

KG-PRE-view: Democratizing a TVCG Knowledge Graph through Visual Explorations



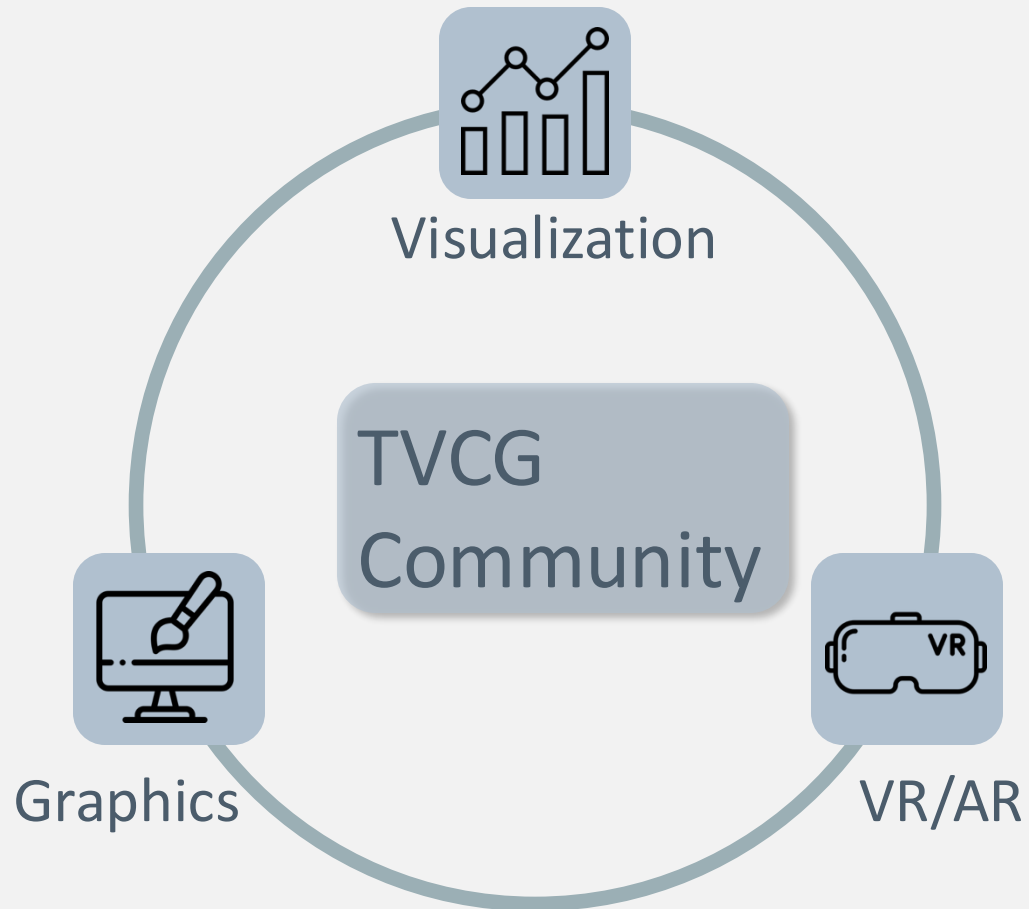
THE OHIO STATE
UNIVERSITY

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Presenter: Haoyu Li



Background



Retrieval of research ideas

Invitation of peer reviewers

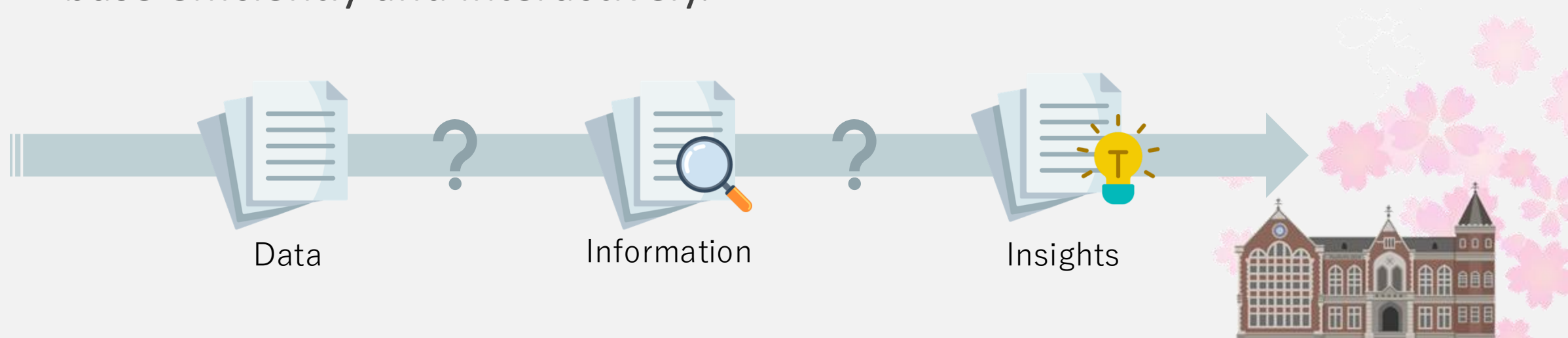
selection of editorial members



Challenge

Using current online access to digital versions of papers directly in real-world scenarios poses two challenges:

- **Pre-processing**: many exploratory questions require **information** extracted from paper contents.
- **Post-querying**: it can be difficult to generate **insights** from knowledge base efficiently and interactively.



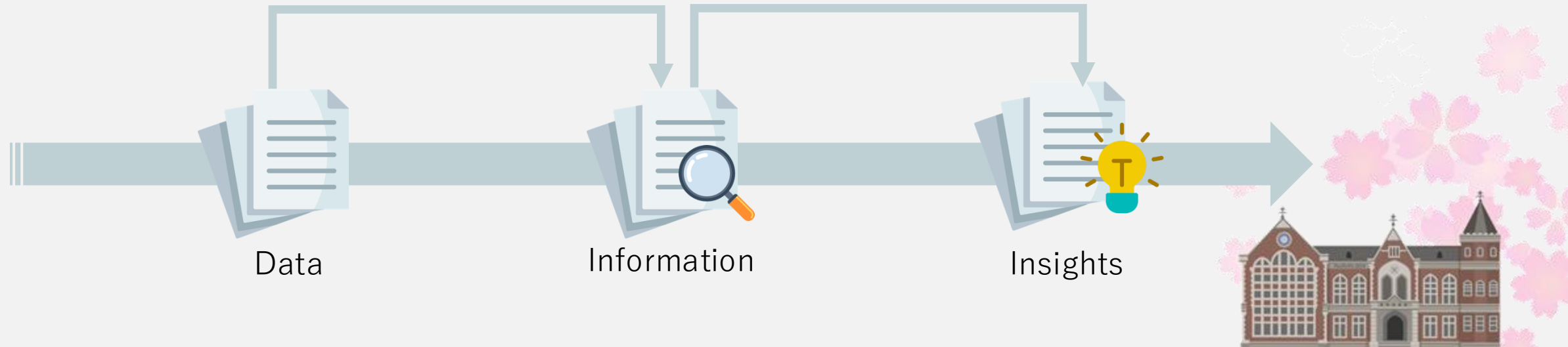
Introduction

? First, construct a knowledge graph for the TVCG community.

What + How

? Second, propose a PRE-view framework for KG visual exploration.

What + How



TVCG-KG Requirement

- KG should contain various types of entities, including *metadata* entity, e.g., Author, Affiliation and *semantic* entity from paper, e.g., Methods, Tasks.
- KG should contain semantic relationships, providing contextual information about entities.
- The format of K should offer flexibility and expressive queries for users to identify target information.



TVCG-KG Construction

Data Preparation

The most up-to-date TVCG dataset:

- Contain 4987 papers from 1995 to Aug, 2023
- Go through the data retrieval, cleaning, and validation.



TVCG-KG Construction

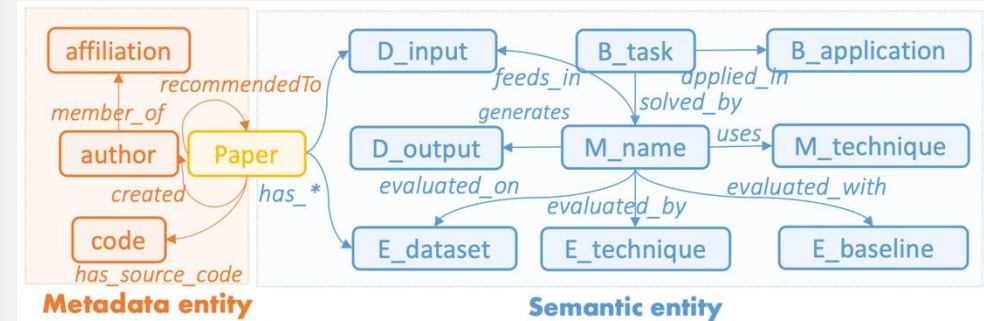
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Ontology Design

- Conducted a survey of 47 survey papers from TVCG.
- identified four semantic dimensions that are widely used: background, data, method, evaluation.



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KG Construction

- Propose an end-to-end pipeline for entity extraction and normalization.
- Experiment several design choices for prompt engineering.
- Utilize Spotlight API to normalize entities.



TVCG-KG Querying

- TVCG-KG can be imported in both RDF triplets or a Property Graph format.
- Various query languages can be utilized to query data from TVCG-KG, the basic semantics of which can be abstracted to resemble a SQL query:

```
SELECT {target}, WHERE {graph pattern}, FILTER {conditions}
```

“provide a list of papers published by Mystery Rivers.”

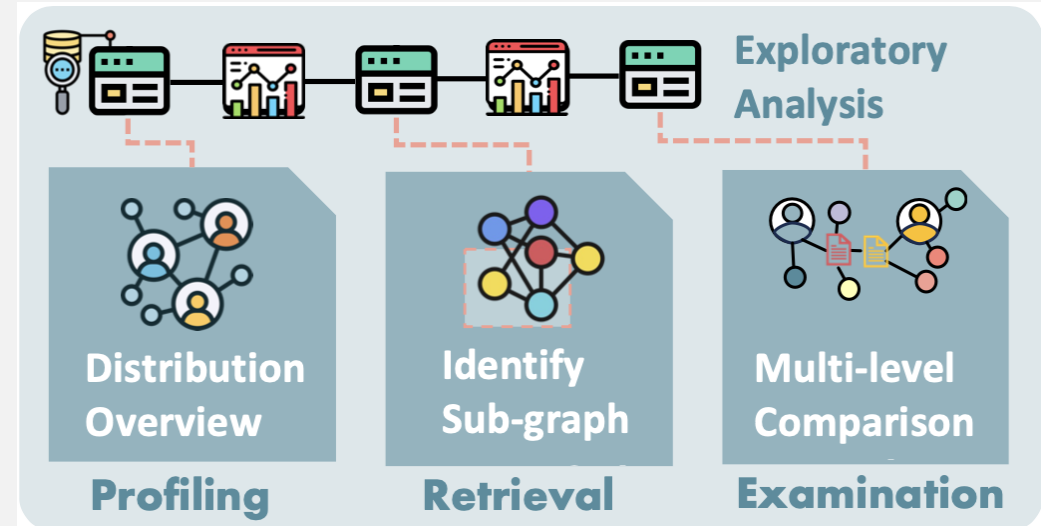
```
SELECT {paper}, WHERE {paper –created- author},  
FILTER {author = “Mystery Rivers”}
```



PRE-view Framework

To fill the gap between Information Metra and knowledge graph explorations, we introduce three visual exploratory tasks:

- **Profiling:** presents the overall structure of the KGs.
- **Retrieval:** helps users retrieve information of interest from KGs.
- **Examination:** delves into the details of target entities.



Various exploratory pipelines can be built as a sequence of tasks.



Evaluation of TVCG-KG

Structure-based
Statistical
Assessment

Data Quality
Evaluation

Task-based
TVCG-KG
Evaluation



Structure-based Statistical Assessment

- Compute several metrics of ontology, data model, and graph structure.
- Compute the number of different entity and relation class.

(a)			(b)	
Ontology	# of entity class	13	technique	19,861
	# of relations	28	author	10,916
	# of relations per class	4.54	application	7,257
Data	# of entities	81,033	task	5,963
Model	# of triplets	406,291	uses	50,124
Graph Structure	Avg. in-degree	2.42	has_technique	49,086
	Avg. out-degree	5.01	seeAlso	47,742
	# of weakly connected components	1	created	42,386



Data Quality Evaluation

- Data Consistency Evaluation:
 - Adopt a 10-fold strategy to evaluate the consistency of triplets.
 - Employ a Knowledge Graph Embedding (KGE) model trained on nine of these folds and predict the left-out fold.
 - Use Hits@K as the evaluation metric to measure the prediction performance. Higher score indicates better performance.
- Interlinking to External knowledge graphs.
 - 89.07% of author entities and 73.89% of paper entities can be mapped to the MAKG.

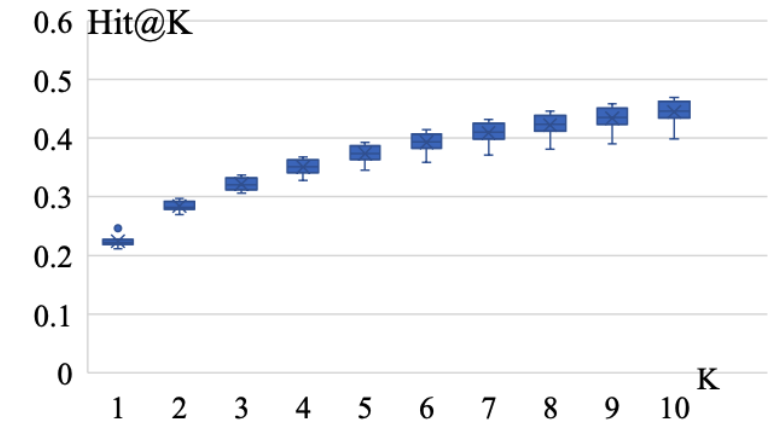
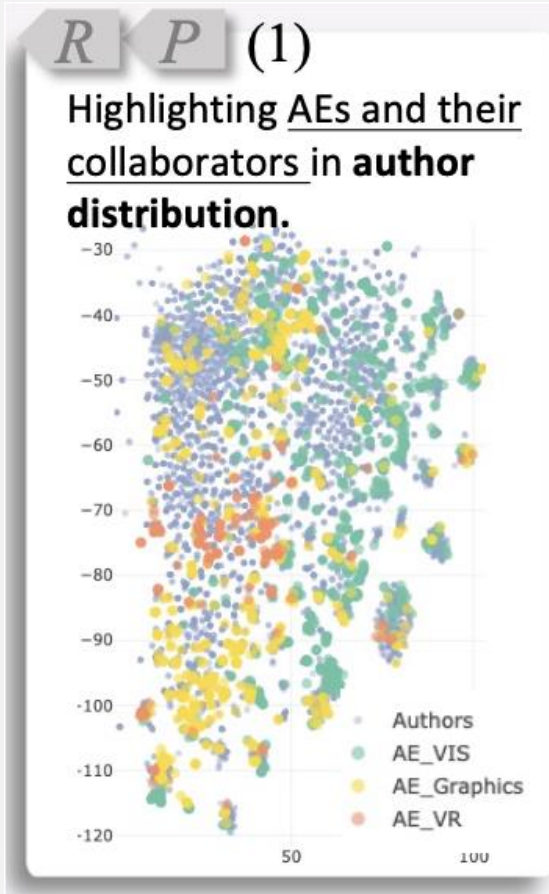


Figure 4: Statistical Distribution of 10 $Hit@K$ Scores for Each K Using a 10-fold Strategy.



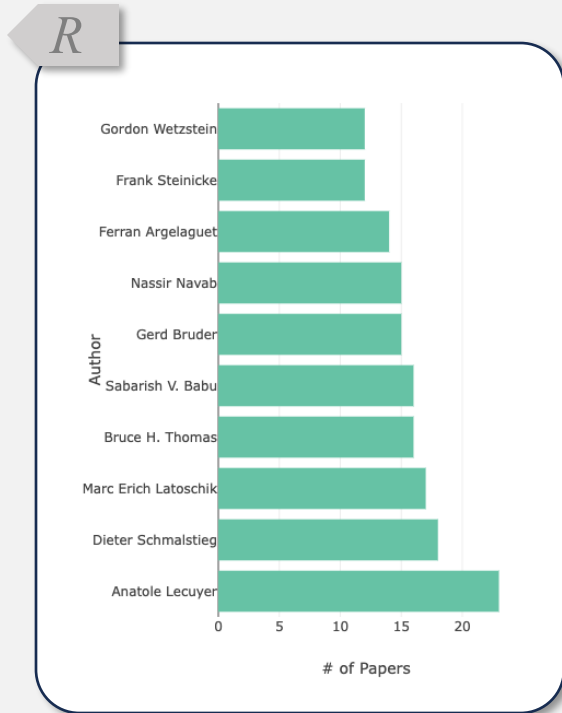
Case study – Author Profiling



- Train a TransE model on the triplets of TVCG-KG to generate embeddings and apply t-SNE to perform dimension reduction.
- To answer “Do the current Associate Editors (AEs) on the editorial board have a comprehensive coverage of the TVCG topics? “
- Color general authors as blue, and highlight AE and their collaborators by research areas (VIS, Graphics and VR).
- Indicating a good coverage of topics from distribution the AE and their collaborators across embedding space.



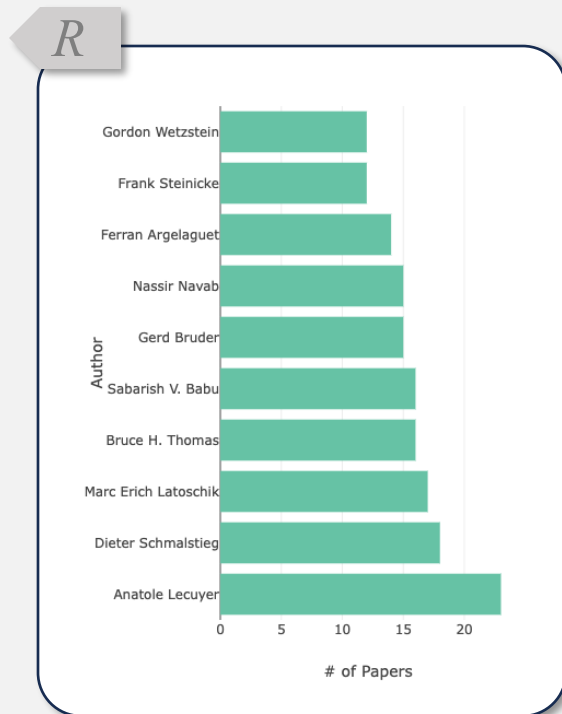
Case study – Author Identification & Analysis



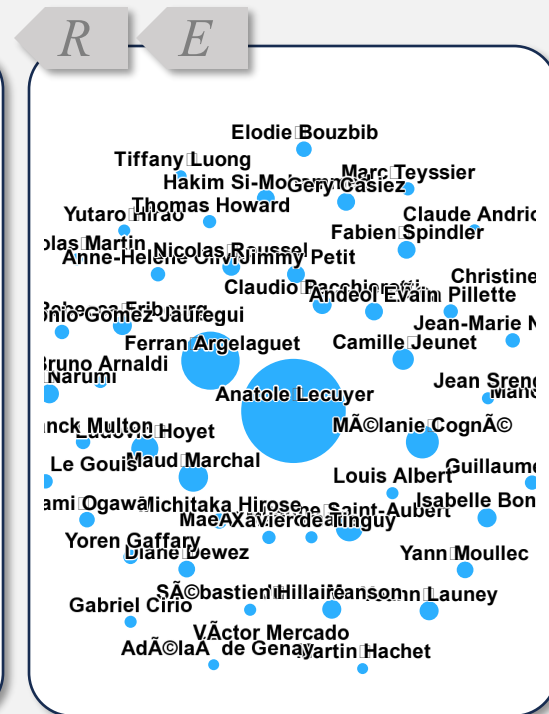
Retrieve **top authors**
in VR/AR/XR.



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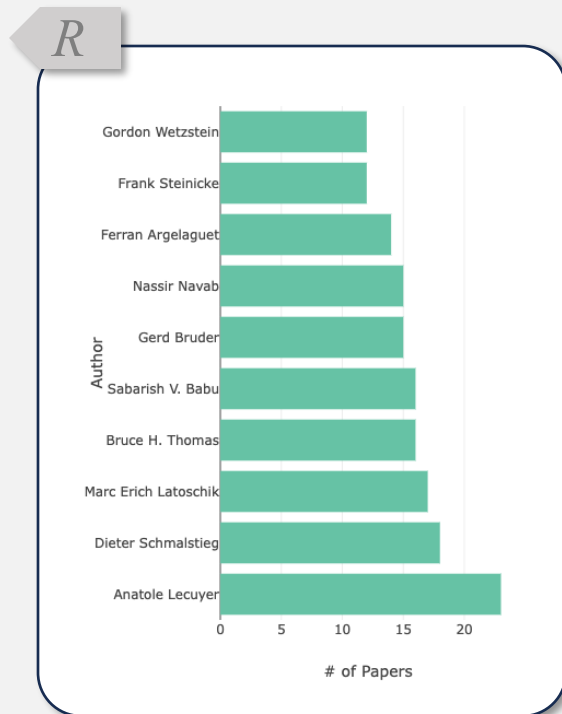
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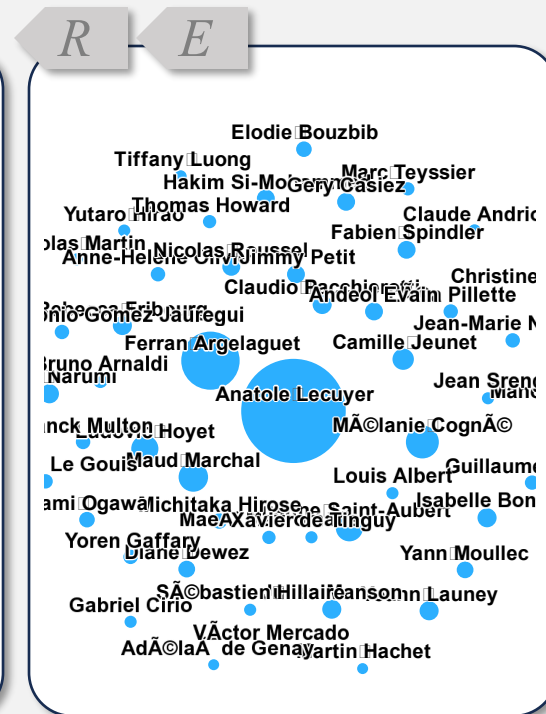
The **collaboration network** around
Anatole Lecuyer.



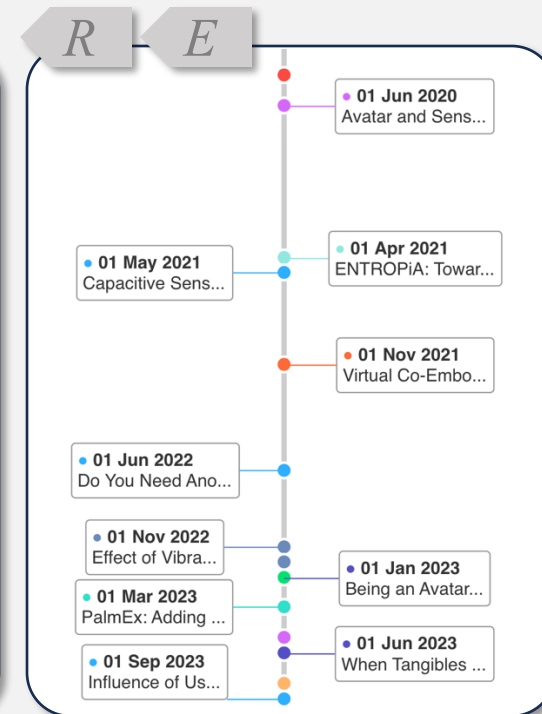
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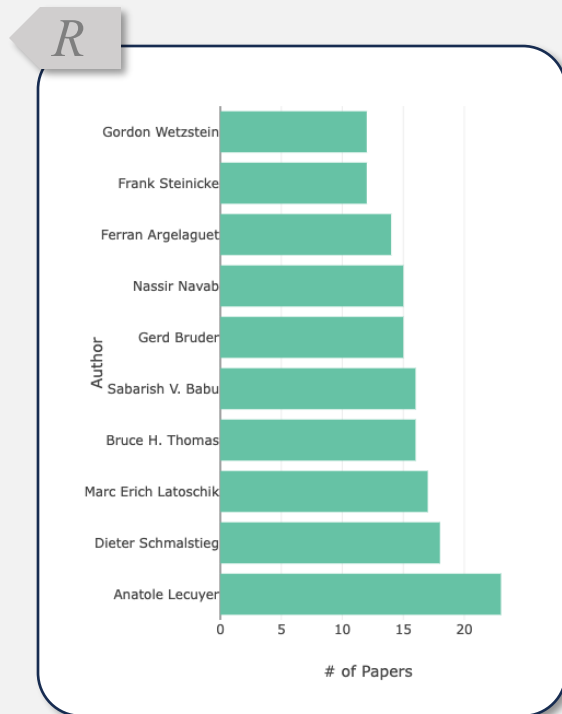
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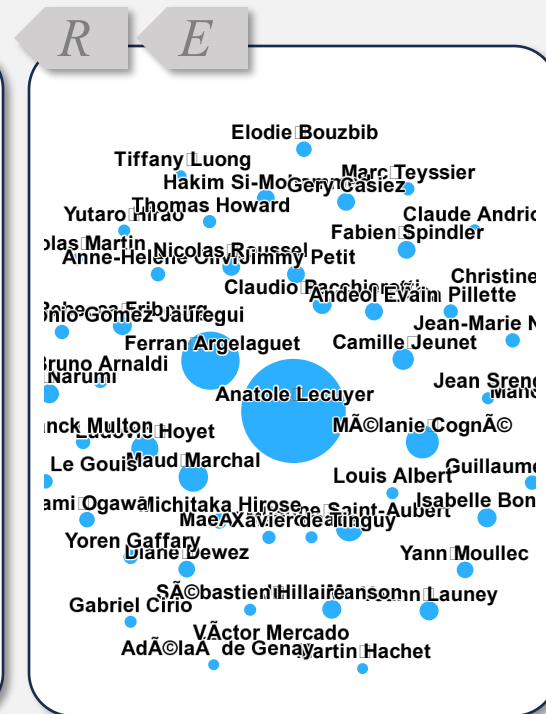
Publication timelines
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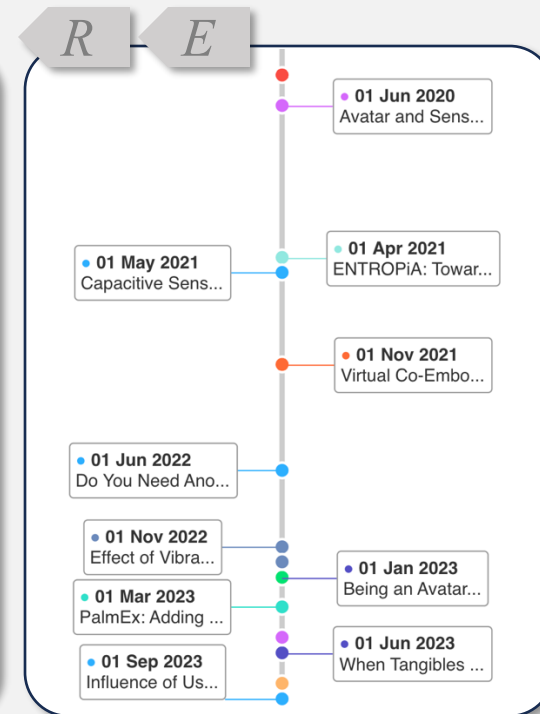
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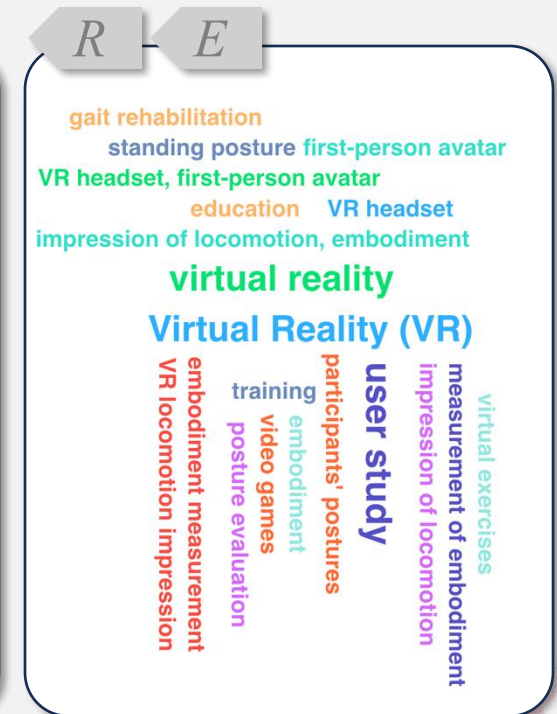
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network around
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Publication timelines
of Anatole Lecuyer in
TVCG.



Research interests of
Anatole Lecuyer.



Thank you

If you have any questions, feel free to contact us

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