



TJU_GSummary at TAC 2011:

Category Oriented Extractive Content Selection for Guided Summarization

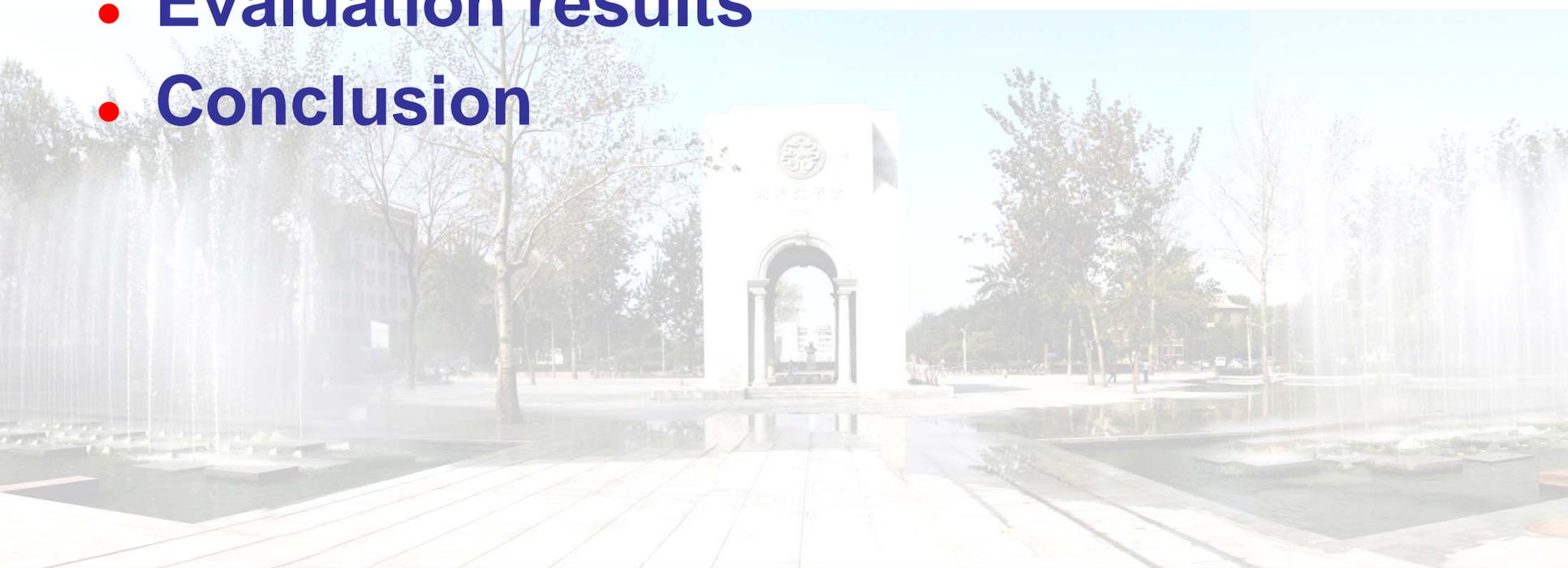
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Outline

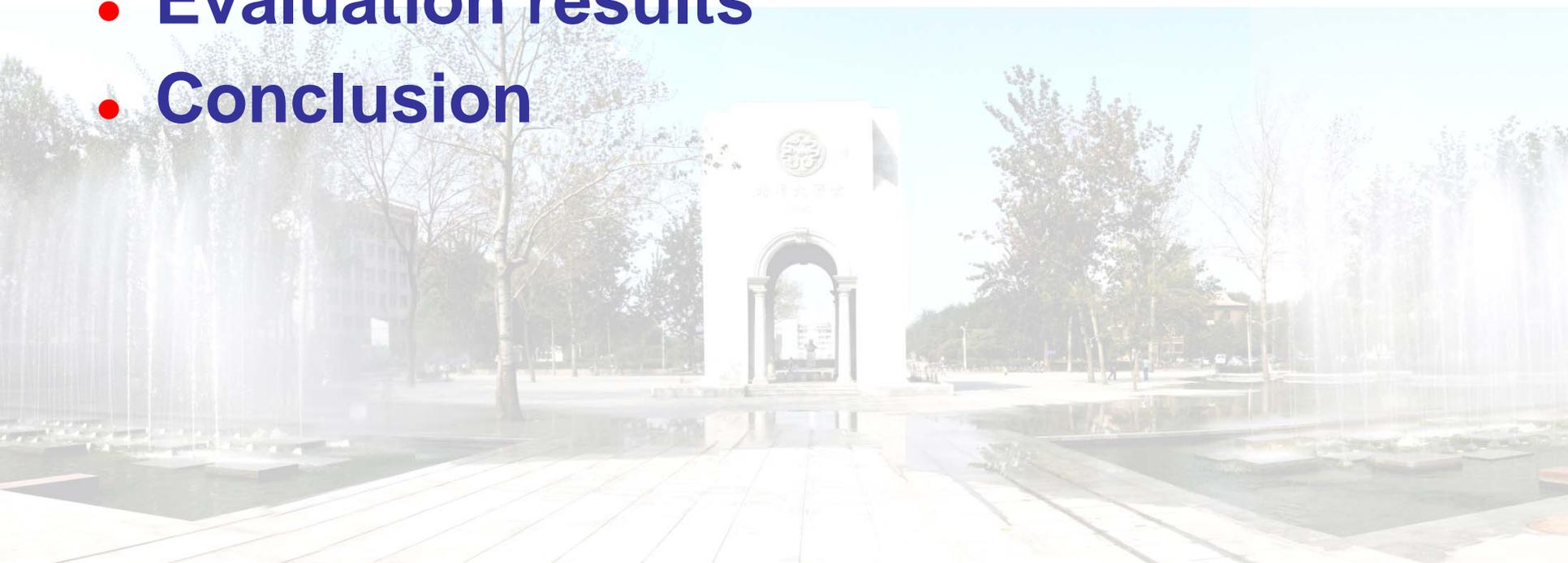
- **Viewpoint about guided summarization**
- **Methods of our systems**
- **Evaluation results**
- **Conclusion**





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Guided summarization

- Originate from the update summarization in DUC 2007
- The main difference
 - Topic is category oriented
 - But there is no topic description
 - Emergency related, such as natural disasters, accidents, attacks, public safety and so on.
- Accord with emergency management and crisis response



Question analysis (0)

from view of crisis management

First property: crises could be categorized, and have expected attributes, say aspects of an emergency, such as who, what, why

--This abstract semantic is different to annotate directly

--Make summary contain the different aspects of an emergency



Question analysis (1)

from view of data

Second property: data has the temporal evolution characteristic

--Deal with dynamic document collection of a single topic in continuous periods of time

from view of users

Third property: user needs have evolution characteristic

--Hope to incrementally care the important and novel information relevant to an emergency



New Challenges (extractive or generative)

- Semantic understanding of an emergency
- The capture of evolving information
- The balanced coverage of summary content

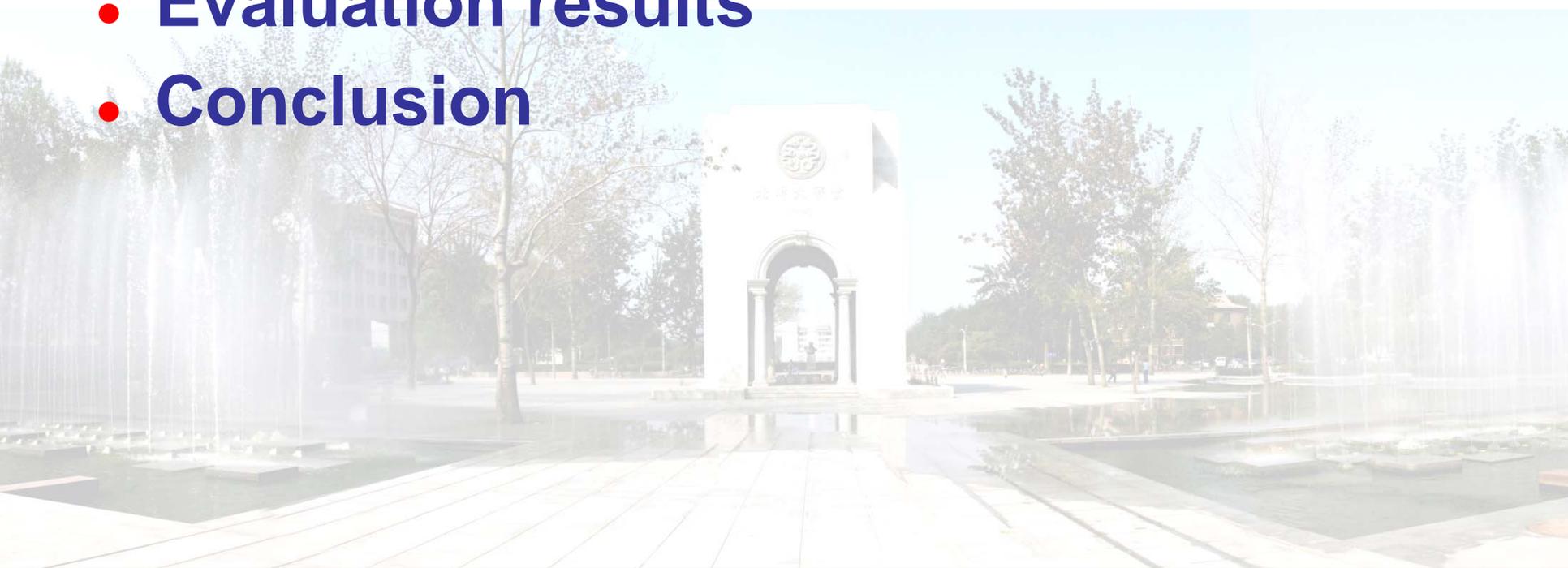
Just focus on the **extractive content selection**

How to model the importance and the redundancy of topic relevance and the content converge under the evolving data and user needs?



Outline

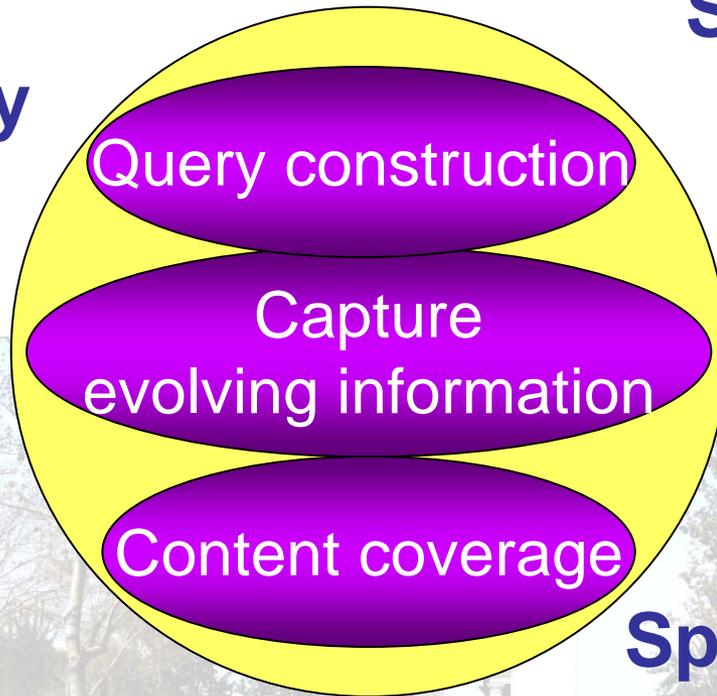
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Explore the new manifold-ranking framework under the context of temporal data points!

Evolutionary manifold-ranking



Semantic of an emergency



Spectral clustering



Combine evolutionary manifold-ranking with spectral clustering to improve the coverage of content selection!



Query construction (0)

- There is no topic description, we need to predict the user needs
- Find the replaceable semantic structure understanding method of an emergency
- Verb and noun mostly trigger the happening of an event



Query construction (1)

- Use the simple statistical method to extract the trigger words
- Try to answer the aspects of what and why
- Do not use any web resources



Evolutionary manifold-ranking

Manifold-ranking ranks the data points under the intrinsic global manifold structure by their relevance to the query

Difficulty: not model the temporally evolving characteristic, as the query is static !

Assumption of our idea

Data points evolving over time have the long and narrow manifold structure



Motivation of our idea



Add relay point of information propagation

Dynamic evolution of query

Relay propagation of information

Iterative feedback mechanism in evolutionary manifold-ranking

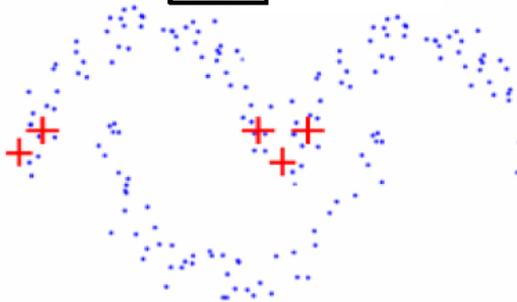
The summary of information propagation in previous time slices

The first sentence of documents in current time

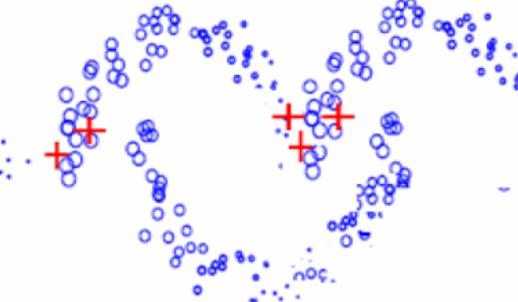
Relay point of information propagation

(a) Three moons ranking problem

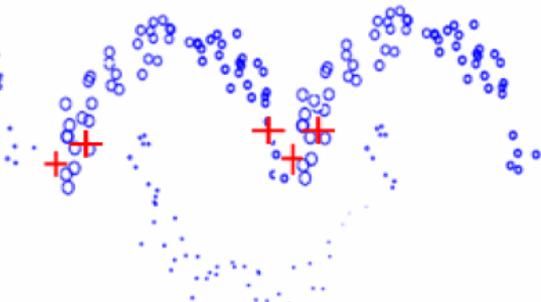
+ query



(b) Ranking by Eudidean distance

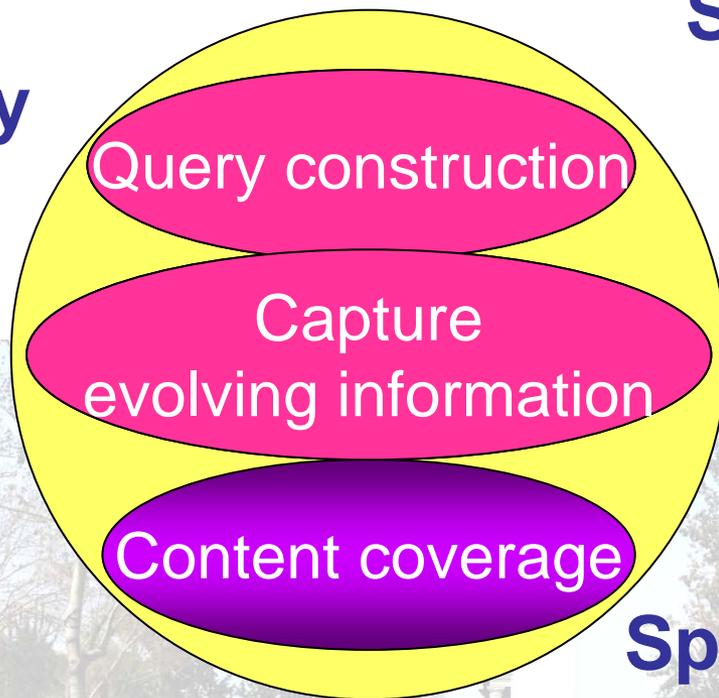
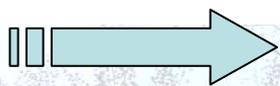


(c) Ideal ranking





Evolutionary manifold-ranking



Semantic of an emergency



Spectral clustering





Spectral clustering based on eigenvector selection (0)

- **Why choose the spectral clustering?**
 - Automatically determine the number of clusters
 - Cluster the data points with arbitrary shape
 - Converge to the globally optimal solution
- **Center object of spectral clustering**
 - Graph Laplacian transformation
 - Select normalized random walk Laplacian
 - Have good convergence

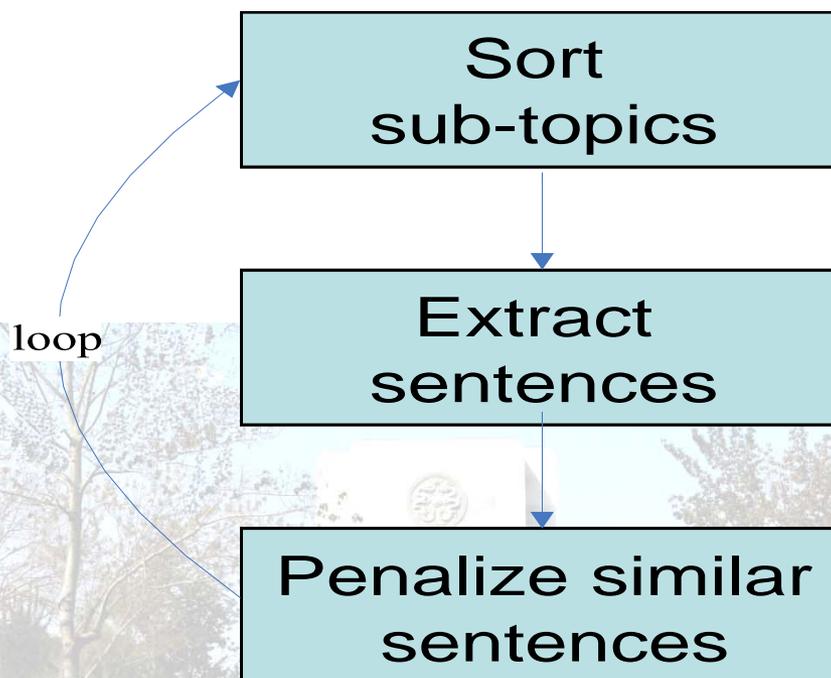


Spectral clustering based on eigenvector selection (1)

- Use the eigenvalue and eigenvector structure of a similarity matrix to partition data points into disjoint clusters
- However, not all eigenvectors are essential to clustering,
- Eigenvector selection based on **entropy ranking**



Sentence selection

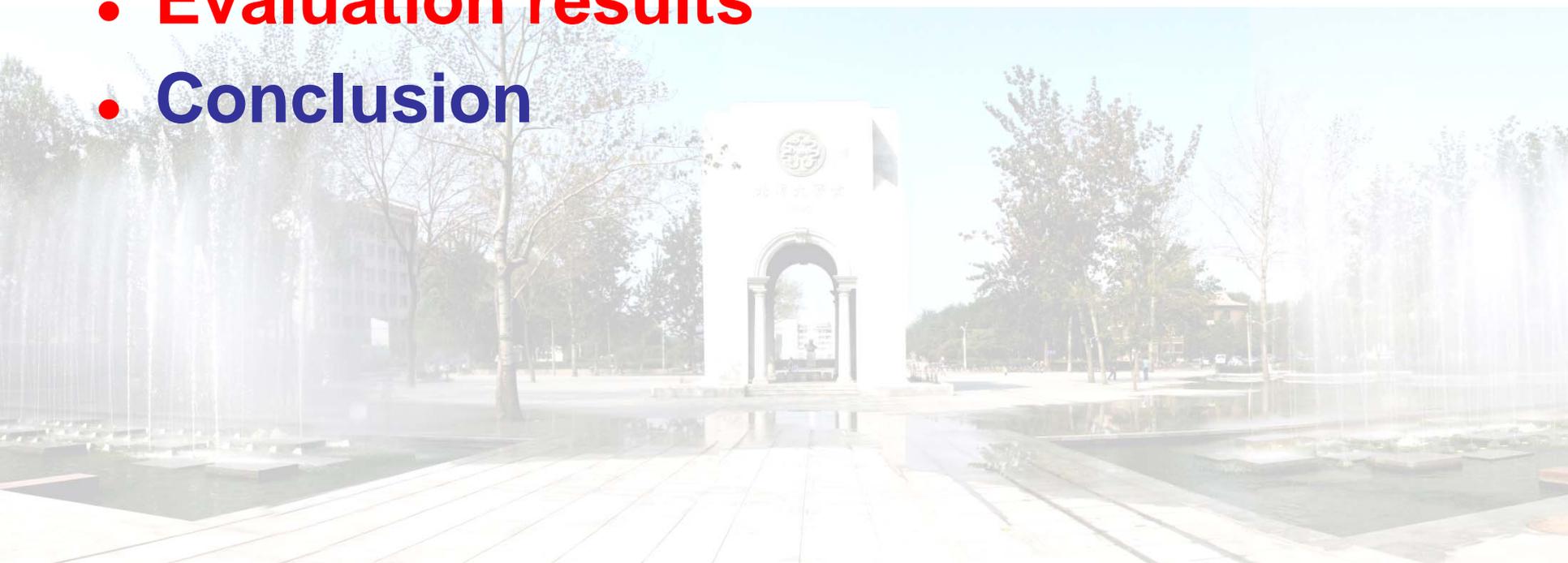


no sub-topics → a greedy algorithm



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System design

- **Experiments on TAC 2010**
 - (1) query construction;
 - (2) evolutionary manifold-ranking;
 - (3) spectral clustering based on eigenvector selection
- **Run systems**
 - RUN1(ID=13): (1) + (2);**
 - RUN2(ID=7): (1) + (2) + (3);**



Evaluation results (0)

Table 1: The Evaluation Results of System 13,7 under PYRAMID metrics

Metrics	Our A		Best A	Model A	Rank	
	Score(ID=13)	Score(ID=7)	Score(ID)	Low,High(ID)	ID=13	ID=7
AMS	0.413	0.392	0.477 (22)		22/50	30/50
ANSCU	5.523	5.091	6.227(22,43)	9.182,11.455(F,G)	14/50	30/50
ANP	1.409	1.841	2(33)		11/50	2/50
MMS3M	0.409	0.387	0.471(22)	0.705,0.888(F,G)	22/50	30/50
ALQ	2.75	2.614	3.75(32)	4.591, 5 (F,(D,C,E))	32/50	36/50
AOR	3.114	2.773	3.159(25)	4.682,4.955 (F,(D,C))	3/50	31/50
Metrics	Our B		Best B	Model B	Rank	
	Score(ID=13)	Score(ID=7)	Score(ID)	Low,High(ID)	ID=13	ID=7
AMS	0.33	0.338	0.353 (9)		10/50	6/50
ANSCU	3.614	3.75	4.023(9)	5.409,8.091(B,D)	11/50	7/50
ANP	0.795	0.841	1(43)		8/50	5/50
MMS3M	0.326	0.333	0.346(12)	0.554,0.823(B,D)	10/50	7/50
ALQ	2.773	2.727	3.455(1)	4.727,5(F,H)	27/50	29/50
AOR	2.364	2.477	2.591(35)	4.318,4.909(F,G)	21/50	11/50



Evaluation results (1)

Table 2: The Evaluation Results of System 13,7 under ROUGE,BE metrics

Metrics	Our A		Best A	Model A	Rank	
	Score(ID=13)	Score(ID=7)	Score(ID)	Low,High(ID)	ID=13	ID=7
ROUGE-2	0.10934	0.09687	0.13447(43)	0.1282(D)	16/50	29/50
ROUGE-SU4	0.14340	0.13053	0.16519(43)	0.16412(D)	19/50	30/50
BE	0.06332	0.05707	0.08553(43)	0.09085(D)	29/50	32/50
Metrics	Our B		Best B	Model B	Rank	
	Score(ID=13)	Score(ID=7)	Score(ID)	Low,High(ID)	ID=13	ID=7
ROUGE-2	0.06759	0.06889	0.09589(43)	0.11474(E)	31/50	30/50
ROUGE-SU4	0.10692	0.10753	0.13080(43)	0.14941(E)	30/50	29/50
BE	0.03682	0.04047	0.06480(43)	0.07970(E)	35/50	32/50

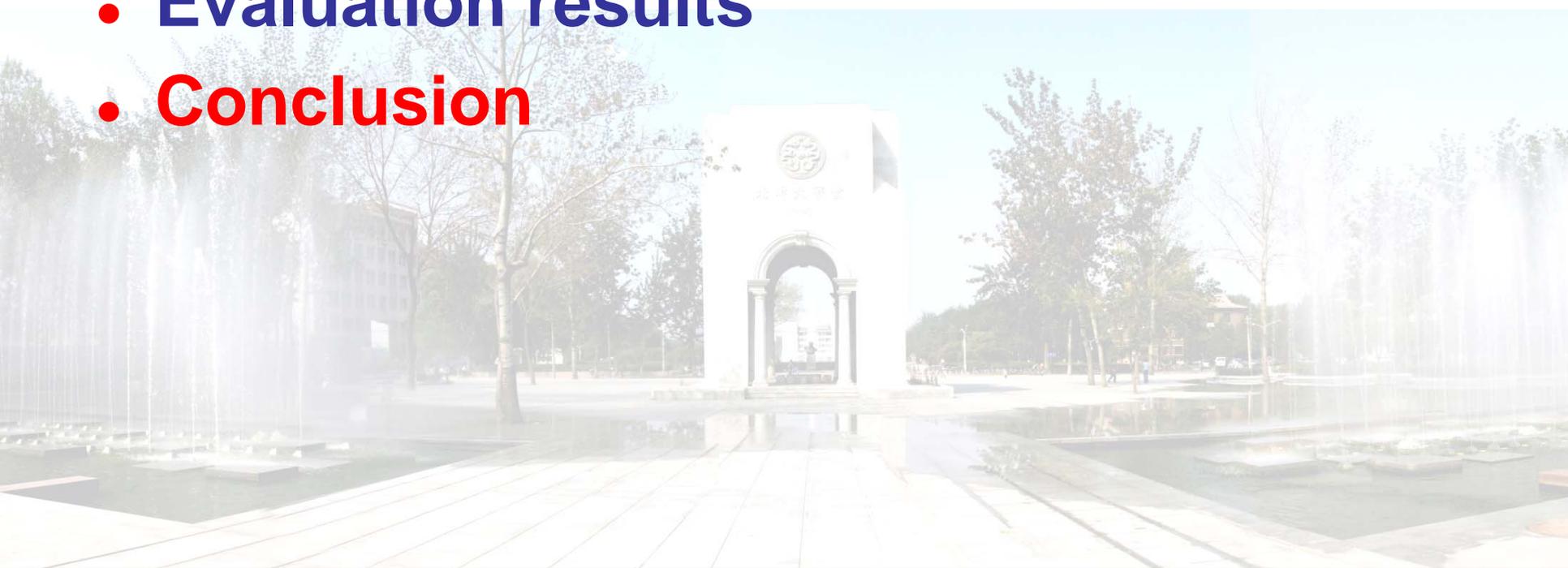
Difficult to say good or bad!

What is the appropriate metrics?



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Personal viewpoint

- **ROUGE and BE → content selection of generative summary**
Relatively short SCU
- **PYRAMID → content selection of extractive summary**
Long SCU

Hope: extend the number of time slice of evolving data



Conclusion (0)

- **Propose the category oriented extractive content selection method for guided summarization**
 - semantic understanding of an emergency
 - evolutionary manifold-ranking;
 - spectral clustering based on eigenvector selection



Conclusion (1)

- **Future work**

- label the semantic structure of an emergency
- mining the temporal characteristic of the evolving information
- better optimization method of parameters

- **Common topic**

Further explore the appropriate evaluation method for update summary



Thank you!

Any question?

