Pāņditya: Visualizing Sanskrit Intellectual Networks

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Abstract

 $P\bar{a}nditya$ is an interactive web-based visualization tool for mapping and exploring rela-1 tionships between Sanskrit authors and their works. Built on the structured data of the 2 Pandit Prosopographical Database of Indic Texts, it offers scholars and students an in-3 tuitive and extensible open-source interface for dynamically investigating commentarial 4 networks, which are an essential aspect of Sanskrit intellectual history. Originally devel-5 oped as an offline tool (Pandit Grapher), Panditya has been re-imagined as an accessible 6 online resource leveraging modern web technologies for everyday scholarly use. Beyond 7 visualizing these networks, *Pānditya* also links to online Sanskrit e-texts through its sister 8 project SETI, and may eventually be extended to illustrate additional phenomena such 9 as parallel passages. 10

11 1 Introduction

12 1.1 Problem and Solution

Sanskrit intellectual history is vast and deeply intertextual, making it challenging to gain a holis-13 tic view of relationships between works and authors, even after years of study. Pandit—a digital 14 humanities project that curates, organizes, and shares structured data on South Asian literary 15 and intellectual history through an interactive, scholar-driven prosopographical database (Bron-16 ner and others, 2024)—provides a strong foundation for addressing this challenge. Until very 17 recently (see § 1.3), Pandit's wiki-like website lacked any intuitive, at-a-glance representation of 18 these relationships. $P\bar{a}nditya$ fills this gap by layering on an interactive network visualization, 19 enabling users to trace scholarly relationships with ease and engage more deeply with the mate-20 rial. By bridging structured data and human comprehension, $P\bar{a}nditya$ serves as both a research 21 aid and a pedagogical tool, enabling more effective exploration of Sanskrit intellectual history 22 while lowering barriers for newcomers. 23

24 **1.2** Related Network Visualization Projects

Interactive visualizations of network data are increasingly common in the humanities, but authorial and commentarial relationships are rarely complex enough to necessitate such an approach.
Likewise, network visualizations have not traditionally been used for organizing access to digital literary canons. Instead, most digital humanities projects have focused on other domains:
Social networks often reconstructed through documented correspondence family ties or

Social networks, often reconstructed through documented correspondence, family ties, or
 professional relationships, mirroring interest in modern social media analysis:

- Six Degrees of Francis Bacon, which maps early modern Britain's (1500–1700) social networks through textual analysis of primary sources (Warren and others, 2016).
- Mapping the Republic of Letters, which reconstructs Enlightenment intellectual networks through letter-writing archives (Edelstein and others, 2017).

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35	2. Citation networks in modern academic work (scientometrics):
36 37 38 39	 Connected Papers, which positions papers in a 2D similarity space based on citation patterns and semantic analysis (Tarnavsky Eitan and others, 2025). VOSviewer (Van Eck and Waltman, 2010) and CitNetExplorer (Van Eck and Waltman, 2014), which analyze large-scale scientific literature through co-citation, biblio-
40	graphic coupling, and co-authorship patterns .
41	3. Computed intertextuality phenomena, such as parallel passages and topic modeling:
42	• Open Knowledge Maps, which organizes academic papers into topic-based clusters
43	rather than direct citation networks (Kraker and others, 2025).
44	• Paper Machines, an open-source Zotero extension that enables researchers to analyze
45	bibliographic metadata and full texts using topic modeling and visualization tools (Jo
46	and Johnson-Roberson, 2012).
47	• The Viral Texts Project, which maps reprinting networks in 19th-century English-
48	language newspapers and magazines (Smith and others, 2015).
49	• BuddhaNexus, which traces intertextual connections across Sanskrit, Pali, Tibetan,
50	and Chinese using FastText embeddings, with a primary focus on Buddhist scriptures
51	(Nehrdich, 2020).

Perhaps the most closely related project is non-academic in nature. *The Oracle of Bacon* is a playful exploration of movie actor networks based on the "six degrees of Kevin Bacon" concept, which posits that any actor in (America-centric) show business can be connected to Kevin Bacon within six hops or fewer (Reynolds and Tjaden, 2025). This project directly inspired the v1 tool *Pandit Grapher*, as evidenced by the latter's use of the phrase "bacon hops." Surprisingly, commercial streaming services, as well as subsidiary navigation platforms like IMDb and JustWatch, have yet to adopt similar visualizations for content browsing.

59 1.3 Project History

The first version of the project, *Pandit Grapher*, required users to install Python, manually execute scripts, and export data for visualization in *Gephi* (Neill, 2021). While potentially useful for technically proficient users, this process was inaccessible to most Sanskrit scholars.

To address this barrier, the second iteration, $P\bar{a}nditya$, was developed as a fully online and 63 interactive visualization tool. Built with D3.js for dynamic graph rendering and Flask as a 64 backend framework, $P\bar{a}nditya$ significantly lowers the technical barrier, making this sort of in-65 teraction with Sanskrit scholarly networks more accessible and engaging. Users can explore 66 and customize visualizations in real-time using only a web browser, adjusting parameters and 67 filtering connections as needed. This transition from an offline, static workflow to an interac-68 tive web-based tool greatly enhances its potential as a reference and research instrument. At 69 the same time, the v1 feature of exporting data for use with offline tools like *Gephi* will be 70 retained—currently only on the backend, but soon also on the front end—as it is particularly 71 useful for visualizing large and dense graphs with hundreds or thousands of nodes. 72

In late April 2025, the *Pandit* project independently introduced a graph visualization feature 73 on its entity pages, similar in concept to $P\bar{a}nditya$. Notable advantages of the Pandit imple-74 mentation include: (1) broader coverage of entity types with options for targeted filtering; and 75 (2) polished interface elements such as a collapsible sidebar, refined zoom controls, and helpful 76 tooltip displays. Alongside this official implementation, $P\bar{a}nditya$ will continue in its current 77 role as an independent, open-source platform well suited to experimentation, offering space for 78 rapid prototyping, alternate feature sets, and exploratory visualization work. Mutual acknowl-79 edgment and ongoing collaboration between the two platforms can help ensure continued benefit 80 to the broader community. 81

82 1.4 Name Derivation

The name *Pānditya* derives from *pandita* ("scholar"), which is basis for the *Pandit* project's own name. *Pānditya*'s grammatical status as a *vrddhi* derivative (meaning "scholarship") also symbolizes its creative derivation from the predecessor project. The full name, *Pāndityatāraka*, can be taken to mean either "that which helps one cross (to the far shore of) Sanskrit scholarly learning" or "a tool for navigating *Pandit* data."

88 2 Data

⁸⁹ References to files below correspond to the project's GitHub repository at

90 https://github.com/tylergneill/panditya.

91 **2.1 Source**

Pāņditya is built upon a structured subset of the Pandit database, derived from a snapshot
taken on December 23, 2024. This source dataset contains 67,529 entities and 163 fields
(data/2024-12-23-pandit-entities-export.csv).

95 2.2 Reduced Entity Model

The dataset was filtered to focus on works (Content type==Work), their authors (Content type==Person), and selected additional information: alternate names (Aka and author Social Identifiers), work Discipline, and basic dates (Highest year and Lowest year). Entirely omitted were entity types such as Manuscript (of which there are 7,532), manuscript Extract (5,911), and modern scholarship Print (35,686), along with their associated attributes. Additional fields within the Work and Person types, such as Genre classifications or various interpersonal relationships, respectively, were also excluded for now.

¹⁰³ The reduced dataset is modeled with the following simplified Python class structure (see ¹⁰⁴ data_models.py for the full implementation):

```
class Entity:
105
        def __init__(self, entity_id:
                                         str):
106
            self.id: str = entity_id
107
            self.type: str
108
            self.name: str
109
            self.aka: str
110
            self.highest_year: Optional[int]
111
            self.lowest_year: Optional[int]
112
113
   class Work(Entity):
114
        def __init__(self, entity_id: str):
115
            super().__init__(entity_id)
116
            self.type: str = "work"
117
            self.author_ids: List[str]
118
            self.base_text_ids: List[str]
119
            self.commentary_ids: List[str]
120
            self.discipline: Optional[str]
121
            self.author_highest_year: Optional[int]
122
            self.author_lowest_year: Optional[int]
123
124
   class Author(Entity):
125
        def __init__(self, entity_id: str):
126
            super().__init__(entity_id)
127
            self.type: str = "author"
128
            self.social_identifiers: Optional[str]
129
```

130 self.work_ids: List[str]

self.disciplines: Optional[str]

This streamlined approach improves usability while preserving essential scholarly connections.
Users can still explore finer details through linked *Pandit* pages as needed.

134 2.3 Synthetic Attributes

131

Some clarification is needed for how certain of these attributes are newly constructed. Since the dataset only specifies a work's base text (via the field Commentary on (work ID)), its commentaries must be inferred by reversing this relation, iterating through all works to associate each with its corresponding commentaries. The resulting structure enables traversal of authorial and commentarial relationships in any direction, allowing for the construction of subgraphs that expand from arbitrary nodes.

Discipline is initially associated only with works—and even then, only sparsely. Based on these, a synthetic disciplines list is generated also for each author, as applicable, including counts of associated works per discipline (e.g., Maṇḍana Miśra's disciplines value is "Mīmāṃsā (3), Advaita Vedānta (1), Vyākaraṇa (1)").

Dates are tracked for both works and authors. When a work lacks its own date information, it may inherit the associated author's range, labeled accordingly (e.g., author_highest_year). In this way, the sparseness of date information for works can be partly overcome.

148 2.4 Obtaining and Processing Data

To support periodic updates to the underlying *Pandit* database, *Pānditya* is equipped with a simple Extract-Transform-Load (ETL) pipeline that processes *Pandit* data. However, the data must first be obtained from the *Pandit* database, which is less straightforward.

152 2.4.1 Exporting from Pandit Database

On the *Pandit* website's "Advanced Search" page, a "Download CSV" button allows users to download search results. When filtering by entity type via the left sidebar, this feature makes it appear possible to export arbitrarily large sets of entities, such as all Persons or all Works. However, such large requests do not currently complete on the project's production server. With the support of the *Pandit* team, work on $P\bar{anditya}$ has so far proceeded on the basis of a full export initiated by a team member with access to the internal development server and manually transferred to me via Google Drive.

That said, automating updates via the production server still appears feasible through the following approach:

- 162 1. Use the lightweight JSON API by appending the ?_format=json parameter to any node 163 URL (e.g., https://panditproject.org/node/89000?_format=json).
- 2. Periodically query all relevant entities (Works and Persons) known from prior data, check the changed attribute's timestamps for recent updates, and update records accordingly.

¹⁶⁶ 3. Leverage the sequential nature of numerical identifiers to detect newly published entities.

¹⁶⁷ Such automation has not yet been implemented at the time of writing.

168 **2.4.2** Extract

- ¹⁶⁹ Out of the original 67,529 rows and 163 columns, only the following were retained:
- 12,700 rows with Content type "Work", and 3,797 "Person" rows limited to individuals listed as authors of at least one work.
- 14 columns with a primary focus on authorial and commentarial relationships:
- ID, Title, Author (person IDs), Authors (person), Commentary on (work ID),
 Commentary on (work), Aka, Social identifiers, Discipline, Highest Year, and
 Lowest Year, which are kept as-is.

- Attributed author (person ID) and Attributed author (person), which are cur-
- rently merged into Author (person IDs) and Authors (person), respectively.
- Content type, which is ultimately dropped.
- This filtering is automated with the script utils/extract.py, producing the output data/2024-12-23-works-raw.csv.

181 2.4.3 Manual Cleaning

- ¹⁸² Before transformation, minor manual cleaning was required to remove a few spurious entities.
- Details are documented in data/manual_cleaning.md, and the cleaned dataset is saved as data/2024-12-23-works-cleaned.csv.

185 2.4.4 Transform

The next step, implemented in utils/transform.py, converts the cleaned dataset into structured Work and Author objects, organizing them into an in-memory lookup table that reflects the entity model described in § 2.2.

189 **2.4.5 Load**

Finally, utils/transform.py saves the processed data in a human-readable and retrievalefficient JSON format (data/2024-12-23-entities.json). This dataset can then be loaded by other code components using utils/load.py.

¹⁹³ 2.5 Component and Other Network Analysis

¹⁹⁴ The module utils/analyze.py analyzes and categorizes network components, i.e., communities

- ¹⁹⁵ of connected nodes, within the dataset, offering insights into how works and authors interconnect.
- ¹⁹⁶ Table 1 summarizes the distribution of components, and full lists of component members are
- 197 available at https://panditya.info/notes/data.

Component Type	Number of Components	Total Nodes
Isolated Nodes (single works only)	3,005	3,005
Small Communities (2–4 entities)	1,608	$3,\!605$
Medium Communities (5–9 entities)	90	565
Large Communities (10–25 entities)	24	344
Second-Largest Community	1	73
Central Community	1	8,905

Table 1: Summary of network component analysis.

A detailed discussion of these and other connection patterns is beyond the scope of this paper. However, the following key observations are noteworthy:

- The presence of a large central community based solely on accepted commentarial relationships underscores the fundamental role of commentaries in Sanskrit intellectual history.
- If additional intertextual phenomena, such as parallel passages, were included, many more so-called "isolated" works—especially those written by authors with only one or two extant texts—would be found to engage in broader intellectual discourse.
- This network structure may evolve as further updates to *Pandit* incorporate new philological findings.
- 207 Two additional practical takeaways from this analysis are:
- Isolated or nearly isolated items are relatively common.
- Conversely, participation in the central community is also widespread, making the visualization of subgraphs with 6-7+ hops impractical for many inputs.

Beyond this component analysis, utils/analyze.py also explores preliminary metrics such as centrality, influential nodes, and temporal patterns. These remain proofs of concept for now,

²¹³ but future refinements could significantly enhance the tool's analytical depth.

214 3 Web Application

Panditya is built on a modular and scalable web architecture:

- 1. A *Flask* backend using flask_restx with Swagger-based API docs.
- 217 2. A REST API that serves entity metadata and builds graphs on demand.
- 218 3. A *D3.js* front end that dynamically renders and updates network visualizations in real time.
- 4. Version control for both code (*GitHub*) and containerized deployments (*Docker Hub*).
- 5. Deployment on a cloud server (*Digital Ocean*).

221 3.1 Flask App

- ²²² The *Flask* app serves as the core backend, handling data requests and visualization processing:
- Loads entity data from the ETL pipeline.
- Exposes API endpoints (see § 3.2) used by the front end.
- Serves the main route (/) with HTML form and graph controls.
- Provides the /view route for external linking to specific graphs.
- Serves informational pages such as /about, /notes/technical, and so on.

228 3.2 Backend REST API

- 229 Key endpoints include:
- GET /api/entities/<type> Retrieves works, authors, or all entities.
- GET /api/entities/labels Maps ID numbers to human-readable labels.
- POST /api/graph/subgraph Generates subgraphs from selected entities and hop counts using simple breadth-first traversal, with optional exclusion of specified nodes.
- For users who wish to interact directly with the API, the following is also available:
- /api/docs Interactive Swagger docs with example queries.

236 3.3 Use of D3.js

- D3.js's forceSimulation models nodes as solid, mutually repelling objects connected by flexible
 links. Four adjustable forces determine the dynamic layout and can be tuned in-browser:
- forceCollide Controls how nodes resist overlapping with local neighbors. Higher values increase virtual node size.
- forceManyBody (.strength) Controls global node repulsion. Higher values increase repulsion strength.
- forceLink (.distance) Controls the spacing of connected nodes. Higher values increase link distance.
- forceCenter (.strength) Controls the tendency of nodes to return to the graph's center.
 Higher values increase centralization.
- D3.js also provides built-in support for dragging and zooming, simplifying interactions. Additionally, it facilitates the implementation of $P\bar{a}nditya$'s node context menus (see §3.7).

249 3.4 GitHub Repository and Local Deployment

- ²⁵⁰ For local development, users can follow these steps:
- Clone the repository from GitHub: https://github.com/tylergneill/panditya.
- Set up a virtual environment using Python 3.11+.
- Install dependencies from the provided requirements files: requirements.txt, requirements_etl.in, and requirements_offline.in.
- Start the server using the included Makefile: make run.

²⁵⁶ 3.5 Versioned Deployment on Digital Ocean

²⁵⁷ Deployment is managed using Docker and Digital Ocean.

- New builds are created for both development and production releases (see GitHub PR titles and descriptions for versioning details); each production release is also marked with a corresponding Git tag.
- Images are pushed to Docker Hub and deployed as containers on the Digital Ocean server.
- Nginx and Gunicorn handle traffic management and load balancing.
- Daily backups of the Digital Ocean "Droplet" ensure system stability and data integrity.

The production deployment is accessible at https://panditya.info, running release 2.4.9 at the time of writing (Neill, 2025).

266 **3.6 User Input Flow**

Using the web form, users select works and/or authors via auto-complete drop-downs, which use
IAST and are sorted in Sanskrit alphabetical order. These selections are grounded in *Pandit*data and disambiguated primarily by *Pandit* ID numbers.

Users should be aware of transliteration ambiguities. $P\bar{a}nditya$ currently relies on the mostly 270 precomposed IAST inherited from the *Pandit* database. As a result, search results may vary 271 when entering forms such as "Śamkara" (m with underdot), "Śamkara" (m with overdot), 272 "Śankara" (velar nasal), "Śamkara" (decomposed diacritics, i.e., S´ am_ kara), or simply 273 "samkara" or "sankara". The Select2 JavaScript module used for dropdowns generally han-274 dles these variants well, and $P\bar{a}nditya$ itself supplements search with auxiliary information such 275 as dates and alternate names, producing more verbose, disambiguated entries like "Śamkara 276 (85218) [710] [Śamkarācārya, Śankara ācārya]". Together, these measures are robust against 277 most orthographic variation, with the exception of decomposed diacritics. In the future, support 278 for a hub transliteration scheme such as SLP1 may improve consistency and search reliability 279 across edge cases. When in doubt, users can consult *Pandit*'s native search interface to help 280 identify the unambiguous ID number. 281

Once entities are selected, users then specify a hop count to determine graph expansion. Optionally, users can also exclude specific entities from expansion to reduce clutter in highly connected subgraphs.

For programmatic use cases, the /view route supports direct queries using these same inputs as URL parameters, e.g., Bhagavadgitābhāṣya, suppressing expansion on Bhagavadgītā and Śamkara: https://panditya.info/view?works=88637&hops=2&exclude_list=85218,42214.

288 3.7 Node Context Menu

²⁸⁹ After a graph is displayed, right-clicking a node reveals the following options:

- More info Lists additional fields when available: alternate names (aka), social
 identifiers (for authors), dates, and discipline (for works) or disciplines (for authors).
- 293 2. View on Links to *Pandit* entity pages and relevant online e-text repositories (see § 4).

3. Recenter - Sets the selected node as the new center and regenerates the graph, expanding
 outward by 1-3 hops.

4. Exclusions – Collapses the selected node, with future options for full removal and reexpansion.

298 **3.8 Force Controls**

Sliders beneath the graph allow users to adjust each of the four D3.forceSimulation forces (see §3.3). Additionally, a "Freeze" toggle temporarily disables all forces, enabling manual rearrangement, which can facilitate close inspection and/or screenshots.

302 3.9 Screenshots



Figure 1: Searching for an entity using auto-complete drop-down

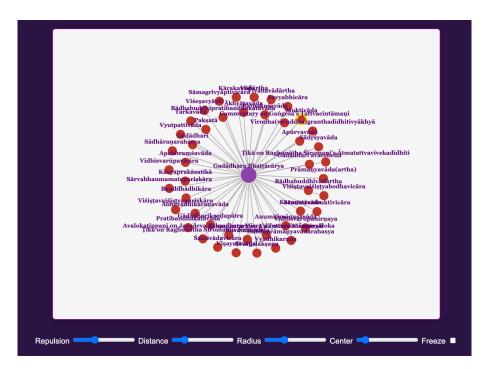


Figure 2: Generating basic graph

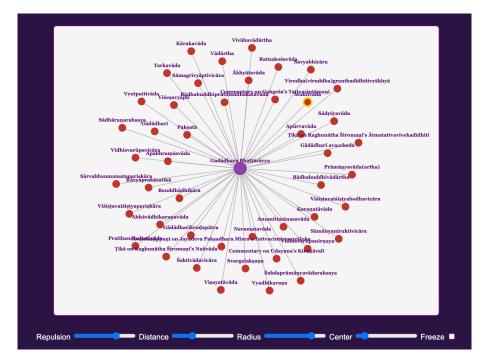


Figure 3: Adjusting force controls

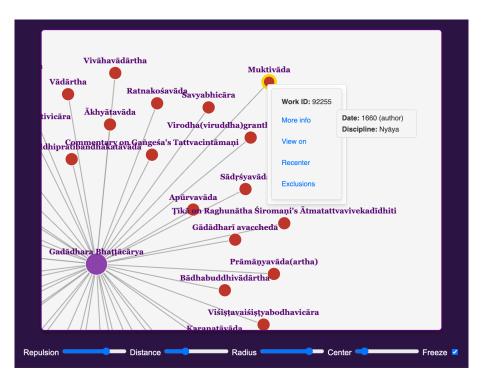


Figure 4: Opening context menu with right-click ("More info")

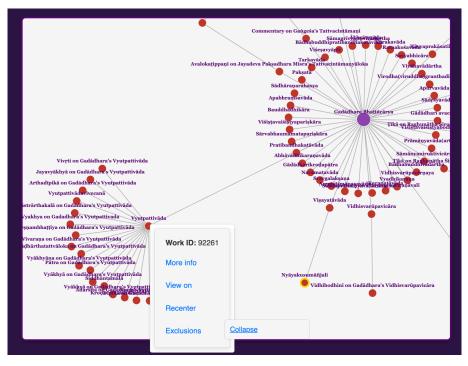


Figure 5: Using context menu to collapse prolific node

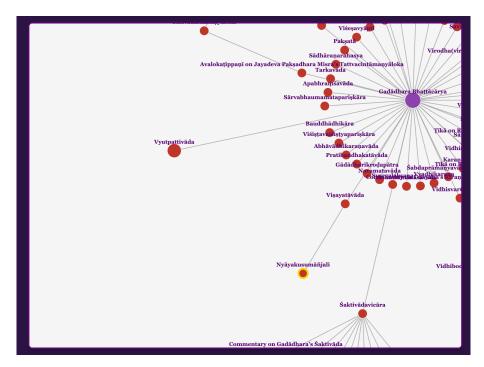


Figure 6: New graph with collapsed node

303 4 The Sanskrit E-Text Inventory (SETI)

In order to enable hyperlinks from $P\bar{a}nditya$ work node context menus to online Sanskrit e-texts, a separate but related effort, the Sanskrit E-Text Inventory (SETI), aggregates metadata from multiple repositories and aligns it with Pandit identifiers. Currently included repositories are GRETIL, Digital Corpus of Sanskrit (DCS), SARIT, The Sanskrit Library (and TITUS), sister projects Vātāyana and Pramāņa NLP, and the Muktabodha collection of digitizations from the Kashmir Series of Texts and Studies (KSTS).

This complementary data layer surfaces throughout $P\bar{a}nditya$: in the ETL pipeline's trans-310 form step, in the REST API (/api/seti routes), and in the D3 front end. For work nodes 311 highlighted in gold, the context menu's "View on" option reveals available e-text collections, 312 which are further differentiated into one to three levels of access, ranging from web read-313 ing platforms to raw GitHub data. Component collections and their contained e-texts can 314 also be programmatically queried and compared via /api/seti GET routes: by_collection, 315 by_collection/overlap, by_collection/unique, and by_work. In addition, the HTML route 316 /seti/by_collection/<collection>/visualize behaves like /view (see § 3.6), returning a 317 preloaded graph visualization in the browser—here, for entire collections at once. For exam-318 ple, https://panditya.info/seti/by_collection/SARIT/visualize shows the SARIT col-319 lection. Note, however, that only items present in both the source collection and the *Pandit* 320 database can be visualized in *Pānditya*. 321

This integration of $P\bar{a}nditya$ and SETI constitutes a powerful navigation tool and highlights

both the breadth of online Sanskrit resources and the structural role of shared identifiers. Further

details, including a system design diagram, numerical coverage overview, and access to the public

master spreadsheet, are available at panditya.info/seti and in the accompanying blog post.



Figure 7: Viewing Dharmakīrti e-texts via SETI and the "View on" context menu option

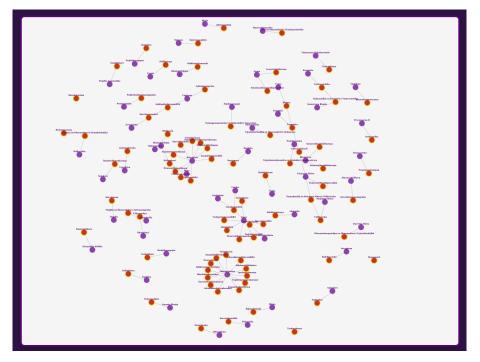


Figure 8: Visualizing SARIT

326 5 Conclusion and Future Directions

³²⁷ With its targeted use of interactive visualization, $P\bar{a}nditya$ bridges a tangible gap between ³²⁸ structured data and human comprehension. It enhances access to the underlying *Pandit* dataset,

empowering users to more effectively explore and understand Sanskrit intellectual relationships. Its three main use cases at present include:

- A gamified reference tool for students and scholars, encouraging deeper engagement with
 Sanskrit works and authors.
- 2. A mechanism for improving *Pandit* data by highlighting inconsistencies and missing connections, while also fostering greater interest in the project.
- 335 3. An electronic catalog tool for finding and navigating to Sanskrit e-texts online.
- ³³⁶ Future expansions may include:
- Building additional features on top of *Pandit* data, such as quick search for related works by both discipline and date, or visualization of family trees.
- Feeding insights from *SETI* back into *Pandit* in the form of new entities and connections, thereby increasing visualization coverage within *Pānditya*.
- Illustrating intertextual network analysis, using weighted edges to represent textual interactions as computed by projects like *BuddhaNexus* (Nehrdich, 2020) and *Vātāyana* (Neill,
- ³⁴³ 2022), with links to interactive intertextuality reports where available.
- ³⁴⁴ Beyond these enhancements, several known development priorities remain:
- Strengthening *Pandit-Panditya* integration to streamline data refreshes.
- Improving navigation with more flexible input methods (e.g., transliteration options) and alternative modalities like alphabetical browsing.
- Implementing error-checking to prevent the creation of overly large subgraphs.
- Exposing v1 export functionality through the front-end interface.

With its rapid development cycle and adaptable design, *Pāṇḍitya* aims to serve as a vital tool for the Sanskrit scholarly community. It complements the *Pandit* project from which it derives, opening new paths for engaging with the vast intellectual heritage of the Sanskrit tradition.

353 Acknowledgments

Pāņditya gratefully acknowledges Pandit as its source and follows it in adopting the Cre ative Commons BY-NC-SA 4.0 license, viewable at https://creativecommons.org/licenses/
 by-nc-sa/4.0/deed.en.

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