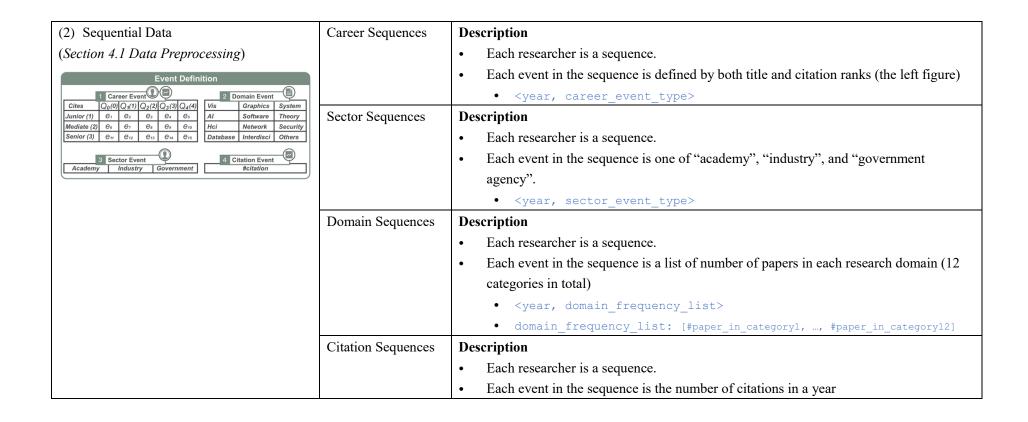
Supplementary Material

- Data Formats
- Implementation Details

Data Formats

Categories	Questions	Data Formats	
(1) Raw Data	Career Data	Description	
(Section 3.2 Data Description)		Background: we collect both LinkedIn data and other sources (researchers' personal	
		websites and their institutional webpages) into the same career format.	
		Data unit: the data unit is a researcher.	
		For every researcher, we collected the education and working experiences.	
		• For each type of experience, there is a list of events recording information including:	
		Event: <working job="" location="" organization,="" period,="" title,=""></working>	
	Bibliographic Data	Description	
		Background: Aminer holds almost all data in the VIS field.	
		Data unit: the data unit is a paper. We further arrange the data into researcher-based.	
		Attributes we use in a paper include:	
		<pre>• <year, author_list,="" venue_name=""></year,></pre>	
	Citation Data	Description	
		Background: we used Google Scholar API to get the citation data by year.	
		Data unit: the data unit is a researcher.	
		For each researcher, the numbers of citations are listed by year.	
		<pre>• <year, number_of_citations=""></year,></pre>	



Implementation Details

Implementation Categories	Steps	Remark
(1) Data Collecting	1-1. Extracted names of 1200+ VIS researchers.	By using DBLP data to apply the constraints:
(Section 3.2 Data Description)		those who have published more than two TVCG
		papers in which the largest time gap is more than
		five years as potential VIS researchers.
	1-2. Manually checked these names and filtered out those from	We manually checked names based on researchers'
	other fields (about 90 researchers) and finally obtained over 1,100	profiles, such as publications and homepages.
	VIS researchers.	
	1-3. Searched for three data sources based on researchers' names.	We use researchers' other information (e.g.,
		institutions) to distinguish those with the same names
		manually.
(2) Data Preprocessing	2-1. Career data: semi-automatically tagged the job titles and	We conduct the tagging in four steps (take job title for
(Section 4.1 Data Preprocessing)	sectors.	an example):
		1. Maintained a table to classify subjective job titles
		(e.g., assistant professor) into different categories
		in career data, which was generated based on
		domain knowledge.
		2. Used this table to classify more job titles
		3. For those new titles that were not included in the
		table, we manually checked them, either adding
		them into the table or just labeling the specific title
		with a category.
		4. Used both job title categories and the researchers'
		tenure in academic research to determine the title

		ranks (i.e., junior, intermediate, and senior).
	2-2. Bibliographic data: semiautomatically tagged the papers into different research domains based on an existing classification (Please refer to our paper Section 4.1 Data Preprocessing)	Sector tagging was conducted in a similar way. We took similar steps as the tagging of job titles above: 1. Some main venues were already classified in the existing classification. We did not need to do more. 2. For those venues not in the classification table, we also maintained a table to classify them into different domain categories using keywords semi-automatically.
	2-3. Bibliographic data: extracted the dynamic ego-networks of a researcher by year using paper author list.	
	2-4. Citation data: classify them into five categories using Quartile (Please refer to our paper <i>Section 4.1 Data Preprocessing</i>).	
	2-5. Construct four sequences based on the results of $2-1 \sim 2-4$. Please refer to our paper Section 4.1 Data Preprocessing (the second paragraph) and Fig. 2-B for more details.	
(3) MIA framework (Section 4.3 Multi-factor Impact Analysis (MIA))	Please refer to Section 4 Data Analysis for more details.	