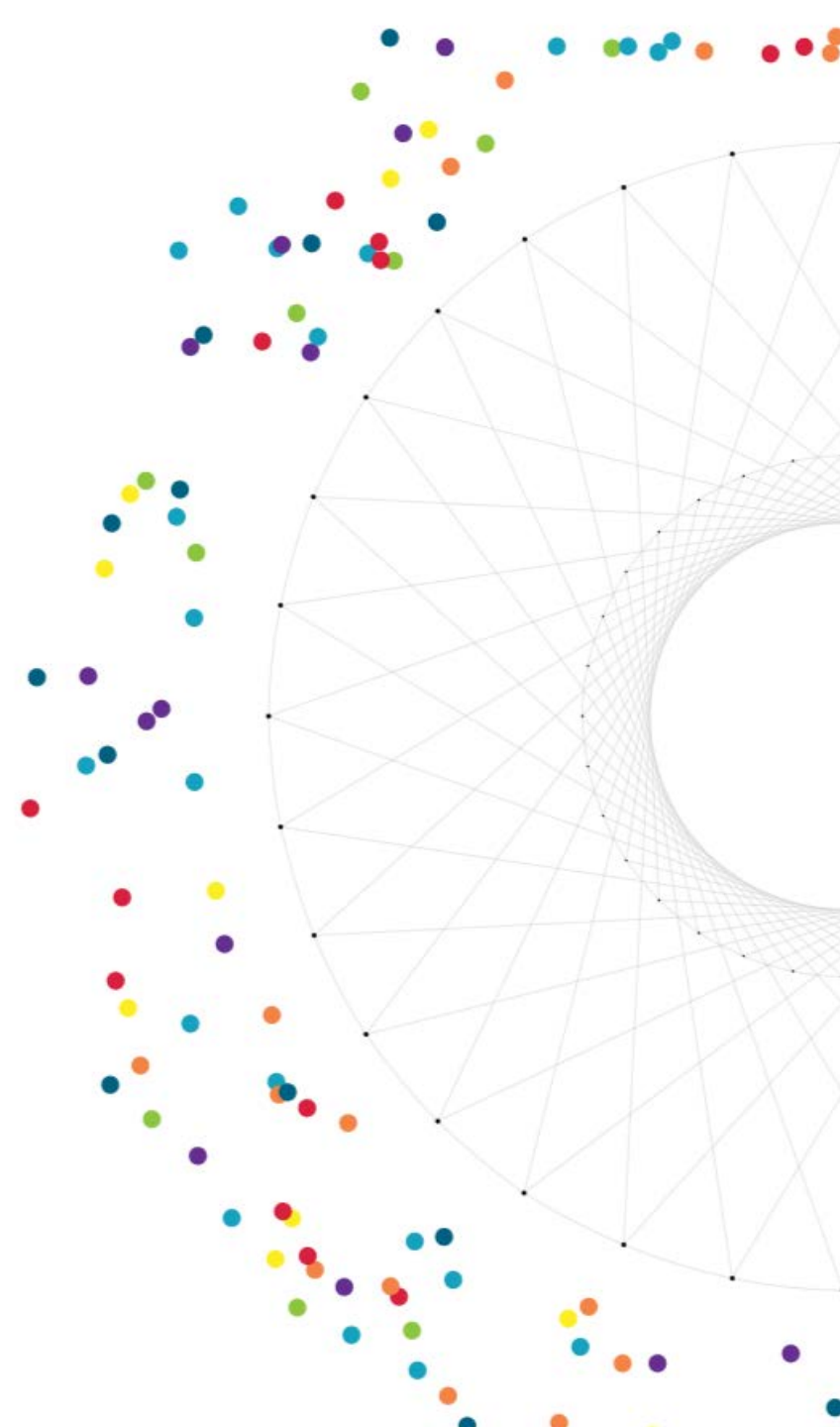
### 1. Coefficient of Determination ( $R^2$ )

$$SS_{\text{tot}} = \sum_i (y_i - \bar{y})^2, \quad SS_{\text{res}} = \sum_i (y_i - f_i)^2 = \sum_i e_i^2$$

$$R^2 \equiv 1 - \frac{SS_{\text{res}}}{SS_{\text{tot}}}$$

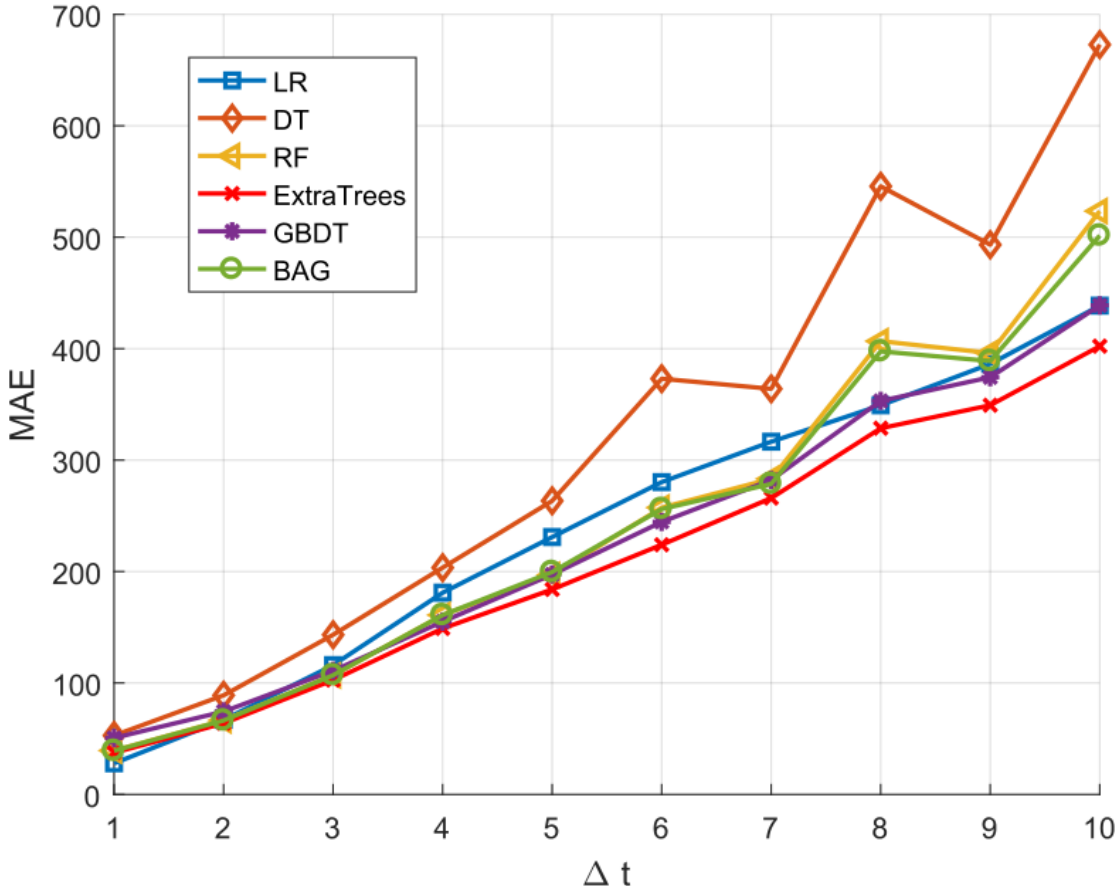
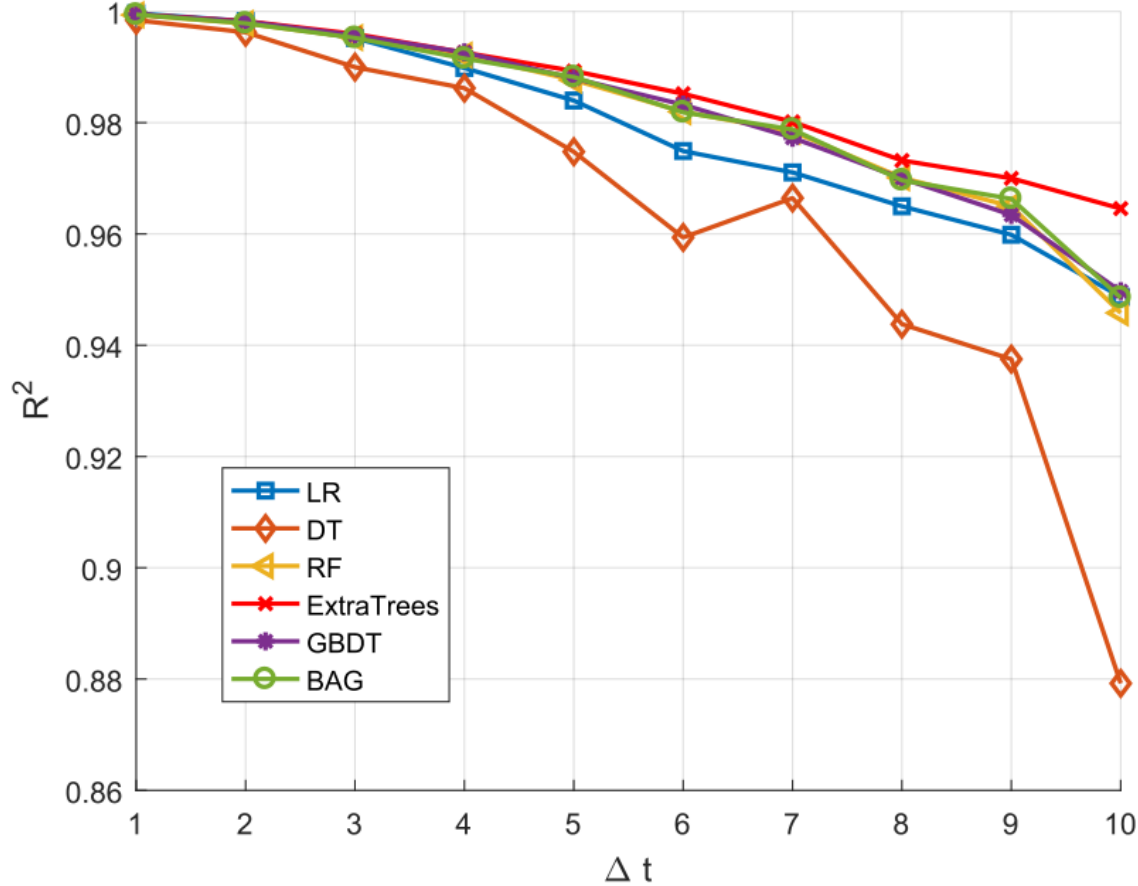
### 2. Mean Absolute Error (MAE)

$$\text{MAE} = \frac{\sum_{i=1}^n |y_i - x_i|}{n} = \frac{\sum_{i=1}^n |e_i|}{n}$$



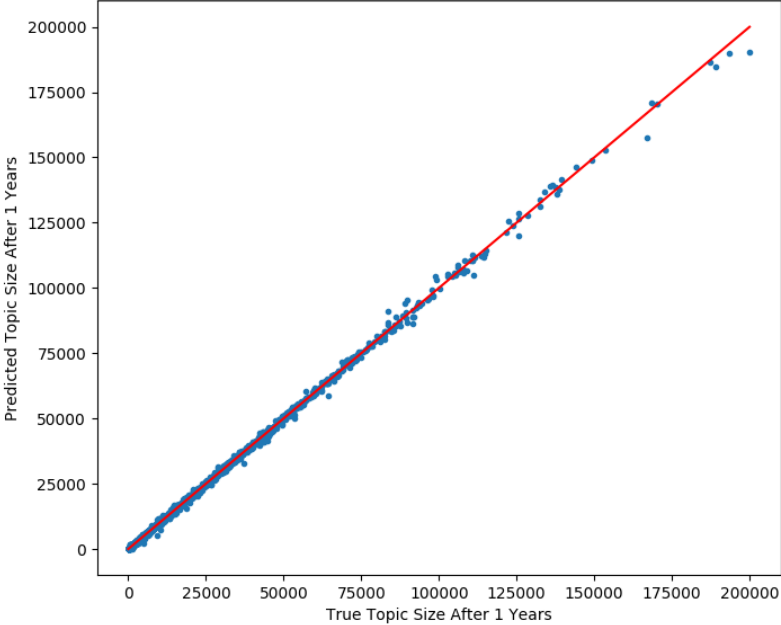


Prediction

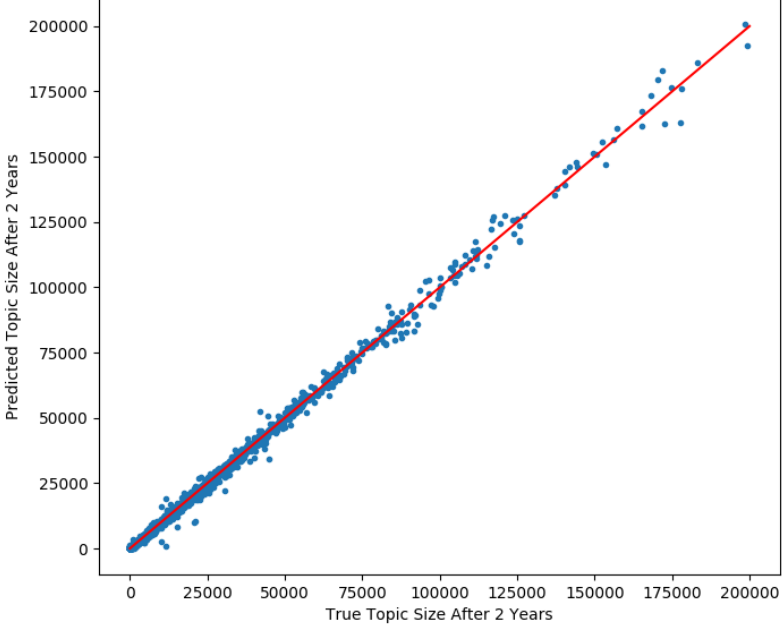


Prediction

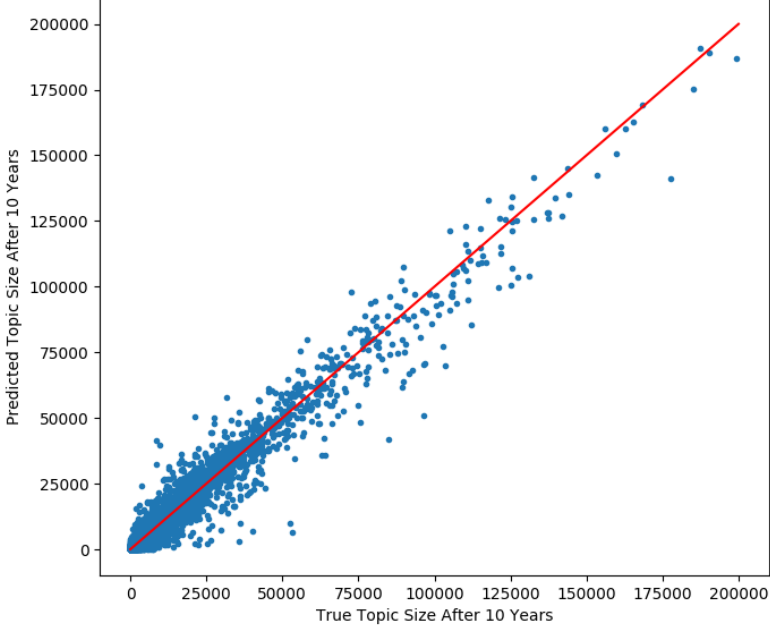
$\Delta t=1$



$\Delta t=5$




$\Delta t=10$



## Top 100 Hot Topic in 5 years

Exome sequencing Topic model MIDI Megabyte Citizen science CycL  
 Software-defined networking Analytics GLONASS Dystopia  
 5G RNA-Seq 3D printing Duplex TaqMan Rubric Python Smart grid AdaBoost  
 .bss Public records 6LoWPAN Internet of Things Pyrosequencing SPARQL  
 Crowdsourcing OpenFlow Memristor Green computing Optogenetics Radio  
 Omics Exome F1 score Small cell CRISPR Arduino Pre-clinical development  
 Machine to machine Lifelog Climateprediction.net Net income  
 Multi-user MIMO Cyber-physical system Hardware security module  
 Cloud computing security Selective laser melting Malware Interquartile range  
 Broadcast communication network Visible light communication Android  
 Latent Dirichlet allocation Information sensitivity Rescue robot Display resolution  
 Systems Modeling Language Three-dimensional integrated circuit Harmony search  
 Static single assignment form Cellulosic ethanol Visual odometry Nanopore Data set  
 Artificial bee colony algorithm Cascading failure Dashboard Geolocation  
 Tiger Quadrature mirror filter Agile software development Transmitter power output  
 Homomorphic encryption Near field communication Bioconjugation  
 Microsoft Office Vehicular communication systems Circulating tumor cell  
 Command-line interface

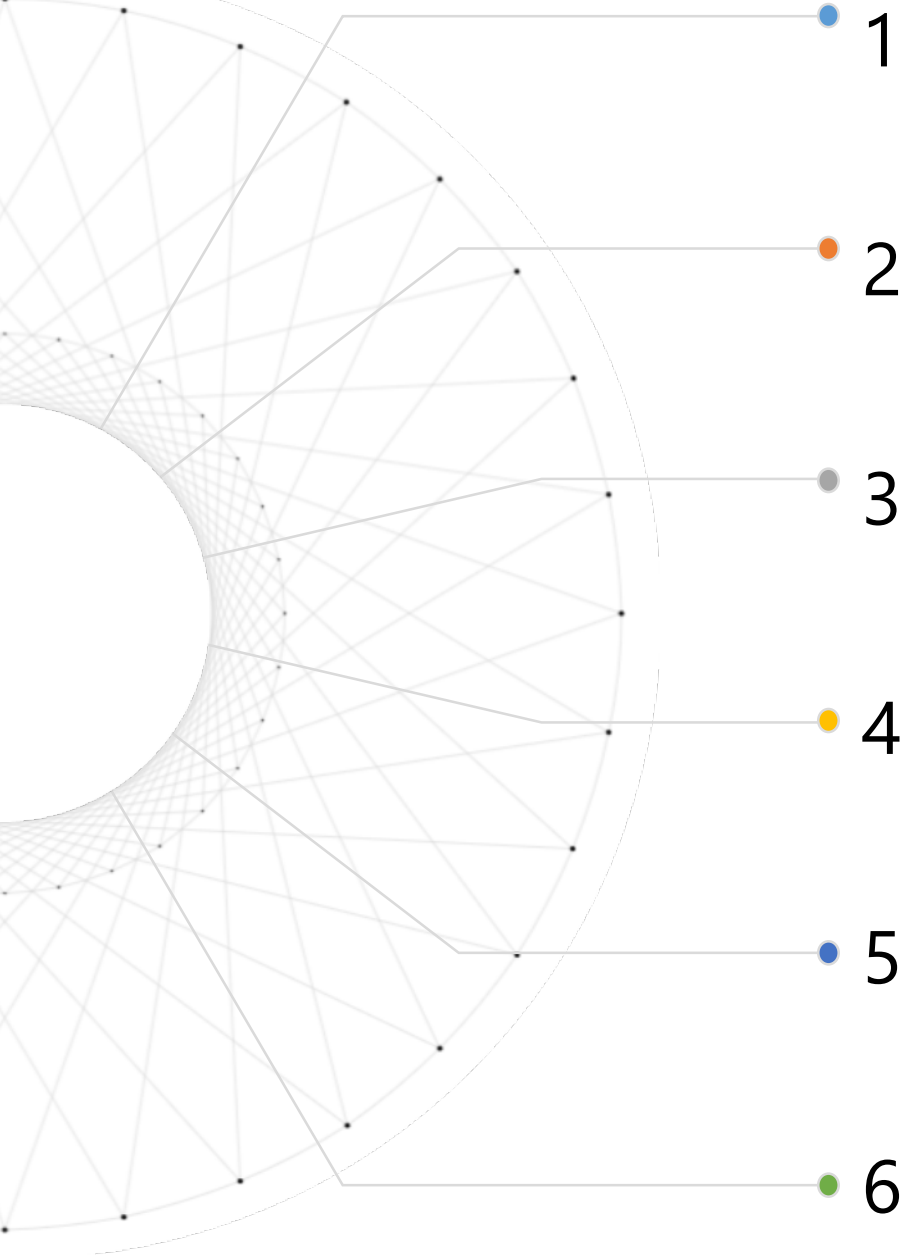


# Conclusion

## PART FIVE

---

## Conclusion



1

**Insight the topic structure of CS field.**

2

**Find the connection between hot topics.**

3

**Extract the factors that can influence topics' development.**

4

**Generate a time series dataset containing all 12243 topics**

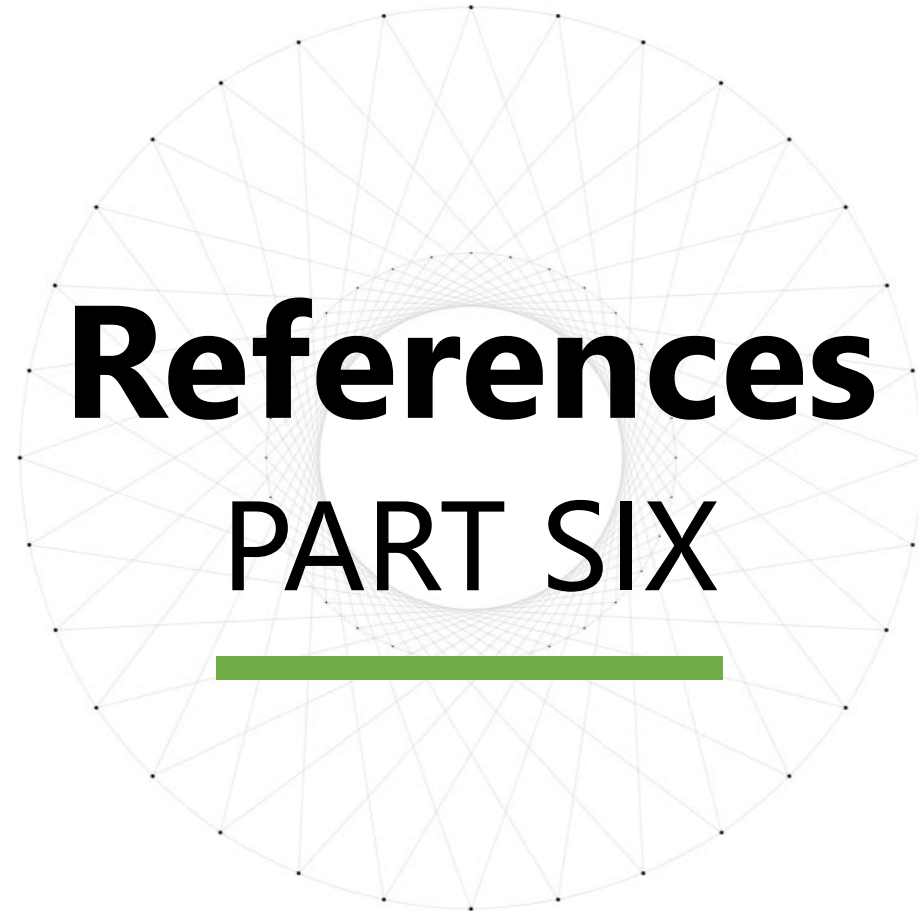
5

**Compare the performance between different models.**

6

**Predict the top 100 hot topics in 5 years.**





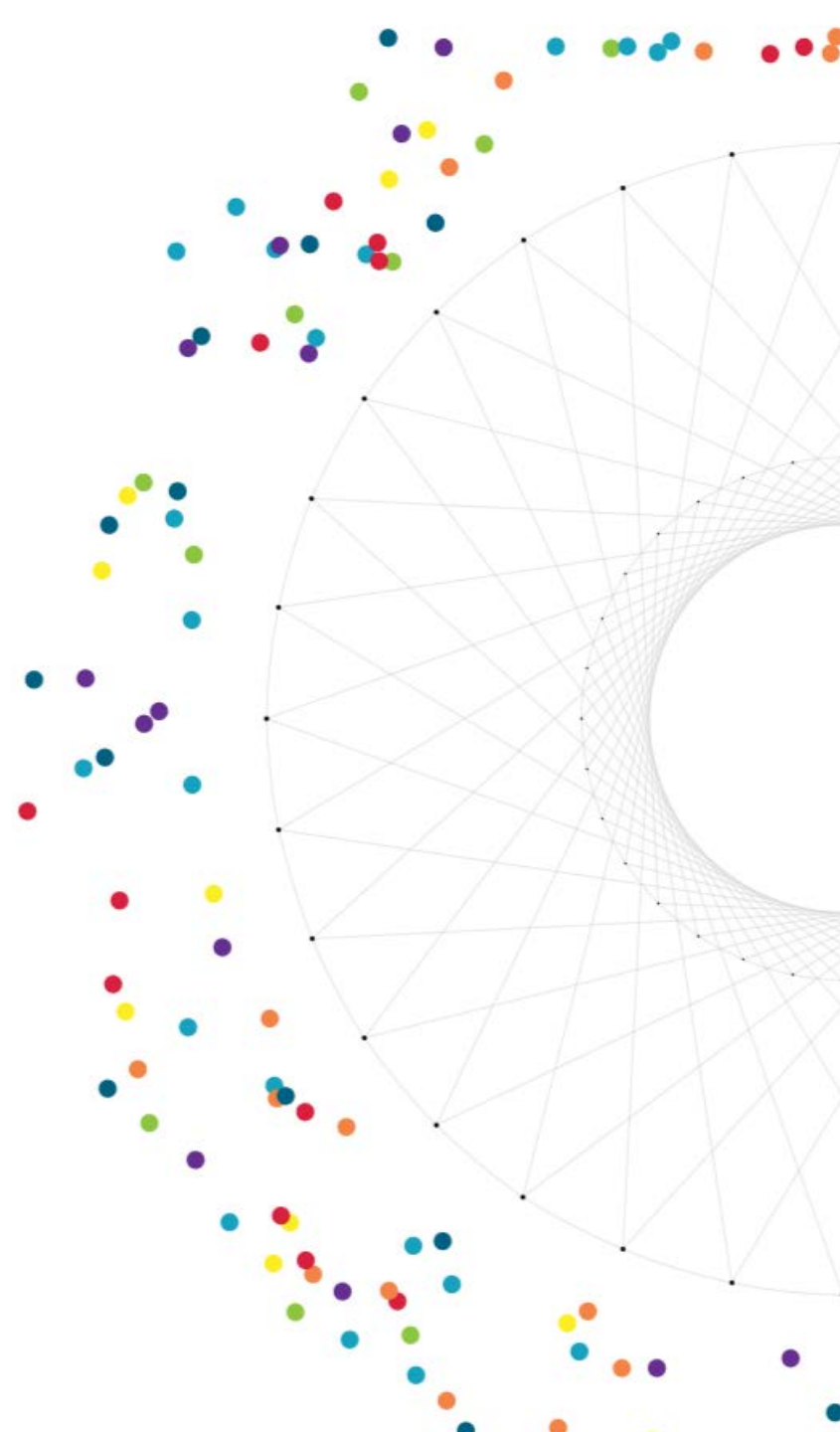
# References

## PART SIX

---

## References

- [1] (2016) Microsoft academic graph. [Online]. Available: <https://www.microsoft.com/en-us/research/project/microsoft-academic-graph/>
- [2] T. Qian, Q. Li, B. Liu, H. Xiong, J. Srivastava, and P. C. Y. Sheu, "Topic formation and development: a core-group evolving process," *World Wide Web*, vol. 17, no. 6, pp. 1343–1373, 2014.
- [3] E. Sarigol, R. Pfitzner, I. Scholtes, A. Garas, and F. Schweitzer, "Predicting scientific success based on coauthorship networks," *EPJ Data Science*, vol. 3, no. 1, p. 9, 2014.
- [4] J. Gehrke, P. Ginsparg, and J. Kleinberg, "Overview of the 2003 kdd cup," *Sigkdd Explorations*, vol. 5, no. 2, pp. 149–151, 2003.
- [5] S. Xiao, J. Yan, C. Li, B. Jin, X. Wang, X. Yang, S. M. Chu, and H. Zha, "On modeling and predicting individual paper citation count over time," pp. 2676–2682, 2016.
- [6] Y. Dong, R. A. Johnson, and N. V. Chawla, "Can scientific impact be predicted," *IEEE Transactions on Big Data*, vol. 2, no. 1, pp. 18–30, 2016.
- [7] A. Saha and V. Sindhwani, "Learning evolving and emerging topics in social media: a dynamic nmf approach with temporal regularization," pp. 693–702, 2012.
- [8] D. M. Blei, A. Y. Ng, and M. I. Jordan, "Latent dirichlet allocation," *Journal of Machine Learning Research*, vol. 3, pp. 993–1022, 2003.





**Q&A**

