

**I ILLINOIS**

School of Information Sciences

**Natural language processing  
to promote transparency  
of clinical publications**

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# Agenda

- Natural language processing for enhancing research transparency
- Recent work
  - Assessing clinical trial publications for reporting guideline adherence
  - Supporting meta-research investigations with natural language processing

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Kilicoglu, H. “Biomedical text mining for research rigor and integrity: tasks, challenges, directions.” *Briefings in Bioinformatics*, 2018. 19(6): 1400-1414.

# “Reproducibility Crisis”

- Causes
  - Poor experimental design and oversight
  - Publication bias for positive, statistically significant results
  - Novelty over reproducibility
  - “Publish or perish”
- Interventions
  - Standards and guidelines for reporting, data and code sharing
  - Cyberinfrastructure support for reproducibility

# Research Transparency

The reporting of experimental materials and methods in a manner that provides enough information for others to independently assess and/or reproduce experimental findings

Antin PB, Baldwin TO, Freeze HH, Haywood JR, Simon SI. Enhancing research reproducibility: recommendations from the Federation of American Societies for Experimental Biology. FASEB. 2016.

# Disentangling the Terminology

**Rigor**<sup>1</sup> Robust and unbiased experimental design, methodology, analysis, interpretation, and reporting of results

**Integrity**<sup>2</sup> Honest/verifiable methods, accurate reporting of results with adherence to rules, guidelines, and professional norms

<sup>1</sup> <https://grants.nih.gov/reproducibility/index.htm>

<sup>2</sup> [https://grants.nih.gov/grants/research\\_integrity/whatis.htm](https://grants.nih.gov/grants/research_integrity/whatis.htm)

# Disentangling the Terminology

- Reproducibility** Same methods and data as in the original research
- Replicability** Same methods as in the original research applied to newly collected data
- Translatability** Different experimental design and data to reach the same conclusion

National Academies of Sciences, Engineering, and Medicine. Reproducibility and replicability in science. National Academies Press; 2019.

# Biomedical Language Processing (bioNLP)

- Transform text into computable representations
  - Scientific publications, clinical notes, drug labels, etc.
  - Support knowledge discovery and clinical decision making
- Tools/techniques adapted from open-domain NLP
- Knowledge representation/ontologies, corpus annotation
- Tasks
  - Text classification, entity recognition, acronym/abbreviation resolution, relation extraction

# BioNLP and Research Transparency

- Textual artifacts are core to biomedical communication lifecycle
  - Manual analysis is time-consuming
- NLP provides scalability
  - Scrutinize reports of conducted research
  - Manage published literature more effectively to improve quality of proposed research
- It can complement efforts in standardization and guideline development

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- Natural language processing for enhancing research rigor, integrity, and transparency
- Recent work
  - Assessing clinical trial publications for reporting guideline adherence
  - Supporting meta-research investigations with natural language processing

Kilicoglu, H, Rosemlat G, Peng Z, Malički M, Schneider J, ter Riet, G. “Annotating Clinical Trial Publications to Assess CONSORT Adherence: A Feasibility Study.” *World Conference on Research Integrity (WCRI 2019)*.

# Study Summary

- Goal
  - Develop text-mining methods to automatically recognize the CONSORT checklist items in randomized controlled trial reports (RCTs)
- Approach
  - An annotation study
  - Comparison of baseline rule-based and weakly supervised machine learning methods

# Reporting Guidelines

- Promote transparent, complete and accurate reporting
- EQUATOR Network
  - CONSORT, ARRIVE, STROBE, PRISMA
- Improve reporting transparency
  - May be easier to reproduce
- Adherence remains inadequate

# CONSORT Statement

- **CON**solidated **S**tandards **O**f **R**eporting **T**rials
- Reporting guidelines for parallel group RCTs
- 25-item checklist and flow diagram
- Endorsed by over 600 journals
  - Lancet, BMJ, NEJM, etc.
- Extensions
  - Abstracts
  - Cluster randomized trials
  - Non-inferiority or equivalence trials

# CONSORT Checklist Examples

Checklist Item	Section	Example Sentence
Objective (2b)	Introduction	<i>We studied the effects of metformin in obese children aged 6–12 years who were believed to be at particular risk because they manifested a significant degree of insulin resistance.</i>
Allocation concealment (9)	Methods	<i>The pharmacy produced identical, sequentially numbered, randomly assigned boxes of study medication, containing either magnesium sulphate or placebo.</i>
Outcome results (17a)	Results	<i>No difference between bosentan and placebo treatments was observed in the time to healing of the cardinal ulcer (HR 0.91 (95% CI 0.61 to 1.35), <math>p=0.63</math>, figure 3).</i>
Limitations (20)	Discussion	<i>The main limitation of our trial is the small sample size of patients with bacteraemia, in whom results suggest an important advantage for vancomycin.</i>
Protocol access (24)	Other	<i>The trial protocol has been published previously.<sup>11</sup></i>

# Automating Adherence Assessment

- Text-mining techniques
  - Locate key statements for checklist items in a manuscript/publication
  - Give alerts in their absence
- Benefits for journal editors, peer reviewers, authors, systematic reviewers
- Commercial/academic software for some items
  - Penelope.ai, StatReviewer, RobotReviewer, ExaCT

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- Labeled data needed to train and evaluate text-mining tools

# Article Selection

- Cochrane RCT search strategy maximizing sensitivity and precision
  - Exclude meta-analyses, systematic reviews
  - 2011 to present
  - 11 journals (9 CONSORT-endorsing)
- 563 articles retrieved
- 50 articles sampled
  - PubMed Central XML download
  - Sentence splitting
  - Section extraction

# Annotation

- Sentence-level, multi-label annotation
  - 25 checklist items → 37 fine-grained categories
- 6 annotators
  - Experts in text mining/informatics, linguistics, meta-research, and clinical trials
- 50 articles annotated
  - 1 exploratory annotation
  - 30 double-annotated and adjudicated
  - 19 single-annotated and corrected

# Corpus Statistics

- 50 articles, 10779 sentences

	Total	Mean (Range)	Median (IQR)
Annotated sentences	4845	96.9 (61-158)	92.5 (80.0-109.8)
Annotations	5679	113.6 (66-197)	110.5 (93.8-126.5)
Items per article		27.5 (15-35)	28 (25-31)

# Corpus Statistics

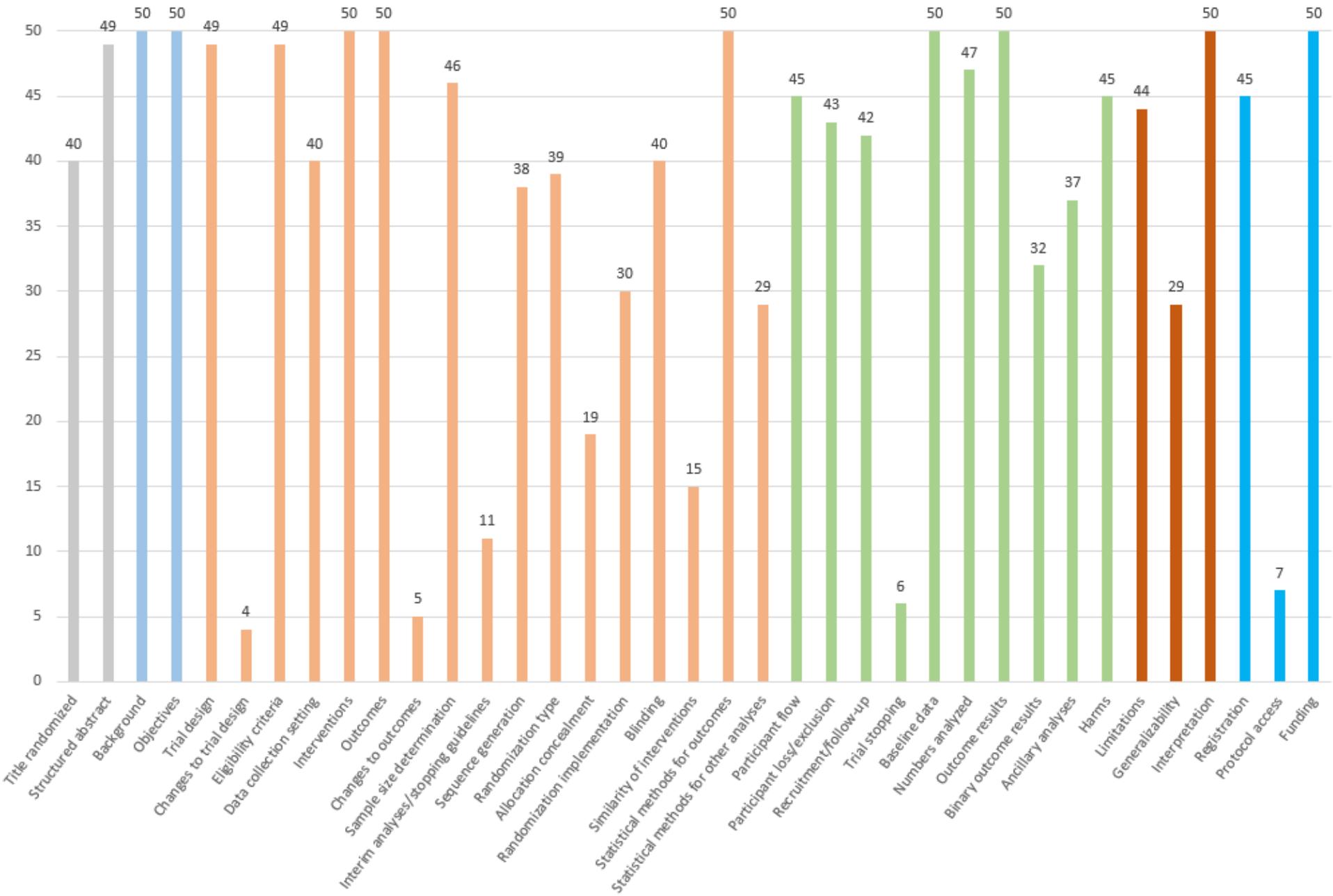
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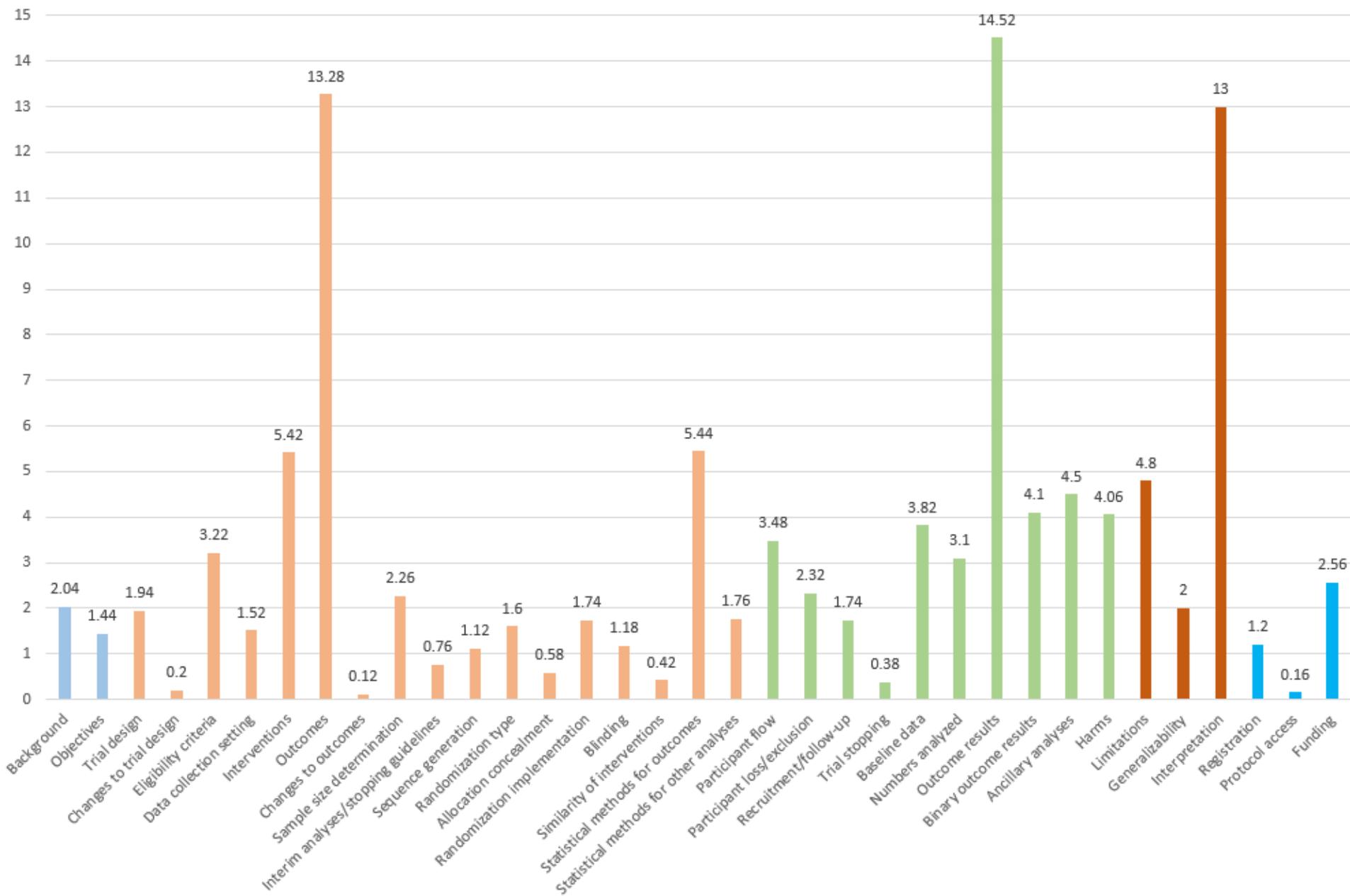
*Patients were randomly assigned, using a computer-generated randomization schedule, from a central location utilizing an interactive voice response system with blinded medication kit number allocation in a 2:1 ratio to identical-appearing tablets of HZT-501 (800mg ibuprofen and 26.6mg famotidine) or ibuprofen (800mg) thrice daily for 24 weeks.*

- Trial design, Sequence generation, Allocation concealment, Randomization implementation, Similarity of interventions

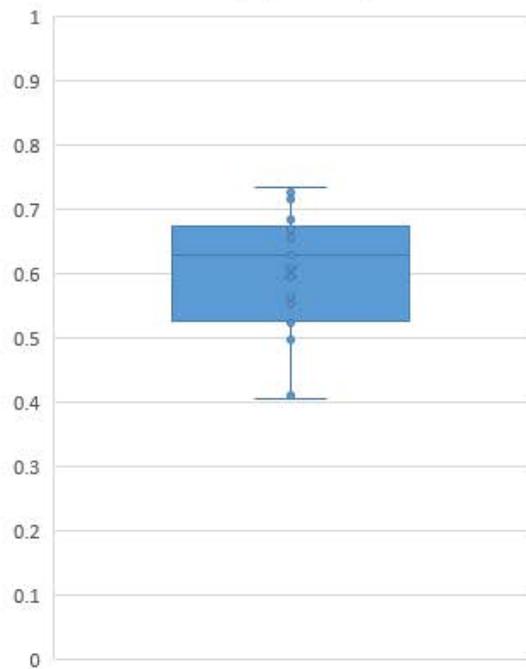
Number of articles with the CONSORT item



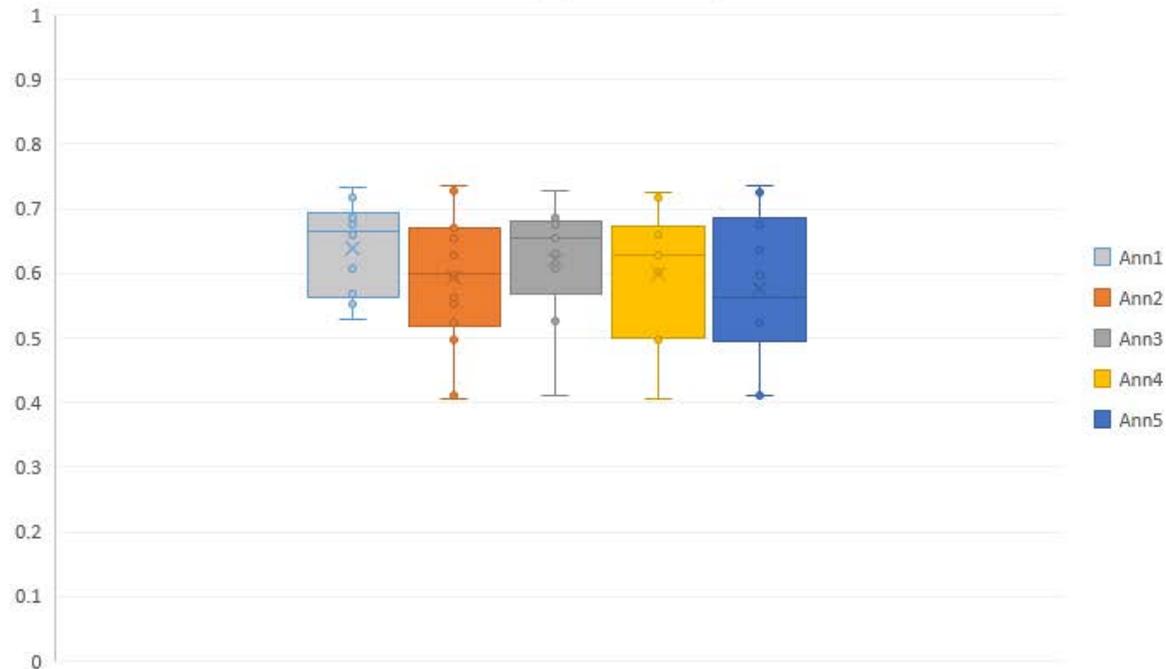
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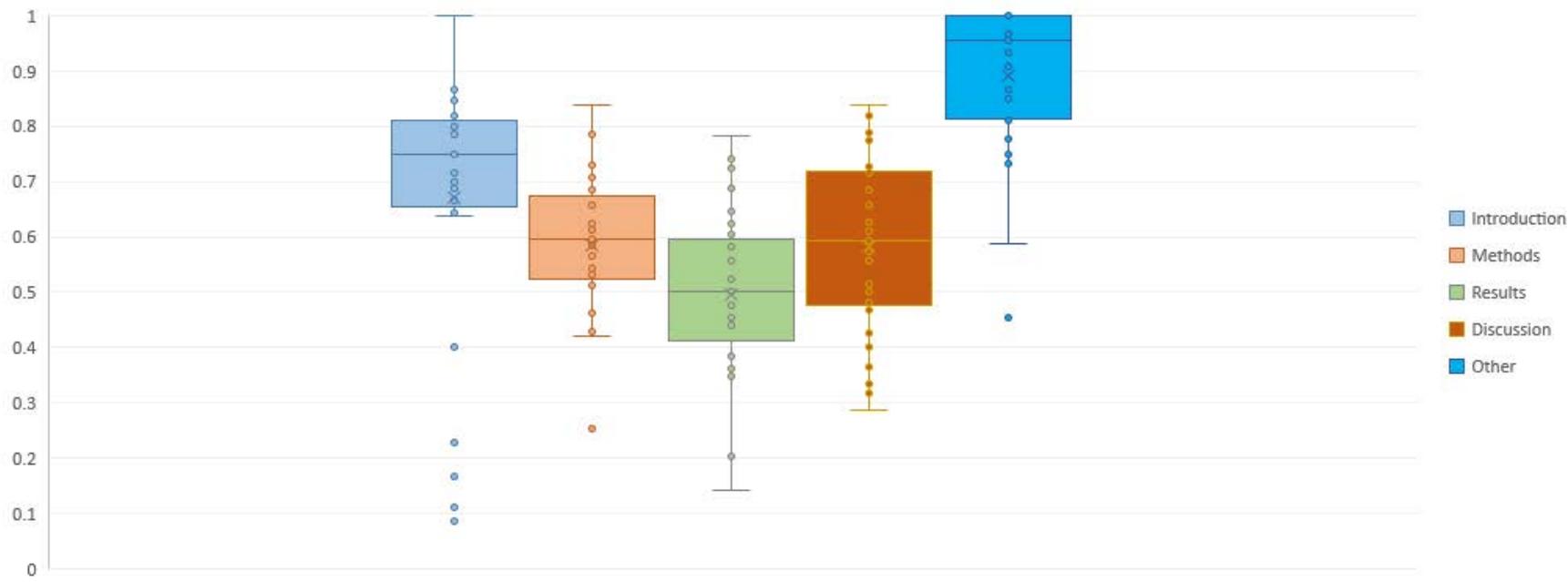
MASI (by article)



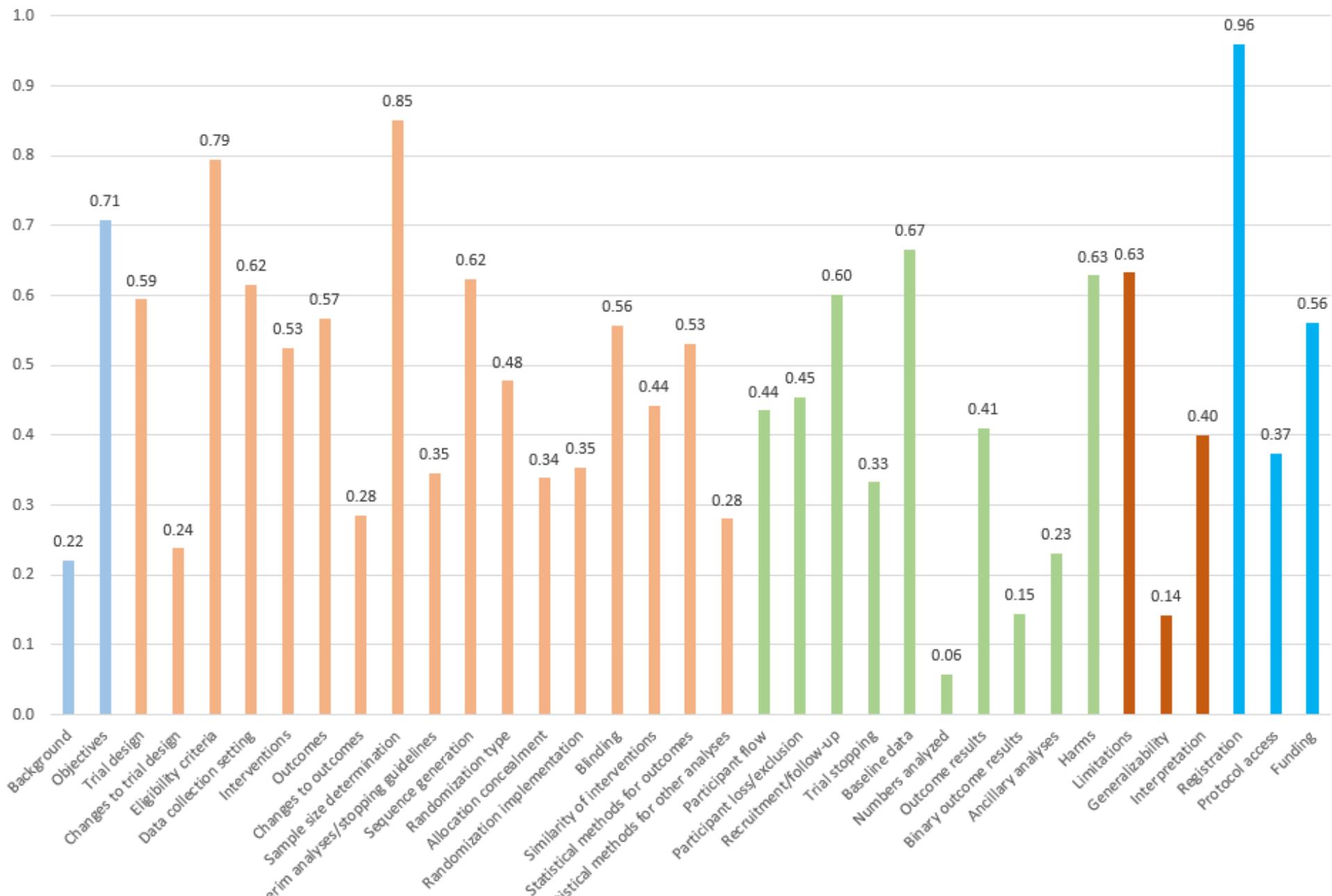
MASI (by annotator)



MASI (by section)



Inter-annotator agreement (Krippendorff's  $\alpha$ ) by CONSORT item



# Baseline Classification Experiments

- Applied to Methods sections and Methods-specific items
  - Trial design, Sample size determination, Blinding procedure, etc.
- Rule-based methods
  - Automatic analysis of frequent subsection headers and phrases
  - ~15K unlabeled clinical trial publications
  - Phrase-based classification
    - “*masked to treatment*” → Blinding procedure
  - Subsection header-based classification
    - “*change*” ... “*plan*” → Changes to trial design

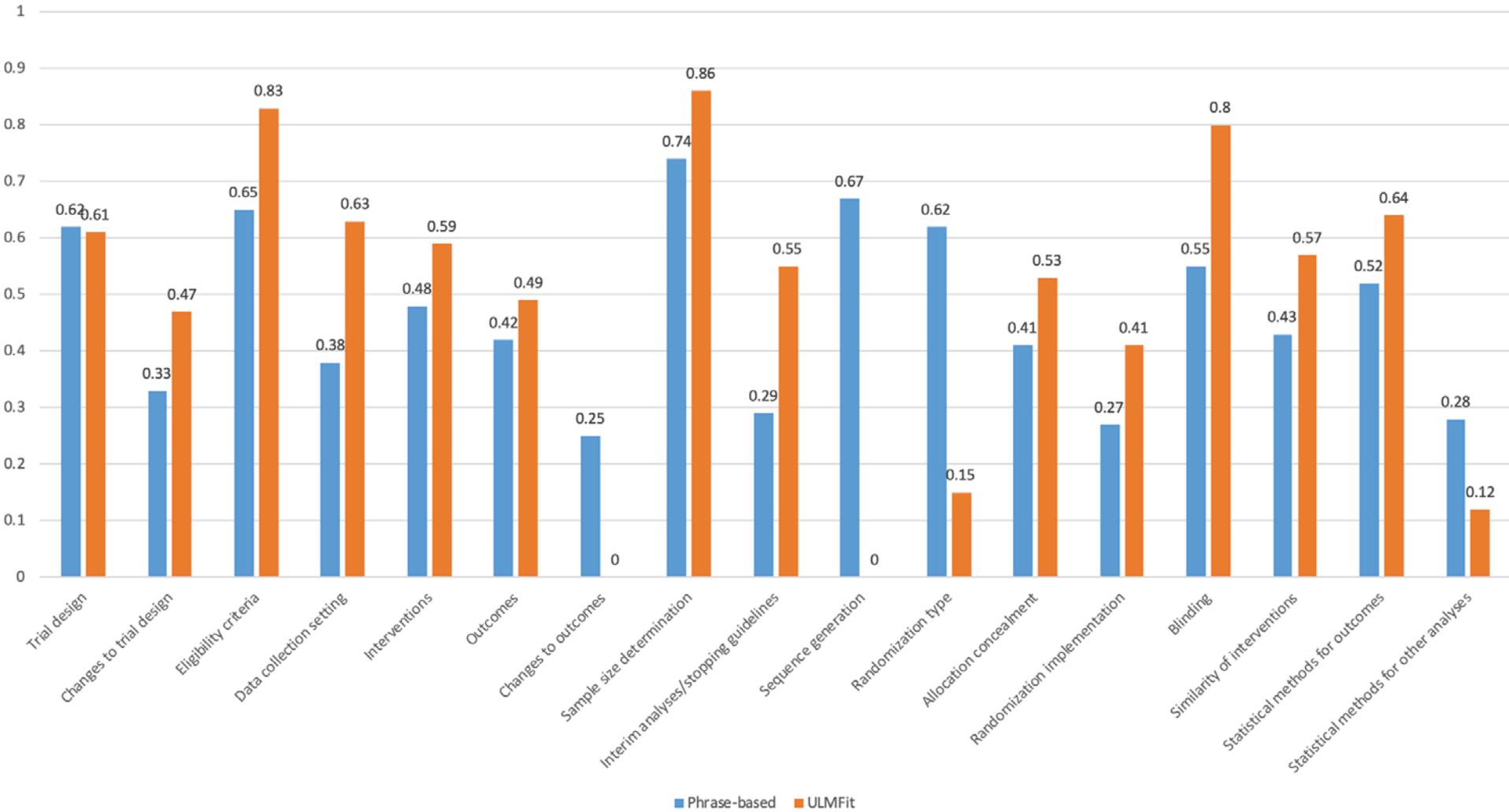
# Baseline Classification Experiments

- ULMFit (Universal Language Model Fine-Tuning)
  - Model pretrained on a large corpus, Wikitext-103 (103M tokens)
  - Fine-tuned on the task corpus
- Training data (with some noise) automatically generated
  - Phrase-based classification
  - Subsection header-based classification
  - RobotReviewer predictions (for Eligibility criteria, Interventions, Outcomes)
  - Nearest neighbor based on sentence vector similarity
- Validation and testing with the manually annotated dataset

# Preliminary Evaluation Results

- Checklist item-level evaluation
  - Macro-precision (p), macro-recall (r), macro-F1 (f)
  - Phrase-based (p: 0.54, r: 0.48, f: 0.46)
  - ULMFit (p: 0.51, r: 0.54, f: 0.49)
- Article-level evaluation
  - CONSORT item present in the article or not?
  - Phrase-based (p: 0.88, r: 0.80, f: 0.84)
  - ULMFit (p:0.87, r: 0.83, f: 0.85)

## Comparison of Baseline Methods: F<sub>1</sub> score



# Conclusion

- Cognitively challenging annotation task
  - Large number of fine-grained categories (37)
- Inter-annotator agreement varied significantly for items ( $\alpha$  range: 0.06-0.96)
  - Broad (Interpretation)
  - Similar (Outcome result, Binary outcome result, Ancillary analyses)
- The manually annotated corpus can be used as a benchmark
- Simple phrase-based method yields moderate results
- ULMFit has similar performance

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# Peer Review and Limitation Reporting

- *Spin*: reporting practices that distort the interpretation of the results of a study [Chiu et al., 2017]
  - Failure to acknowledge the limitations of the study
  - Inappropriate overstatement of claims
- Hypotheses
  - Compared to subsequent publications, discussion sections of submitted manuscripts
    - Discuss fewer limitations
    - Make stronger claims

Keserlioglu K, Kilicoglu H, ter Riet, G. “Impact of peer review on discussion of study limitations and strength of claims in randomized controlled trial reports: a before and after study.” *Research Integrity and Peer Review*, 2019(4):19.

# Approach

- NLP methods applied to discussion sections of manuscript/publication pairs
  - Recognize limitation sentences (91.5% accuracy) [Kilicoglu et al., 2018]
  - Identify speculative sentences (“hedging”) and measure their level of speculative-ness (“normalized hedging score”) (93% accuracy) [Kilicoglu and Bergler, 2009]
- 446 RCT reports from BMJ Open and 27 BMC journals
  - Open peer review

# Results

- Limitation reporting increases by 56% in peer review
  - But the difference is small in absolute terms (2.48 vs. 3.87 sentences)
  - Greater impact on manuscripts with zero or low number of limitation sentences
- No support for the hypothesis that the peer review leads to increased hedging of claims
  - Authors are asked to both temper their statements and hedge less, resulting in minimal changes

# Translatability of Animal Studies

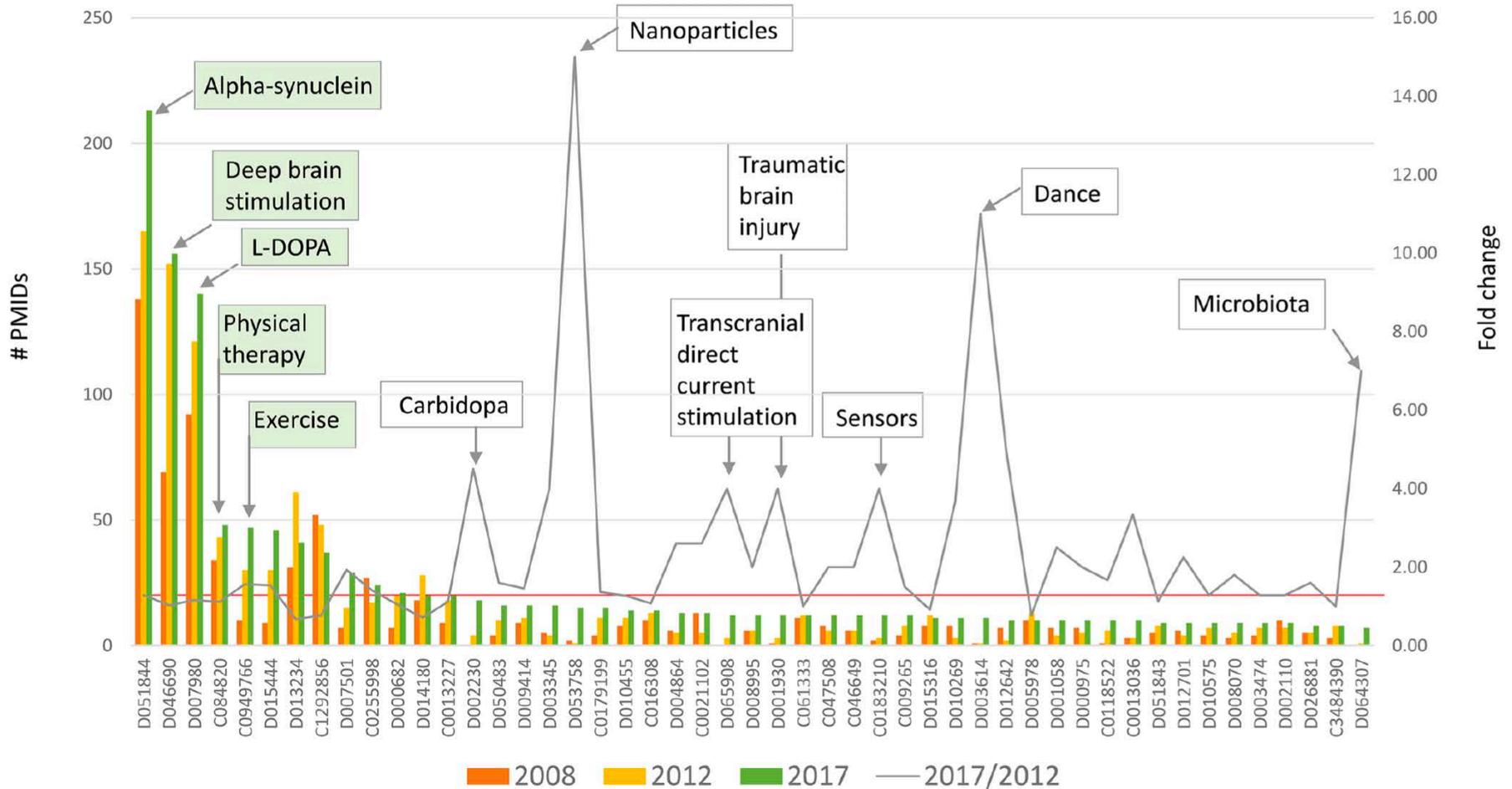
- No disease-altering therapies for the most common neurodegenerative diseases are available
  - Promising results in animal models for the same conditions
- Assess therapeutic approaches in the context of
  - Species, models, molecular targets, outcomes
  - Large-scale analysis to reveal patterns of animal use

Zeiss CJ, Shin D, Vander Wyk B, Beck AP, Zatz N, Sneiderman CA, Kilicoglu H. “Menagerie: A text-mining tool to support animal-human translation in neurodegeneration research.”  
*Under review.*

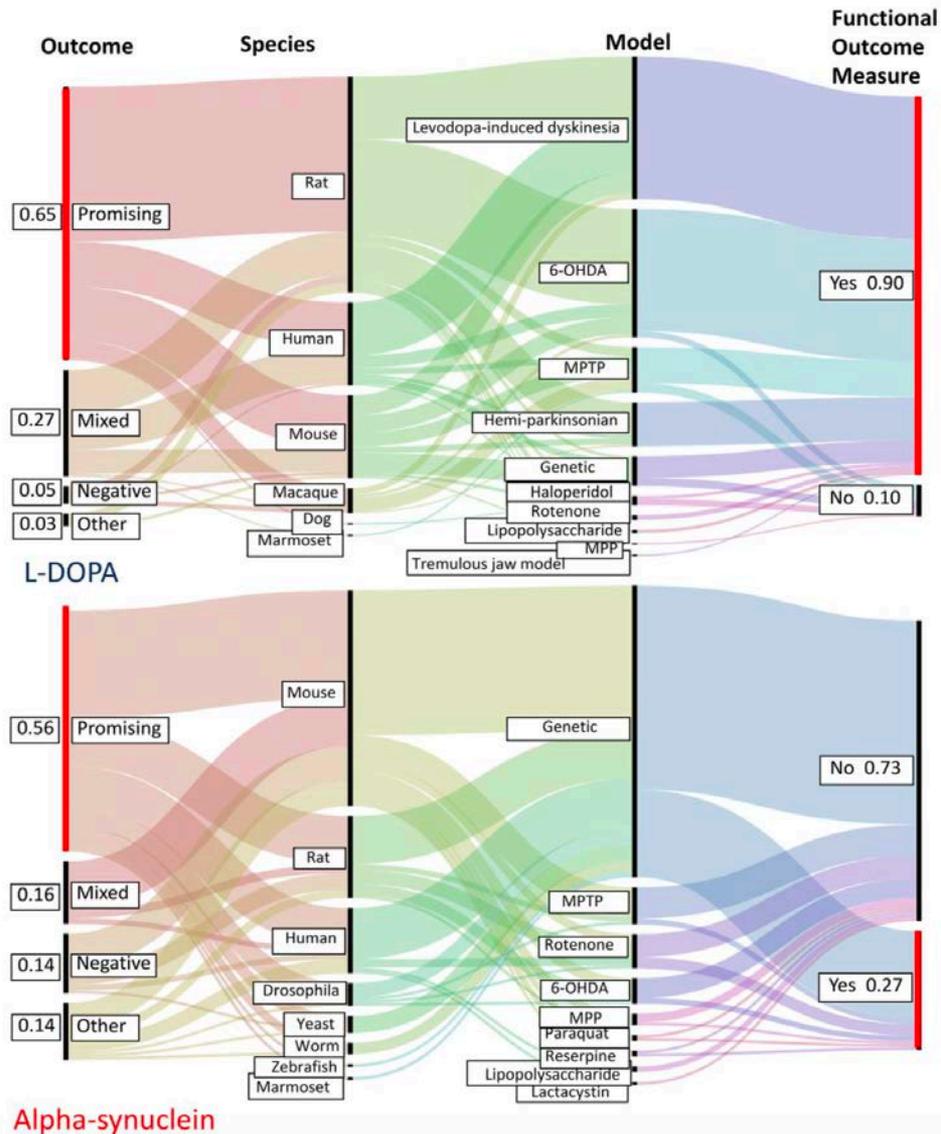
# Case Study: Parkinson's Disease

- Text mining approach to extract
  - Interventions, species, models, molecular targets, study outcome, functional outcome measures
  - Off-the-shelf NLP tools augmented with rules, term lists, and supervised machine learning
- Annotation and intrinsic evaluation on 504 abstracts
- Large-scale trend analysis on 3-year data
  - ~15K abstracts (2008, 2012, 2017)
  - e.g., Species=MOUSE and Model=GENETICALLY ALTERED MODEL and StudyOutcome=PROMISING

# Therapeutic Intervention Trends



# Integrating Information from Modules



# Concluding Remarks

- NLP/text mining can aid in assessing and improving research practices
- But many challenges lie ahead
  - Availability of full-text articles and other textual artifacts
    - Restrictions on text mining
  - Annotated corpora
    - Difficulty of manual annotation
  - Information in modalities other than narrative text
    - Tables, figures, supplementary data
  - Accuracy of methods