

Continuous voting by approval and participation

Renato Fabbri *

* IFSC/USP, Participa.br, labMacambira.sf.net

Draft

In finding the adequate way to prioritize proposals, the Brazilian participation community agreed about two indexes of approval and of participation. Both practice and literature is constantly handled by the experts involved, and the formalization of such model and metrics seems novel. Also, the relevance of this report is strengthened by the nearby use of these indexes by the Brazilian General Secretariat of the Republic to raise and prioritize proposals about public health care in open processes.

social participation | recommendation systems | online voting | statistics

Online decision making is a kind of recommendation system with special appeal for online social participation and electronic governments. This poses challenges on the design of such processes regarding validity, security and the adequate indicators. Indeed, the processes themselves vary, and the fact that the indexes presented here seem not to be formalized and published is an evidence that such online decision making is very recent phenomena.

The main contribution of this report is a modeling for an online voting process [1, 2, 3] with the following characteristics:

- proposals might be inserted by voters after the voting phase started.
- Voting might be extended as a permanent process. In other words, voting on and adding new proposals might be open continuously.
- A proposal is presented to a voter one by one as random outcomes of all proposals.
- Each vote might be of one and only type among: “approve”, “disapprove” and “indifferent”.
- Voters vote without authentication.
- Intended mostly for national rankings, but can also be local or have foreign participation.
- Should result in a ranking of proposals to assist public management.

This setting requires care about security and validity. Some of which are:

- adequate sampling of individual proposals and overall ranking.
- Registration of the IP address and time of votes to ease detection of automated and other fraudulent efforts.
- Reasonable use of the outcomes from the voting process. This requires probing the survey being conducted and its purposes. The indexes here presented target indicatives for the Brazilian federal government about the most important health care proposals. Given the unauthenticated voting, the outcomes might be regarded as reference rankings if data is minimally shared and checked for inadequate data entry (such as voting by automated scripts or a persistent participant introduced bias).

Approval and participation indexes. The approval index α_i and the participation index γ_i of the proposal i was defined as:

$$\begin{aligned}\alpha_i &= \frac{v_i^+ - v_i^-}{\eta_i} \\ \gamma_i &= \frac{v_i^+ + v_i^-}{\eta_i}\end{aligned}\quad [1]$$

where v_i^+ , v_i^- and η_i are approval count, disapproval count and exhibition count, respectively. Note that $\alpha_i \in [-1, 1]$, $\gamma_i \in [0, 1]$, and $v_i^o = \eta_i - v_i^+ - v_i^-$ is the count of the “indifferent” manifestations received by proposal i . Also, such α_i and γ_i indexes are expected, for each proposal i , to be a constant plus a sampling estimate error that should be smaller as η_i raises. This error is thought to be acceptable if η_i is above a threshold $\bar{\eta}$ established by the participation community and public managers. As an initial decision, the staff agreed to use $\bar{\eta}$ as to select 10 – 20% of all proposals. A threshold $\bar{\gamma}$ can be used as a required level of engagement for proposals to be relevant, while the threshold $\bar{\alpha}$ is used to classify the outcome as “approved”, “disapproved” and “clash”. More specifically:

$$\begin{aligned}\eta_i > \bar{\eta} &\Rightarrow i \text{ is sampled} \\ \gamma_i > \bar{\gamma} &\Rightarrow i \text{ is relevant} \\ |\alpha_i| \leq \bar{\alpha} &\Rightarrow i \text{ is a clash} \\ \alpha_i > \bar{\alpha} &\Rightarrow i \text{ is approved} \\ -\alpha_i > \bar{\alpha} &\Rightarrow i \text{ is disapproved}\end{aligned}\quad [2]$$

If a proposal is both sampled and relevant, than it is prioritized. The coherent values of $\bar{\alpha}_i = 0.5$ (or $1/3$) and $\bar{\gamma}_i = 0.5$ were chosen as standards of the decision model. These are likely to change with implementation and management.

Selected decision framework examples. Many of the online decision processes conceived and practiced resemble our model and have similar measurements to the α_i and γ_i indexes. This section presents a collection of models more familiar to the Brazilian participatory community, with focus on the mechanisms, not on historical notes.

Pairwise [4] is part of the tackled paradigm: the ranking procedure accepts new proposals while the voting occurs. Even so, pairwise voting is comparative, voter chooses between two proposals at each vote, and this does not fit proposed procedure.

Appgree software [5] ranks proposals by sampling voters in cycles, each with fewer proposals. This is adequate for a range of decision making cases and showcases statistical estimates utility. The system has a separate proposition phase, and relies on an organized group engagement and user identities, which also does not fit current needs.

Liquid Feedback [6] is a very renowned and bleeding edge solution for collective decision making. It relies on delegating your voting count on specific subjects to other people you know or trust. Therefore, it does not fit current needs. Even so, this framework have precious considerations for our case, such as about ranking and presenting proposals to voters in the most useful ways.

A Brazilian solution, used in diverse software and specially important as the output of a nation-wide decision making need, is the Agora Algorithm [7]. It presents a decision procedure in phases (agenda proposition, deliberations proposition and commenting, voting) with resolution outcomes. Although coherent, this framework requires authentication and might need experimentation and tuning in order to be effective with more than dozens or a few hundreds of participants.

There is a number of other solutions for online collaborative prioritization, such as IdeaScale, Kidling, or any flavor of

an Analytic Hierarchy Process (AHP). Authors hope to better formalize possible solutions (and found implementations), maybe through recommender systems theory [8].

Discussion

The above estimates are the best fit the researchers could deliver, suitable for current needs and not found (yet) in literature. The following questions should be answered in near future:

- Are there more adequate metrics for ranking proposals in the given setting?
- What are strong and weak aspects of the approach for collective recommendation?
- What thresholds will be the choice of community and will they be adjusted in time?
- Are there really no previous formalized model of this setting? If there is, what comparisons can we make on design, metrics and outcomes?
- To which extent will participation community and public managers legitimize this approach?
- What is the impact of this technological approach in public health care, social participation and the scientific community?

Most importantly, this report is being delivered to the civil society and scientific community for consideration. Given the large number of possibilities for the collective ranking procedure, and the proliferation of solutions, research efforts might aim the organization of such procedures.

ACKNOWLEDGMENTS. Author is grateful to CNPq (process 140860/2013-4, project 870336/1997-5), UNDP (contract 2013/00056, project BRA/12/018), SNAS/SGPR, and the Postgraduate Committee of the IFSC/USP.

1. Issue #19 for implementing prioritization algorithm. <https://gitlab.com/participa/proposal-app/issues/19>
2. Issue #29 for implementing prioritization algorithm. <https://gitlab.com/participa/proposal-app/issues/29>
3. Table developed for discussions about approval and participation indexes. Ronald Costa. https://docs.google.com/spreadsheets/d/1Q2uXcuKY0uL0nJdpjXoUaQhjmj6mNv2St7F-pBi_31w
4. Wiki surveys: Open and quantifiable social data collection, Salganik, Matthew J and Levy, Karen EC, arXiv preprint arXiv:1202.0500, <http://arxiv.org/abs/1202.0500>
5. Appgree software for collective ranking of proposals. <http://www.appgree.com>

6. Through Liquid Democracy to Sustainable Non-Bureaucratic Government, Paulin, Alois, CeDEM 2014, p. 205–217, 2014, <http://research.apaulin.com/research/2014/cedem14-ld-snbg>
7. Agora Algorithm, Brazilian social participation community, <https://vimeo.com/39135943>
8. Konstan, Joseph A., et al. "Teaching recommender systems at large scale: evaluation and lessons learned from a hybrid MOOC." Proceedings of the first ACM conference on Learning@ scale conference. ACM, 2014. <http://dl.acm.org/citation.cfm?id=2566244>