

CART: A Tool for Making Paper Relevancy Screening Easier

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Summary

CART (Culling Abstracts for Relevancy in Teams) provides a way for a team of individuals to complete the first step common to most systematization of knowledge (SoK) projects¹: culling, in teams, a large set of research papers (i.e., titles and abstracts) based on per-project relevancy criteria. Existing software solutions either do not fit this need in a usable way—leading to error-prone voting—or do not offer the necessary functionality to work on the problem effectively.

CART provides both functionality and usability: *Functionally*, CART accommodates teams of any size (i.e., accommodating race conditions), review requirements of any number (e.g., mandating that each paper is reviewed by n team members), and dynamic participation rates (e.g., team members may participate in as little or as much reviewing as they like). *Usability-wise*, CART provides a gamified user interface, making the process of reviewing papers easier, and therefore more accurate and efficient.

Statement of Need

A systematization of knowledge or systematic review paper is a popular type of paper found across multiple disciplines: medicine, computer science, sociology, psychology, economics, political science, marketing, and genetics. This type of paper most commonly involves a two-stage process. In the first stage, researchers gather a large set of papers—easily over 10,000—and then manually filter those papers (based on paper titles and abstracts) according to predefined relevancy criteria ([Franz et al., 2021](#); [Reitering & Mazurek, 2023](#); [Singhal et al., 2023](#); [Thomas et al., 2021](#)). For example, imagine a researcher who wanted to learn about legally-sufficient data sanitization tools ([Bellovin et al., 2019](#)). A paper on the prevalence of a particular web tracking technique like canvas fingerprinting ([Reitering & Mazurek, 2021](#)) would be marked as *irrelevant*, but a paper on statutory differential privacy requirements ([Reitering & Deshpande, 2023](#)) would be marked as *relevant*. After this first step is complete—often resulting in a set of around 100 papers—most researchers then move on to analyzing papers in-depth, the second stage, using qualitative coding techniques well-fit by existing software.

Surprisingly, software to support this first step of paper culling is lacking. Existing tools are either too generalist (i.e., resulting in ineffective paper reviewing) or too fine-tuned (i.e., resulting in tradeoffs impacting functionality). Generalist tools like Microsoft Excel or Google Sheets could allow teams of researchers to check the relevancy of snippets of text, but these tools can be clunky (i.e., given a scale of more than 10,000 papers in consideration) and not usable (i.e., reviewing hundreds of rows in a spreadsheet is error-prone and fatiguing). Other tools, like Atlas.ti ([Smit & Scherman, 2021](#)) or maxQDA ([Franzosi et al., 2013](#)), could allow teams of researchers to “code” for relevancy, but do not naturally accommodate dynamic

¹These projects are also called systematic reviews.

40 participation (e.g., team members might not be able to review the same amount of papers) or
41 variable review rates (e.g., sometimes you want all team members to review all papers, and
42 sometimes you want at least two, but you do not set who those team members are). Enter
43 CART.

44 CART

45 CART provides a way for teams to accomplish the goal of culling a large set of research papers.
46 Team members may contribute as little or as much as they want, the experience of reviewing
47 papers is gamified, and CART prioritizes protection against data loss, error, and data corruption.
48 Most importantly, CART is fully functional without requiring server space (although this is
49 certainly possible with an appropriately provisioned environment) and easily tunable for a
50 project's specific needs.

51 CART utilizes Flask, ngrok, and a web interface to allow one team member to create a server
52 that other team members can access. As shown in [Figure 1](#), CART has two primary functions.

Architecture of CART.

Figure 1: Architecture of CART.

53 *First*, CART runs a Flask application responsible for serving and updating papers (i.e., represented
54 as an abstract, title, URL, and a few metadata fields). A team member will only be served
55 an abstract if: (1) the paper is not one that the team member has seen before; (2) the
56 paper has not yet met the required amount of votes per paper; (3) the paper has not been
57 flagged as “not” a research paper (i.e., a proceedings introduction, extended abstract, or
58 similar document); and (4) the paper is not currently being viewed by another user. After a
59 vote is cast, CART will update a log noting who voted, what their vote was, and how many
60 votes have been cast for the paper. This process of serving and updating papers will loop until
61 there are no more eligible papers available.

Web interface tabs: home, progress, history, information, and account.

Figure 2: Web interface tabs: home, progress, history, information, and account.

62 *Second*, CART creates a gamified user experience for reviewing papers. The web interface makes
63 voting on a paper simple (i.e., button click or keyboard press), allows users to set personal
64 goals, and rewards (with confetti) users on every 50th (changeable) vote submitted. The web
65 interface also includes tabs ([Figure 2](#)) for **voting** on papers (home), viewing the team's **progress**
66 (bar chart, total vote counts), viewing personal voting **history** (last 50 edits, permitting a
67 change-of-vote), logging information **about** the project (collaborative editing with backups,
68 where users may write guidelines for voting), and viewing **account** details (logging in and out).
69 Moreover, the web app is optimized to provide information in a timely fashion (e.g., storing
70 ‘open’ papers in logs rather than reviewing the full corpus of possible papers to find open
71 papers) to enable speedy reviewing.

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