

¹ **textNet: Directed, Multiplex, Multimodal Event Network Extraction from Textual Data**

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⁶ Introduction

⁷ Network measurement in social science typically relies on data collected through surveys and interviews. Document-based measurement is automatable and scalable, providing opportunities for large scale or longitudinal research that are not possible through traditional methods. A ⁸ number of tools exist to generate networks based on co-occurrence of words within documents (such as the [Nocodefunctions](#) app ([Levallois et al., 2012](#)), the “[textnets](#)” package ([Bail, 2024](#)), ⁹ [InfraNodus](#) ([Paranyushkin, 2018](#)), and many more). But there is, to our knowledge, no open-¹⁰ source tool that generates network data based on the syntactic relationships between entities within a sentence. *textNet* allows a user to input one or more PDF documents and create arbitrarily complex directed, multiplex, and multimodal network graphs, enabling rich analysis of the relationships between verb attributes and tenses, entity types, structural motifs, and other network characteristics. For instance, [Zufall and Scott](#) demonstrate the use of *textNet* to identify which actors are involved in start-up versus ongoing management tasks, characterize patterns of information and funding flows, and compare the distribution of management tasks in networks from regions known to have contrasting characteristics ([2024](#)). *textNet* also works on arbitrarily long documents, making it well suited for research applications using long texts such as government planning documents, court proceedings, regulatory impact analyses, and environmental impact assessments.¹¹

¹² *textNet* has applications in governance network scholarship, as demonstrated by [Zufall and Scott \(2024\)](#) and by ongoing work on water resources governance at the UC Davis Center for ¹³ Environmental Policy and Behavior. Additional potential applications include legal scholarship, ¹⁴ social-ecological network analysis, government planning documents, court proceedings, archival ¹⁵ research, communication and media research, and other fields interested in exploring events ¹⁶ and entity relationships in textual data.¹⁷

³⁰ Statement of Need

³¹ Network extraction from documents has typically required manual coding. Furthermore, ³² existing network extraction methods that use co-occurrence leave a vast amount of data on ³³ the table, namely, the rich edge attribute data and directionality of each verb phrase defining ³⁴ the particular relationship between two entities, and the respective roles of the entity nodes ³⁵ involved in that verb phrase. We present an R package, *textNet*, designed to enable directed, ³⁶ multiplex, multimodal network extraction from text documents through syntactic dependency ³⁷ parsing, in a replicable, automated fashion for collections of arbitrarily long documents. The ³⁸ *textNet* package facilitates the automated analysis and comparison of many documents, based ³⁹ on their respective network characteristics. Its flexibility allows for any desired entity categories, ⁴⁰ such as organizations, geopolitical entities, dates, or custom-defined categories, to be preserved.

41 **Directed Graph Production**

42 As a syntax-based network extractor, *textNet* identifies source and target nodes. This produces
43 directed graphs that contain information about network flow. Methods based on identifying
44 co-occurring nodes in a document, by contrast, produce undirected graphs. Co-occurrence
45 graphs also have the tendency to generate saturated subgraphs, since every co-occurring
46 collection of entities has every possible edge drawn amongst them. By contrast, *textNet* draws
47 connections not between every entity in the document or even the sentence, but specifically
48 between pairs of entities that are mediated by an event relationship.

49 **Multiplex Graph Output**

50 Syntax-based measurement encodes edges based on subject-verb-object relationships. *textNet*
51 stores verb information as edge attributes, which allows the user to preserve arbitrarily complex
52 topological layers (of different types of relationships) or customize groupings of edge types to
53 simplify representation.

54 **Multimodal Graph Output**

55 Multimodal networks, or networks where there are multiple categories of nodes, have common
56 use cases such as social-ecological network analysis of configurations of actors and environmental
57 features. Existing packages such as the *manynet* package (Hollway, 2024) provide analytical
58 functions for multimodal network statistics. *textNet* provides a structure for tagging and
59 organizing arbitrarily complex node labeling schemes that can then be fed into packages for
60 multi-node network statistical analysis. Node labels can be automated (e.g., the default entity
61 type tags for an NLP engine such as *spaCy* (Honnibal et al., 2021)), customized using a
62 dictionary, or based on a hybrid scheme of default and custom labels. Any node type is possible
63 (e.g., species, places, people, concepts, etc.) so this can be adapted to domain-specific research
64 applications by applying dictionaries or using a custom NER model.

65 **Installation**

66 The stable version of this package can be installed from Github, using the *devtools* package
67 (Wickham et al., 2022):

68 `devtools::install_github("ucd-cepb/textnet")`

69 The *textNet* package suggests several convenience wrappers of packages such as *spacyr* (Benoit
70 et al., 2023), *pdftools* (Ooms, 2024), *igraph* (Csárdi et al., 2024), and *network* (Butts et al.,
71 2023). To use the full functionality of *textNet*, such as pre-processing tools and post-processing
72 analysis tools, we recommend installing these packages, which for *spacyr* requires integration
73 with Python. However, the user may wish to preprocess and parse data using their own NLP
74 engine, and skip directly to the *textnet_extract()* function, which does not depend on *spacyr*
75 or Python integration.

76 **Overview and Main Functions**

77 The package architecture relies on four sets of functions around core tasks:

- 78 ■ [OPTIONAL] Pre-processing: *pdf_clean()*, a wrapper for the *pdftools::pdf_text()* function
79 which includes a custom header/footer text removal feature; and *parse_text()*,
80 which is a wrapper for the *spacyr* package and uses the *spaCy* natural language pro-
81 cessing engine (Honnibal et al., 2021) to parse text and perform part of speech tagging,
82 dependency parsing, and named entity recognition (NER). Alternatively, the user can
83 skip this step and load parsed text directly into the package. Externally produced data
84 must be converted to standards outlined in the package manual.
- 85 ■ Network extraction: *textnet_extract()*, which generates a graph database from parsed text
86 based upon tags and dependency relations. The object returned from *textnet_extract()*

87 consists of a nodelist, an edgelist with a rich set of edge attributes, a verblist, and a list
 88 of potential coreferences for disambiguation.

- 89 ▪ Disambiguation: tools for cleaning, recoding, and aggregating node and edge attributes,
 90 such as the `find_acronyms()` function, which can be paired with the `disambiguation()`
 91 function to identify acronyms in the text and replace them with the full entity name.
- 92 ▪ Exploration: the `export_to_network()` function for exporting the graph database to
 93 `igraph` and `network` objects, `top_features()` for viewing node and edge attributes, and
 94 `combine_networks()` for aggregating multiple document-based graphs based on common
 95 nodes.

96 The figure below summarizes the functionality of `textNet` and the flow of function outputs.
 97 Optional data cleaning features are shown with dotted arrows.

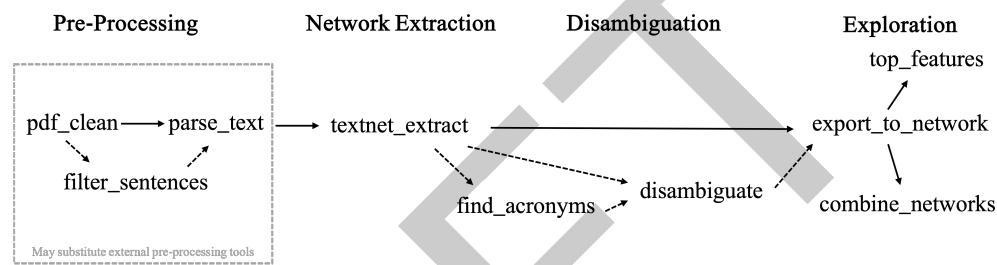


Figure 1: Workflow of `textNet` Functions

98 Potential Further Analyses

99 `textNet` is compatible with standard network analysis tools in R. Functionality from `ggraph`
 100 ([Pedersen & RStudio, 2024](#)), `sna` ([Butts, 2024](#)), `igraph` ([Csárdi et al., 2024](#)), `network` ([Butts
 101 et al., 2023](#)), and other network visualization and analysis packages can be used to further
 102 explore the extracted networks.

103 The `ggraph` package has been used to create the network visualization seen here, using a
 104 weighted version of a `igraph` constructed using the “old_new_parsed” sample data in `textNet`.

New Network



Figure 2: Representation of the Event Network of the New Plan

105 The network-level attributes output from `export_to_network` can also be analyzed against
 106 exogenous metadata that has been collected separately by the researcher regarding the different
 107 documents and their real-world context. The extracted networks can also be analyzed through
 108 a variety of tools, such as an Exponential Random Graph Model or a Temporal Exponential
 109 Random Graph Model.

110 **Vignette**

111 More information about the entity network extraction algorithm and an example start-to-finish
 112 data processing and analysis workflow can be found in the vignette for this package. The
 113 vignette uses sample data that travels with the `textNet` package.

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