

TRUTH-TRACKING VIA APPROVAL VOTING: SIZE MATTERS

MEETING IN DEAUVILLE

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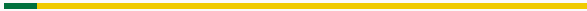
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LAMSADE - PSL University

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INTRODUCTION



EPISTEMIC SOCIAL CHOICE - TRUTH TRACKING

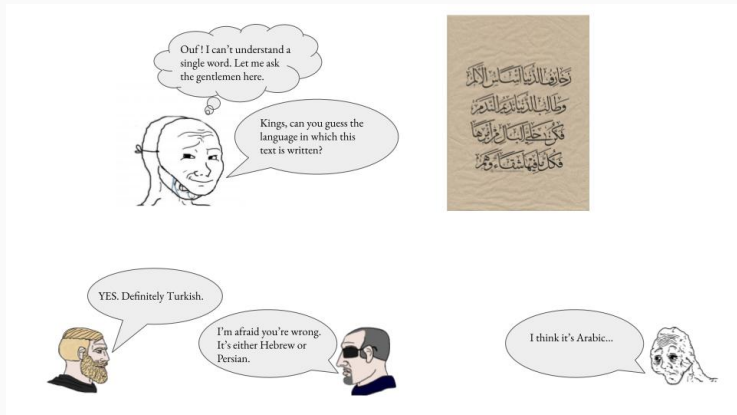


Figure 1: Real World Application of Epistemic Social Choice

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- A profile of n approval ballots $A_i \subseteq X$.

(+) Noise model: probability distribution over the set of possible approval ballots.

\implies : **Estimate** the ground truth given the approval ballots by **Maximum Likelihood Estimation**.

TRUTH-TRACKING AND BALLOTS' SIZES



When voters answer truthfully:

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- A voter who knows the correct answer would select a single alternative.

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⇒ : **Smaller** ballots are of a better quality (are **more accurate**).

⇒ : Assign **more weight** to **smaller** ballots: *size-decreasing* rules.

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- There exist n parameters $\phi_i \in]0, +\infty[$ (**reliability**) and a function $d : X \times \mathcal{P}(X) \mapsto \mathbb{R}$ (**distance to the ground truth**) such that:

$$P_{\phi_i, d}(A_i | a^* = a) = \frac{1}{\beta_i} \phi_i^{d(a^*, A_i)}, \forall a \in X$$

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We seek to estimate the ground truth by *Maximum Likelihood Estimation*:

$$\hat{a} = \arg \max_{a \in X} P_d(A_1, \dots, A_n | a^* = a)$$

Given the noise expression:

$$P_{\phi_i, d}(A_i | a^* = a) = \frac{1}{\beta_i} \phi_i^{d(a^*, A_i)}, \forall a \in X$$

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Remark

The noise is neutral (invariant by permutation of alternatives) if and only if the distance $d(a^, A)$ only depends on:*

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- $|A| \in \{0, \dots, m\}$: *The size of the ballot.*

Example

Hamming distance:

$$d_H(a, A) = |\bar{a} \cap A| + |a \cap \bar{A}| = 1 + |A| - 2|a \cap A| = \psi_d(|a \cap A|, |A|)$$

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Theorem

For $n \geq 3$, the maximum likelihood estimation rule ζ_d is a size-decreasing approval rule if and only if:

$\Delta\psi_d : j \mapsto \psi_d(0, j) - \psi_d(1, j)$ is decreasing

Example

Consider the Jaccard distance:

$$d_J(a, A) = \frac{|\bar{a} \cap A| + |a \cap \bar{A}|}{|\bar{a} \cap A| + |a \cap \bar{A}| + |a \cap A|}$$

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This yields:

$$\Delta\psi_{d_J}(j) = \psi_{d_J}(0, j) - \psi_{d_J}(1, j) = 1/j$$

ζ_{d_J} is a size-decreasing approval rule with **weights** $w_j = 1/j$.

Example

- The Dice distance:

$$d_D(a, A) = 1 - \frac{2|a \cap A|}{|\bar{a} \cap A| + |a \cap \bar{A}| + 2|a \cap A|}$$

yields a size-decreasing approval rule with **weights** $w_j = 2/j+1$.

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- Any strictly concave transformation g of the Hamming distance:

$$d_g(a, A) = g(|\bar{a} \cap A| + |a \cap \bar{A}|)$$

yields a size approval rule with **weights** $w_j = g(j+1) - g(j-1)$.

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This implies that $\Delta\psi_{d_H}(j) = 2$ is not decreasing !

More generally, any distance which is **separable**:

$$\psi_d(|a \cap A|, |A|) = f(|a \cap A|) + g(|A|)$$

does not yield a size-decreasing rule !

HETEROGENEOUS RELIABILITY

Now each voter has her own reliability parameter:

$$P_{\phi_i, d}(A_i | a^* = a) = \frac{1}{\beta_i} \phi_i^{d(a^*, A_i)}$$

such that:

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Example

How does the Hamming distance decompose into two separable terms ?

$$d_H(a, A) = |\bar{a} \cap A| + |a \cap \bar{A}| \implies \psi_{d_H}(|a \cap A|, |A|) = \underbrace{1 + |A|}_{g(|A|)} \overbrace{-2|a \cap A|}^{f(|a \cap A|)}$$

Theorem

If for every $1 \leq t < k \leq m - 1$ we have that:

$$g(k) - g(t) \geq \frac{k-t}{2} [f(1) - f(0)]$$

Then:

$$\frac{\partial \mathbb{E}_{\phi, d}[|A_i|]}{\partial \phi} \geq 0$$

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Good news: This sufficient condition holds for the Hamming distance (and for many other distances also). But we can still go further !

HETEROGENEOUS RELIABILITY: CONDORCET NOISE

The Mallows noise with Hamming is equivalent to the Condorcet noise:

$$P_{p_i}(a \in A_i | a = a^*) = P_{p_i}(a \notin A_i | a \neq a^*) = p_i = \frac{\phi_i}{1 + \phi_i}, \forall a \in X$$

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For $m \geq 2$, we have that:

$$\mathbb{E}_p[|A_i|] = (m - 1) - (m - 2)p$$

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Why is this interesting?

$$p_i = \frac{m - 1 - \mathbb{E}_{p_i}[|A_i|]}{m - 2}$$

EXPERIMENTS



We conducted experiments on data collected in [Shah et al., 2015]:

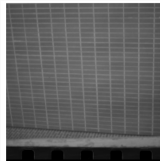
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- ☒ Leopard
- ☐ Tiger
- ☐ Puma
- ☒ Jaguar
- ☐ Lion(ess)
- ☒ Cheetah

(a) Animals



- ☒ Gravel
- ☐ Grass
- ☒ Brick
- ☒ Wood
- ☐ Sand
- ☐ Cloth

(b) Textures



- ☐ Hebrew
- ☐ Russian
- ☒ Japanese
- ☒ Thai
- ☒ Chinese
- ☐ Tamil
- ☐ Latin
- ☐ Hindi

(c) Languages

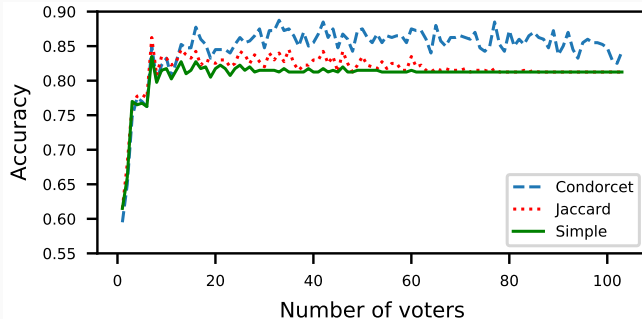


Figure 3: Guessing Animals from Images

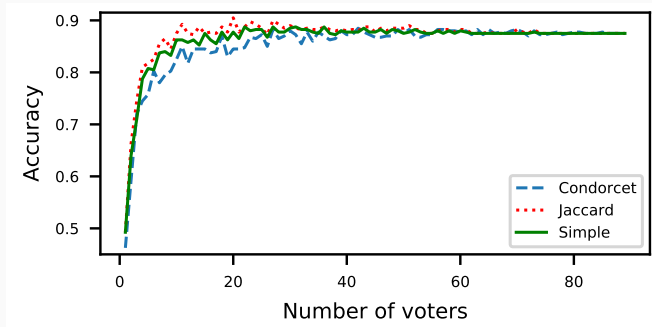


Figure 4: Guessing Textures from Images

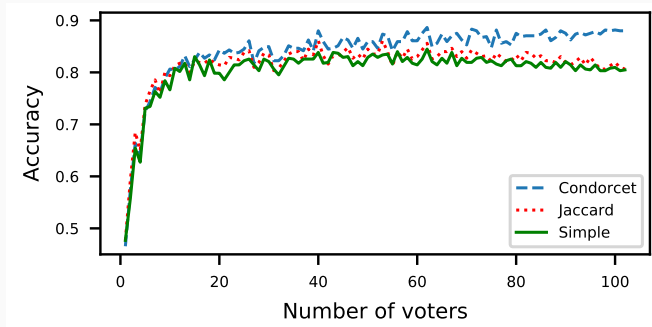


Figure 5: Guessing Languages from Images

CONCLUSION

MAIN IDEAS

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- **Homogeneous reliability:** We characterize all the functions d whose associated optimal rule is size-decreasing and we give the optimal weights.

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- **Condorcet noise:** In case d is the Hamming distance we give an explicit formula to estimate the reliability of a voter from her ballot's size.
- **Experiments:** We tested different size-decreasing rules on different image annotation datasets.

THANKS !



Shah, N., Zhou, D., and Peres, Y. (2015).

Approval voting and incentives in crowdsourcing.

In *ICML*.