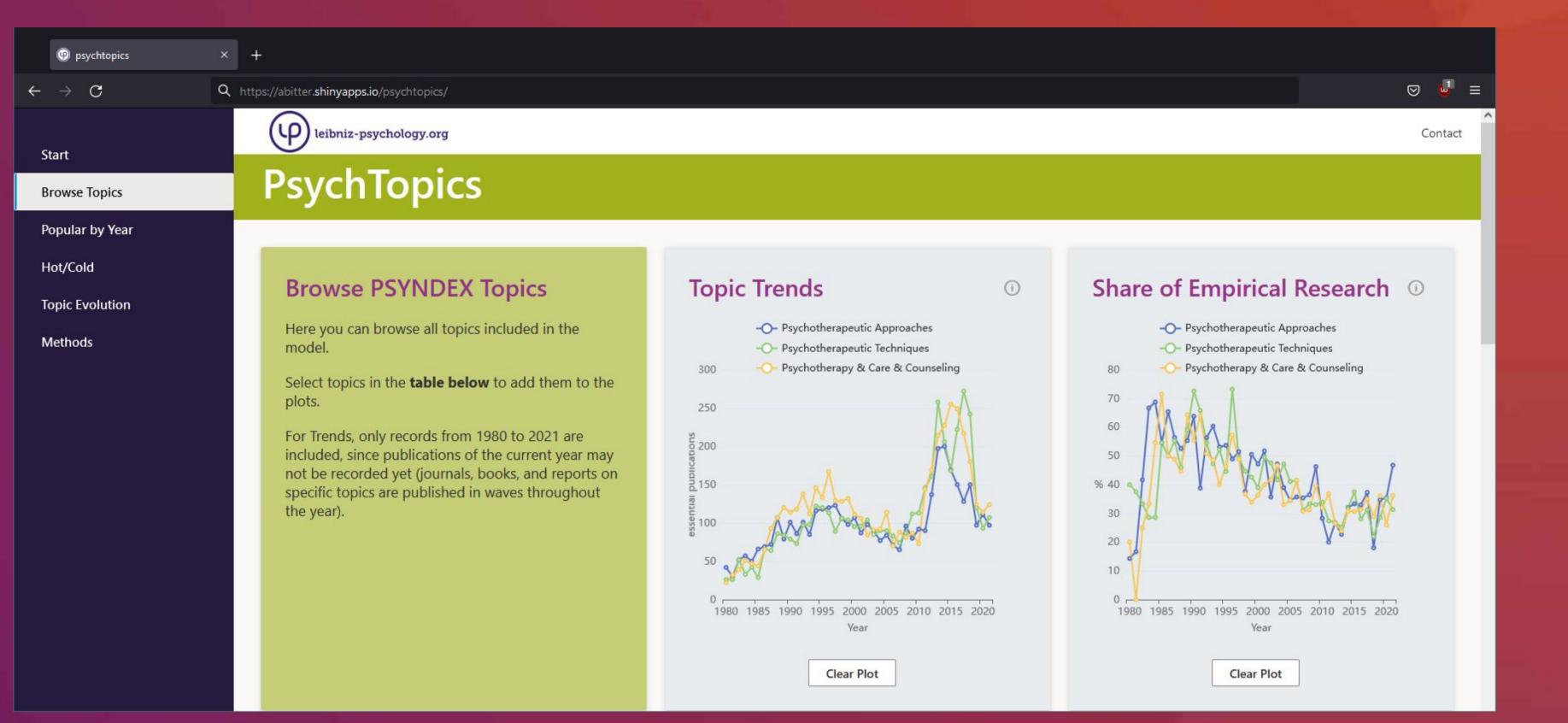
Scientific topics can be monitored with RollingLDA and R Shiny



How to keep up with the vast amount of scientific information published every day?

We present a framework and app for continuous topic detection

Finding Scientific Topics in Continuously Growing Text Corpora

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Study Design

- We compare the evolution of 42 RollingLDA model variants to a single IdaPrototype reference model (for 2020).
- The best fitting model is determined using cosine similarity to the reference model, topic quality metrics, and external validation:

Start	K	Similarity*	Coherence	Exclusivity	Correlation**	Mean (of <i>z</i> -scores)
2010	200	0.623898	-123.997870	4.137017	0.960064	0.188 719
2005	200	0.621397	-123.516668	3.949559	0.962599	-0.054622
1995	200	0.621219	-123.226158	3.881941	0.966658	0.176869
2010	300	0.621108	-123.386484	4.320748	0.946135	-0.008355
2015	200	0.620810	-123.740794	4.410456	0.944504	-0.302611

in scientific publications.

We employ RollingLDA, a topic modeling variant designed for "living" text corpora.

Our open source Shiny App can be modified to your needs.

Table 1: Comparison of RollingLDA model variants. The reference model for 2020 (cf. Sect. 3.3.1) comprised 250 topics. The best fitting model variant is printed in bold. Notes: *mean cosine similarity to the topics of the reference model. **correlations between actual classification category frequencies and classification shares in the topics (external validation).

Results

- Using RollingLDA with annual updates of the corpus, 82% of the reference model's topics could be detected.
- Missed topics were either of low prevalence in the reference model (13.6%) or included in other topics (4.4%).

Conclusion

- The model integrates new publications while keeping time series of topic trends consistent.
- This can help various stakeholders like researchers or policy-makers to evaluate how research fields evolve over time.
- The presented framework has a high degree of automation once the initial model is created.



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abitter.shinyapps.io/psychtopics/

Find the source code:



github.com/leibniz-psychology/psychtopics

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