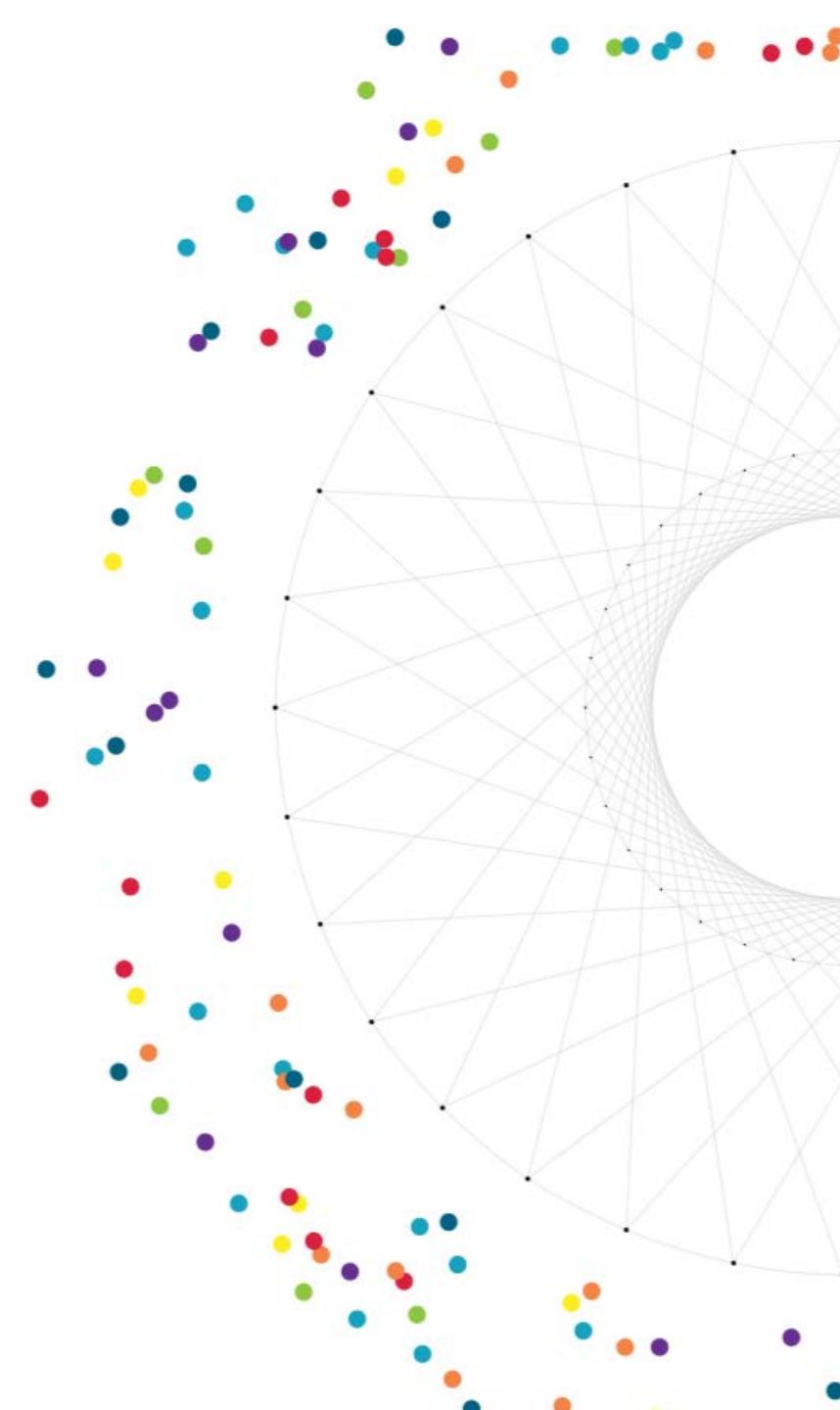


Creating Knowledge Map For Topic Search

Te Lin
5140219432

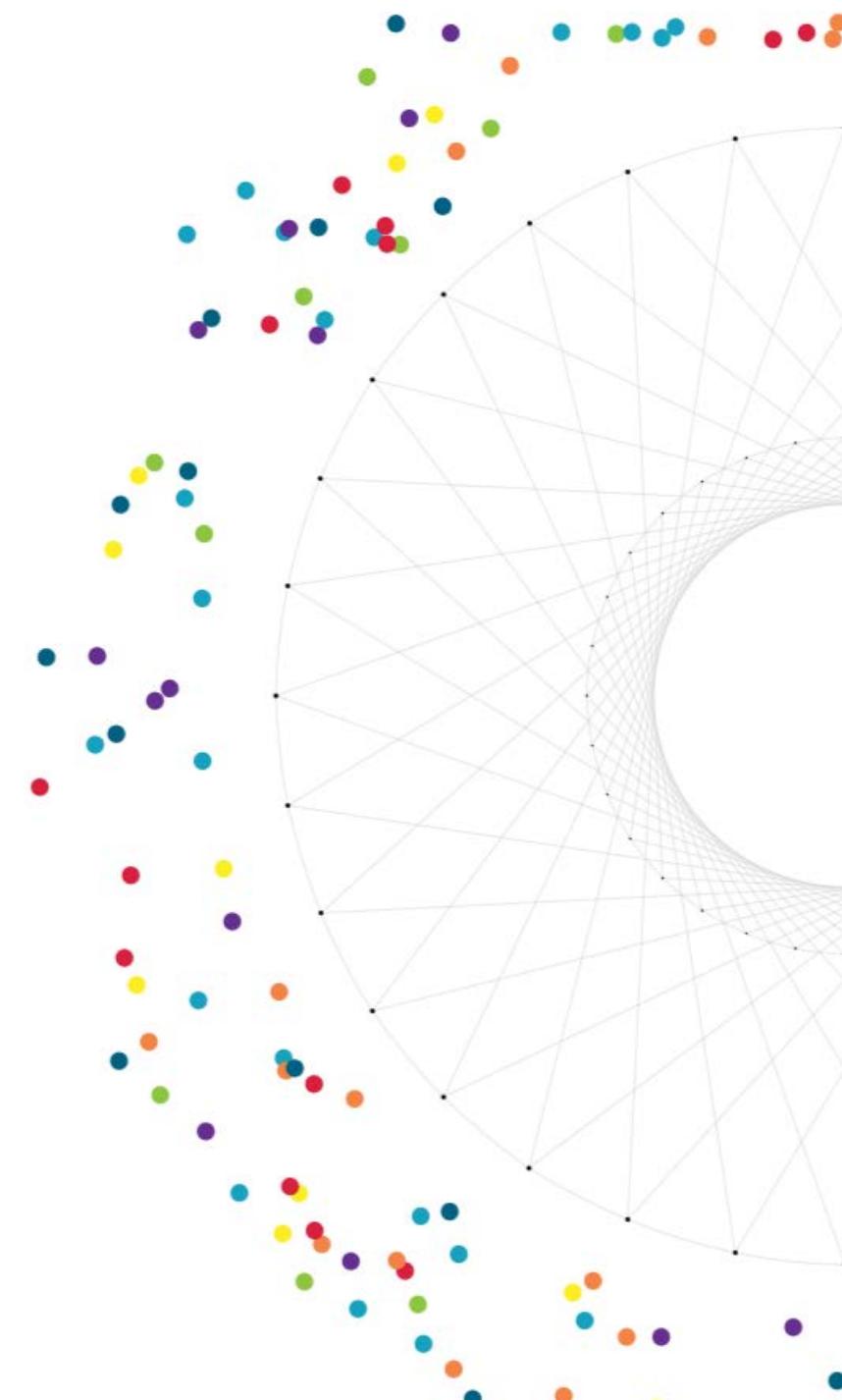
Outlines

- Motivation for Knowledge Map
- Construction
- Manipulating the Scale
- Visualization
- Future Work



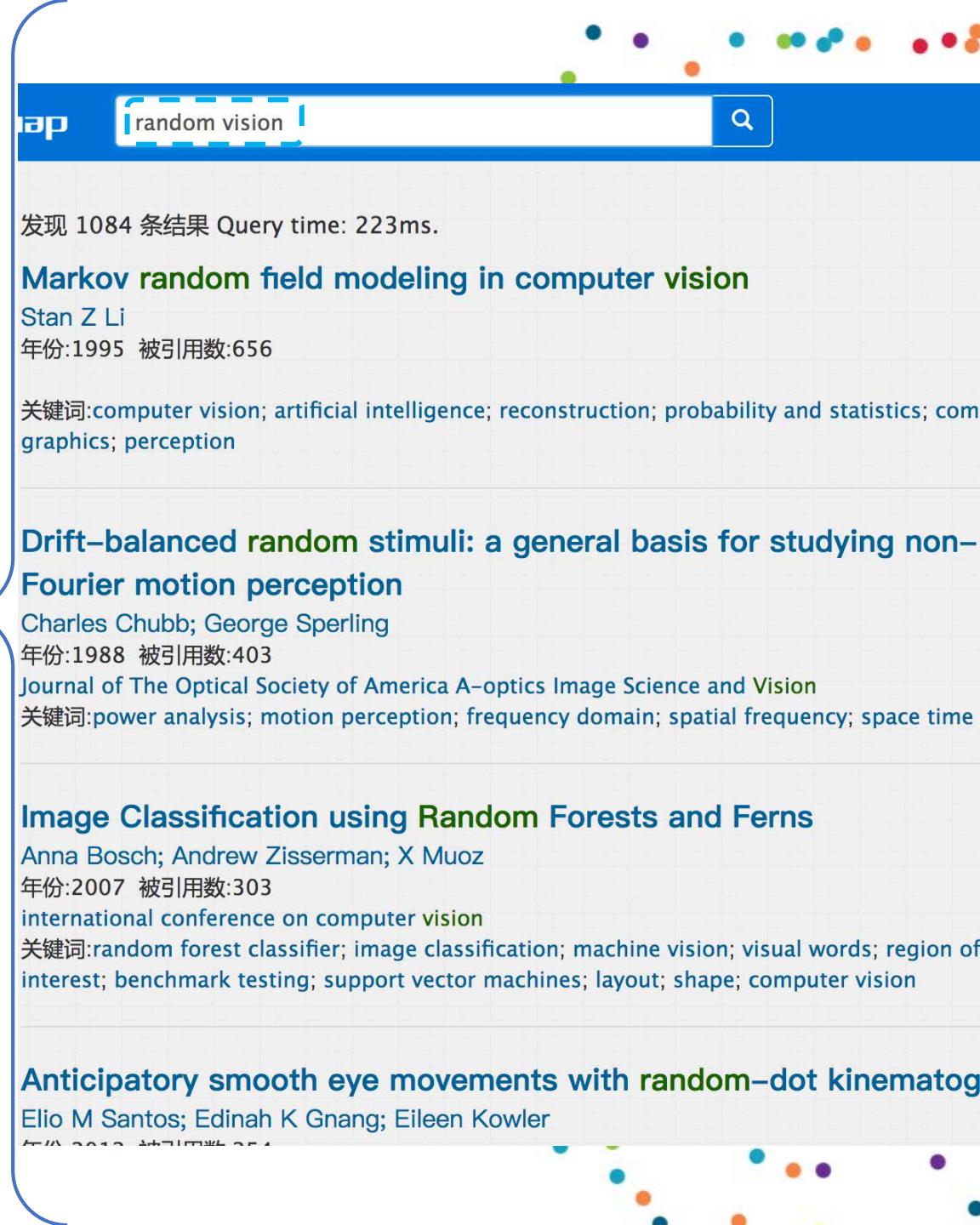
Guide

- Motivation for Knowledge Map
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1. When users are searching the topic “random vision”

- No topic showing
- No information about topic related
- None clear idea of relation and hierarchy



发现 1084 条结果 Query time: 223ms.

Markov random field modeling in computer vision
Stan Z Li
年份:1995 被引用数:656
关键词:computer vision; artificial intelligence; reconstruction; probability and statistics; computer graphics; perception

Drift-balanced random stimuli: a general basis for studying non-Fourier motion perception
Charles Chubb; George Sperling
年份:1988 被引用数:403
Journal of The Optical Society of America A-optics Image Science and Vision
关键词:power analysis; motion perception; frequency domain; spatial frequency; space time

Image Classification using Random Forests and Ferns
Anna Bosch; Andrew Zisserman; X Muoz
年份:2007 被引用数:303
international conference on computer vision
关键词:random forest classifier; image classification; machine vision; visual words; region of interest; benchmark testing; support vector machines; layout; shape; computer vision

Anticipatory smooth eye movements with random-dot kinematog
Elio M Santos; Edinah K Gnang; Eileen Kowler
年份:2010 被引用数:254

2. When users are searching the topic “machine learning”

- Redundancy
- Vague hierarchy & relation
- None informative



父话题:

Computer Science

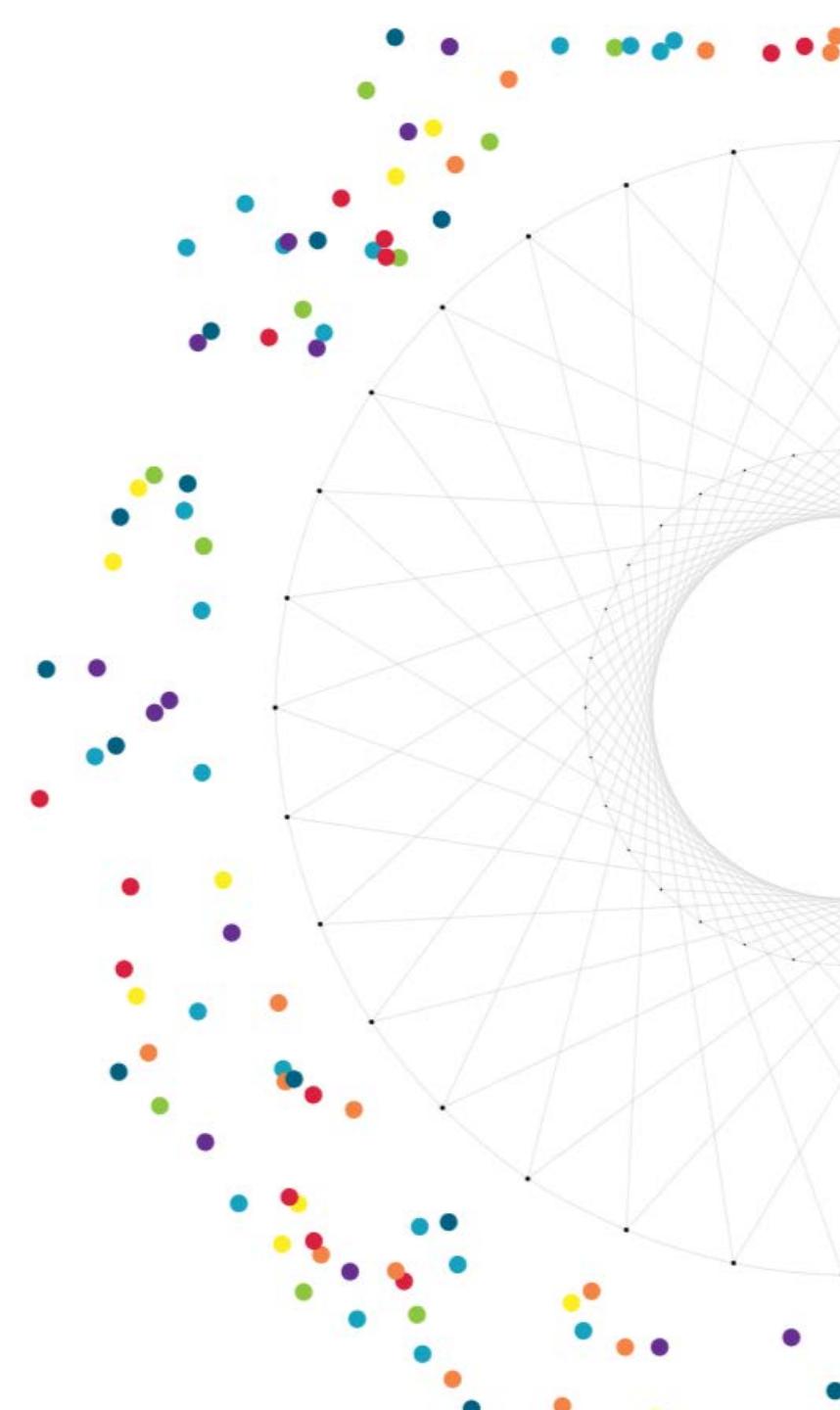
子话题:

Instance-based learning, Data stream mining, Feature selection, Heuristic, Gene expression programming, Swarm intelligence, Connectionism, Fuzzy control system, Recommender system, Unsupervised learning, Linear discriminant analysis, Intelligent word recognition, Feature learning, Probabilistic classification, Conceptual clustering, Convolutional neural network, Evolutionary computation, Intelligent character recognition, Semi-supervised learning, Computer-automated design, Algorithmic learning theory, Biological neural network, Computational intelligence, Computational learning theory, Anomaly detection, Learning to rank, Hidden Markov model, Artificial neural network, Computational neuroscience, Reinforcement learning, Deep learning, Soft computing, Automated



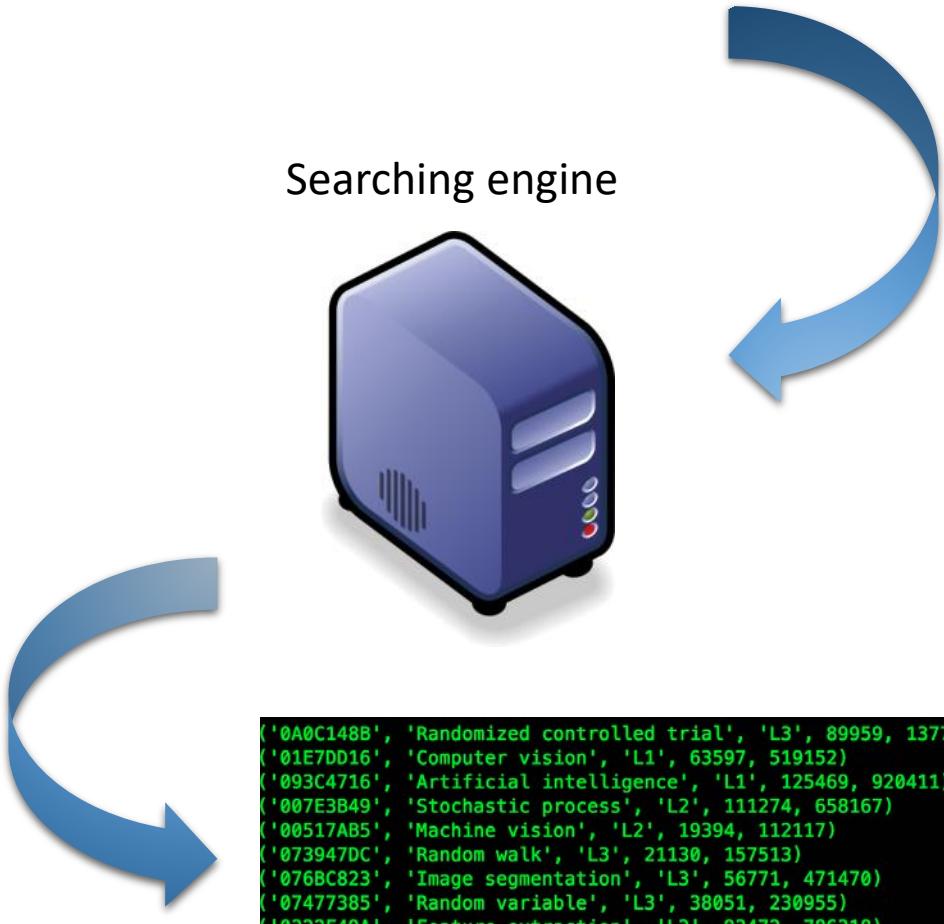
Guide

- Motivation for Knowledge Map
- **Construction**
- Manipulating the Scale
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- Future Work

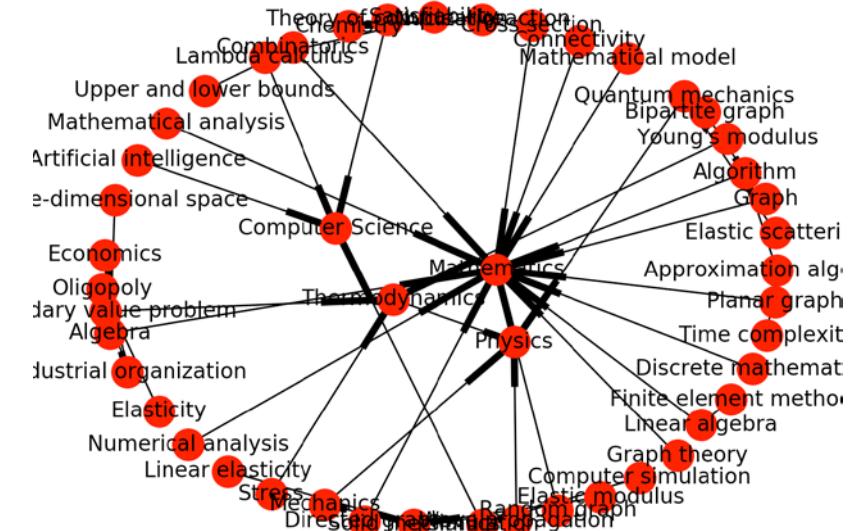


1.Extraction & Ranking

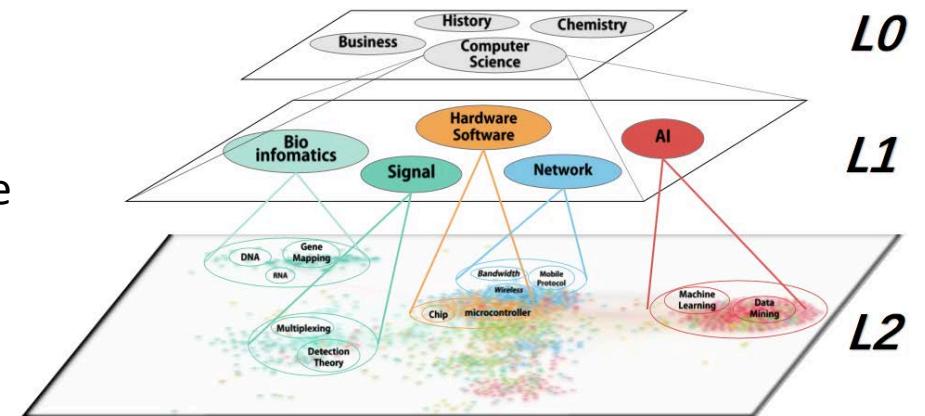
“Random vision”



- ```
'0A0C148B', 'Randomized controlled trial', 'L3', 89959, 1377309
('01E7DD16', 'Computer vision', 'L1', 63597, 519152)
('093C4716', 'Artificial intelligence', 'L1', 125469, 920411)
('007E3B49', 'Stochastic process', 'L2', 111274, 658167)
('00517AB5', 'Machine vision', 'L2', 19394, 112117)
('073947DC', 'Random walk', 'L3', 21130, 157513)
('076BC823', 'Image segmentation', 'L3', 56771, 471470)
('07477385', 'Random variable', 'L3', 38051, 230955)
('0322F49A', 'Feature extraction', 'L3', 93472, 786318)
('008FB4C8', 'Image processing', 'L2', 102494, 566598)
('07FC5C84', 'Vision', 'L3', 30238, 247463)
```

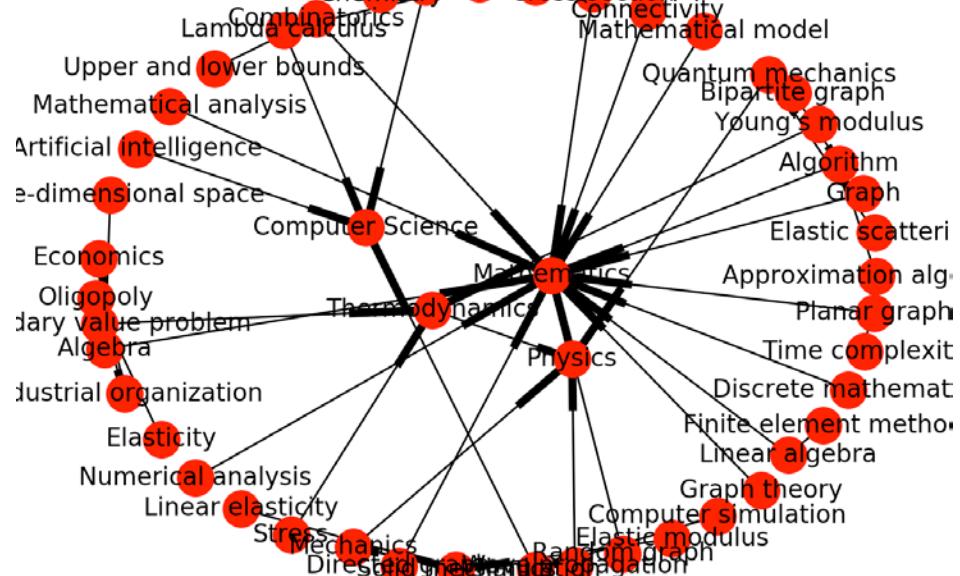


## Hierarchy

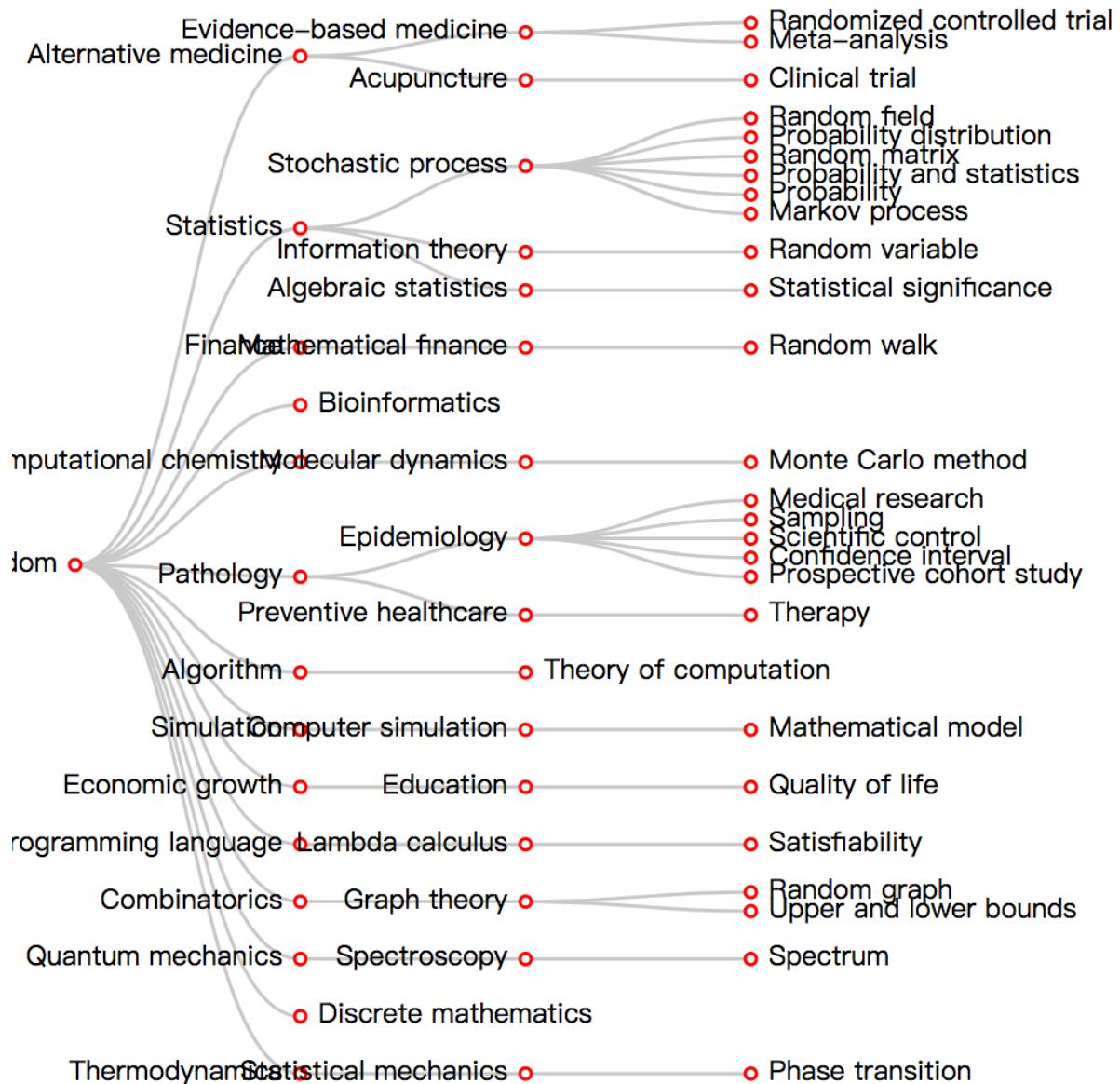


## 2. Construction

# Topic Graph



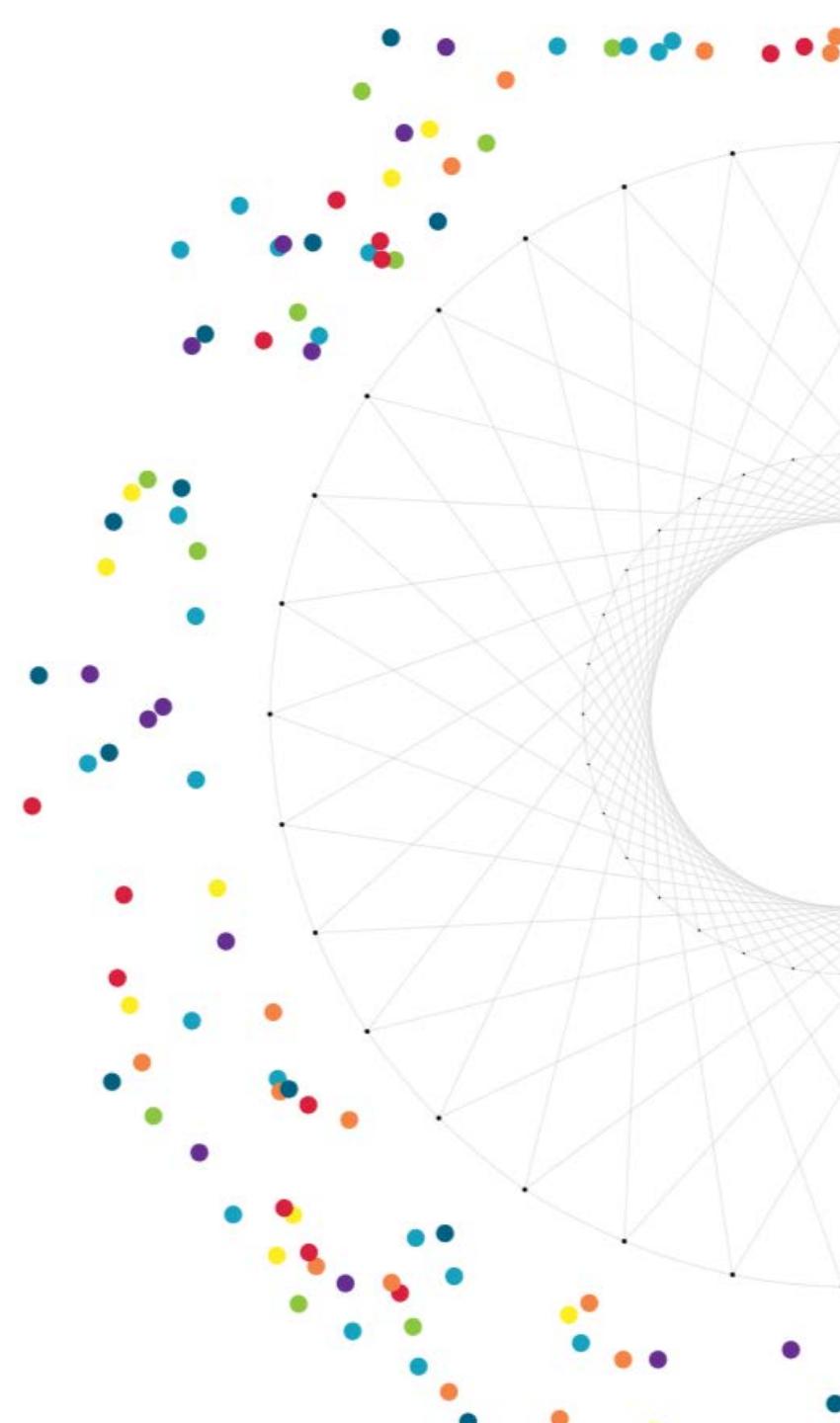
## Transform



# Topic Tree

# Guide

- Motivation for Knowledge Map
- Construction
- **Manipulating the Scale**
- Visualization
- Future Work

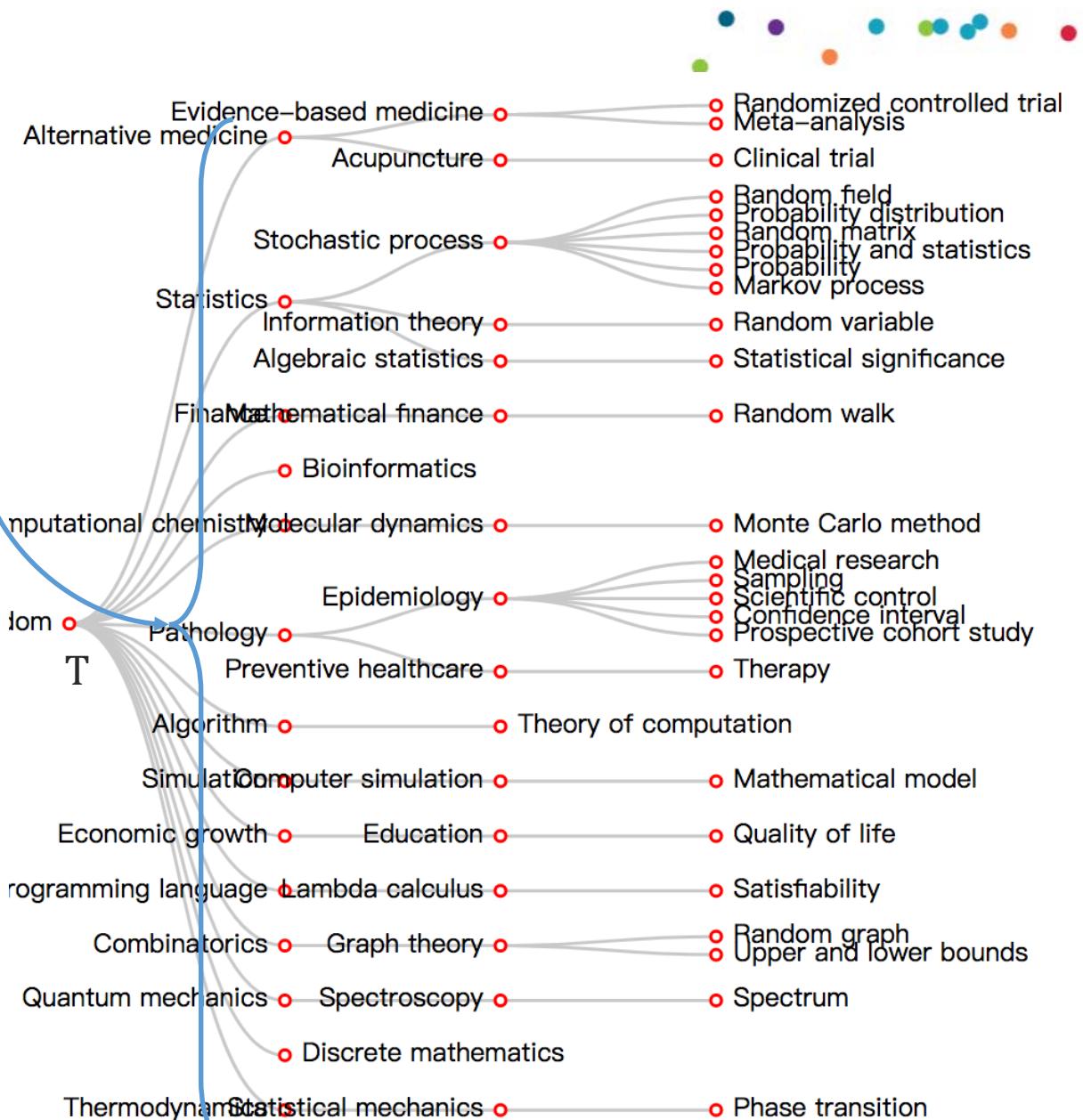


# 1. Filtering the topic preliminarily

For a knowledge tree  $T$ , if we delete the root node, we can obtain tree sequences  $T_1, T_2, \dots, T_M$

- For root node in each subtree, the  $T_i$  will be delete the hierarchy of the root i is ' $L_1'$ , ' $L_2'$ '
- Whose scale is too small and its hierarchy can't represent the knowledge.

New tree sequences  $T_1, T_2, \dots, T_N$



## 2.Computing the scale

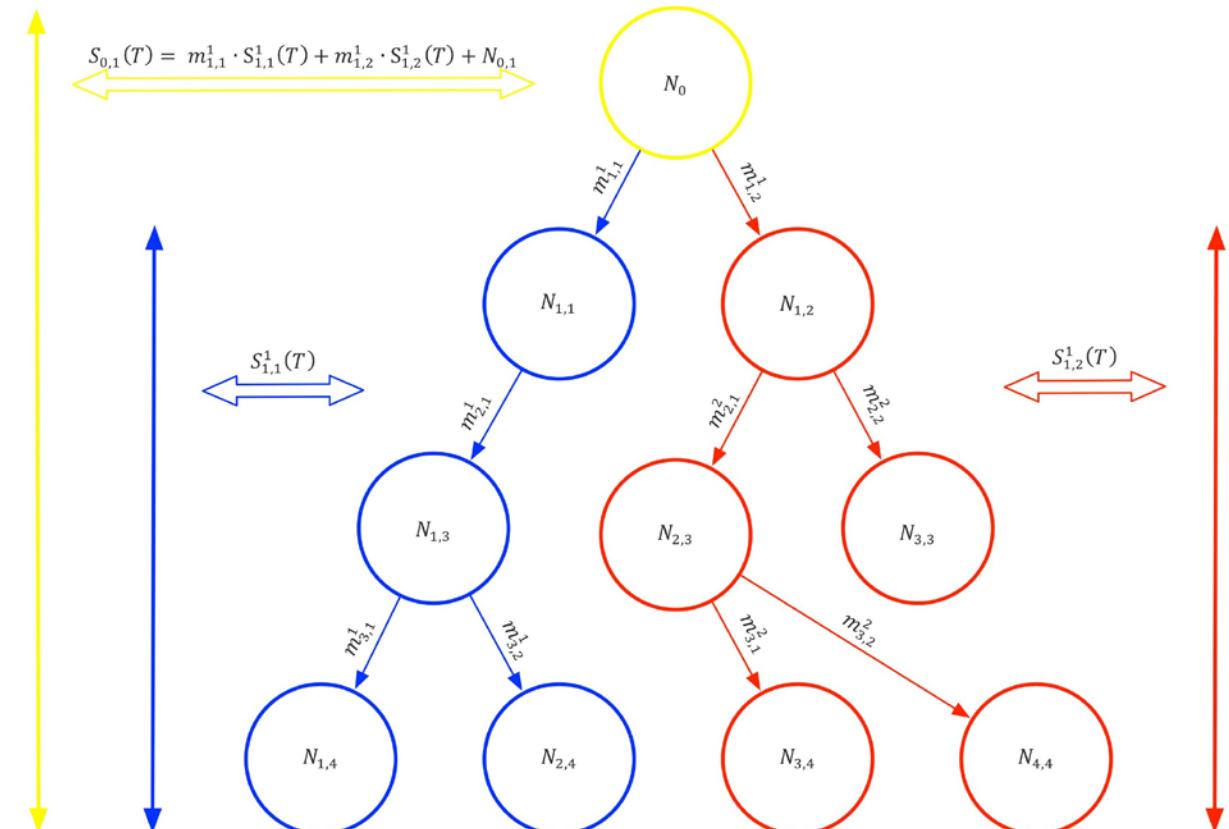
Recursive property  
define  $S(T)$ ,  $M(T)$ ,  $N$

$$S_{j-1,M(j-1)}^Q(T) = \sum_{i=1}^n m_{j,i}^{M'(j)} \cdot S_{j,i}^{M'(j)}(T) + N_{j-1,M(j-1)}$$

$$N_{j,M(j)} = \left( \alpha_1 L_{j,M(j)} + \alpha_2 \frac{1}{D_{j,M(j)}} \right) \cdot \frac{F_j}{\sum_{i=1}^U F_i}$$

$$F_j = \frac{f_j - f_{min}}{f_{max} - f_{min}} \quad m_{j,i}^{M'(j)} = \frac{\delta}{\delta_{i,j}}$$

Where  $S_{j-1,M(j-1)}^Q(T)$  is the sum scale of node M in level  $j - 1$ , node Q is the predecessor of node M,  $N_{j-1,M(j-1)}$  is the scale of Node M in level  $j - 1$ .



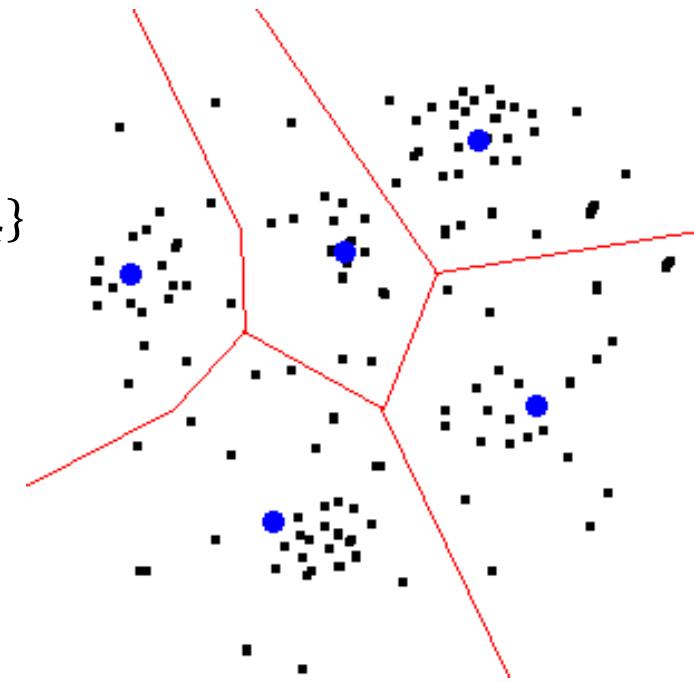
### 3.Clustering subtrees

*K-means Algorithm:*

1. Initialize assignment variables  $\{r_{ik}\}$  and cluster centers  $\{\mathbf{m}_1, \mathbf{m}_2, \dots, \mathbf{m}_c\}$
2. Repeat
  - **E-step**: with fixed  $\{\mathbf{m}_1, \mathbf{m}_2, \dots, \mathbf{m}_c\}$ , assign each sample to its nearest cluster center, i.e.,  $r_{ik} = 1, if k = \operatorname{argmin}_l ||x_i - \mathbf{m}_l||$
  - **M-step**: with fixed  $\{r_{ik}\}$ , recalculate each cluster center, i.e.,

$$\mathbf{m}_k = \frac{\sum_i x_i r_{ik}}{\sum_i r_{ik}}$$

3. Until Convergence



$$J_e \triangleq \sum_{i=1}^c \sum_{\mathbf{x} \in H_i} \|\mathbf{x} - \mathbf{m}_i\|^2$$

$[S_1^1, S_2^1, \dots S_N^1]_1, [S_1^2, S_2^2, \dots S_N^2]_2, \dots [S_1^k, S_2^k, \dots S_N^k]_k \rightarrow$  Largest sequence  $S'_1, S'_2, \dots S'_w$



New tree sequences  $T_1, T_2, \dots T_w$

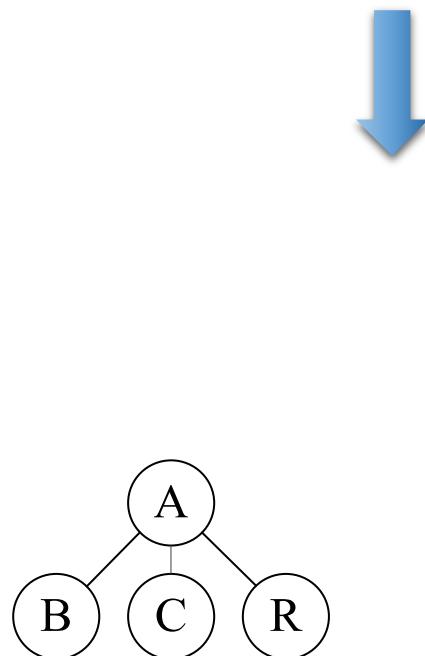
## 3. Recursion

- Filtering the topic preliminarily
  - Computing the scale
  - Clustering subtrees

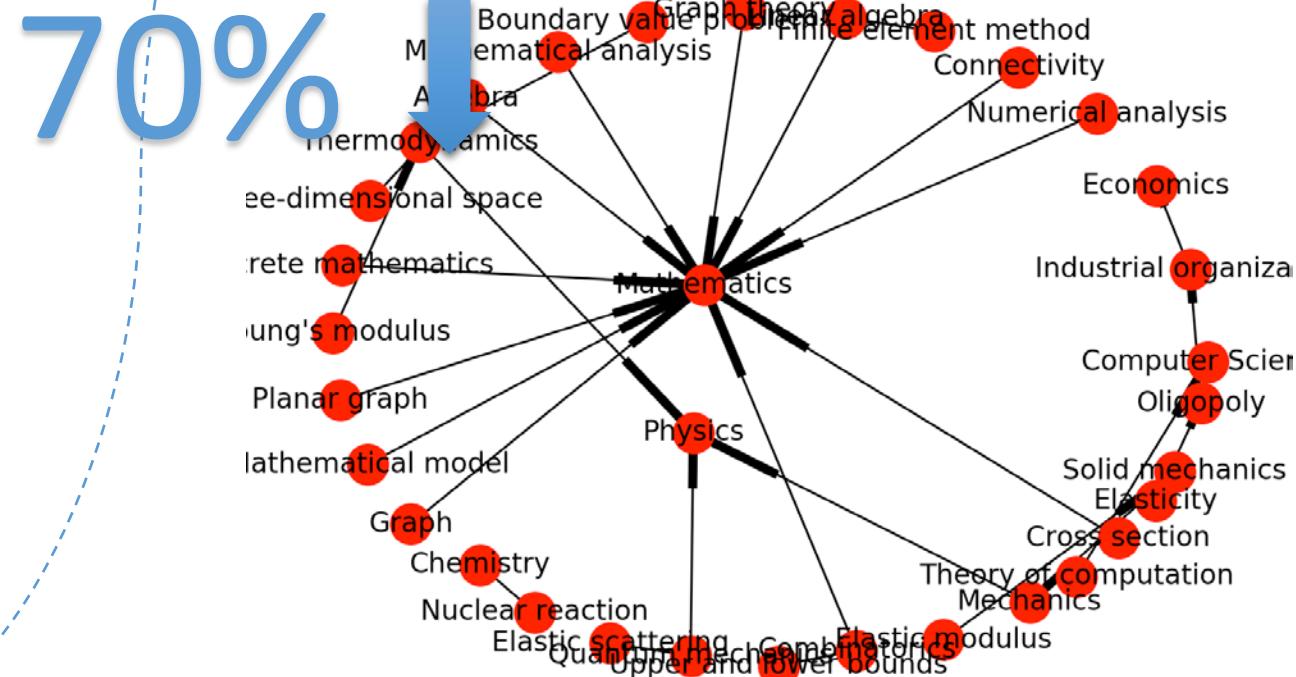
For the selected subtree sequences  $T_1, T_2, \dots, T_w$ , we delete each root of them, thus obtaining forest sequences.

Repeating this proceed until the leaves

New tree sequences  $T_1, T_2, \dots T_w$

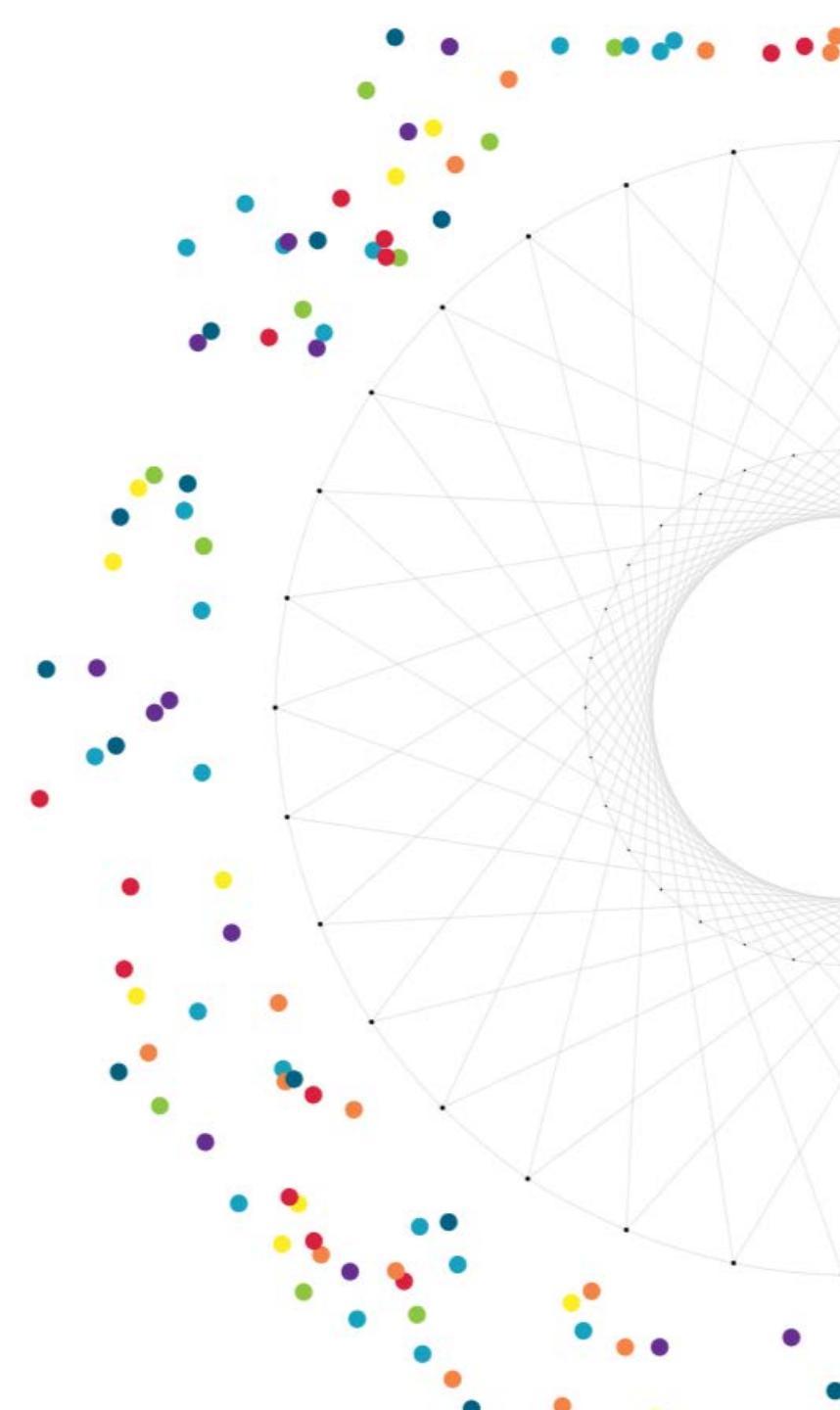


forest sequences  $F_i$



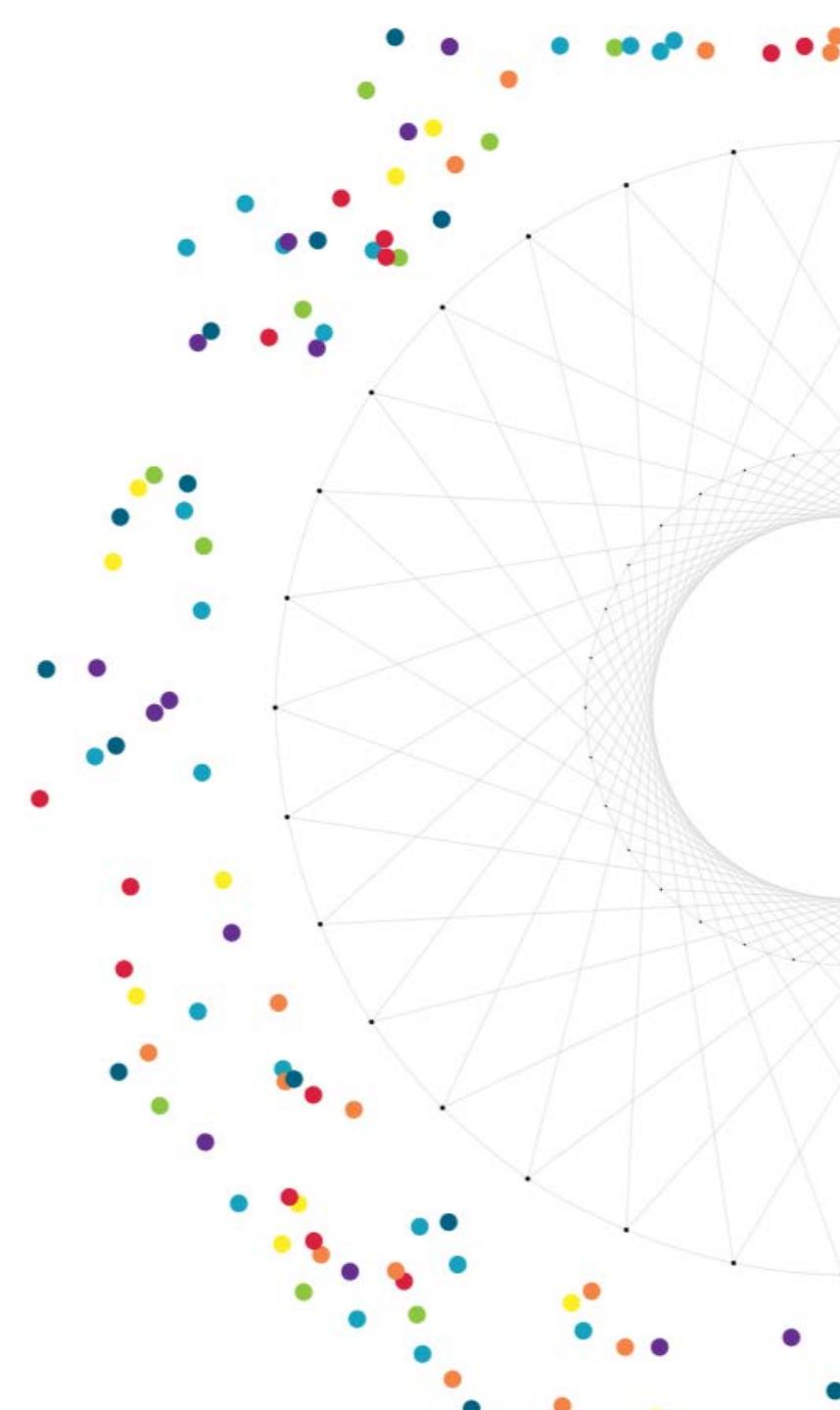
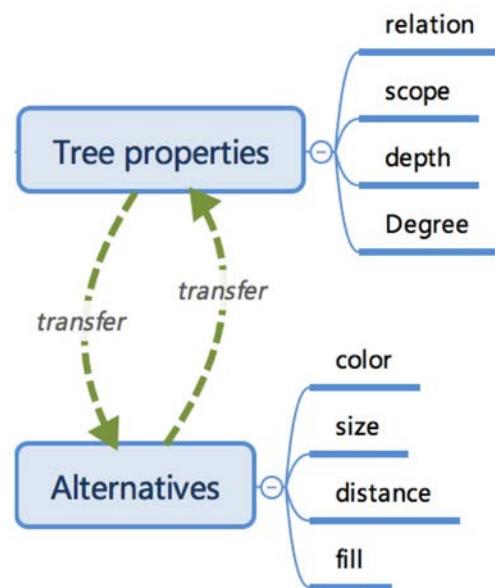
# Guide

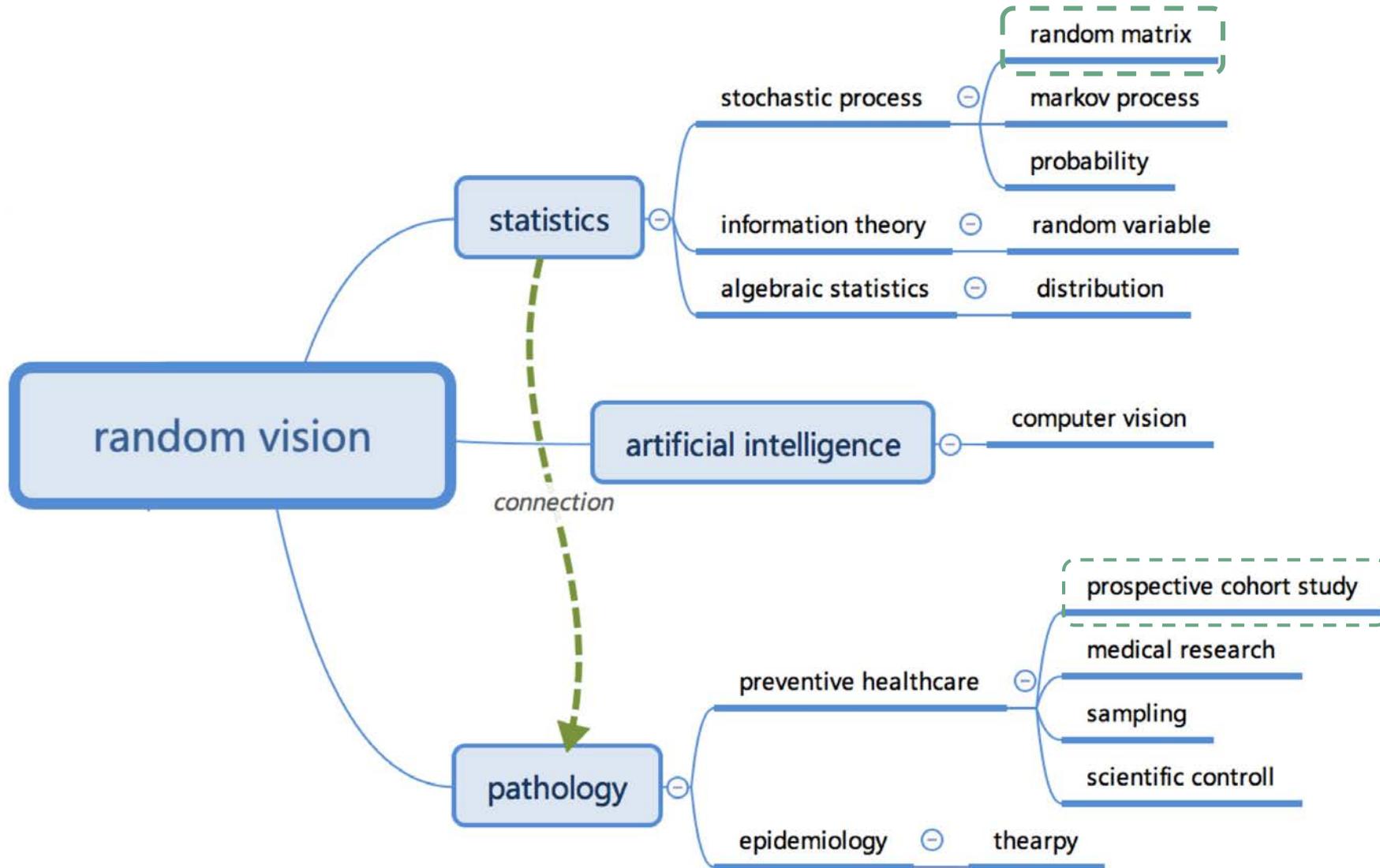
- Motivation for Knowledge Map
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- Our goal
- More informative
- Let user know what topic he is searching

- Different variables
- Depth
- Size
- Color
- fill of the stripe and frame

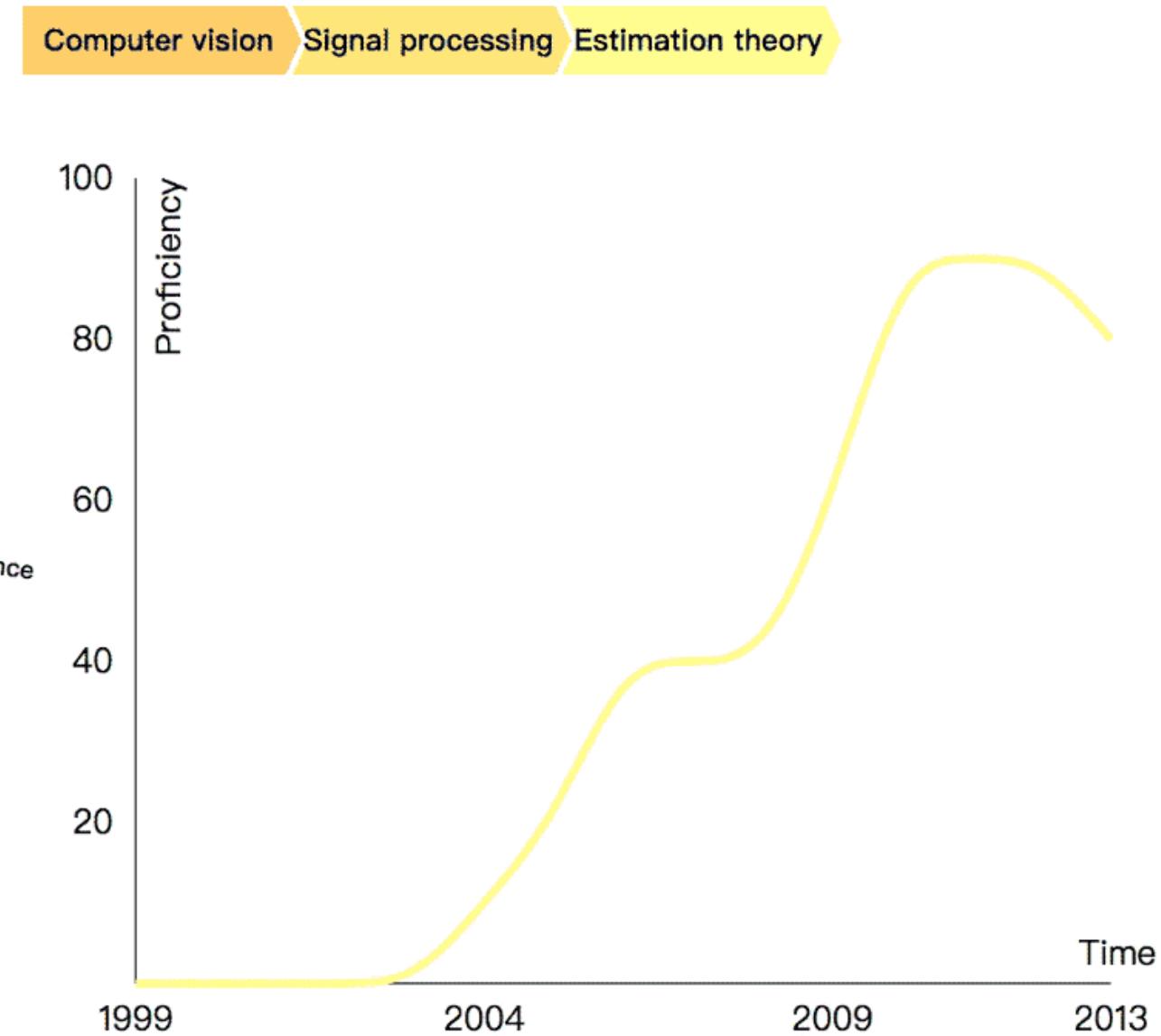
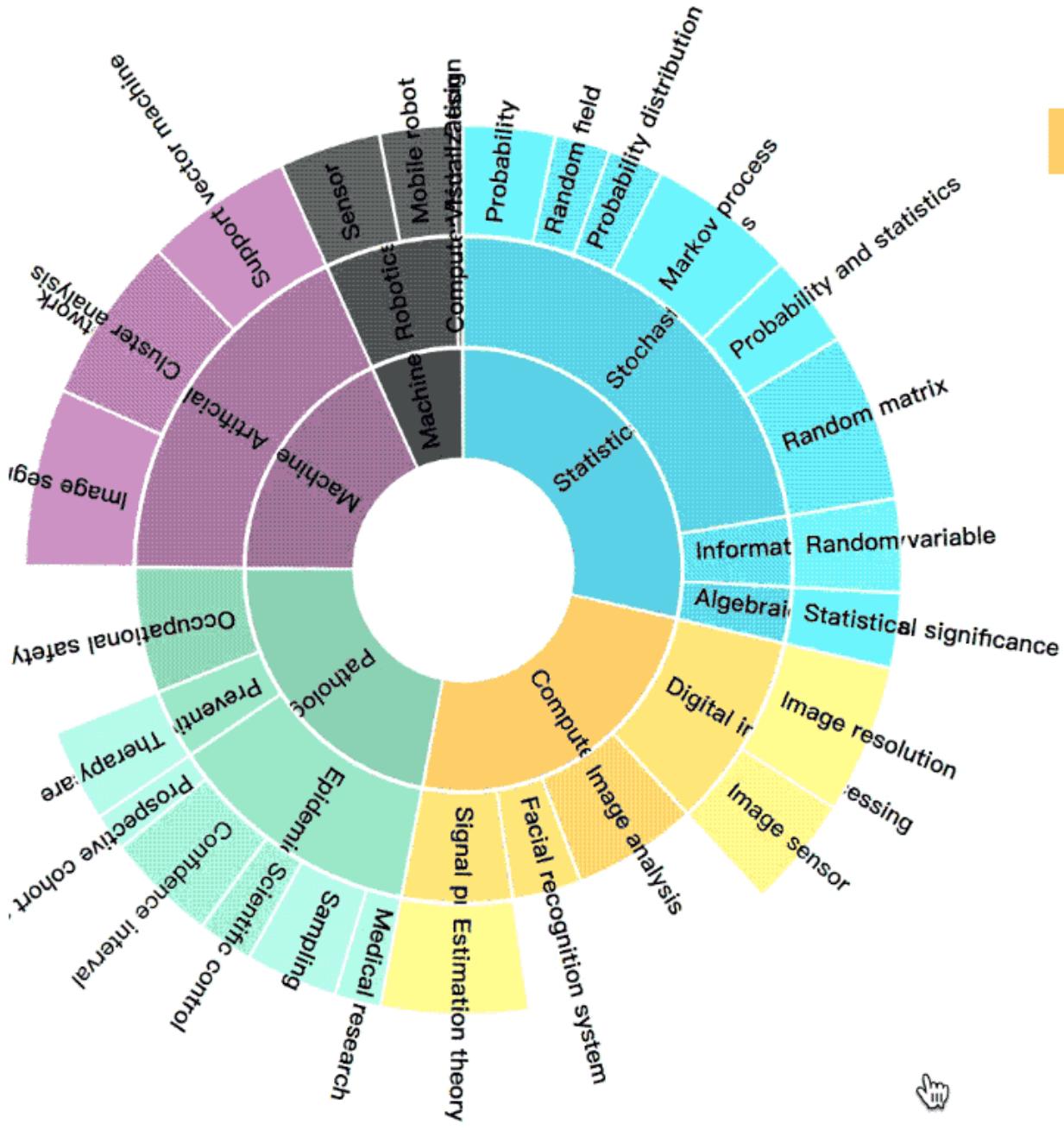


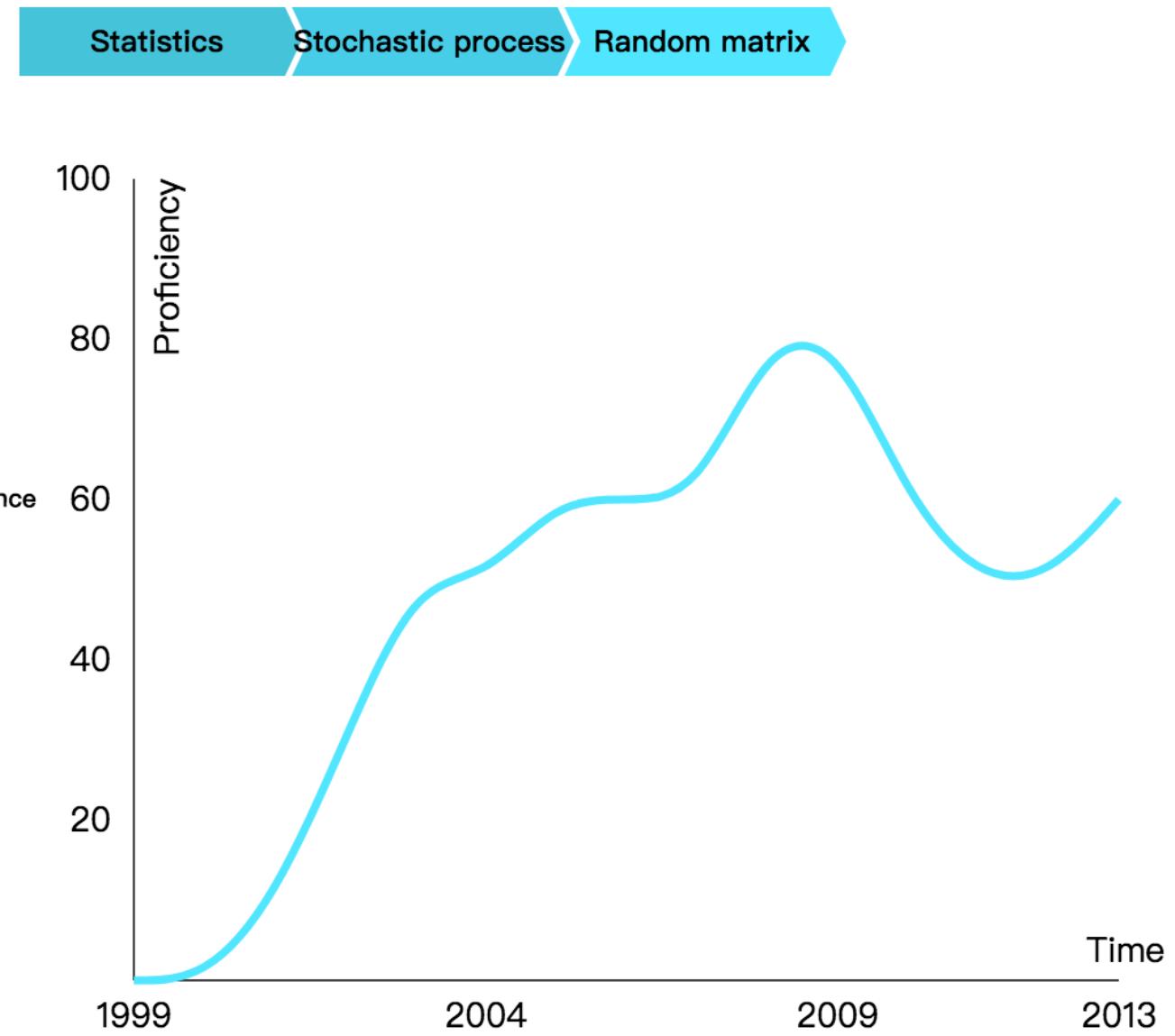
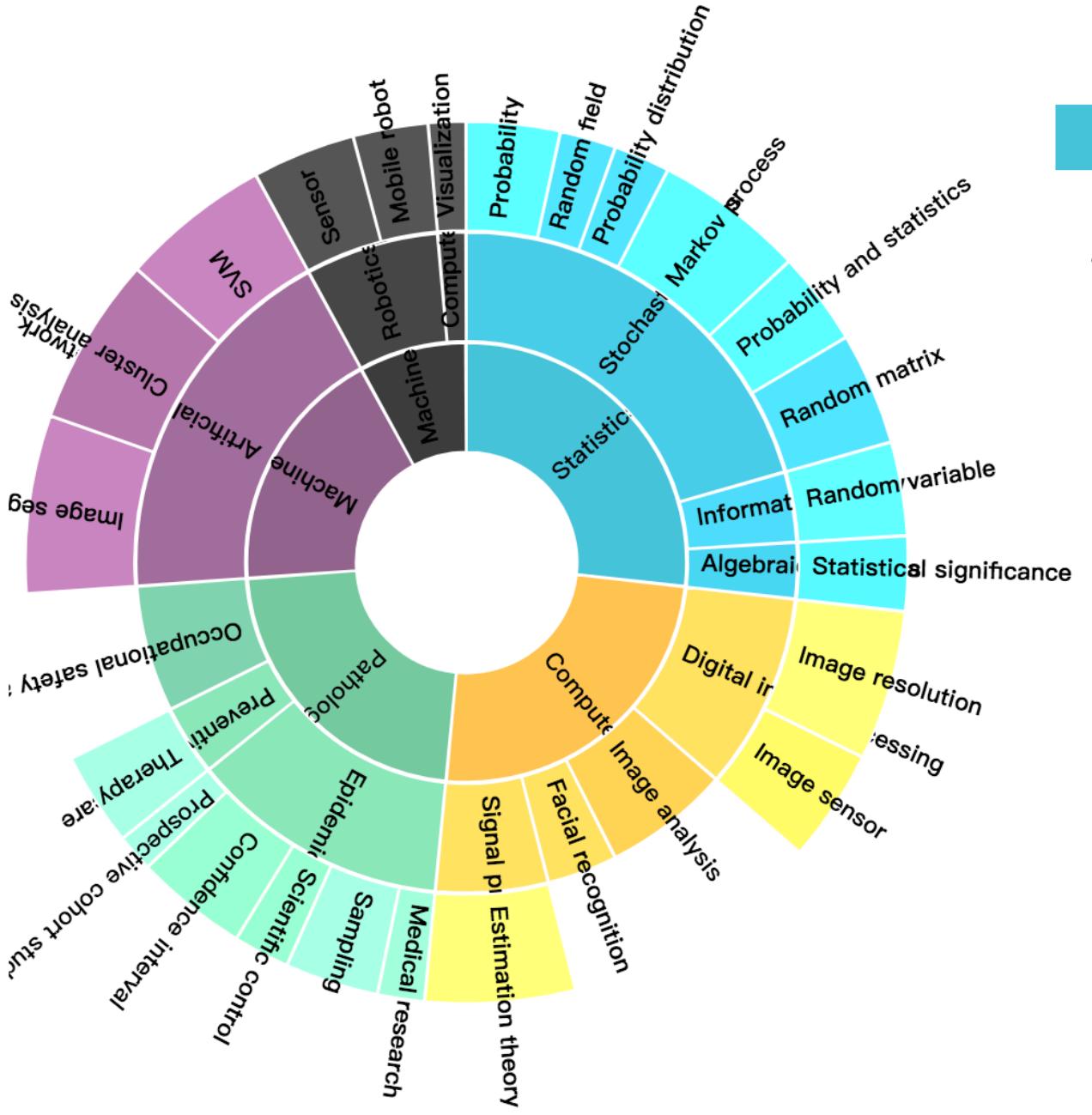


# D3.js Sunburst



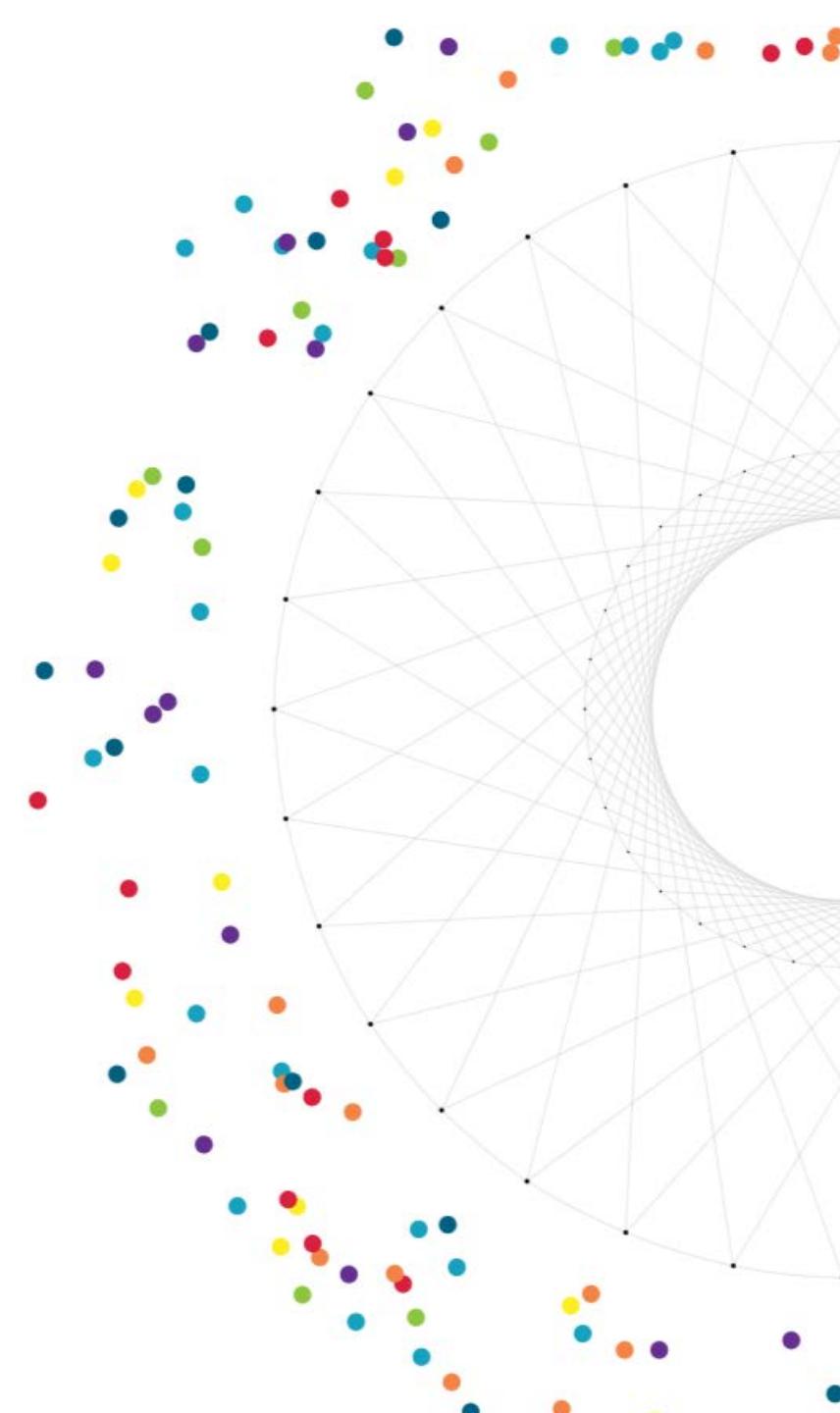
Data-Driven Documents





# Guide

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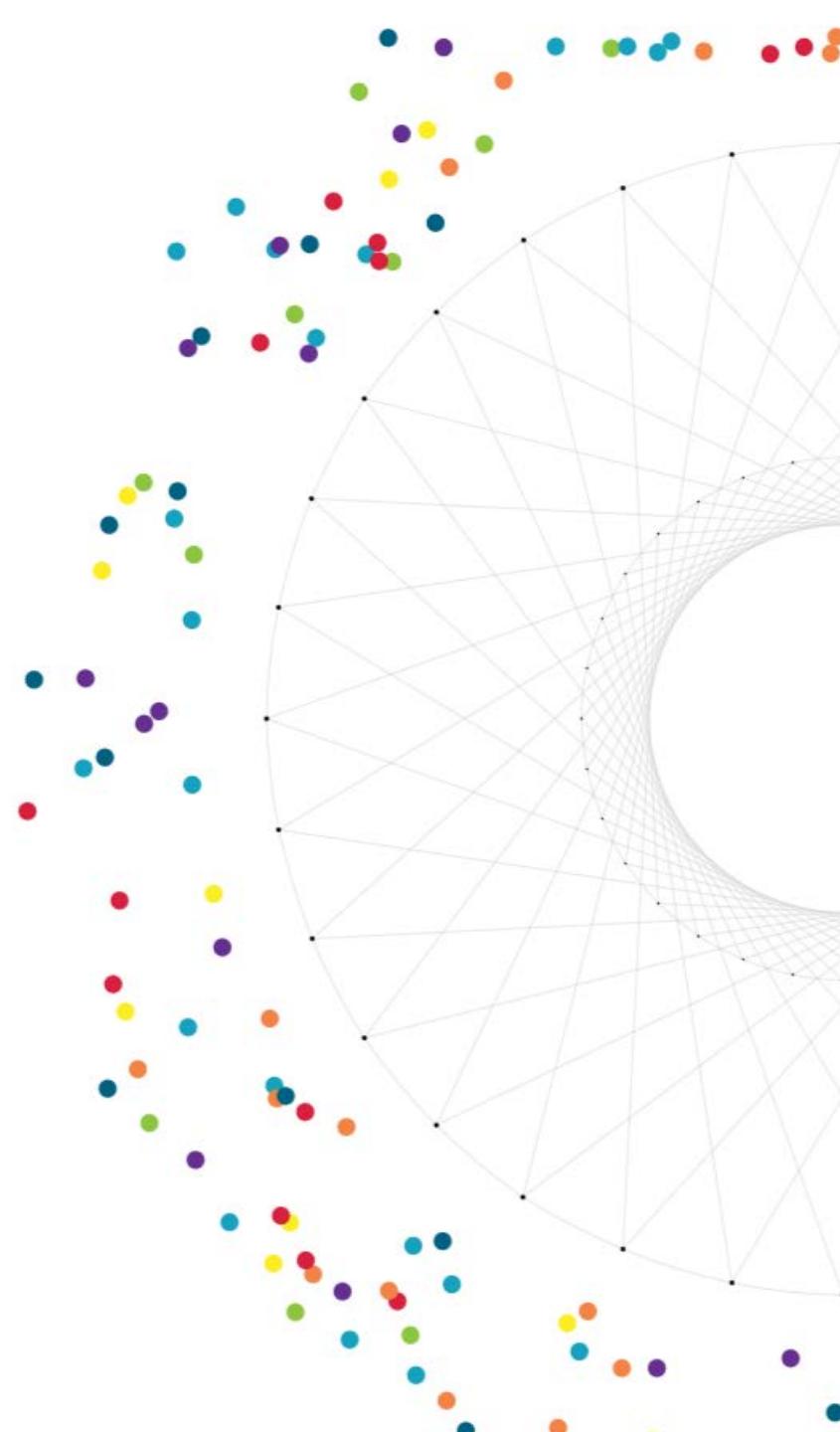


## 1. Theoretical Part

- (i) More Clustering method to be tested
- (ii) Seeking useful features to be added into the feature vector to accomplish the clustering.

## 2. Visualization

- (i) Changing the chart form
- (ii) More dynamic effect
- (iii) More topics information





**Thank you!**